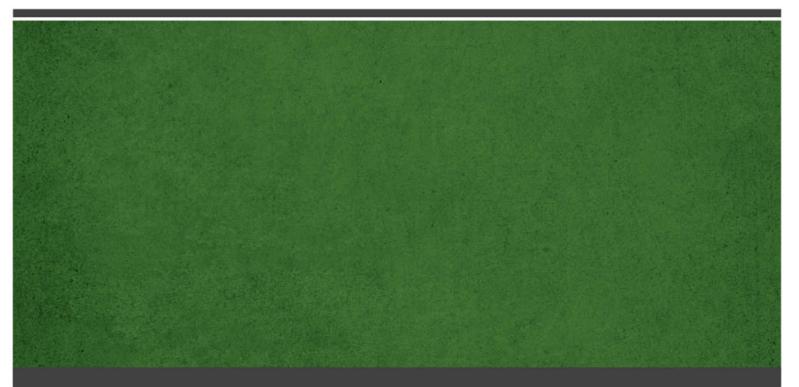
3	National Advisory Committee on Immunization (NACI)
5 6 7	Guidelines for the Economic Evaluation of Vaccination Programs in Canada
	Draft 1 st Edition For Public Consultation 2022



PROTECTING AND EMPOWERING CANADIANS TO IMPROVE THEIR HEALTH





1 Table of Contents

2

3	TABLE OF CONTENTS	2
4	Acknowledgements	3
5	NACI Economic Guidelines Task Group Members and Liaisons	3
6	Reviewers	6
7	Contributors	8
8 9	Conflicts of Interest Declaration Abbreviations	
10	Glossary	
11	Introduction	
12	Guideline Statements	
13	Guidelines in Detail	31
14	1. Decision Problem	31
15	2. Types of Evaluations	34
16	3. Study Populations	
17	4. Comparators	
18	5. Perspectives	40
19	6. Time Horizon	49
20	7. Discounting	52
21	8. Modelling	55
22	9. Effectiveness	64
23	10. Measurement and Valuation of Health	70
24	11. Resource Use and Costs	75
25	12. Analysis	

2	13. Uncertainty	90
3	14. Equity	
4	15. Reporting	
5	Appendix 1: Impact Inventory Table	
6	Appendix 2: Reference Case	
7	References	
8		

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1 Conflicts of Interest Declaration

2 3 4 5	As part of standard procedures for identifying and addressing affiliations and interests, Economic Guidelines Task Group members completed individual disclosure forms which were assessed by PHAC to ensure no undue influence or perceived conflict of interest.
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1 Abbreviations

2	AQoL	Assessment of Quality of Life
3 4	CADTH	Canadian Agency for Drugs and Technologies in Health
5 6 7 8 9 10 11 12	СВА	Cost-benefit analysis
	CEAC	Cost-effectiveness acceptability curve
	CEAF	Cost-effectiveness acceptability frontier
	CHU9D	Child Health Utility 9-Dimensions
13 14	CoP	Correlate of protection
15 16	COVID-19	Coronavirus infection disease 2019
17 18 19 20	CUA	Cost-utility analysis
	DSA	Deterministic Sensitivity Analysis
21 22	EEFA	Ethics, equity, feasibility, and acceptability
23 24	EQ-5D	EuroQol 5-Dimensions questionnaire
25 26	EQ-5D-Y	EuroQol 5-Dimensions questionnaire youth
27 28	Hib	Haemophilus influenzae type b
29 30	HIV	Human Immunodeficiency Virus
31 32	HPV	Human papilloma virus
33 34	HRQoL	Health-related quality of life
35 36	HUI	Health Utilities Index
37 38	ICER	Incremental cost-effectiveness ratio
39 40	NACI	National Advisory Committee on Immunization
41 42	PedsQL	Pediatric Quality of Life Inventory
43 44 45 46 47 48 40	PCV	pneumococcal conjugate vaccine
	PSA	probabilistic sensitivity analysis
	QALY	Quality-adjusted life year
49 50	RCT	Randomized controlled trial

- 1 2 3 4
 - SF-6D Short Form 6-Dimensions
- 4 TB Tuberculosis

DRAFT

1 Glossary

Age-shifting of infection: A phenomenon that can occur when a particular age cohort
 of the population is vaccinated, which reduces the force of infection in that cohort and
 shifts the average age at infection.

Agent-based model: A type of dynamic microsimulation model that allows individuals to
 act autonomously based on defined behavioural rules. See definition for individual-based
 model.

9

5

Basic reproduction number: The average number of secondary cases infected by an
 infectious person in a completely susceptible population

12

13 Catch-up strategy: A strategy of vaccinating individuals who did not receive a particular 14 vaccination at the recommended age. This strategy may be used in individuals who have 15 not been previously eligible for vaccination, who have missed a scheduled vaccine dose, 16 or who have not completed a vaccine series. 17

18 Coverage: The estimated percentage of eligible individuals who have received a
 19 particular vaccine.

20

Canadian Agency for Drugs and Technologies in Health (CADTH): Independent,
 not-for-profit organization tasked with providing Canada's health care decision-makers
 with evidence related to the optimal use of drugs and medical devices within the public
 health care system.

25

Closed (population) model: A model that follows a fixed cohort or cohorts of
 individuals. Individuals are not able to enter or exit the model through births, deaths or
 immigration over time.

29

30 Community Immunity: A state where a sufficient proportion of a population is immune 31 to an infectious disease, either from vaccination or prior infection, thereby preventing 32 outbreaks from occurring and making spread between individuals less likely. The term is 33 commonly used to refer to the indirect protection unvaccinated individuals receive due to 34 the presence of immune individuals in a population. This term is also referred to as herd 35 immunity.

37 Confounding bias: A distortion in the estimate of the relationship between an exposure
38 and an outcome in a study, resulting from a third variable, the confounder, which is
39 related both to the exposure and the outcome.

40

41 **Consumption:** The value of goods and service bought by individuals.

42

43 **Continuous (time) model:** A model in which events can occur at any point in time. 44

45 Correlate of protection (CoP): An immune biomarker that predicts vaccine efficacy in
 46 vaccinated individuals and can be used as a surrogate endpoint in studies of vaccine
 47 efficacy or effectiveness.

48

1 **Cost-effectiveness acceptability curve (CEAC):** A graphic summary of the uncertainty 2 in results of an economic evaluation, where a range of cost-effectiveness thresholds is 3 plotted against the probability thatan intervention is cost-effective. 4 5 Cost-effectiveness acceptability frontier (CEAF): A graphical summary of the 6 uncertainty in results of an economic evaluation, which indicates the strategy that is 7 economically preferred at different threshold values for cost-effectiveness and the 8 probability of that strategy being cost-effective. As the threshold increases the 9 economically preferred treatment may change, the switch point being where the 10 threshold value increases beyond the relevant ICER reported for the intervention of 11 interest. CEAFs are most useful when three or more alternatives are being compared, in 12 which case there may be two or more switch points at different threshold values. 13 14 Cost-benefit analysis (CBA): An economic evaluation in which both costs and 15 outcomes are expressed in monetary terms. 16 17 Cost-effectiveness analysis (CEA): An economic evaluation in which health outcomes 18 are expressed in natural units (e.g., infections avoided). 19 20 Cost-utility analysis (CUA): An economic evaluation in which health outcomes are 21 expressed in quality-adjusted life years (or other generic measure of health-related 22 utility). It is sometimes referred to as a cost-effectiveness analysis (CEA), or CEA with 23 QALYs. This is the form of economic evaluation favoured by public health care decision-24 makers in Canada. 25 26 **Decision problem:** An explicit statement of the interventions, study populations, 27 outcome measures, and perspective adopted in an economic evaluation, related 28 specifically to the decision(s) that the evaluation is designed to inform. 29 30 **Deterministic model:** A model that describes what happens on average and in which 31 events cannot occur randomly (by chance). For a defined set of parameters and starting 32 conditions these models will always generate the same results each time they are run. 33 34 Deterministic sensitivity analysis (DSA): A method used to explore uncertainty in 35 results of a model-based economic evaluation, where one or more parameters are 36 changed across a pre-specified range while holding the remaining parameters fixed to 37 determine the extent to which the parameter values impact the results of the analysis. 38 39 **Discount rate:** Costs and health outcomes occurring in the future are generally 40 considered to be valued less than those occurring presently, and so they are discounted 41 in an economic evaluation to ascertain their present day value. The factor by which costs 42 and health outcomes are discounted is expressed as the discount rate. 43 44 **Discrete (time) model:** A model in which events can only occur at pre-specified points 45 in time. 46 47 **Disease control:** The state in which incidence, prevalence, morbidity, or mortality of a 48 particular disease has been reduced locally but continued efforts are required to 49 maintain this reduction. 50

- 1 **Disease elimination:** Local incidence of a particular infection has been reduced to a 2 level below that necessary to sustain ongoing transmission in a given geographic area, 3 but continued efforts are required to maintain this reduction. 4 5 **Disease eradication:** The incidence of an infection has been permanently reduced to 6 zero worldwide, the causative organism is no longer present in the wild and efforts are 7 no longer required to maintain this reduction. 8 9 **Distributional cost-effectiveness analysis:** An extension to the conventional CEA 10 framework that quantifies the distributional impacts of health interventions based on 11 different equity criteria such as socioeconomic status or disease severity. 12 13 Dominance: Refers to a scenario in which a strategy results in greater benefits and 14 fewer costs compared to its alternative 15 **Dose completion:** The accumulation of the required number of doses of a vaccination 16 regimen during a specified time period. 17 18 Dynamic (transmission) model: A model in which the force of infection can vary over 19 time. Incidence is a function of the number (or proportion) of infected and susceptible 20 individuals and the transmissibility of the virus. May also be referred to as a model with 21 an endogenous force of infection. 22 23 24 Effectiveness: The extent to which an intervention provides the desired outcome(s) in 25 the relevant study population in a real-world setting. 26 27 Effective reproduction number: The average number of secondary cases infected by 28 an infectious person in a population where some of the individuals are immune due to 29 vaccination or infection. 30 31 Efficacy: The benefit of an intervention produced in an experimental and controlled 32 setting, such as in a randomized controlled trial (RCT). 33 34 Efficiency frontier: A graphical summary of cost-effectiveness results comparing 35 multiple interventions. The plot compares the effect on the y-axis and the costs on the x-36 axis. The frontier links the interventions that are not dominated. An intervention located 37 on or below the frontier can be considered reasonably efficient. 38 39 **Epidemiologic equilibrium:** A situation where the rate of new infections circulating in a 40 population is equal to the rate of recovery from the infection, resulting in a stable or 41 unchanging state. 42 43 Expected value of perfect information (EVPI): The maximum price that a decision-44 maker would be willing to pay to have perfect information regarding all parameter values 45 that influence which intervention is preferred based on results of a CEA. This represents 46 the value (in monetary terms) of removing all uncertainty about the parameters in the 47 analysis. EVPI can also be expressed for the total population who stand to benefit over 48 the expected lifetime of the intervention (known as population EVPI).
- 49

1 Expected value of partial perfect information (EVPPI): The maximum price that a 2 decision-maker would be willing to spend in order to gain perfect information for one or 3 more inputs to an economic model. 4 5 **Extended dominance:** A scenario where a strategy can be excluded when it costs more 6 and provides fewer benefits than a combination of two other alternatives. 7 8 Extended cost-effectiveness analysis: An extension to the conventional CEA 9 framework that guantifies the distributional impacts of health interventions based on 10 equity criteria as well as financial risk protection. 11 12 **Externalities:** Costs and consequences of an intervention such as a vaccination 13 program that fall on other members of the population beyond those producing, 14 purchasing or consuming the intervention (e.g., community immunity, age-shifting of 15 disease). 16 17 Equity: The absence of unfair and avoidable or remediable differences in health among 18 population groups defined by any relevant characteristic (e.g., medical, social, economic, 19 demographic, geographic). Horizontal equity refers to individuals with like characteristics 20 (of ethical relevance) being treated the same way, while vertical equity allows for 21 individuals with different characteristics (of ethical relevance) to be treated differently in 22 order to achieve more equitable outcomes. 23 24 **First-order uncertainty:** Uncertainty related to random variability. This type of 25 uncertainty is also referred to as stochastic uncertainty. 26 27 Force of infection: The rate at which susceptible individuals become infected per unit 28 time. It is a function of the number of infectious individuals in the population at a given 29 time and the transmissibility of the infection. 30 31 Health technology assessment (HTA): The multi-disciplinary evaluation of various 32 domains of a health technology in order to inform its use, which may include clinical 33 effectiveness, cost-effectiveness, social impacts, ethical impacts, among others. 34 35 Health equity: See definition for Equity. 36 37 Health-related quality of life (HRQoL): A combination of a person's physical, mental 38 and social functioning. 39 40 **Health utility:** A measure of health-related quality of life that represents preference 41 values that individuals attach to their overall health status. Conventionally the valuations 42 are anchored by 0 (representing a health state equivalent to being dead) and 1 43 (representing a health state equivalent to perfect health). Health utilities are also referred 44 to as preference-based measures of health-related quality of life. 45 46 Herd immunity: See definition for community immunity. 47 48 **Heterogeneity:** Differences between individuals that can, in part, be explained. This 49 differs from the random chance that individuals with the same underlying characteristics 50 will experience a different outcome. 51

- 1 **Incremental costs:** Difference in mean expected costs associated with the use of an 2 intervention compared with the use of an alternative. This is a key output of an economic 3 evaluation. 4 5 **Incremental cost-effectiveness ratio (ICER):** A ratio that is calculated by dividing the difference in mean expected costs by the difference in mean expected health outcomes 6 7 or effects between two alternatives being compared in an economic evaluation. The 8 comparator usually represents the current standard of care. 9 10 Incubation period: The time from infection to onset of clinical disease. 11 12 **Indigenous Peoples:** The earliest known people groups of any land around the world. 13 For the purposes of these guidelines, the term 'Indigenous Peoples' refers to individuals 14 who are First Nations, Inuit, and Métis. 15 16 Infectious period: The time from the end of latent or pre-infectious period until the host 17 is no longer able to transmit the infection to other individuals. 18 19 Individual-based model (or Microsimulation): A model in which the individual, rather 20 than the group, is the unit that is modelled. Microsimulation models that do not allow for 21 interactions among individuals are classified as static microsimulation models. 22 Microsimulation models that do allow for interactions among individuals or with the 23 environment (such as the healthcare system) are classified as dynamic microsimulation 24 models. An agent-based model is a type of dynamic simulation model 25 26 Intergenerational equity: The concept that people of different generations should 27 benefit equitably from policy decisions such as expenditures on vaccination programs. 28 29 Latent period: The time period from when a host acquires an infection until they are 30 able to transmit it to another host. It is sometimes referred to as the pre-infectious 31 period. 32 33 Methodologic uncertainty: Uncertainty related to the different methods that can be 34 used to conduct an economic evaluation.
- 36 **Microsimulation:** See definition for individual-based model.

35

- 37 38 National Advisory Committee on Immunization (NACI): A national advisory 39 committee of experts in the fields of pediatrics, infectious diseases, immunology, 40 pharmacy, nursing, epidemiology, pharmacoeconomics, social science and public 41 health. NACI makes recommendations for the use of vaccines currently or newly 42 approved for use in humans in Canada, including the identification of groups at risk for 43 vaccine-preventable diseases for whom vaccination should be targeted. 44 45 **Net health benefit:** A summary statistic, expressed in QALYs, that represents the 46 impact on population health related to a given intervention, adjusted for the expected 47 costs if purchasing care at the rate of a marginally cost-effective strategy. It is calculated 48 by subtracting the ratio of the expected costs by the health opportunity cost. 49 **Net monetary benefit:** A summary statistic that represents the value of an intervention 50 as the impact on population health, expressed in monetary terms, adjusted for the
- 51 expected costs if purchasing care at the rate of a marginally cost-effective strategy. It is

1 calculated by multiplying the expected QALYs by the health opportunity cost and 2 subtracting the expected costs associated with the intervention. 3 4 **Open (population) model:** A model that allows new individuals to enter the model 5 either through births or in-migration, or to exit the model through deaths or out-migration 6 over time. 7 8 **Parameters:** Variables that determine the rates of movement between model states or 9 probabilities of events within a model. 10 Parameter uncertainty: Uncertainty in parameter estimates that are used to populate a 11 model. This type of uncertainty is also referred to as second-order uncertainty. 12 13 **Pathogen variations:** Differences between strains of a pathogen related to serotypes, 14 serogroups, or genotypes. 15 16 Perspective: The viewpoint from which an economic evaluation will be conducted. The 17 perspective determines the outcomes and costs that will be included in the analysis. 18 19 **Population-based model:** A model in which groups of individuals are assigned to 20 compartments or health states based on their health status or other characteristics. 21 Individuals in each compartment move according to parameter values defined at the 22 aggregate level and the model records the number of individuals in each compartment 23 over time. This type of model can also be referred to as an aggregate model. 24 25 **Positive time preference:** The preference for present benefits over benefits occurring in 26 the future. 27 28 **Probabilistic analysis:** A method used to quantify parameter uncertainty in an 29 economic analysis where a probability distribution is assigned to each uncertain 30 parameter and values are randomly sampled from each distribution repeatedly to 31 generate a distribution of outcomes that can be analyzed 32 33 **Probabilistic sensitivity analysis (PSA):** See definition for probabilistic analysis. 34 35 **Productivity:** A measure of how efficiently production inputs, such as labour and capital, 36 are being used in an economy to produce a given level of output. 37 38 Randomized controlled trial (RCT): A comparative study, designed to ascertain the 39 efficacy of a health intervention, in which units such as individuals are randomly 40 assigned to either the intervention or control group. 41 42 **Real-world evidence:** Evidence used for decision-making that is collected through non-43 experimental studies. 44 45 **Reference case:** A set of methods for conducting an economic evaluation specified by 46 the decision maker. The purpose of the reference case analysis is to ensure consistency 47 between methods underpinning analyses and the decision-making process that is 48 employed. 49 50 Quality-adjusted life year (QALY): A summary outcome measure used to quantify the 51 health outcomes associated with a particular intervention. QALYs combine the impact of

- 1 benefits related to both survival and health-related quality of life expressed as health 2 utilities, and allow comparisons between interventions across disease states. 3 4 Scenario analysis: An analysis that tests alternate model scenarios underpinned by 5 different plausible structural assumptions. 6 7 Second-order uncertainty: See definition for parameter uncertainty. 8 9 Selection bias: Bias in a non-randomized study resulting from systematic differences in 10 sampling of individuals who are in the exposed group compared to those who are not. 11 resulting in a distribution of exposures and outcomes that is no longer representative of 12 the source population. Selection bias can also occur in randomized studies due to 13 attrition post-randomization. 14 15 **Serotype replacement:** The expansion in non-vaccine serotypes of a pathogen 16 resulting from the removal of vaccine-specific serotypes from the population that 17 compete with them for colonisation of hosts. 18 19 **Spillover effects:** The effects of conditions and treatments on different aspects of the 20 welfare of other individuals such as family members, including caregiver health effects, 21 informal care time costs, or both, 22 23 Static model: A model in which the force of infection is constant over time or dependent 24 only on characteristics of each individual, and not on the number of other individuals who 25 are infectious. May also be referred to as a model with an exogenous force of infection. 26 **Stochastic model:** For the purposes of these guidelines, a model that accounts for first-27 order uncertainty where events are programmed to occur randomly. 28 29 Structural uncertainty: Uncertainty related to the structure of a model and other 30 unparameterized sources of uncertainty. Scenario analysis is one approach for 31 assessing this type of uncertainty. 32 33 34 **Time horizon:** The time period over which outcomes and costs are quantified in an 35 economic evaluation. 36 37 Value of information analysis: An analysis used to estimate of the value, in terms of 38 cost and health outcomes, of collecting more data on key parameters influencing a 39 funding decision. It is most useful where the output of an economic evaluation is 40 uncertain, but close to a decision threshold and a key parameter on which the output is
- 41 based is uncertain. See definitions for Expected value of perfect information (EVPI) and 42 Expected value of partial perfect information (EV/PPI)
- 42 Expected value of partial perfect information (EVPPI).

Introduction 1

2 This is the first edition of the National Advisory Committee on Immunization (NACI) 3 Guidelines on the Economic Evaluation of Vaccination Programs in Canada; in the text 4 hereafter, they will be referred to as the Guidelines. These Guidelines have been 5 established to articulate best practices for conducting and reporting economic 6 evaluations of vaccination programs in Canada, be they regional, provincial, or national. 7 Adherence to common best practices can allow decision-makers in Canada's publicly 8 funded health system to have access to consistent and credible information to inform 9 funding decisions related to vaccination programs. These Guidelines focus on 10 information specific to the vaccination programs. The Canadian Agency for Drugs and 11 Technologies in Health's Guidelines for the Economic Evaluation of Health 12 Technologies: Canada¹ present more general information applicable to health 13 technologies in Canada, and where CADTH's Guideline Statements are applicable to 14 economic evaluations of vaccination programs, they have been included in these Guidelines.

- 15
- 16

17 The main feature that distinguishes vaccination programs from other health technologies 18 is their population-level effects, which result from their potential to affect both vaccinated

19 and unvaccinated individuals. These Guidelines present detailed information on how to

20 incorporate these population-level effects into economic evaluations of vaccination

21 programs, including methods for addressing their non-health sector impacts.

22

23 The recommendations contained in these Guidelines were formulated by NACI's

24 Economic Guidelines Task Group, which consisted of Canadian and international

25 experts in infectious diseases and health economics. This group engaged in a series of

26 discussions that led to decisions made by consensus, which were supported by literature

27 reviews for selected topics. A peer-review and public consultation process was

28 undertaken after completing an initial draft of the Guidelines and informed the final

29 recommendations and text contained in this document.

30

31 A social decision-making framework has been adopted for these Guidelines. The basis

32 of this framework is that the health decision-maker acts on behalf of a socially legitimate

33 higher authority (e.g., a democratically elected government) to achieve an explicit policy

34 objective (e.g., improving overall population health). The function of an economic

1 evaluation within this framework is to inform social decisions.²⁻⁴ 2 3 Economic evaluations have traditionally focused on the trade-offs between incremental 4 costs and incremental effects of different health interventions to enable decision-makers 5 to make judgments related to efficiency. These Guidelines expand on the traditional 6 approach by presenting recommendations on integrating equity considerations into 7 economic evaluations of vaccination programs. The integration of equity in economic 8 evaluations is consistent with NACI's Ethics, Equity, Feasibility, and Acceptability 9 (EEFA) Framework, which provides a mechanism for decision-makers to systematically 10 consider important programmatic factors, alongside effectiveness and cost-11 effectiveness, when making recommendations about vaccination programs.⁵ 12 13 These Guidelines recommend adoption of two reference case analyses for the economic 14 evaluation of vaccination programs: one conducted from the publicly funded health 15 system perspective and the other conducted from the societal perspective. The latter is 16 to account for the full range of benefits associated with vaccination programs, including 17 those that accrue to non-health sectors. The purpose of these reference cases is to 18 encourage the use of a standard set of methods when conducting economic evaluations 19 of vaccination programs and to ensure that decision-makers are able to compare results 20 between different vaccination programs. 21 22 Recommendations are presented for the following aspects of economic evaluations of 23 vaccination programs: decision problem, types of evaluations, study populations, 24 comparators, perspectives, time horizon, discounting, modelling, effectiveness, 25 measurement and valuation of health, resource use and costs, analysis, uncertainty, 26 equity, and reporting. Guidance on each of these topics is contained in a separate 27 chapter. Guidelines Statements are presented at the beginning of this document and at 28 the beginning of each chapter for ease of use, followed by a detailed discussion of the 29 recommendations. The Guidelines are written for end-users, including researchers and 30 decision-makers, who are technically proficient in the methods of economic evaluation, 31 and as such, background on these methods has been omitted. Similarly, the Guidelines 32 omit detailed background information on scientific and technical subjects related to 33 vaccines and immunization, as it is expected that researchers undertaking economic 34 evaluations of vaccination programs will consult with subject matter experts in this area.

1 The references contained in this document provide sources for researchers to obtain

- 2 additional information when required.
- 3

4 The guidance presented in this document represents NACI's current recommendations 5 for conducting economic evaluations of vaccination programs. NACI and the Economic 6 Guidelines Task Group have attempted to reflect current best practices, but the 7 recommendations contained in these Guidelines will evolve alongside scientific and 8 methodological advancements in this area. Topics for which there is no current 9 consensus on best practices and require further research have been identified in the 10 Guidelines. As such, the function of these Guidelines is not only to recommend current 11 practices for the economic evaluation of vaccination programs, but also to suggest 12 directions for future research and that will contribute to advancing methods used in this 13 area. 14 15 The remit of these Guidelines is to specify methods for conducting economic evaluations

16 of vaccination programs, and not to provide guidance or insights into the decision-

- 17 making process. As such, considerations or factors related to making funding decisions
- 18 about vaccination programs are not included in the Guidelines.

1	Guid	leline Statements
2	1.	Decision Problem
3	1.1	"The decision problem addressed by the economic evaluation should be clearly
4		stated." [CADTH Guideline Statement]
5		
6	1.2	"The decision problem statement should provide a comprehensive specification
7		of the interventions to be compared, the setting(s) in which they are to be
8		compared, the perspective of the evaluation, which costs and outcomes are to be
9		considered, the time horizon, and the intended population for the evaluation."
10		[CADTH Guideline Statement]
11		
12	1.3	A separate decision problem statement is required for each perspective and for
13		each analysis related to a distinct population group for which the vaccination
14		program may be intended.
15		
16	1.4	In addition to specifying the intended population for the vaccination program, the
17		decision problem must also identify other population groups that could be
18		affected by the vaccination program, including the population at risk for the
19		disease of interest, and any populations that may be indirectly affected by the
20		vaccination program, either through externalities or spillover effects.
21		
22	2.	Types of Evaluations
23	2.1	In the reference cases, the economic evaluation should be cost-utility analyses
24		(CUA) with outcomes expressed as quality-adjusted life-years (QALYs). Any
25		departure from this approach should be clearly justified. [CADTH Guideline
26		Statement with amendment]
27		
28	2.2	A cost-benefit analysis (CBA) may be used alongside the reference case CUAs in
29		situations where the vaccination program may be compared to a non-health
30		intervention.
31		
32	3.	Study Populations
33	3.1	Researchers should identify the intended population(s) for the vaccination
34		program, the population at risk for the disease of interest, and any populations

		24 Guidelines for the Economic Evaluation of Vaccination Programs in Canada: Public Consultation
1		that may be indirectly affected by the vaccination program, either through
2		externalities or spillover effects.
3		
4	3.2	Researchers should present an overall analysis that includes the costs and
5		health outcomes for all of the affected populations. When relevant, researchers
6		should also summarize the results separately for each affected group (e.g.,
7		intended population, population experiencing externalities or spillover effects)
8		that was included in the overall analysis.
9		
10		
11	3.3	Where there are factors that could lead to differences in costs and outcomes
12		related to the vaccine program across subgroups, researchers should conduct
13		separate economic evaluations for each subgroup. These factors could include
14		demographic factors, behavioural factors, disease-related factors, and
15		effectiveness of the vaccine or comparator intervention(s).
16		
17	4.	Comparators
18	4.1	The choice of comparator(s) should be related to the scope of the decision
19		problem. As such, the comparators should reflect the intended population for the
20		vaccination program and the jurisdiction for which the decision is being made.
21		[CADTH Guideline Statement with amendment]
22		
23	4.2	Researchers should consider both preventive and treatment-based approaches
24		when selecting comparators for economic evaluations of vaccination programs.
25		Preventive interventions could include vaccine-based measures, screening
26		programs, preventive medication-based interventions, and preventive non-
27		medical interventions.
28		
29	5.	Perspectives
30	5.1	Two reference case analyses should be presented as part of the economic
31		evaluation of vaccination programs: one conducted from the publicly funded
32		health system perspective, and the other conducted from the societal
33		perspective.
34		

		25 Guidelines for the Economic Evaluation of Vaccination Programs in Canada: Public Consultation
1	5.2	"Both costs and outcomes should be consistent with the stated perspective."
2		[CADTH Guideline Statement]
3		
4	6.	Time Horizon
5	6.1	In the reference cases, the time horizon should be long enough to capture all
6		relevant differences in the future costs and outcomes associated with the
7		interventions being compared. Thus, the time horizon should be based on the
8		condition and the likely impact of the intervention. [CADTH Guideline Statement
9		with amendment]
10		
11	6.2	Researchers should justify their choice of time horizon. Where it spans a long
12		period of time (i.e., multiple decades), researchers should report ICER estimates
13		from various time points throughout the time horizon.
14		
15	7.	Discounting
16	7.1	In the reference cases, costs and outcomes that occur beyond one year should
17		be discounted to present values at a rate of 1.5% per year. [CADTH Guideline
18		Statement with amendment]
19		
20	7.2	"The impact of uncertainty in the discount rate should be assessed by comparing
21		the results of the reference cases to those from non-reference case analyses,
22		using discount rates of 0% and 3% per year." [CADTH Guideline Statement]
23		
24	8.	Modelling
25	8.1	"Model conceptualization and development should address the decision
26		problem." [CADTH Guideline Statement]
27		
28	8.2	"Researchers should consider any existing well-constructed and validated models
29		that appropriately capture the clinical or care pathway for the condition of interest
30		when conceptualizing their model." [CADTH Guideline Statement]
31		
32	8.3	The model structure should reflect the natural history of disease, the clinical or
33		care pathway, and account for susceptibility, infectiousness, and immunity,
34		related to the infection.

1		
2	8.4	Relevant behavioural dynamics including contact patterns between individuals
3		and behaviours related to infection prevention and control should be incorporated
4		into the model where appropriate.
5		
6	8.5	Dynamic models should be considered in economic evaluations of vaccines that
7		are associated with externalities such as prevention of human-to-human
8		transmission of infection and age-shifting of disease.
9 10	8.6	Other model attributes including whether the model is deterministic or stochastic,
11		population-based or individual-based, and open or closed should be considered
12		in the context of the decision problem.
13		
14	8.7	Researchers should transparently report on model calibration and validation
15		processes that were undertaken and on their results.
16		
17	9.	Effectiveness
18	9.1	"A comprehensive search of the available data sources should be conducted to
19		inform the estimates of effectiveness and harms associated with the
20		interventions. Report the included studies and methods used to select or
21		combine the data." [CADTH Guideline Statement]
22		
23	9.2	"The data sources should be assessed based on their fitness for purpose,
24		credibility, and consistency. Describe the trade-offs among these criteria and
25		provide justification for the selected source(s)." [CADTH Guideline Statement
26		with amendment]
27 28	9.3	The following criteria should be considered when assessing estimates of vaccine
29		effectiveness: vaccine effectiveness by dose; expected vaccine coverage;
30		pathogen variation-specific (i.e., serotypes, serogroups, strains) effectiveness;
31		and geographic and host factors that may affect effectiveness.
32		
33	9.4	Researchers should ensure that immune biomarkers used as surrogate
34		outcomes in studies of vaccine efficacy or effectiveness meet the criteria for
35		correlates of protection.

1		
2	10.	Measurement and Valuation of Health
3	10.1	In both reference cases, the quality-adjusted life year (QALY) should be used as
4		the method for valuing health outcomes.
5		
6	10.2	"Health preferences should reflect the general Canadian population." [CADTH
7		Guideline Statement]
8		
9	10.3	In the reference cases, researchers should use health preferences obtained from
10		an indirect method of measurement that is based on a generic classification
11		system (e.g., EuroQol 5-Dimensions questionnaire [EQ-5D], Health Utilities Index
12		[HUI], Short Form 6-Dimensions [SF-6D], Child Health Utility 9-Dimensions
13		[CHU9D], Assessment of Quality of Life [AQoL]). Researchers must justify where
14		an indirect method is not used. [CADTH Guideline Statement with amendment]
15		
16	10.4	"The selection of data sources for health state utility values should be based on
17		their fitness for purpose, credibility, and consistency. Describe the trade-offs
18		among these criteria and provide justification for the selected sources." [CADTH
19		Guideline Statement]
20		
21	11.	Resource Use and Costs
22	11.1	For each reference case analysis, researchers should systematically identify,
23		measure, value, and report all relevant resources consumed or saved as a result
24		of the delivery or implementation of the vaccination program under consideration.
25		
26	11.2	Where possible, researchers should value relevant resources identified for all
27		sectors in monetary terms. In situations where this is not possible, researchers
28		should present the relevant resources that have been identified in the Impact
29		inventory table for economic evaluations of vaccination strategies for
30		consideration by decision-makers.
31	11.3	"Resource use and costs should be based on Canadian sources and reflect the
32		jurisdiction(s) of interest (as specified in the decision problem)." [CADTH
33		Guideline Statement]

34

		28 Guidelines for the Economic Evaluation of Vaccination Programs in Canada: Public Consultation
1	11.4	When valuing and monetizing resources, researchers should select data sources
2		that most closely reflect the opportunity cost, given the perspective of the
3		analysis. [CADTH Guideline Statement with amendment]
4		
5	11.5	Researchers should assess sources used for cost data based on their fitness for
6		purpose, credibility, and consistency. The selection of data sources should be
7		based on trade-offs between these criteria.
8		
9	12.	Analysis
10	12.1	Incremental cost-effectiveness ratios (ICERs) and, where useful for interpretation,
11		net monetary benefits or net health benefits, should be calculated for both
12		reference case analyses.
13		
14	12.2	"For analyses with more than two interventions, a sequential analysis of cost-
15		effectiveness should be conducted following standard rules for estimating ICERs,
16		including the exclusion of dominated interventions." [CADTH Guideline
17		Statement]
18 19	12.3	The expected values of costs and outcomes, where possible, should be
20		generated probabilistically to reflect the overall uncertainty in the model
21		parameters.
22		
23	13.	Uncertainty
24	13.1	Researchers should address parameter uncertainty using a probabilistic
25		reference case analysis, where possible, as well as deterministic sensitivity
26		analyses.
27		
28	13.2	"Methodological uncertainty should be explored by comparing the reference case
29		results to those from a non-reference case analysis that deviates from the
30		recommended methods in order to examine the impact of methodological
31		differences." [CADTH Guideline Statement]
32 33	13.3	Cost-effectiveness acceptability curves (CEACs) and cost-effectiveness
33 34	13.3	acceptability frontiers (CEAFs) should be used to represent the uncertainty in the
54		

		29 Guidelines for the Economic Evaluation of Vaccination Programs in Canada: Public Consultation
1		estimates of costs and outcomes when these estimates have been generated
2		probabilistically. [CADTH Guideline Statement with amendment]
3	40.4	
4 5	13.4	When the decision problem includes the option of commissioning or conducting
6		future research, value-of-information analysis may be helpful to characterize the value of these options and design future research and may be included in the
7		reference case analysis. [CADTH Guideline Statement with amendment]
8		
9	13.5	Scenario analyses should be used to assess structural uncertainty. [CADTH
10		Guideline Statement with amendment]
11		
12	14.	Equity
13	14.1	Researchers and decision-makers should work together to establish which equity
14		dimensions and goals should be included in the economic evaluation of the
15		vaccination program being considered. Equity should be considered in the
16 17		context of NACI's Ethics, Equity, Feasibility, and Acceptability (EEFA) framework.
18	14.2	Analyses that incorporate relevant equity concerns should accompany the
19		reference case analysis (e.g., distributional cost-effectiveness analysis, extended
20		cost-effectiveness analysis, or other emerging methods), and presented
21		alongside the reference case.
22		
23	15.	Reporting
24	15.1	"The economic evaluation should be reported in a transparent and detailed
25		manner with enough information to enable the reader or user (e.g., decision-
26		maker) to critically assess the evaluation. Use a well-structured reporting format."
27		[CADTH Guideline Statement]
28	45.0	
29 20	15.2	"A summary of the evaluation written in non-technical language should be
30 21		included." [CADTH Guideline Statement]
31 32		
33	15.3	"Results of the economic evaluation should be presented in graphical or visual
34		form, in addition to tabular presentation." [CADTH Guideline Statement]
35		

1	15.4	"Details and/ or documents describing quality assurance processes and results
2		for the economic evaluation should be provided. An electronic copy of the model
3		should be made available for review with accompanying documentation in
4		adequate detail to facilitate understanding of the model, what it does, and how it
5		works." [CADTH Guideline Statement]
6		
7	15.5	"Funding and reporting relationships for the evaluation should be described, and
8		any conflicts of interest disclosed." [CADTH Guideline Statement]
9		
10	15.6	Researchers should use NACI's Guidelines for Reporting Economic Evaluations

of Vaccination Programs in Canada, and complete the *Impact inventory table for* economic evaluations of vaccination strategies, which is found in Appendix 1.

11 12

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1 Guidelines in Detail

2 1. Decision Problem

3 1.1 "The decision problem addressed by the economic evaluation should be clearly 4 stated." [CADTH Guideline Statement] 5 6 1.2 "The decision problem statement should provide a comprehensive specification 7 of the interventions to be compared, the setting(s) in which they are to be 8 compared, the perspective of the evaluation, which costs and outcomes are to be 9 considered, the time horizon, and the intended population for the evaluation." 10 [CADTH Guideline Statement] 11 12 1.3 A separate decision problem statement is required for each perspective and for 13 each analysis related to a distinct population group for which the vaccination 14 program may be intended. 15 16 1.4 In addition to specifying the intended population for the vaccination program, the 17 decision problem must also identify other population groups that could be affected by the vaccination program, including the population at risk for the 18 19 disease of interest, and any populations that may be indirectly affected by the 20 vaccination program, either through externalities or spillover effects. 21 22 The decision problem being addressed by an economic evaluation of a vaccination 23 program should address concerns relevant to decision-makers and be clearly articulated 24 at the outset of the analysis, while ensuring consistency with other vaccine evaluations 25 where possible. Decision-makers who assess economic evaluations of vaccination 26 programs in Canada include NACI, provincial/ territorial immunization technical advisory 27 groups, and provincial/ territorial Health Ministries. NACI develops non-binding, 28 evidence-informed recommendations to facilitate timely decision-making for publicly 29 funded vaccine programs at provincial and territorial levels. Some provinces and 30 territories have formal immunization technical advisory groups while others do not. 31 Formal advisory groups and Health Ministries make decisions on whether a vaccination 32 program will be funded in a given jurisdiction, and how it will be implemented. 33

1 The decision problem should provide a detailed description, and justification, of the 2 vaccination program being evaluated, including: 1) the perspectives from which the 3 analysis is being carried out; 2) the type of economic evaluation being conducted; 3) 4 which costs and outcomes will be quantified in the analysis; 4) the time horizon over 5 which the analysis will be carried out; 5) the comparators that will be considered; and 6) 6 the populations affected by the vaccination program. These populations include the 7 intended population for the vaccination program, and where applicable, the population at risk for the disease of interest, and the population that may experience spillover effects 8 9 (e.g., informal caregivers). 10

11 It should state all possible population subgroups that the decision-maker is considering

12 vaccinating (e.g., age groups, clinical risk groups, people in certain professions,

13 geographical areas, individuals who possess certain biomarkers or genetic profiles), as

14 well as the potential vaccine delivery setting (e.g., physician clinics, pharmacies,

15 schools, workplaces). All options of interest to the decision-maker should be evaluated

16 together using the principles of full incremental analysis.

17

18 Researchers should seek out and engage with decision-makers to gain an

19 understanding of the concerns they are intending to address with the introduction of the

20 vaccination program. Some aspects of the decision problem that are particularly relevant

21 to decision-makers include, but are not limited to: 1) the time horizon of the evaluation;

22 2) possible impacts of the vaccination program beyond the health sector; and 3) health

23 inequities that could potentially be affected by introduction of the vaccination program.

24

25 Ensuring a time horizon that is relevant to decision-makers is particularly important when 26 a vaccination program results in protection for unvaccinated individuals through

27 community immunity (as known as herd immunity) with the potential for disease

28 elimination. Often a very long time horizon (sometimes many decades) is required for

29 the full costs and effects of a vaccination program to become apparent. Researchers

30 should note that these long time horizons may not reflect present-day outcomes and

31 costs that are relevant to decision-makers. In these cases, researchers should ensure

32 that results of an economic evaluation are reported from several time points to allow

33 decision-makers to determine when payoffs of the program become positive.

34

1 Given the broader, non-health-related outcomes that are associated with many 2 vaccination strategies, a perspective broader than the health system perspective will 3 usually be relevant. In these guidelines, health system refers to both healthcare 4 treatment services and Public Health. Researchers should attempt to gain an 5 understanding of the broader costs and benefits related to the vaccination program that 6 may fall outside of the health sector, and that are relevant to the decision-maker. For 7 example, because measles can lead to neurologic damage, preventing measles through 8 childhood vaccination improves educational outcomes. Similarly, preventing influenza in 9 the population through a universal vaccination program leads to productivity-related 10 benefits. Further details on this topic are found in Chapter 5 on Perspectives. 11

12 Certain groups are vulnerable to infectious disease and the adverse impacts of 13 infectious disease control policies due to historical harms and socially constructed 14 barriers. Vaccines have been identified as a strategy to potentially reduce specific 15 inequities relating to risk of infection or burden of the disease in question. Researchers, 16 in collaboration with decision-makers, should identify specific groups that may especially 17 benefit from the vaccination program. For example, individuals of lower socioeconomic 18 status and those belonging to minorities experience a higher incidence of cervical cancer 19 and greater mortality related to the disease, which could be prevented through a human 20 papilloma virus (HPV) vaccination program.⁶ Conversely, researchers, in collaboration 21 with decision-makers, should also consider whether some groups may not benefit from 22 the vaccination program, thus potentially increasing health inequities. A further 23 discussion on this topic is found in Chapter 14 on Equity.

24

The type of economic evaluation should be specified and justified. The type of economic evaluation, along with the perspective, will determine which costs and outcomes should be included (and how). The included outcome measures, which should be the same for each comparator, should be explicitly stated in the decision problem and listed by sector (e.g., health outcomes, educational achievement). Similarly, the included costs should be explicitly stated and listed by cost category (e.g., healthcare costs, education-related costs, productivity-related costs).

32

A clear description of the vaccine being evaluated including the dosage of vaccine, the
 number of doses required, dose schedule, whether any booster doses are required,

1 2 expected dose completion, handling of vaccine wastage, assumptions on waning, 3 coverage estimates, and setting of vaccine delivery should be provided along with 4 detailed descriptions of comparators. Comparators could include other existing 5 preventive vaccines, non-vaccine-based preventive approaches, and current treatment 6 approaches including best supportive care. 7 2. Types of Evaluations 8 9 2.1 In the reference cases, the economic evaluation should be cost-utility analyses 10 (CUA) with outcomes expressed as quality-adjusted life-years (QALYs). Any 11 departure from this approach should be clearly justified. [CADTH Guideline 12 Statement with amendment] 13 14 2.2 A cost-benefit analysis (CBA) may be used alongside the reference case CUAs in 15 situations where the vaccination program may be compared to a non-health 16 intervention. 17 18 In the reference cases, the economic evaluation should be a cost-utility analysis (CUA) 19 with outcomes expressed as quality-adjusted life-years (QALYs). There is recognition, 20 however, that there are populations in whom CUAs cannot be robustly conducted 21 because valid instruments for direct utility elicitation do not exist, such as children under 22 8 years of age. In these cases, alternative analytic approaches such as cost-23 effectiveness analysis (CEA) with a relevant outcome measure in natural health units 24 should be justified. 25 26 In addition to the reference case CUAs, a cost-benefit analysis (CBA) may be presented 27 in cases where broader impacts beyond health are important factors for decision-28 makers. CBA has been proposed as a method to evaluate vaccination programs 29 associated with consequences that fall outside of the health sector.7-9 NACI's 30 recommendation to conduct a reference case CUA from the societal perspective should 31 enable researchers to account for non-health sector benefits by monetizing them, and 32 including them in the incremental costs and subsequently the numerator of the 33 incremental cost-effectiveness ratio (ICER) estimate. This approach, however, does not

- enable decision-makers to compare non-health benefits of alternative programs, or to
 compare vaccination programs to non-health programs since the denominator of the
 ICER estimate is reported in quality-adjusted life years (QALYs). In cases where a
 decision-maker may be interested in comparing the economic attractiveness of a
 vaccination program to a non-health intervention (e.g., school lunch program),
 researchers could present a CBA alongside the societal perspective reference case
- 7 analysis to enable such a comparison.⁷ Researchers should be aware that different
- 8 approaches can be used to monetize benefits in a CBA, and that this could lead to wide
- 9 variations in the results of a CBA.⁸ The choice of a particular approach needs to be
- 10 specified and justified.

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1	3. Study Populations		
2	3.1	Researchers should identify the intended population(s) for the vaccination	
3		program, the population at risk for the disease of interest, and any populations	
4		that may be indirectly affected by the vaccination program, either through	
5		externalities or spillover effects.	
6			
7	3.2	Researchers should present an overall analysis that includes the costs and	
8		health outcomes for all of the affected populations. When relevant, researchers	
9		should also summarize the results separately for each affected group (e.g.,	
10		intended population, population experiencing externalities or spillover effects)	
11		that was included in the overall analysis.	
12			
13			
14	3.3	Where there are factors that could lead to differences in costs and outcomes	
15		related to the vaccine program across subgroups, researchers should conduct	
16		separate economic evaluations for each subgroup. These factors could include	
17		demographic factors, behavioural factors, disease-related factors, and	
18		effectiveness of the vaccine or comparator intervention(s).	
19			
1J			
20			

21 The results of any economic evaluation of a vaccination program depend on the impact 22 of the vaccination program on three populations: 1) the intended population(s) for the 23 vaccination program; 2) the population at risk for the disease of interest; and 3) 24 population(s) that may experience externalities or spillover effects. In cases where a 25 vaccination program is associated with externalities, both the intended population for the 26 vaccination program and the population expected to experience externalities should be 27 identified in the decision problem. Researchers should identify any externalities 28 associated with vaccination programs (e.g., community immunity, age-shifting of 29 disease), and the population(s) they are expected to affect. For example, a measles 30 vaccination program intended for infants and children may result in community immunity 31 that could potentially lead to population-wide disease elimination. Another example is 32 varicella vaccine intended for young children for the prevention of chickenpox, which 33 could increase the incidence of herpes zoster in the general population. Further details

1 on incorporating externalities into economic evaluations are provided in Chapter 8 on 2 Modelling. Additionally, researchers should identify any population(s) that may 3 experience spillover effects (e.g., caregivers) related to the vaccination program. 4 5 Researchers should provide a detailed description of each population being considered 6 in the analysis that includes age, gender, and geographic location. Researchers should 7 also describe any other factors that determine eligibility for the vaccination program 8 being evaluated, and factors that may affect the magnitude of the externalities 9 experienced. 10 11 Researchers should present an overall analysis that includes the costs and outcomes for 12 all of the affected populations, including the group(s) identified for the vaccination 13 program, and any groups that may experience externalities or spillover effects. Such an 14 analysis should be presented for each implementation strategy that is being considered 15 (e.g., universal vaccination, vaccination of high-risk groups only, vaccination of children 16 only). When relevant, researchers should also summarize the results separately for each 17 affected group (e.g., intended population, population experiencing externalities or 18 spillover effects) that was included in the overall analysis. 19 20 However, in situations where heterogeneities may affect the results of an economic 21 evaluation have been identified between groups of individuals, economic evaluations for 22 different strategies that improve coverage in each of the subgroups and present 23 outcomes stratified by subgroup should be undertaken. This should ideally based on an 24 underlying mathematical model that considers all the subgroups and interactions 25 between them. Important heterogeneities with respect to vaccination programs could 26 include demographic factors (e.g., age, gender, geographic location), behavioural factors 27 (e.g., expected uptake of the vaccination program, risk-taking behaviours), disease-28 related factors (e.g., natural history of the disease, risk of disease transmission), 29 effectiveness of the vaccine or comparator intervention(s), and health utilities or costs

30 associated with the health states or interventions included in the analysis.

1 4. Comparators

	4.00	onparators
2	4.1	The choice of comparator(s) should be related to the scope of the decision
3		problem. As such, the comparators should reflect the intended population for the
4		vaccination program and the jurisdiction for which the decision is being made.
5		[CADTH Guideline Statement with amendment]
6		
7	4.2	Researchers should consider both preventive and treatment-based approaches
8		when selecting comparators for economic evaluations of vaccination programs.
9		Preventive interventions could include vaccine-based measures, screening
10		programs, preventive medication-based interventions, and preventive non-
11		medical interventions.
12		

13 When selecting comparators for economic evaluations of vaccination programs,

14 researchers should consider all current interventions, those that may become available

15 in the near future, and those that may be displaced by the vaccination program being

16 evaluated. Interventions used for both prevention or treatment of the disease of interest

17 should be considered. Preventive interventions could include vaccine-based measures,

18 screening programs, preventive medication-based interventions, and preventive non-

19 medical interventions. Often, more than one comparator will be relevant for the

20 economic evaluation, and therefore, all relevant comparators should be included.

21

22 Vaccine-based measures could include alternative vaccines against the same pathogen 23 (e.g., parenteral trivalent inactivated vaccine versus intranasal live attenuated vaccine 24 for influenza, whole-cell versus acellular vaccines for pertussis, mRNA versus viral 25 vector vaccines for coronavirus infection disease 2019 (COVID-19)), or vaccines with 26 additional valents (e.g., 10-valent and 13-valent pneumococcal conjugate vaccines 27 (PCV10 and PCV13)). Vaccine-based measures could also include different 28 implementation or delivery with the same vaccine product. Examples include universal 29 vaccination versus vaccination of high-risk groups only; vaccination of the intended age 30 group with no catch-up strategy versus vaccination of the intended age group with a 31 catch-up strategy for other age groups; vaccination strategy with no booster doses 32 versus strategy with booster doses; vaccination strategies based on a differing number 33 of doses or differing administration schedules; or different settings for delivery such as a

1	school-based strategy versus a public health clinic-based strategy versus mass
2	vaccination strategy in hot spots or warehouses. ¹⁰⁻¹³
3	
4	Screening programs (also known as secondary prevention) could include regular exams
5	and tests to detect disease in its earliest stage. They may be changed by the
6	introduction of vaccination programs. For example, HPV vaccination may change the
7	value and necessity for routine cytological smears for cervical cancer screening.
8	
9	Preventive medication-based interventions may involve pre- or post-exposure
10	administration of medications. Examples include anti-malarial medications for Canadian
11	travellers to malaria-endemic regions, and pre- and post-exposure prophylaxis for
12	human immunodeficiency virus (HIV).
13	
14	Preventive non-medical interventions could include physical measures such as condoms
15	to prevent sexually transmitted infections, face masks to prevent transmission of
16	respiratory infections, or behavioural modifications such as physical distancing and hand
17	washing to prevent infections that are transmitted through close personal contact
18	between individuals.
19	
20	When treatment-based comparators are being considered, researchers should be aware

- 21 that best supportive care should be considered the relevant comparator in cases where
- 22 no curative treatments exist for the disease of interest.

1	5. Pe	erspectives
2	5.1	Two reference case analyses should be presented as part of the economic
3		evaluation of vaccination programs: one conducted from the publicly funded
4		health system perspective, and the other conducted from the societal
5		perspective.
6		
7	5.2	"Both costs and outcomes should be consistent with the stated perspective."
8		[CADTH Guideline Statement]
9		
10	Two re	eference case analyses should be conducted as part of the economic evaluation of
11	vaccir	nation programs: one conducted from the publicly funded health system
12	perspe	ective, and the other conducted from the societal perspective. In these guidelines,
13	health	system refers to both healthcare treatment services and Public Health.
14		
15	Public	cly Funded Health System Perspective
16		
17	For th	e reference case analysis conducted from the publicly funded health system, the
18	scope	of the perspective should be defined to include a single provincial/territorial
19	public	ly funded health system, multiple regional publicly funded health systems, or a
20	nation	al system. Researchers should include: 1) health outcomes experienced by
21	vaccir	nated individuals and their informal caregivers; and 2) costs incurred by the health
22	syster	n. It must be recognized that when the reference case analysis includes multiple
23	public	ly funded health systems, the publicly funded cost items may vary from jurisdiction
24	to juris	sdiction (e.g., prescription medications), or even within a jurisdiction (long-term
25	care).	Variations in what items are included across systems should be made
26	transp	parent.
27		
28	In cas	es where vaccines are associated with externalities, the health outcomes and
29	costs	considered in the analysis also include those experienced by unvaccinated
30	individ	luals since vaccine plays a critical role in population health. ¹⁴ Population-level
31	health	outcomes that should be considered include: 1) incidence of infection and disease
32	in vac	cinated and unvaccinated individuals; 2) changes in the age distribution of
33	indivic	luals who are infected as a result of age-shifting related to the vaccination program
34	(when	this has consequences on the overall disease burden as a result of age-

1 dependency in severity of disease); 3) emergence of new diseases related to variations

2 of the pathogen (i.e., serotypes, serogroups, strains) or unrelated pathogens that may

- 3 replace the one(s) targeted by the vaccine; and 4) disease eradication.
- 4

5 Population-level costs that should be considered from this perspective include:

6 vaccination program implementation, delivery and sustainment costs including public

7 health campaigns; transaction costs related to introduction of new vaccines or switching

8 between vaccines; costs related to screening, diagnosis, and treatment of disease; and

9 epidemiological surveillance, contact tracing, case investigations, and outbreak

10 investigations. Guidance on quantifying the costs associated with these outcomes is

11 found in Chapter 11 on Resource Use and Costs.

12

13 **Societal Perspective**

14

15 A societal perspective reference case analysis is also recommended because many 16 vaccines prevent diseases that have impacts in areas beyond health. For example, the 17 Haemophilus influenzae type b (Hib) vaccine administered to infants prevents 18 neurological sequelae (e.g., deafness, blindness, developmental delays), all of which 19 would affect a child's school attendance, future productivity and consumption, as well as 20 broader well-being.^{15,16} Even relatively mild diseases such as childhood diarrhea 21 resulting from rotavirus infection can lead to impacts outside of health. In many cases, 22 medical attention is not required to treat these infections; nonetheless, a parent is 23 required to take time off work to care for the sick child.^{17,18} Finally, diseases such as 24 COVID-19 have tremendous health and economic impacts that extend to every area of 25 the economy,¹⁹ and their impacts could be mitigated through vaccination programs.^{20,21} 26 Failing to consider the full range of benefits associated with vaccines underestimates the 27 role of health as a driver of economic activity and well-being, and could lead to 28 undervaluation of vaccination programs.¹⁴ 29 30 The societal perspective analysis captures all the health outcomes and health system

31 costs from the health system perspective. In addition, it captures impacts that fall outside

32 of the publicly funded health system, including: healthcare costs not publicly funded by

33 the health system, direct out-of-pocket costs, productivity, consumption, education,

34 social services, and environment. Longer term impacts such as the effect of childhood

- 1 illness on their neurodevelopmental impairment, educational attainment and subsequent 2 long-term productivity (and consumption) should also be considered where relevant and 3 feasible. These potential impacts are listed along with examples in Table 1: Impact 4 inventory table for economic evaluations of vaccination strategies. This table was 5 adapted from the impact inventory published by the 2nd Panel on Cost-Effectiveness²² to 6 also include broader impacts associated with vaccines described in the literature.14,22-25 7 The table provides a comprehensive list of health and non-health impacts that could 8 result from vaccination programs. The intent is to allow researchers to consider the 9 impacts systematically when planning for, and conducting economic evaluations of 10 vaccination programs. Specific guidance on guantifying these impacts and their 11 associated costs is found in Chapter 11 on Resource Use and Costs.
- 12
- 13 Researchers should complete and present Table 1 as part of their analysis to explicitly
- 14 indicate which impacts are included and excluded in the economic evaluation for each of
- 15 the two reference case analyses. The comments column could be used to provide
- 16 justification for including or excluding certain impacts or to provide additional information.



Table 1: Impact inventory table for economic evaluations of vaccination strategies

Area of Impact	Definitions/Examples	Included in Re	ference Case?	Comments
		Publicly funded health	Societal perspective	
		system perspective		
Health				
Health outcomes	Individual health outcomes for persons inten	ded for vaccinatio	n	
	Mortality			
	Health-related quality of life			
	Safety (i.e., adverse events)			
	Irreversible health impacts not captured by QALYs (e.g., infertility associated with sexually transmitted infections			
	Individual health outcomes for informal care	givers		
	Health-related quality of life			
	Population health outcomes			
	Incidence of disease in vaccinated and unvaccinated individuals			
	Changes in age distribution of individuals who develop infection and disease			
	Changes in infection and disease incidence related to variations of pathogen			

Area of Impact	Definitions/Examples	Included in Re	ference Case?	Comments
		Publicly funded health system perspective	Societal perspective	
	or other pathogens that replace ones targeted by vaccine Disease eradication			
Health system costs	Healthcare treatment costs			
	Publicly funded healthcare services (e.g., physician visits, diagnostic tests, drug treatment where applicable, hospitalization, formal caregiving, ^a rehabilitation in a facility or at home, ^a home care, ^a long-term care in nursing homes ^a)			
	Future related and unrelated healthcare costs			
	Public health costs			
	Program-related costs (e.g., implementation, delivery and recurrent costs, public health campaigns, health promotion activities, transaction costs, population-based screening, epidemiologic surveillance, contact tracing, investigation and management of outbreaks)			
	Intervention-related costs (e.g., cost of vaccine doses, distribution such as			

Area of Impact	Definitions/Examples	Included in Re	ference Case?	Comments
		Publicly funded health system perspective	Societal perspective	
	transportation and cold storage, administration including personnel, wastage and ancillary supplies)			
Healthcare costs NOT funded by the health system	Drug treatments (in some cases) Formal caregiver services, ^a rehabilitation in	N/A		
System	a facility or at home, ^a home care, ^a long- term care in nursing homes ^a (in some cases)	N/A		
	Miscellaneous out-of-pocket costs (e.g., non-prescription medications)	N/A		
	Ancillary costs (e.g., private insurance copayments, dental care, vision care, assistive devices, physiotherapy, etc.)	N/A		
Non-Health Areas				
Direct out-of-pocket costs	Transportation costs	N/A		
	Accommodation costs	N/A		
Productivity loss	Paid work			

Area of Impact	Definitions/Examples	Included in Re	ference Case?	Comments
		Publicly funded health system perspective	Societal perspective	
	Time off work resulting from treatment, illness, disability, or death	N/A		
	Presenteeism	N/A		
	Lifetime productivity consequences of childhood disease		_	
	Unpaid work			
	Time off work in informal labour market resulting from treatment, illness, disability, or death	N/A		
	Uncompensated household production (e.g., Cooking, cleaning, shopping, raising children, other tasks related to household management)	N/A		
				I
	Time off work resulting from caring for sick individuals	N/A		
	Caregiver presenteeism	N/A		
	Macroeconomic consequences	<u> </u>	L	1
	Labour supply shocks, widespread business closures	N/A		

Area of Impact	Definitions/Examples	Included in Re	ference Case?	Comments
		Publicly funded health system perspective	Societal perspective	
Consumption	Future individual non-medical consumption	N/A		
	Changes in household consumption	N/A		
	Health impacts of consumption (e.g.,	N/A		
	associated with job loss)			
Education	Level of educational achievement as a result of physical health, mental health, and cognition	N/A		
	Costs of special education needs as a result of illness/disability	N/A		
Social services and community services	Social services and community services (e.g., disability support, programs to improve access to vaccination programs for adults)	N/A		
	Child and Youth Services (e.g. awareness programs, family respite, programs to improve access to vaccination programs for children and youth)	N/A		
Environment	Environmental impact of vaccination programs and comparators (e.g.,	N/A		

Area of Impact	Definitions/Examples	Included in Reference Case? Publicly Societal		Comments
		funded health system perspective	perspective	
	manufacturing, distribution, and implementation)			
Other Areas	Consider areas such as legal/criminal or housing when applicable	N/A		

^a Some of these costs may or may not be incurred by the publicly funded health system, depending on the precise nature of these costs and the relevant jurisdiction

1 6. Time Horizon

2	6.1 In the reference cases, the time horizon should be long enough to capture all
3	relevant differences in the future costs and outcomes associated with the
4	interventions being compared. Thus, the time horizon should be based on the
5	condition and the likely impact of the intervention. [CADTH Guideline Statement
6	with amendment]
7	
8	6.2 Researchers should justify their choice of time horizon. Where it spans a long
9	period of time (i.e., multiple decades), researchers should report incremental
10	costs, incremental effects, and ICER estimates from various time points
11	throughout the time horizon.
12	
13	Models used to estimate the cost-effectiveness of vaccination programs can be closed
14	or open models. Closed models follow a cohort of individuals over a length of time and
15	do not allow for the entry of new individuals into the model. Most Markov (state-
16	transition) models are closed models. Closed models are usually static, meaning that
17	they do not account for disease transmission dynamics between individuals. Open
18	models, on the other hand, do allow for entry of new individuals into the model over time
19	(e.g., via new births, immigration), specifically to account for disease transmission
20	dynamics within a population over time. ²⁶
21	
22	Since closed models of vaccination programs follow a single group of individuals, these
23	models should follow the group for a long enough time horizon to capture all important
24	differences in future costs and outcomes related to the vaccination strategies being
25	compared.
26	
27	Open models may have time horizons that extend beyond the life of any individual alive
28	at the start of the simulation, and so may require a time horizon that spans multiple birth
29 20	cohorts. This is particularly true for vaccines that provide population-level protection
30 21	through community immunity over multiple birth cohorts. For example, a cohort of
31 22	individuals vaccinated against measles today may prevent transmission of this infection
32	to another cohort years later. Individuals who are not vaccinated would benefit from this

protection for the rest of their lives, in turn not infecting future cohorts with measles who
 would also benefit for the rest of their lives.^{27,28}

3

4 There are typically three phases in open models: 1) the run-in phase; 2) the evaluation 5 phase; and 3) the steady-state phase. The run-in phase must account for epidemiologic 6 characteristics of the disease prior to introduction of the vaccine in order to realistically 7 and accurately predict uptake of the vaccine. The epidemiologic estimates used during 8 the run-in phase should be validated based on historical data pertaining to the disease of 9 interest. More information on validation can be found in Chapter 8 on Modelling. The 10 evaluation phase begins when the vaccination program is implemented in the intended 11 population, and should be long enough to account for externalities associated with the 12 program. The steady-state phase begins once epidemiologic variation terminates.²⁹ 13 14 For the measles example above, and for similar vaccines, the model time horizon should

15 continue until the undiscounted ICER reach a steady-state. This is when the ratio

16 between cumulative incremental costs and cumulative incremental health outcomes

17 (QALYs) between the interventions being compared stabilizes. In these cases, the

18 appropriate duration of the model time horizon should be ascertained during, rather than

19 prior, to the analysis.⁹

20

21 Stability of the undiscounted incremental estimates and ICER as a criterion should mean 22 that the time horizon of the model will be long enough to capture the full costs and 23 benefits of community immunity as well as any other externalities (e.g., age-shifting of 24 disease) associated with a vaccination program. Researchers should note that models 25 might achieve epidemiologic steady-state prior to the incremental estimates and ICER 26 estimates stabilizing. For example, one program may continue accruing costs or QALYs 27 relative to the other even after epidemiologic equilibrium has been realized. In these 28 cases, the model time horizon needs to continue until the ICER estimate has stabilized. 29

For some vaccinations programs, modelling a very large number of birth cohorts may be required to achieve stable ICER estimates, but this approach may not be practical or appropriate for the decision-making process.²⁸ For instance, researchers should note that modelling a large number of birth cohorts is not required in situations where the vaccination program is not expected to result in disease elimination or to take many

1 years to deliver its full impact, such with some seasonal illnesses (e.g., current vaccines 2 against influenza), or with infections whose source is non-human and transmission 3 between individuals is not possible (e.g., tetanus). If the model is not run until the 4 incremental estimates and ICER have stabilized, researchers should justify why this is 5 the case, and define the run time in terms of time horizon or number of cohorts, and 6 provide justification for this choice.³⁰ 7 8 In justifying the time horizon and number of cohorts, researchers should be aware of the 9 trade-offs between bias and uncertainty. Shorter time horizons may introduce bias into 10 cost-effectiveness estimates since they do not allow for enough time to account for 11 epidemiologic changes resulting from the vaccination program. Shorter time horizons 12 place a greater weight on upfront costs related to initiation of the vaccination program 13 relative to later or annual costs, and reduce the consequences of discounting on

measured outcomes. Shorter time horizons also may not quantify all of the benefits
accrued to the final cohorts vaccinated. This may not be an issue for large-scale vaccine

- 16 programs where the payer borrows to fund the program, and costs are annuitized.
- 17

18 For models with long time horizons, researchers should consider the potential for future 19 changes that might alter the costs and benefits of the vaccine (e.g., technological change, long-term estimates of vaccine effectiveness, demographic projections).^{10,31,32} 20 21 While some of this uncertainty may be accounted for in the discount rate (in particular, 22 by the "catastrophic risk"— the risk of an unanticipated event removing much of the 23 value of the intervention), researchers may wish to consider context-specific, long-term 24 uncertainties such as the emergence of treatment-resistant disease. Where the time 25 horizon spans a long period (i.e., multiple decades), researchers should report ICER 26 estimates from various time points throughout the time horizon.

27

In some jurisdictions, the effects of high, upfront implementation costs for vaccination
programs can be tempered by annuitization, reflecting the borrowing practices of
government agencies to fund wide-scale programs. Regardless of the time horizon
length, analyses should indicate the rates used for annuitization and amortization (if
applicable). It may be appropriate to conduct the analysis with and without annuitization
of upfront implementation costs.

1 7. Discounting

- 7.1 In the reference cases, costs and outcomes that occur beyond one year should
 be discounted to present values at a rate of 1.5% per year. [CADTH Guideline
 Statement with amendment]
- 6 7.2 "The impact of uncertainty in the discount rate should be assessed by comparing
 7 the results of the reference cases to those from non-reference case analyses,
 8 using discount rates of 0% and 3% per year." [CADTH Guideline Statement]
- 9

5

Discounting costs, health outcomes, and non-health outcomes in economic evaluations reflects the societal preference for present consumption over future consumption. This is because discounting reduces the value of future costs and outcomes compared to their present value. The discount rate accounts for the social rate of time preference, growth rates in healthcare resources and the consumption value of health, and uncertainty about whether future health outcomes will be realized.^{33,34}

16

17 Discounting in economic evaluations of vaccination programs can have a profound effect 18 on the cost-effectiveness of programs, particularly in situations where the expected 19 benefits of the vaccination program may not become apparent for years or even 20 generations (e.g., prevention of cervical cancer through an HPV vaccination program). 21 Pediatric populations are particularly susceptible to these effects. Discounting of health 22 and non-health outcomes that accrue in the distant future may lead to a considerably 23 reduced present value of outcomes. For example, in a cost-effectiveness analysis of an 24 HPV vaccination program, the authors report an undiscounted ICER of \in 7.600/QALY. 25 which increased to €59,100/QALY when a 4% discount rate was applied to both costs 26 and benefits (3,462 undiscounted QALYs versus 438 discounted QALYs).35 27 28 The two most common approaches to discounting in economic evaluations of 29 vaccination programs are: 1) constant discounting, where the same fixed discount rate is 30 applied to both outcomes and costs; and 2) differential discounting, where a lower 31 discount rate is applied to outcomes compared to costs.³³⁻³⁶

- 32
- 33 The approach most commonly employed in economic evaluations of vaccination
- 34 programs is constant discounting, which is also the approach most commonly used for

1 non-vaccine health interventions.³⁴ Some national immunization technical advisory group 2 guidelines and health technology assessment (HTA) guidelines, recommend differential 3 discounting approaches in uncertainty analyses or in special circumstances.^{37,38} 4 Arguments in favour of constant discounting of outcomes and costs include consistency 5 and horizontal equity.³⁴ The consistency argument posits that health technologies 6 associated with the same outcomes and costs over the same analytic time horizon 7 receive equal priority by decision-makers, regardless of the time at which they are 8 initiated.³⁹ This is because of the constant value of health over time. The horizontal 9 equity argument posits that all individuals who potentially benefit from a vaccination 10 program are treated equally, regardless of when they experience the benefits relative to 11 when the program was initiated. Constant discounting prevents vaccination programs 12 that span multiple generations from being given preference over programs that span a 13 shorter time.33

14

15 One argument for differential discounting with a lower discount rate for health and non-16 health outcomes compared to costs is to normatively place more weight on future 17 benefits. Another main argument for differential discounting is the increasing value that 18 future health expectedly represents, or the changing thresholds for judging costeffectiveness.⁴⁰ Discount rates could be adjusted to reflect these changes, although they 19 20 could also be dealt with more explicitly in an analysis. With respect to vaccination 21 programs, long time horizons— often generations, are required to achieve outcomes 22 related to indirect effects of community immunity, which benefit not only the vaccinated 23 individuals but also future cohorts through disease elimination or eradication.³⁴ Using 24 constant discounting, particularly with higher discount rates, may render the present 25 value of these programs close to zero. Differential discounting, on the other hand, 26 increases the present value of outcomes occurring in the distant future compared to a 27 constant discounting strategy.

28

A downside of differential discounting is that strategic use of time horizons and the
number of included cohorts could alter cost-effectiveness estimates. O'Mahony et al., for
instance, provide an example comparing constant and differential discounting
approaches in a cost-effectiveness analysis of an HPV vaccination program in 12-yearold girls. The authors considered 1, 10, 20, and 30 birth cohorts. They discounted health
outcomes and costs with an equal rate of 4%, and with differential rates of 1.5% and 4%

1 respectively. As expected, they demonstrated that the ICER decreased as the number of 2 cohorts increased with the differential discounting strategy, but not with the constant 3 strategy.⁴¹ Although normative and analytical solutions to this problem have been 4 formulated.^{41,42} it does raise potential concerns that unjustified analytic choices in 5 economic analyses could lead to variations in results. This underscores the need for 6 appropriate guidance on the use of differential discounting. 7 8 Researchers should discount health outcomes, non-health outcomes, and costs 9 occurring beyond one year at a rate of 1.5% per year in the reference case analyses. 10 This value represents the real cost of long-term borrowing for Canadian provinces, who 11 are the authorities responsible for funding the majority of the Canadian healthcare 12 system,⁴ and approximates the rate at which society is willing to trade-off consumption 13 today for consumption in the future.³⁴ 14 15 Sensitivity analyses should be conducted using rates of 0% and 3% per year applied to 16 both outcomes and costs to account for any uncertainty in the discount rate. The low 17 discount rate in the reference case analyses mitigates some of the concerns regarding 18 present values of expected outcomes in the distant future. In situations where effects of 19 a vaccination program span multiple generations and may be affected by the discounting 20 strategy or rate used in the analysis, presenting undiscounted results will help decision-21 makers to assess potential intergenerational effects. A discussion of intergenerational 22 equity is found in Chapter 14 on Equity.

1	8. M	odelling	
2	8.1	"Model conceptualization and development should address the decision	
3		problem." [CADTH Guideline Statement]	
4			
5	8.2	"Researchers should consider any existing well-constructed and validated models	
6		that appropriately capture the clinical or care pathway for the condition of interest	
7		when conceptualizing their model." [CADTH Guideline Statement]	
8			
9	8.3	The model structure should reflect the natural history of disease, the clinical or	
10		care pathway, and account for susceptibility, infectiousness, and immunity related	
11		to the infection.	
12			
13	8.4	Relevant behavioural dynamics including contact patterns between individuals	
14		and behaviours related to infection prevention and control should be incorporated	
15		into the model where appropriate.	
16			
17	8.5	Dynamic models should be considered in economic evaluations of vaccines that	
18		are associated with externalities such as prevention of human-to-human	
19		transmission of infection and age-shifting of disease.	
20			
21	8.6	Other model attributes including whether the model is deterministic or stochastic,	
22		population-based or individual-based, and open or closed should be considered	
23		in the context of the decision problem.	
24			
25	8.7	Researchers should transparently report on model calibration and validation	
26		processes that were undertaken and on their results.	
27			
28	This c	hapter presents considerations related to constructing models used to estimate the	
29	cost-effectiveness of vaccination programs. A look at dynamic versus static models in		
30	the co	ontext of infectious disease modelling is presented, followed by an overview of	
31		model attributes. Finally, recommendations related to model calibration and	
32	valida	tion are presented.	
33			

1 Model Structure and Attributes

2

3 The model's structure and attributes should reflect the natural history of disease, and 4 include all relevant health states and transitions between these states. There are two 5 primary considerations when conceptualizing a model used to estimate the cost-6 effectiveness of a vaccination program: whether transmission of infection between 7 individuals is important in estimating the effects of a vaccination program; and whether 8 individual behaviours and characteristics are important in understanding outcomes 9 related to a vaccination programs. Researchers should refer to more detailed model taxonomies by Brennan et al.,⁴³ Kim and Goldie,²⁶ Stahl⁴⁴ and Mac et al.⁴⁵ for additional 10 11 details if required. 12 13 Further guidance on constructing models for economic evaluation of vaccination 14 programs can be found in Chapter 13 on Uncertainty (e.g., ensuring that the model 15 structure accounts for factors related to transmission of infection between individuals, the 16 natural history of the disease being modeled, as well as the direct and indirect effects of 17 the vaccination program).^{28,46} 18 19 Endogenous vs. exogenous infection rate 20 Models need to capture externalities related to vaccination programs such as community 21 immunity and age-shifting of disease. In these guidelines, the terms "dynamic model" 22 and "static model" refer to the nature of the incidence rate being dynamic or not (i.e. 23 changing over time based on the proportion of the population that is infectious). They 24 may also be referred to as having an "endogenous" or "exogenous" rate of infection,

25 respectively.

26

27 Static models, which typically use a constant risk of exposure, do not explicitly represent 28 dynamic infection transmission. These models are acceptable for use in economic 29 evaluations of vaccination programs where there is no human-to-human transmission 30 (e.g., tetanus or rabies).⁹ They are also acceptable in situations where the intended 31 group for vaccination is not epidemiologically influential with respect to transmission 32 (e.g., hepatitis A vaccination of healthcare workers, influenza or pneumococcal vaccination in the elderly).^{9,28} Static models may also be acceptable for infections where 33 34 the individual is already a "host" (e.g., some pneumococcal strains; varicella-zoster virus

1 where incidence of infection is more a random event in a person's life after long-standing 2 colonization). Finally, a static model is acceptable when: 1) a vaccination program is 3 demonstrated to be cost-effective, and a dynamic model would only serve to reinforce 4 this conclusion by accounting for infections prevented through indirect protection or 5 secondary transmission; or 2) a vaccination program is not demonstrated to be cost-6 effective, but there are epidemiological or modelling data available that will allow 7 estimation of the magnitude of community immunity or secondary transmission in the 8 same or very similar setting.28,47

9

10 Although the scenarios above outline situations where static models may be acceptable 11 for estimating the cost-effectiveness of vaccination programs, researchers should be 12 aware of their limitations. First, when a static model has demonstrated the cost-13 effectiveness of a vaccine despite not accounting for the effects of community immunity 14 or secondary transmission, the true cost-effectiveness of the intervention may be 15 underestimated, and could result in biased resource allocation decisions.²⁸ Second, 16 when epidemiological or modelling data are used to estimate the magnitude of 17 community immunity or secondary transmission in the context of static models, there 18 may be biased cost-effectiveness estimates when the data used are from a different 19 population than the one considered in the model and when there are other important 20 differences. Also, if the data used are from epidemiologic equilibrium, the fluctuation in 21 prevalence in the initial post-vaccination period will not be accounted for. This limitation 22 is particularly important for vaccination programs with a positive time preference given 23 that the initial time period is when most of the costs and benefits related to the vaccine 24 are accrued.⁴⁷ Examples of static models include decision trees, cohort-based Markov 25 models, and discrete event simulations.

26

Dynamic models, which explicitly represent infection transmission, should be considered
in economic evaluations of vaccination programs where human-to-human transmission
is an important factor. For example, dynamic models should be employed when a largescale vaccination program is expected to change the force of infection leading to control,
elimination, or eradication of a disease by preventing its transmission.

32

33 Dynamic models should also be employed when serotype replacement and age-shifting
34 of disease could potentially result from a vaccination program. Vaccines that are specific

1 for certain pathogen variation (i.e., serotype, serogroup, or strain) may reduce one 2 variation of the disease, but in the presence of multiple variations, the prevalence of 3 infection from non-vaccine variations may still increase.^{48,49} For diseases caused by 4 multiple variations of a pathogen, researchers should include each variation separately 5 within the model so that infection and disease related to the emergence of new 6 variations can be accounted for. Situations where a vaccination program leads to an 7 increase or decrease in the average age of individuals affected by an infection may lead 8 to a corresponding increase in disease severity, treatment costs, and mortality, which 9 should also be accounted for in a cost-effectiveness analysis.^{50,51} Examples of dynamic 10 models include dynamic cohort models and individual-based simulation models.

11

12 When choosing between a dynamic or static model, researchers should consider the 13 trade-offs between the need to represent transmission, and the additional complexities 14 associated with dynamic models. In some situations, the decision to select one model 15 type over another may not always be straightforward. Dynamic models are conceptually 16 and computationally more complex than static models. Decision-makers who are the 17 end-users of the results generated must be able to understand and interpret the 18 structure of the model. They also need to trust that the results are a reasonable 19 representation of what would be expected to transpire in the real-world setting after the 20 implementation of the vaccination program. There is also a trade-off between the 21 complexity (and realism) of a model, and the ease by which it can be understood, 22 communicated and validated. In some cases, transmissibility between individuals may 23 result in spread of an infection, but the nature of the vaccination program may negate 24 the need to represent transmission in an economic model. For example, for a universal 25 vaccination program that is expected to achieve a high level of coverage in the 26 population, a static model may be adequate in predicting its effects. For further guidance 27 on whether to use a static or dynamic model when estimating the cost-effectiveness of a 28 vaccination program, researchers could consult published schematic diagrams that 29 delineate considerations related to this choice by Jit and Brisson and the World Health Organization (Figure 4, Table 8).^{9,47} 30

31

It should be noted that there are "hybrid" models between dynamic and static models, in
which researchers do not fully account for infection transmission. Rather, they estimate

 and incorporate the costs and benefits of preventing those cases into the analysis. Other Attributes Although the fundamental choice facing researchers who are modelling the cost- effectiveness of vaccines is between selecting static versus dynamic modelling techniques, they must also consider other attributes related to the model structure. Considerations related to these attributes are discussed below. Deterministic versus Stochastic In deterministic models, events depend on pre-specified parameters and model structure; in other words, first-order uncertainty is not accounted for since events cannot occur randomly (by chance). In stochastic models, on the other hand, events are programmed to occur randomly, accounting for first-order uncertainty.^{26,52} For a discussion of second-order (parameter) uncertainty, researchers should refer to Chapter 14 on Uncertainty. Average parameter values used in deterministic models may realistically approximate the processes being modeled if the population at risk is large, and the infection is not close to elimination or global eradication (e.g., HPV). For small populations, (e.g., college outbreak of meningococcal B infection), or when modelling the rise of an emerging infection or a rare infection that is on the verge of elimination (e.g., measles and polio in some countries) models that incorporate individual variability and first-order uncertainty (e.g. individual-based models) are more appropriate since they are able to account for random transmission events that are important in these situations ^{8,28} Aggregate wordels (also referred to as population-based or cohort models) such as Markov cohort models and dynamic compartmental models, groups of individuals are aggregated into compartments representing health states based on their characteristics. 	1	the average number of secondary infections averted through the prevention of a case
4 Other Attributes 5 Although the fundamental choice facing researchers who are modelling the cost- effectiveness of vaccines is between selecting static versus dynamic modelling techniques, they must also consider other attributes related to the model structure. 9 Considerations related to these attributes are discussed below. 10 Deterministic versus Stochastic 11 Deterministic models, events depend on pre-specified parameters and model 14 structure; in other words, first-order uncertainty is not accounted for since events cannot 15 occur randomly (by chance). In stochastic models, on the other hand, events are 16 programmed to occur randomly, accounting for first-order uncertainty. ^{26,52} For a 17 discussion of second-order (parameter) uncertainty, researchers should refer to Chapter 18 14 on Uncertainty. 19 Average parameter values used in deterministic models may realistically approximate 19 the processes being modeled if the population at risk is large, and the infection is not 10 close to elimination or global eradication (e.g., HPV). For small populations, (e.g., 12 college outbreak of meningococcal B infection), or when modelling the rise of an 14 emerging infection or a rare infection that is on the verge of elimination (e.g., measles	2	and incorporate the costs and benefits of preventing those cases into the analysis.
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33 aggregated into compartments representing health states based on their characteristics.		
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- 1 Changes over time represent shifts in the proportion of the population in each health
- 2 state based on average parameter values.^{26,53}
- 3

4 In individual-based models (also called micro-simulations or agent-based models), the 5 individual, rather than the group, is the unit that is modelled. Models that simulate 6 transmission between infected and susceptible individuals are dynamic, in that they have 7 a changing risk of infection over the simulation, whereas those that assume an 8 exogeneous risk of infection independent of the whether there are infected people in the 9 population are static.²⁶ This type of model is generally more complex and requires more 10 data than a population-based model, and can be programmed stochastically so that an 11 individual's probability of future events accounts for uncertainty related to randomness.⁴⁵ 12 13 Individual-based models are also appropriate when there are significant heterogeneities

between individuals in a population. These heterogeneities may be related to genetic factors, socioeconomic status, age, access to healthcare services, occupational risk, and behaviour changes in response to disease outbreaks, just to name a few. See Chapter 14 on Equity for more equity-relevant differences. These models may be programmed such that the individuals are able to alter their behaviours over time based on their previous interactions.⁵²

20

Individual-based models are also appropriate when there are significant heterogeneities between individuals in a population. These heterogeneities may be related to genetic factors, socioeconomic status, age, access to healthcare services, and behaviour changes in response to disease outbreaks, just to name a few. See Chapter 14 on Equity for more equity-relevant differences. These types of models account for these characteristics and the effect that they could have on outcomes related to the introduction of a vaccination program.⁵⁴

28

29 Population-based models, on the other hand, are appropriate for vaccination programs

30 for relatively homogeneous groups of individuals (e.g., a pneumococcal vaccination

31 program for elderly individuals in one geographic area)⁵⁵ since they have similar

32 characteristics that could be reasonably represented by average values as they

33 transition through different health states. Note that population-based models can

1 nonetheless incorporate some heterogeneity through stratifying by risk, and/ or

2 incorporating assortative mixing by age groups and on other risk factors.

3

4 When modelling heterogeneities between groups or individuals (including equity-relevant

5 differences), researchers should consider how much detail is required to adequately

6 model the cost-effectiveness of a vaccination program, and consider the trade-offs

7 between different model types that could be used to account for these heterogeneities.

8

9 Open versus Closed

10

Models can represent open or closed populations. Open models allow new susceptible individuals, through births and immigration, to enter and exit the model over time, while closed models do not. Although open models may be computationally more complex, they allow researchers to estimate the evolution of the population intended for vaccination and account for its characteristics such as risk exposure, age, and disease severity.^{26,56}

17

18 Open models are generally useful for projecting changes in healthcare costs and 19 treatment outcomes for infectious diseases at different time points over the model time 20 horizon,⁵⁷ and should be used when the effects of vaccination programs in one cohort 21 will affect other population cohorts (e.g., childhood immunization programs for diseases 22 such as measles or polio). Closed models are appropriate when examining vaccination 23 programs in small groups of individuals that are unlikely to be epidemiologically 24 influential in the wider population (e.g., hepatitis A vaccination program for healthcare 25 workers), or where the effects of the vaccine are short-lived (e.g., seasonal influenza 26 vaccination program). Note that closed models with long time horizons may undercount 27 potential costs and health benefits.

28

29 Discrete versus Continuous Time

30

31 Continuous time models are recommended when multiple events need to be modelled

32 simultaneously. One case may be in disease outbreaks where, for example,

transmission of infection between individuals may depend on multiple factors such as

34 contact patterns between individuals, as well as the number of infectious individuals in a

given population.⁴³ Although continuous time models may provide more accurate results
in such situations, these models are computationally more complex. They require use of
ordinary differential equations for which solutions may be difficult to obtain. Results of
continuous models may be approximated by employing discrete time models with a
small time steps and appropriately rescaling parameters.^{26,43}

6

7 Model Calibration

8

9 Model calibration is the process used to infer unknown model parameters by adjusting 10 them to ensure that model outputs represent a good fit with observed data (calibration 11 targets).⁵⁸ In infectious disease modelling, many parameters may be unknown or cannot 12 be directly estimated based on available data. These could include parameters related to 13 the natural progression of the infection or disease, details related to sexual behaviours in 14 the case of sexually transmitted infections, and data related to uptake and distribution of results of screening interventions.⁵⁹ Calibration targets that are selected should be 15 16 independent data that are accurately reported with a high degree of both internal and 17 external validity. When appropriate, these data should be stratified by relevant 18 subgroups to ensure adequate model performance across key population strata.⁶⁰ 19 Researchers could also consider eliciting expert opinion when selecting calibration 20 targets.

21

22 Researchers should be aware that because subjective decisions are required during the 23 calibration process, such as selecting calibration targets, goodness-of-fit measures, and 24 calibration method, there is uncertainty related to the calibration methods that are 25 employed. These uncertainties can lead to considerable differences in results of 26 economic evaluations. Although calibration is often computationally intensive, when 27 possible, researchers should consider using more than one approach for model 28 calibration, and multiple goodness-of-fit statistics.⁶¹ Researchers should retain 29 uncertainty in the calibration estimated parameters, which then can be used in 30 probabilistic analysis.

31

Difficulty calibrating multiple model parameters may indicate that the model structure or
 its underlying assumptions are incorrect. It may also suggest a limited understanding of
 the natural history of the disease being modelled, or of the behaviours that affect its

1 transmissibility, detection, or treatment. Alternatively, it may reveal biases, inconsistency, 2 or imprecision in the data being used as calibration targets. As such, it should not be 3 minimized or ignored, but rather used to help establish future research priorities.²⁸ 4 Model Validation 5 Validation is the process that is used to ensure the accuracy of results generated from 6 models used in economic evaluations. The validity of a model should be examined within 7 a relevant decision-making context so that decision-makers are able to determine 8 whether the model under consideration addresses the decision problem at hand.⁶² 9 Researchers should assess various aspects of model validity using different methods. 10

11 Face validity concerns whether a model reflects the current understanding and evidence 12 related to the disease and vaccination program being considered. It involves the 13 subjective assessment of a model's structure, assumptions, data sources, and results. 14 This is best conducted by clinical experts in the field, and can also be done by 15 comparing the model structure to accepted clinical disease algorithms. Internal validity is 16 often referred to as verification, and refers to whether the model behaves as it should. It 17 involves verifying that the mathematical equations used in the model have been 18 programmed correctly. It ensures that there are no computational errors in the model. 19 Cross-validation involves comparing the results generated from one model, and 20 determining the extent to which they correspond to results of other models.⁶³ External 21 validity involves comparing results generated from a model with existing data from 22 independent sources such as clinical trials, epidemiologic studies, routinely available 23 population statistics such as mortality data, or electronic health records. External 24 validation is not possible in situations where the model makes use of all relevant known 25 data. It may be difficult in situations where these types of data do not exist, or when they are not sufficiently detailed to allow appropriate comparison.⁶⁴ Predictive validity refers to 26 27 whether a model is performing its intended purpose, which is to predict outcomes related 28 to a vaccination program. It is also the most difficult type of validation to perform since 29 results must relate to events or studies conducted in the future. This type of validation is 30 usually not applicable to decision-making related to a new vaccination program.⁶⁴ 31 However, it may be relevant when developing a model based on older models. 32 Researchers can assess the older models prior to re-use. As with model calibration, 33 researchers could consider eliciting expert opinion when undertaking model validation 34 processes.

1	9.	Effectiveness		
2	9.1	"A comprehensive search of the available data sources should be conducted to		
3		inform the estimates of effectiveness and harms associated with the		
4		interventions. Report the included studies and methods used to select or combine		
5		the data." [CADTH Guideline Statement]		
6				
7	9.2	"The data sources should be assessed based on their fitness for purpose,		
8		credibility, and consistency. Describe the trade-offs among these criteria and		
9		provide justification for the selected source(s)." [CADTH Guideline Statement with		
10		amendment]		
11				
12	9.3	The following criteria should be considered when assessing estimates of vaccine		
13		effectiveness: vaccine effectiveness by dose; expected vaccine coverage;		
14		pathogen variation-specific (i.e., serotypes, serogroups, strains) effectiveness;		
15		and geographic and host factors that may affect effectiveness.		
16 17	9.4	Researchers should ensure that immune biomarkers used as surrogate		
18		outcomes in studies of vaccine efficacy or effectiveness meet the criteria for		
19		correlates of protection.		
20				
21	This chapter details factors that should be considered when assessing the effectiveness			
22	of vaccines, and considerations related to data synthesis, interpretation and use of			
23	surrogate outcomes, and extrapolation of effectiveness estimates.			
24				
25	Asse	essing Estimates of Vaccine Effectiveness		
26				
27	There	There are several factors specific to vaccines that should be considered when		
28	interpreting effectiveness data. These factors are discussed below.			
29				
30	Researchers should be aware of differences between efficacy and effectiveness related			
31	to vaccines. Efficacy is established through randomized controlled trials (RCTs), which			
32	evaluate changes in immune markers, reductions in disease severity, and improvements			
33	in health outcomes in vaccinated individuals. Effectiveness of vaccines in individuals is			
34	often different from efficacy. For example, there are often higher rates of vaccine series			
35	comp	pletion in RCTs compared to the real-world setting; there are limitations of the RCT		

1 design in capturing community immunity; and there are other differences between RCT

2 populations and the real-world populations in which the vaccine is used.

3

4 Vaccine series completion is an important consideration for the many vaccines that 5 require administration of multiple doses at defined time intervals. For example, the HPV 6 vaccine was initially administered on a 3-dose schedule, although a 2-dose schedule is 7 now recommended for some. For the 3-dose schedule, the second dose is given 1-28 months after the first dose, and the third dose 6 months after the first dose.⁶⁵ 9 Researchers should be mindful that individuals who do not receive all doses of a 10 recommended vaccine series might experience lower rates of vaccine effectiveness than 11 those who receive the full series. Researchers should assess both clinical trial data and 12 expected real-world dose completion estimates, as both have strengths and limitations. 13 Real-world data may be obtained from acceptability studies on vaccine series completion 14 or from data on completion of other vaccine series used in similar populations with a 15 similar number of doses. Researchers should keep in mind that residual confounding 16 may affect results of observational studies that examine the relationship between dose 17 completion rates and vaccine effectiveness. Specifically, factors that predict for lower 18 probability of dose completion may also increase the underlying risk of infection (e.g., 19 earlier sexual exposure in girls who receive fewer than three doses of HPV vaccine).¹³ 20 Researchers should use expected real-world dose completion estimates based on the 21 relevant jurisdiction(s) and intended population for the vaccination program for the 22 reference case analyses.

23

24 In terms of community immunity, RCTs may underestimate a vaccine's population-level 25 effects. That is, community immunity is not observed in RCT participants since they 26 represent a very small proportion of the population. Community immunity is dependent 27 on the distribution of immunity conferred by the vaccine and natural infection within the 28 population, the transmissibility of the infection, and contact patterns of individuals in the 29 population.⁶⁶ Population-level effectiveness is usually established through observational 30 studies, which would normally capture the indirect effects of a vaccine. Researchers 31 should be aware, however, that studies using surveillance data are subject to the same 32 limitations as other observational studies, and may not be appropriate to extrapolate to 33 different settings.²⁸ In such cases, dynamic models parameterised using local 34 epidemiological data can be used to estimate indirect effects of vaccines. When

1 assessing whether to include estimates of vaccine efficacy or effectiveness from RCTs

2 or from observational studies in the reference case analyses, researchers should justify

3 which data sources best represent results in populations most similar to the

4 population(s) affected by the vaccination program to be implemented.

5

6 Geographic variation should be considered with respect to vaccine efficacy and

7 effectiveness. Several studies have found that vaccine efficacy and effectiveness can

8 vary by country. Several factors have been postulated to account for these differences,

9 including: 1) differences in serotype or strain prevalence; 2) the role of climate and daily

10 mean temperatures; 3) population heterogeneities with respect to social and

11 demographic factors that influence vaccine efficacy and effectiveness; 4) co-

12 administration of other vaccines (e.g., oral rotavirus and polio vaccines co-

13 administration); and 5) differences in prevalence of other endemic infections.^{67,68}

14 Host factors should also be considered when evaluating fitness for purpose of vaccine

15 efficacy and effectiveness data. Examples of such factors include age, genetic

16 susceptibility to infection, inborn errors of immunity, the effect of nutrition on host

17 responses, previous sensitization to organisms antigenically related to the pathogen,

18 comorbidities, particularly those that can affect immune response, secondary

19 immunodeficiencies due to medications, and possible genetic differences in response to

20 a particular vaccine.⁶⁹ RCTs tend to include only healthy adults, whereas real-world

21 studies include at-risk populations that would otherwise be excluded from RCTs

22 including pregnant persons, children and those who are immunocompromised.

23

24 Vaccination coverage may differ between groups of individuals or by geographic area.

25 For example, diphtheria, pertussis and tetanus (DTaP) vaccine coverage of four or more

26 doses in two-year-old children differs between Canadian provinces, with Newfoundland

27 and Labrador achieving the highest coverage (89%) and Manitoba the lowest (66%),

28 based on the 2017 Canadian Immunization Coverage Survey.⁷⁰ Coverage is an

29 important factor in determining effectiveness at the population-level through community

30 immunity. Achieving high levels of vaccination coverage depends on the implementation

31 strategy undertaken when a new vaccination program is introduced, and the ongoing

32 strategies employed to scale up and sustain the program. For instance, health

33 promotion, information campaigns and other efforts to build community trust may

1 counteract vaccine hesitancy. The success of these strategies will depend on the 2 capacity of the resources deployed, ease of access to vaccines doses in the intended 3 population, preparedness of healthcare providers, and attitudes of both healthcare 4 providers and the public. These are all distinct elements related to vaccine coverage, 5 and different levers can be pulled to achieve better outcomes. Researchers should 6 incorporate these factors into economic evaluations to better align these evaluations with 7 decision-makers' practical needs. Importantly, including these factors focuses decision-8 makers' attention on specific implementation strategies, the relative time and effort 9 needed to execute each one, the inherent trade-offs posed by these alternative courses 10 of action, and their independent and joint effects on population coverage.⁷¹

11

12 Some vaccines provide protection only against some variations of a pathogen. For 13 example PCV13 is active against 13 out of over 90 known pneumococcal serotypes,^{9,72} 14 and 23-valent pneumococcal polysaccharide vaccine (PPSV23) is active against 23 15 pneumococcal serotypes.⁷³ HPV vaccines are available in bivalent and quadrivalent 16 forms, although there are over 100 HPV serotypes.⁷⁴ For these types of vaccines, 17 researchers should ensure that efficacy and effectiveness data being considered are 18 specific to the diseases caused by the variations of the pathogen targeted by the 19 vaccines. Researchers should also be aware that in some cases vaccines that are 20 specific to certain pathogen variations may also confer some degree of protection 21 against variations of the pathogen not covered by the vaccine. An example of this cross-22 protection has been shown with both bivalent and quadrivalent HPV vaccines 23 demonstrating some protection against infections and lesions associated with HPV 31, 33, and 45, which are non-vaccine serotypes.⁷⁴ 24

25

26 Data Synthesis

27

28 Researchers should be mindful of vaccine-specific considerations when combining data

29 from different sources. These include potential geographic and host factors outlined

30 above that may be different between study populations and the population being

31 considered in an economic analysis conducted in the Canadian setting.

32

33 Surrogate Outcomes

34

Whenever possible, the efficacy or effectiveness of vaccines should be determined with
comparative studies (either RCTs or observational studies) that report the incidence of
the infectious disease targeted by the vaccine, in the vaccinated group versus relevant
comparator(s).

5

6 The primary endpoint of these studies should be defined as clinically apparent infection 7 that meets clinical and laboratory diagnostic criteria. In some situations, it may not be 8 possible to measure cases of clinically apparent infection. One example of such a 9 situation arises when the incidence of the infection is too low to measure in a study, 10 typically constrained by its study time period and study population size. This situation 11 occurs with rare infectious diseases (e.g., meningitis due to meningococcal group B infection), or those that rarely afflict the population because current vaccines provide 12 13 effective prevention.^{75,76} Another example arises with seasonal influenza vaccines, many 14 of which receive provisional approval based on immunogenicity alone.77 15

16 In these situations, correlates of protection (CoPs), which are immune biomarkers 17 (antibodies or T-cells) that predict vaccine efficacy in vaccinated individuals, can be used as surrogate endpoints.^{75,78,79} Researchers should be aware that multiple CoPs can exist 18 for a single vaccine,^{80,81} and that different vaccine types and formulations indicated same 19 20 disease may be associated with different CoPs.^{82,83} For multivalent vaccines that provide 21 protection against multiple variations of a pathogen, higher titers of the CoP may be 22 required for protection against some variations compared to others.⁸⁴ Finally, it is 23 important for researchers to identify which dimension of prevention (e.g., preventing 24 infection, preventing disease, reducing severity of disease) is linked to a CoP since 25 correlates may differ quantitatively and qualitatively based on the preventive outcome 26 being considered.85

27

28 Extrapolation

29

30 The duration of clinical trials is often not long enough to ascertain the duration of

31 protection provided by a vaccine, and researchers have to extrapolate estimates of

32 duration of protection from clinical trial data.^{86,87} A number of different modelling

techniques (e.g., logarithmic waning, exponential waning) can be used to generate

34 duration of protection estimates, which can vary widely based on the technique chosen.

Consequently, cost-effectiveness estimates can be sensitive to assumptions on duration of protection.⁸⁶ This has been demonstrated with cost-effectiveness analyses of herpes zoster vaccine (Zostavax[®]) in Belgium, where the authors found that cost-effectiveness estimates varied considerably based on the choice of model used estimate to vaccine efficacy.³¹ Specific guidance on addressing uncertainty of the estimates of duration of protection is provided in Chapter 13 on Uncertainty.

DRAFT

1	10.	Measurement and Valuation of Health	
2	10.1	In both reference cases, the quality-adjusted life year (QALY) should be used as	
3		the method for valuing health outcomes.	
4			
5	10.2	"Health preferences should reflect the general Canadian population." [CADTH	
6		Guideline Statement]	
7			
8	10.3	In the reference cases, researchers should use health preferences obtained from	
9		an indirect method of measurement that is based on a generic classification	
10		system (e.g., EuroQol 5-Dimensions questionnaire [EQ-5D], Health Utilities Index	
11		[HUI], Short Form 6-Dimensions [SF-6D], Child Health Utility 9-Dimensions	
12		[CHU9D], Assessment of Quality of Life [AQoL]). Researchers must justify where	
13		an indirect method is not used. [CADTH Guideline Statement with amendment]	
14			
15	10.4	"The selection of data sources for health state utility values should be based on	
16		their fitness for purpose, credibility, and consistency. Describe the trade-offs	
17		among these criteria and provide justification for the selected sources." [CADTH	
18		Guideline Statement]	
19 20	QALYs are the metric used to quantify health outcomes in a CUA. QALY estimates are		
21	generated by combining data on survival and health-related quality of life (HRQoL). In		
22	order to estimate QALYs, HRQoL data in the form of a summary measure, often referred		
23	to as a health utility, is required. As the CUA implicitly espouses an extra-welfarist		
24	foundation, decision-makers are concerned with HRQoL because the key output of		
25	health interventions is health outcomes.		
26			
27	Healt	h Utility Data	
28		•	
29	The u	tilities obtained from HRQoL instruments should represent the preferences of the	
30	general Canadian population, consistent with the social decision-making standpoint		
31	adopted by these guidelines. Population preferences for health states defined in an		
32	HRQoL instrument are normally elicited from a sample of the general population using		
33		ods such as standard gamble or time trade-off.	
34			

1 Although it is possible to elicit health utilities directly from respondents, instruments 2 designed to capture health utilities indirectly provide a more efficient and consistent 3 method of obtaining this information. Both disease-specific and generic HRQoL 4 instruments are available to obtain indirect health utility measurements. The most 5 commonly used generic instruments are the EuroQol 5-Dimensions questionnaire (EQ-6 5D), Health Utilities Index (HUI), the Short Form 6-Dimensions (SF-6D), and the 7 Assessment of Quality of Life (AQoL). Instruments for children's HRQoL include Child 8 Health Utility 9-Dimensions (CHU9D), KIDSCREEN Qality of Life Questionnaire, 9 Pediatric Quality of Life Inventory (PedsQL) Generic Cores Scales, and EQ-5D-Youth 10 (EQ-5D-Y). Researchers should use HRQoL data obtained from a generic instrument to 11 estimate QALYs to ensure comparability between vaccination programs being 12 considered by decision-makers. Where multiple estimates of utilities are available, 13 source studies should be subjected to formal quality appraisal using a suitable quality

- 14 appraisal tool.⁸⁸
- 15

16 Health utility data used to populate an economic model are often derived from published 17 literature. To ensure consistency within a model, health utility valuations for all health 18 states included in the model should be obtained from the same instrument and use 19 preference weights obtained from the same population, whenever possible.⁸⁹ When this 20 is not possible, researchers should consider trade-offs between the fitness for purpose, 21 credibility, and consistency for the available data. In these cases, researchers may also 22 consider pooling health utility data using techniques such as meta-analysis or meta-23 regression, although the usefulness of these methods may be limited by the considerable heterogeneity in the valuation methods and study populations.⁹⁰ 24 25 Researchers should explore uncertainty in health utilities in sensitivity analyses.

26

27 It must be recognized that there are no valid instruments for directly measuring utility in 28 neonates, newborns, infants or young children, although this is an active area of current 29 research.⁹¹ Moreover, the construct of HRQOL for children differs by age group and is 30 conceptually different than adults.⁹² While several pediatric-specific preference-based 31 measures of health-related quality of life have been developed recently (e.g., EQ-5D-Y, 32 CHU-9D, A-QOL), all have lower age limits and typically rely on tariff sets derived from 33 adult populations. The convergent validity of pediatric-specific and adult preference-34 based HRQOL measures requires study. Despite the limitations, researchers should

1 ideally use utilities for child health states sourced from a pediatric-specific generic 2 instrument, as opposed to using adult utilities. If a pediatric-specific generic instrument is 3 not used for a child health state, this should be justified and its impact tested in 4 sensitivity analysis. The use of generic instruments is encouraged in pediatrics, despite 5 direct elicitation methods being frequently used. Utilities generated from direct elicitation 6 for health states are sensitive to framing. In cases where utilities may be missing due to 7 a child's young age (e.g., under 5 years), assumptions used should be explicit and 8 justified. Preferences should be from a general population, supplemented with child 9 valuations if available. Proxy respondents (e.g., by parents or healthcare providers) are 10 often required in pediatrics because valuation methods can be cognitively difficult or 11 require reading comprehension. However, proxy responses can systematically differ 12 from child self-report where the directionality of the discrepancy is difficult to predict.⁹³ 13 Researchers should use child utilities from instruments that are self-reported where 14 possible, and specify if proxies are used. Further, many vaccines are given in infancy or 15 childhood, some of which prevent diseases in childhood and others in diseases that 16 emerge in adulthood. Researchers should explicitly state which health states in a model 17 are related to child health states and which relate to future adult health states. In 18 economic evaluations where adult and children are modelled, consistency in the use of 19 instrument across ages is encouraged.

20

In addition to including health utility data for the population intended for the vaccination
program and any population(s) that may experience externalities related to the program,
researchers should include health utility data for informal caregivers in cases where
potential spillover effects have been identified that could affect the health states of this
population.

26

27 A more detailed discussion on HRQoL measurement and data can be found in Chapter

28 10 Measurement of Valuation of Health of CADTH's *Guidelines for the Economic*

29 Evaluation of Health Technologies: Canada 4th Edition.¹

30

31 Quality-Adjusted Life Years in Societal Perspective Economic Evaluations

32

33 There is uncertainty about whether QALYs capture only health benefits, or whether they

34 also, implicitly or explicitly, capture non-health-related effects. This uncertainty is

1 particularly germane to CUAs conducted from the societal perspective since these

2 analyses are concerned with not only costs and outcomes borne by the health system,

- 3 but also with costs and outcomes that fall onto non-health sectors. Specifically,
- 4 uncertainties exist around how to include the impacts of productivity and consumption in
- 5 the ICER estimate.
- 6

7 As noted above, QALYs are estimated using survival and HRQoL data. The HRQoL data 8 are elicited based often on health domains included in common HRQoL instruments. 9 However, it is unclear whether, or to what extent, respondents implicitly consider non-10 health-related factors when valuing these health states. In particular, the degree to which 11 respondents consider how changes in productivity and consumption may affect their 12 HRQoL, and the extent to which these considerations are implicitly incorporated into 13 respondents' valuation of their health states have received attention. The available 14 evidence suggests their influence is limited.94-97

15

16 If individuals were to account for the economic effects of productivity when valuing their 17 health states, then including monetary estimates of productivity in the numerator of the 18 ICER estimate, along with costs of other resources, double counts these impacts.⁹⁶ The 19 current consensus is that productivity and income changes are not likely to be captured 20 in QALY estimates.^{98,99} This supports the inclusion of productivity costs in the numerator 21 of the ICER estimate.

22

23 Similarly, questions have been posed about whether respondents in health state 24 valuations consider and value non-medical consumption such as clothing and housing. 25 One argument posits if the utility value of this consumption is not (implicitly) captured in 26 QALYs, it would be inconsistent to include changes in such consumption on the costs 27 side of the evaluation and therefore, these costs should be excluded.⁹⁷ Another 28 argument posits that non-medical consumption (e.g., daily food intake) is necessary to 29 stay alive. Even if respondents would not consider this in their valuations of health 30 states, it should nonetheless be included as a cost. This would be even more evident if 31 respondents would assume usual levels of consumption in answering health state 32 valuation guestions. The same argument applies to other non-medical consumption, 33 which to some extent may also contribute to an individual's HRQoL.95

In contrast to findings related to respondents' consideration of productivity changes when valuing health states, evidence suggests that respondents do consider utility of consumption when valuing health states.¹⁰⁰ This suggests that health improvements may lead to increases in the marginal utility of non-health-related consumption. Although further research is required to corroborate these findings, they do provide justification for including consumption costs in the numerator of the ICER estimate.

- 8 In summary, for the societal perspective reference case analysis, changes related to
- 9 productivity and costs of non-health-related consumption should be included in the
- 10 numerator of the ICER estimate. Further details on quantifying the costs of productivity
- 11 and consumption are found in Chapter 11 on Resource Use and Costs.

DRAFT

1 11. Resource Use and Costs

2	11.1	For each reference case analysis, researchers should systematically identify,
3		measure, value, and report all relevant resources consumed or saved as a result
4		of the delivery or implementation of the vaccination program under consideration.
5	11.2	Where possible, researchers should value relevant resources identified for all
6		sectors in monetary terms. In situations where this is not possible, researchers
7		should present the relevant resources that have been identified in the Impact
8		inventory table for economic evaluations of vaccination strategies for
9		consideration by decision-makers.
10	11.3	"Resource use and costs should be based on Canadian sources and reflect the
11		jurisdiction(s) of interest (as specified in the decision problem)." [CADTH
12		Guideline Statement]
13	11.4	When valuing and monetizing resources, researchers should select cost data
14		sources that most closely reflect the opportunity cost, given the perspective of the
15		analysis. [CADTH Guideline Statement with amendment]
16	11.5	Researchers should assess sources used for cost data based on their fitness for
17		purpose, credibility, and consistency. The selection of data sources should be
18		based on trade-offs between these criteria.
19		
20	Both i	ncreases and decreases in consumption of resources and services may result from
21	vaccir	nation programs. They are related to both the implementation of the vaccination
22	progra	am and its ongoing delivery, as well as to downstream effects of the program.
23	Reso	urce consumption may fall upon vaccinated individuals, the population at risk for
24	the di	sease of interest when the vaccination program is associated with externalities,

25 and the population that experiences spillover effects (e.g., informal caregivers).

26 Furthermore, resources consumed as a result of vaccination programs may fall within

27 the health system sector or outside the health system. Researchers should use the

- 28 Impact inventory table for economic evaluations of vaccination strategies to
- 29 systematically identify all potential resources and services associated with the
- 30 vaccination program under consideration. Once the range of resources and services
- 31 occurring as a result of a vaccination program has been identified, researchers should
- 32 determine which of the resources consumed can be measured and valued in monetary
- 33 terms.^{101,102}
- 34

1 Health System Sector

2

For the health system perspective reference case analysis, researchers should identify and include all resources within the publicly funded health system that are consumed through the delivery of the vaccination program, and resources that are consumed or saved as a result of its implementation. All health system costs incurred over the model time horizon should be included.

8

9 (i) Healthcare Costs

10

11 When assigning local unit costs to resources that have been identified as relevant to the 12 decision problem being modelled, researchers should consult the Guidance Document 13 for the Costing of Healthcare Resources in the Canadian Setting: Second Edition,¹⁰³ 14 which provides key recommendations and data sources for identifying, valuing, and 15 measuring costs within the Canadian healthcare system. For analyses that are 16 conducted within or across multiple jurisdictions, variations in the public financing of 17 specific resources and services should be indicated. Researchers should report whether 18 a single price set is used or multiple jurisdictional price sets set are applied and methods 19 used for assigning price sets to multi-jurisdictional data.

- 20
- 21

22

a. Formal caregiving

23 Individuals requiring a caregiver may receive this care from either a professional 24 caregiver such as a nurse or a nursing assistant who is hired to perform these 25 services, or an informal caregiver, usually a family member. Professional 26 caregiver time should be valued at the hourly wage rate that would be paid to an 27 individual who performs this service. Formal caregiving costs may be incurred or 28 may not be incurred by the publicly funded health system, depending on the 29 precise nature of these costs and the relevant jurisdiction. Informal caregiving is 30 discussed later in the chapter under *Productivity*.

31 32

b. Future Healthcare Costs

1 For vaccination programs that confer a survival benefit in relation to the 2 comparator(s) being considered in the economic evaluation, researchers should 3 include in the reference case analyses future healthcare costs, both related to the 4 infection and disease of interest and unrelated. This recommendation is 5 underpinned by the following considerations: 1) there is an opportunity cost 6 associated with life-prolonging interventions since they increase future health 7 expenditures in those individuals— expenditures which could have been used 8 towards other individuals' healthcare needs; 2) it is often difficult to distinguish 9 related costs from unrelated costs, such as in the case of different diseases that 10 share overlapping physiologic pathways (e.g., diabetes and coronary heart 11 disease), potentially leading to arbitrary decisions about which costs are related 12 versus unrelated; and 3) internal consistency: the benefits related to future 13 medical spending are already included in CUAs through estimates of survival and 14 quality of life, and are based on the assumption that the individual will receive 15 future medical care, both related and unrelated to the condition of interest.95,101,104,105 16

18 Excluding future costs leads to lower incremental cost estimates and ICER 19 estimates for life-prolonging interventions, and may make them seem more 20 economically attractive than those that improve quality-of-life. Including future 21 costs, however, increases incremental cost estimates and ICERs for life-22 prolonging interventions, leading in some cases to a do-nothing option (i.e., 23 where patients do not survive) being more cost-effective than providing treatment for a disease.¹⁰⁶ In some cases even relatively inexpensive life-prolonging 24 25 interventions in patients with high costs of ongoing care may not be cost-effective when future costs are considered in an economic evaluation.¹⁰⁷ Researchers 26 27 should present outcomes and costs in a disaggregated manner so that decision-28 makers are aware of how individual components included in the analysis 29 contribute to the overall cost-effectiveness of the vaccination program. 30 Researchers should present: 1) the expected health outcomes of the vaccination 31 program and the comparator(s); 2) the direct health system costs resulting from 32 the vaccination program and the comparator(s) but excluding costs of future care; 33 and 3) the expected increase in costs of ongoing care resulting from improved survival for the vaccination program and comparator(s).¹⁰⁷ 34

1

Estimates of future healthcare costs may be obtained from data published by the
Canadian Institute for Health Information's National Health Expenditure
Trends.¹⁰⁸ In situations where cost estimates are required for populations with
high costs of ongoing care (e.g., dialysis patients, solid organ transplant
recipients).^{109,110} researchers may have to consult the published medical
literature to obtain these estimates.

8

9 (ii) Public Health Costs

10

11 Public Health costs may represent a large share of the costs associated with vaccination 12 programs, and management of infectious diseases. Accurately guantifying these costs is 13 necessary to ensure that results generated from economic evaluations of vaccination 14 programs are valid, and lead to optimal funding decisions. Public Health costs can be 15 categorized as either program-related costs or intervention-related costs. Program-16 related costs are the costs of program implementation, delivery and sustainment costs. 17 They include costs of public health campaigns and health promotion activities; 18 transaction costs related to introduction of new vaccines or switching between vaccines; 19 and costs related to population-based screening, epidemiological surveillance, contact 20 tracing, case investigations, and outbreak investigations. Specific components that 21 should be considered when quantifying these costs include personnel costs, overhead 22 costs, travel costs, and other service-related and administrative costs.^{103,111} Specific 23 components that should be considered when quantifying costs of disease outbreaks 24 include laboratory serologic testing; personnel time related to contact tracing, symptom 25 screening, travel, monitoring, and follow-up; post-exposure prophylactic vaccines or immune globulin doses and associated administration costs.¹¹²⁻¹¹⁵ Intervention-related 26 27 costs include costs of vaccine doses, distribution (e.g., transportation of vaccines and 28 cold storage), and administration of the vaccine, including any wastage and ancillary 29 supplies required. Researchers should present costs related to different aspects of 30 implementation and ongoing delivery of the vaccination program in a disaggregated 31 manner. Further, researchers should elaborate on the different levels of intensity of the 32 implementation strategy, which is especially relevant for public health campaigns and 33 health promotion activities for instance, as they can produce different levels of benefit.

1 Given the paucity of published data on program-related Public Health resource use and 2 prices in the Canadian setting, researchers may have to rely on data obtained from local 3 Public Health authorities or provincial ministries of health through personal 4 communication. Although costs from local Public Health authorities and provincial 5 ministries are jurisdiction-specific, they may be generalizable to other areas. When 6 determining the applicability of data from one jurisdiction to another, researchers should 7 consider factors such as geographic similarities, population characteristics, and 8 epidemiologic patterns.

9

10 There are limited Canadian data available on intervention-based Public Health resource 11 use and prices. Some provincial Public Health agencies, such as Quebec's Institut national de santé publique du Quebec,¹¹⁶ publish findings of their work online, which 12 13 may include epidemiologic surveillance and cost data relevant to the economic 14 evaluation of a vaccination program. If the required data are not available through 15 publications from provincial Public Health agencies, researchers may have to obtain 16 these data from provincial ministries of health or local Public Health authorities. The 17 actual price paid by governments for vaccine doses is confidential. Researchers should 18 use the manufacturer's list price in the reference case analyses and conduct 19 deterministic sensitivity analyses using plausible discounted prices. Researchers should 20 also consider cost items related to the administration of vaccine doses, as they may vary 21 considerably based on the setting of delivery. For example in Alberta, community-based 22 delivery of HPV vaccine is considerably more expensive than school-based delivery.¹¹⁷ 23 Resources and services related to providing culturally safe access to healthcare and 24 vaccine program communication materials should also be considered in situations where 25 they are applicable.

26

27 (iii) Healthcare Costs Not Funded by the Health System

28

Some services associated with vaccination programs may not be reimbursed or publicly funded by the healthcare system. Services that are excluded from the publicly funded healthcare system may vary by jurisdiction or region. Examples of such costs include long-term care services, private nursing, drug treatments for individuals who do not have coverage through a publicly funded drug insurance program, non-prescription drugs, as well as ancillary costs related to items such as private insurance copayments, dental and

1	vision care, assistive devices, physiotherapy and others. These costs may be funded
2	through private insurance plans, by the individual(s), or a combination of both.
3	Regardless of how these costs are funded, they should be quantified and included in the
4	incremental costs and ICER (where applicable) for the societal perspective reference
5	case analysis.
6	
7	Non-Healthcare Areas
8	
9	Researchers should also identify all resources consumed as a result of the
10	implementation or ongoing delivery of the vaccination program that fall outside of the
11	publicly funded health system and quantify their corresponding costs. For example,
12	relevant non-health sectors for the societal perspective reference case analysis could
13	include: direct out-of-pocket costs (e.g., co-payments, transportation costs, private
14	caregivers), paid and unpaid labour time losses, non-medical consumption, and services
15	not funded by other sectors including education, social services, and environment.
16	Guidance on identifying resources and quantifying costs for non-healthcare sectors is
17	presented below.
18	
19	(i) Direct Out-of-Pocket Costs
20	
21	Estimates of direct out-of-pocket costs (e.g., transportation costs, accommodation costs)
22	should be included in economic evaluations of vaccination programs. Transportation
23	costs include costs related to public transit, including fully accessible barrier-free
24	transportation when necessary, taxis, personal vehicle use, and parking fees. ¹¹⁸
25	
26	(ii) Productivity
27	
28	Researchers should consider the effects of vaccination programs on the productivity of
29	vaccinated individuals and caregivers, and where applicable, on macroeconomic
30	consequences. For the former, vaccine-related productivity improvements may occur
31	through: 1) increased paid and unpaid labour productivity related to either prevention of
32	illness, or decreased severity of illness in vaccinated individuals; and 2) increased
33	productivity of caregivers related to decreased care needs for sick individuals. ^{7,25,119}

34 When productivity gains for life-prolonging interventions are included in an analysis from

the societal perspective, they may attenuate or offset increased incremental costs due to
increased future healthcare consumption in survivors.

- 3 4 a. Individual Productivity 5 6 Productivity costs are output losses associated with productive time spent in paid 7 labour, or unpaid labour (e.g., volunteering, helping, mentoring) including 8 household production (e.g., cooking, cleaning, shopping, raising children).¹⁰¹ 9 There are two primary methods for quantifying lost productivity related to paid 10 11 work: the human capital approach, and the friction cost approach. The human 12 capital approach is based on the cost of forgone productive time, whereas the 13 friction cost approach attempts to estimate overall societal production losses, assuming replacement of ill workers in the formal labour market.^{101,120-122} 14 15 16 The human capital approach is commonly used to value lost production. It 17 typically requires estimates of time lost from paid work, and averages wage rates 18 of the involved individuals. As such, it may be seen as estimating the lost 19 production (or income) from an individual perspective, due to illness, disability, or 20 death. As the human capital method does not account for societal replacement 21 mechanisms, especially for longer periods of absence (e.g., in case of disability 22 or premature death), it has been suggested that it likely overestimates the true 23 cost of lost production from a societal perspective.¹²³ This is a particularly 24 important consideration in situations where childhood death or lifelong disability 25 may be avoided as a result of a vaccination program. 26 27 The friction cost approach, on the other hand, attempts to quantify lost 28 productivity on a societal level based on the assumption that production levels
- can be restored by substituting labour for labour (e.g., in case of unemployment)
 or for capital.¹²⁰ This implies that after some 'friction period' production losses
 cease to occur from a societal perspective. Macro-economic consequences of
 changes in labour supply and unemployment benefits have been estimated to be
 small for typical health care programs. Applying this method requires more

- detailed information on periods of absence, the available labour pool, and the
 relevant friction period in a country or province.¹²²
- While both of these methods primarily focus on valuing lost production in the
 context of paid work, changes in productivity related to unpaid work should also
 be captured. Lost productivity in the context of unpaid work can be captured by
 valuing lost hours with an appropriate value. Estimations of (changes in)
 productive time in unpaid work for the relevant population may be difficult to
 obtain in some cases.¹²⁴ Other than using general estimates from existing
 sources, questionnaires may be used to estimate these changes.¹²⁵
- 11

3

- 12 Researchers should calculate total change in productive time, related to both 13 paid and unpaid work, attributable to the vaccination program. Researchers 14 should account for losses of an individual's productive time related to obtaining a 15 vaccine, seeking treatment, illness, disability, and death of vaccinated or 16 otherwise affected individuals. Changes in productivity associated with 17 vaccination programs should be quantified using the human capital approach. 18 Given that it is the most commonly recommended approach in 19 pharmacoeconomic guidelines across different countries, ¹²⁶ it allows increased 20 comparability between economic evaluations of vaccination programs undertaken 21 in different jurisdictions.
- ∠ı 22

23 For the societal reference case analysis, researchers should include the full-time 24 period over which affected individuals are expected to incur paid production 25 losses. These losses should be valued based on age-specific average income 26 and number of hours worked based on Statistics Canada data^{127,128} combined 27 with the disease-specific likelihood of an individual participating in the labour 28 force. Using the same wage rate for both genders is a correction for 29 measurement bias because females are on average paid less than male for the 30 same work.129

31

In most cases, there will be equity considerations related to whether and how
 productive time is valued. If it is differentially valued based on attributes such as
 age, gender, or health status, results of an economic evaluation could favour

1 groups with the greatest income-earning potential and disadvantage other groups 2 such as children who do not work or individuals with disabilities or severe health 3 conditions that prevent them from holding high-income jobs.¹⁰¹ In these 4 situations, researchers should conduct an additional sensitivity analysis using the 5 average income and the average number of full-time hours worked for all 6 Canadians based on Statistics Canada data.^{127,128} Although the measurement of 7 these losses is imperfect and biased towards high-wage earners, this approach 8 reveals the efficiency losses that decision makers need to be prepared to accept 9 each time they choose an option that is neutral to individual characteristics with 10 respect to production.

12To account for the likely overestimation of production losses associated with the13human capital approach, researchers should include an additional sensitivity14analysis that accounts for production losses for a single year using the average15number of full-time hours worked for all Canadians based on Statistics Canada16data.^{127,128} Average yearly income and average yearly number of hours worked17for all Canadians should be used for this analysis. This approach represents a18naïve friction cost approximation.

Although productivity losses may result from both absenteeism (time off work) and presenteeism (continuing to work but with reduced productivity), researchers are not required to account for the effects of presenteeism in their estimates of lost productivity in the reference case analysis. It is often difficult to collect this information given that it requires survey data from affected individuals and recall can be subjective in many cases.^{125,130}

26

11

19

Lost unpaid production should be valued by estimating lost hours of unpaid work,
and valuing this using the replacement cost method. Although unpaid work may
differ in terms of tasks performed and required skills, for the reference case
analysis, lost hours should be valued using the wage-rate of a professional.
Researchers should exclude costs of leisure time from the economic evaluation
of vaccination programs.

- 33
- 34

b. Informal Caregiver Productivity

1	
2	As described above, individuals requiring a caregiver may receive this care from
3	either a professional caregiver, or an informal caregiver, usually a family
4	member. Two approaches have been proposed for valuing informal caregiver
5	time: 1) the replacement cost approach; and 2) the opportunity cost approach.
6	The replacement cost approach is based on the estimated cost of hiring a paid
7	caregiver should informal care not be available. The opportunity cost approach is
8	based on the cost of displaced productive time that results from time spent
9	providing informal care. ¹³¹ Since individuals may receive a mix of formal and
10	informal care, researchers should use the replacement cost approach to value
11	caregiver time for the societal perspective reference case analysis. Such
12	estimates can be used alongside estimates of potential health spillover effects
13	due to informal care, captured in terms of caregiver QALYs. ¹³²
14	
15	c. Macroeconomic Consequences
16	
17	Although most vaccination programs are unlikely to have large macroeconomic
18	impacts, those that are designed to prevent widespread disease pandemics,
19	such as the 2020 COVID-19 pandemic caused by the SARS-CoV-2 virus, could
20	attenuate important consequences. Macroeconomic impacts include labour
21	supply shocks and widespread business closures, which may affect labour pools
22	and workforce participation rates, and changes in household consumption
23	preferences. ¹³³
24	
25	(iii) Non-Medical Consumption

26

Non-medical consumption represents expenditure on non-health-related items that
 contribute to overall welfare. These items include individual financial expenditures, and
 consumption of public goods and services such as clean water and safe roads.^{95,101}
 Researchers should include consumption costs whenever they will be altered by the
 vaccination program.

32

33 Researchers should use Statistics Canada data on household spending as the

34 information source for non-medical consumption (Table: 11-10-0222-01, formerly

CANSIM 203-0021, "Household spending, Canada, regions and provinces").¹³⁴ In order 1 2 to obtain an estimate of non-medical consumption, researchers should exclude health 3 consumption from total consumption. Estimates of individual consumption should be 4 obtained by adjusting household consumption estimates using an equivalence scale, to 5 account for consumption by household size, reflecting the fact that one-person 6 households would have higher per-person consumption compared to multi-person 7 households.¹³⁵ For vaccination programs that result in changes to consumption. 8 researchers should subtract individual estimates of consumption from individual 9 estimates of productivity during the relevant time period. To ensure consistency between 10 estimates of productivity and consumption, estimates of consumption should not be 11 stratified by gender for the reference case analysis.

12

13 (iv) Education

14

Vaccination programs may affect educational outcomes by preventing diseases that lead to serious morbidities that, in turn, could affect an individual's level of educational achievement. For example, a Danish study found that children who suffered from bacterial meningitis experienced lower levels of educational achievement and economic self-sufficiency in adulthood.¹³⁶

20

21 Higher levels of educational achievement are associated with a greater likelihood of 22 labour market participation and higher labour market earnings.^{137,138} In Canada, it is 23 estimated that each additional year of schooling increases lifetime earnings by 24 approximately 11 to 12%. Assuming that decreasing an individual's education level 25 similarly decreases lifetime earnings, it is estimated that each month of education loss 26 will result in an approximately 1% drop in lifetime earnings.¹³⁹ Changes in earnings 27 related to education achievement should be accounted for in estimates of lost (or 28 gained) productivity, and researchers should ensure that these costs are not double 29 counted when considering educational impacts of vaccination programs. 30 31 In addition to effects on educational achievement and labour market productivity, 32 vaccination programs may result in direct effects on the education sector. For example,

33 children who have suffered from bacterial meningitis may experience cognitive

34 impairment, hearing loss, seizures, and learning disabilities,¹⁴⁰ and may require in-school

1 special education resources. Boards of education and schools may also invest in 2 vaccination delivery programs, as well as ancillary programs to improve the learning 3 environment during a pandemic (e.g., upgraded heating, ventilation, and air conditioning, 4 reduced classroom size, virtual learning). 5

6 Researchers should consider potential education-related outcomes and direct effects on 7 the education sector that could result from the vaccination program and the 8 comparator(s) being considered. Where possible, these effects should be monetized for 9 inclusion in the ICER estimate. For outcomes that may be difficult to monetize (such as 10 disruptions to learning outcomes as a result of school-based vaccine delivery, paediatric 11 disease and disability, or death/disability of a close family member), researchers should 12 nonetheless identify them and include in them in the *Impact inventory table for economic* 13 evaluations of vaccination strategies for consideration by decision-makers.

14

15 (v) Social Services

16

17 Vaccination programs may affect social services, community services, and child and 18 youth services by preventing diseases that lead to serious morbidities. Examples include 19 disability support, awareness programs, family respite, and programs to improve access 20 to vaccination programs. Researchers should identify (and where feasible monetize) 21 consequences of social services.

22

23 (vi) Environment

24

25 Vaccination programs and the comparator(s) included in the analysis may have 26 environmental impacts related to the manufacture or distribution of vaccine doses, as 27 well as to implementation the program. For example, vaccines have been shown to 28 reduce antibiotic use,^{141,142} which may lead to decreased residual antibiotics from 29 sources such as households, the pharmaceutical industry, and hospitals in wastewater, 30 which has been identified as a reservoir of antibiotic resistant organisms.¹⁴³ 31

32 Environmental impacts may should be identified and included in the *Impact inventory*

33 table for economic evaluations of vaccination strategies for consideration by decision1 makers. They should be monetized where possible, although this is sometimes difficult

- 2 to do.
- 3

4 (vii) Other Areas

5

6 Researchers should consider other sectors that may offer services or programs relevant

- 7 to specific vaccination programs. Examples of such sectors could include the
- 8 legal/criminal justice sector (e.g., the cost to the state of handling potential lawsuits
- 9 against vaccine manufacturers resulting from adverse effects of vaccines, or the cost to
- 10 the healthcare system of lawsuits from patients if a vaccine is not introduced), or the
- 11 housing sector (e.g., changes in type of housing or adaptations to housing required
- 12 because of functional disabilities resulting from infection, or to improve ventilation/reduce
- 13 crowding to reduce infection transmission).

DRAFT

1	12.	Analysis	
2	12.1	Incremental cost-effectiveness ratios (ICERs) and, where useful for interpretation,	
3		net monetary benefits or net health benefits, should be calculated for both	
4		reference case analyses.	
5	10.0		
6 7	12.2	"For analyses with more than two interventions, a sequential analysis of cost- effectiveness should be conducted following standard rules for estimating ICERs,	
8			
8 9		including the exclusion of dominated interventions." [CADTH Guideline Statement]	
10			
11	12.3	The expected values of costs and outcomes, where possible, should be	
12		generated probabilistically to reflect the overall uncertainty in the model	
13		parameters.	
14			
15	Resea	archers should generate two sets of estimates of expected values for costs related	
16	to eac	h intervention considered in the economic evaluation: one for the publicly funded	
17	health	system perspective reference case analysis, and the other for the societal	
18	persp	ective reference case analysis. One estimate of expected values for outcomes (i.e.,	
19	QALY	s) should be generated for use in both reference case analyses. These estimates,	
20	where	possible, should be generated probabilistically so that the expected values reflect	
21	the ov	rerall uncertainty in the model parameters. In most cases, the probabilistic analysis	
22	will tal	ke the form of a Monte Carlo simulation, where an appropriate point estimate,	
23			
24			
25	as estimates of incremental costs and incremental effectiveness. All values, including		
26	incremental estimates, must be reported with 95% confidence or credible intervals as		
27	indica	tors of precision. These intervals can be obtained from the 2.5% and 97.5%	
28	bound	Is from the generated simulations. Additional indicators of precision may also be	
29	appro	priate if the distribution of uncertain outcomes is not approximately Gaussian. In	
30	cases	where probabilistic analyses are not possible, estimates of these values should be	
31	0	ated deterministically. This scenario is most likely to occur when the computational	
32	-	required for a probabilistic analysis is a limiting factor, especially for agent-based	
33	model	ls.	
34			

1 For the publicly funded health system reference case analysis, the following costs and 2 outcomes should be incorporated: all costs directly borne by the publicly funded health 3 system in Canada, and QALYs that accrue to individuals who are vaccinated, individuals 4 who experience externalities related to the vaccination program, as well as QALYs that 5 accrue to informal caregivers. For the societal perspective reference case analysis, costs 6 and outcomes from the publicly funded health system perspective should be included, 7 along with the following, at minimum: patient-borne costs, caregiver costs, and 8 productivity costs. Non-health impacts, such as consumption, social services, education, 9 and environment, should also be included when relevant.

10

11 Depending on the position of scenarios in the cost-effectiveness plane, it may not be 12 sensible to calculate ICERs, such as in the case of dominance of the vaccination 13 strategy or the alternative care comparator. In all cases, however, mean values for costs, 14 effectiveness, incremental costs and incremental effectiveness should be reported with 15 95% confidence or credible intervals. When the mean incremental values for costs and 16 effectiveness are both positive, ICERs should be presented (i.e., the ratio of the 17 difference in expected costs to the difference in expected outcomes for the two 18 interventions being compared). Specifically, when two interventions are being compared, 19 there should be an ICER for each reference case perspective. Where there are more 20 than two interventions being considered in the analysis, sequential ICERs should be 21 presented. This approach involves comparing each intervention to the next most costly 22 intervention, and excluding all interventions that are either dominated or subject to 23 extended dominance. Graphically, results should be presented as health production 24 functions or cost-effectiveness efficacy frontiers.

25

26 In cases where subgroup analyses have been conducted, expected values for costs and

27 outcomes as well as ICERs should be generated for each relevant subgroup in

accordance with the guidance presented in this chapter. In cases where multiple

29 regional or provincial/ territorial publicly funded health system perspectives have been

30 analyzed, findings should be reported for each one.

1	13.	Uncertainty	
2	13.1	Researchers should address parameter uncertainty using a probabilistic	
3		reference case analysis, where possible, as well as deterministic sensitivity	
4		analyses.	
5			
6	13.2	"Methodological uncertainty should be explored by comparing the reference case	
7		results to those from a non-reference case analysis that deviates from the	
8		recommended methods in order to examine the impact of methodological	
9		differences." [CADTH Guideline Statement]	
10	40.0		
11	13.3	Cost-effectiveness acceptability curves (CEACs) and cost-effectiveness	
12		acceptability frontiers (CEAFs) should be used to represent the uncertainty in the	
13		estimates of costs and outcomes when these estimates have been generated	
14 4 5		probabilistically. [CADTH Guideline Statement with amendment]	
15 16	13.4	When the decision problem includes the option of commissioning or conducting	
17		future research, value-of-information analysis may be helpful to characterize the	
18		value of these options and design future research and may be included in the	
19		reference case analysis. [CADTH Guideline Statement with amendment]	
20			
21	13.5	Scenario analyses should be used to assess structural uncertainty. [CADTH	
22		Guideline Statement with amendment]	
23			
24	Decisio	on-makers need information about uncertainty related to the results of economic	
25	evaluations of vaccination programs in order to avoid making suboptimal funding		
26	decisio	ons. Specifically, three types of uncertainty should be examined and reported:	
27	parameter, structural, and methodological.		
28			
29	Param	eter Uncertainty	
30			
31		eter uncertainty, also called second-order uncertainty, refers to uncertainty in	
32	•	eter estimates that are used to populate a model. ^{46,144,145} This differs from random	
33		lity, also called first-order uncertainty or stochastic uncertainty, as well as from	
34	heterogeneity. Most guidelines on conducting economic evaluations of healthcare		
35	interventions recommend using probabilistic reference case analysis, and/ or		

1 probabilistic sensitivity analysis (PSA) to explore parameter uncertainty, but in rare 2 situations, this technique may not be feasible with dynamic models. Such situations arise 3 when models are particularly complex (e.g., agent-based simulations), or when only 4 limited computing power is available. In dynamic models, many parameters related to 5 transmission, such as contact patterns between individuals and prevention-related 6 behaviours, may be correlated and these correlations must be preserved in the models 7 to generate sensible results that fit to existing data (e.g., epidemiologic surveillance 8 data). In some cases, correlations between parameters may be unknown,^{28,52} although they can sometimes be established using Bayesian parameter inference methods.^{146,147} 9 10 In these instances, researchers may be required to choose between a complex model 11 structure that does not allow for probabilistic analysis, and a simpler structure that allows 12 exploration of the impact of parameter uncertainty.

13

14 Where feasible, parameter uncertainty should be addressed probabilistically through 15 probabilistic reference case analyses. Parameter ranges chosen to assess uncertainty 16 should, where possible, be based on estimates from observational studies or 17 surveillance data. Results of these analyses should be presented as cost-effectiveness 18 acceptability curves (CEACs) or cost-effectiveness acceptability frontiers (CEAFs). 19 Scatter plots on the cost-effectiveness plane may be provided alongside CEACs and 20 CEAFs. Scatter plots are useful to observe the density and spread of the iterations, and 21 to assess inflection points and the shape of the ellipses produced.

22 In addition to quantifying uncertainty probabilistically in the reference cases, researchers 23 should conduct deterministic sensitivity analyses (DSA) on individual model parameters 24 to gain insight into the isolated effects of variations in these parameters that is provided 25 by deterministic methods. In particular, researchers should conduct a DSA on vaccine 26 price using a number of plausible values since the actual unit price of vaccine doses in 27 Canada is often confidential. DSAs should also be conducted on estimates of vaccine 28 effectiveness as there is often a high degree of uncertainty in these parameters. Ideally, 29 the DSA should be based on the output of the probabilistic analysis rather than 30 assuming base case values (e.g., using partial rank correlation coefficients or linear 31 regression).

- 1 Researchers should consider conducting threshold analyses on the most uncertain 2 parameters that may not be based upon empirical evidence (e.g., implementation 3 parameters such as population coverage), so that decision makers are able to ascertain 4 ranges of parameter values that result in a cost-effective vaccination program. 5 6 Researchers can present results of one-way (or univariate) DSAs using a tornado 7 diagram, and of two-way DSAs using two-way threshold graphs.¹⁴⁵ 8 9 When conducting DSAs researchers should identify parameter regions associated with 10 distinct model behaviours such as epidemic spread or extinction of disease, and state 11 whether the sensitivity analysis has been confined to a single region. If the sensitivity
- 12 analysis spans more than one region, researchers should indicate the probability of
- 13 achieving different disease equilibrium states as the parameter values vary.²⁸

When probabilistic analyses are not undertaken in the context of non-linear dynamic
models, researchers should conduct comprehensive DSAs on uncertain parameters. In
these situations, researchers may consider using novel DSA methods such as stepwise
DSA and distributional DSA.¹⁴⁸

18

19 The effects of parameter uncertainty may be particularly pronounced in dynamic models 20 compared to static models because of non-linearity in these models that can lead to 21 more variable population outcomes model behaviour in different parameter regions. For 22 example, a small change in parameter values may alter the model state from a disease-23 free state to a state of endemic equilibrium when the basic reproduction number (R_0) is 24 near a value of 1. These model behaviours have implications for the effectiveness of 25 vaccination programs. If the program is introduced near a threshold state (e.g., 26 beginning of an epidemic), its indirect effects may be substantial compared to a program 27 introduced at disease equilibrium where its effectiveness may exhibit a linear relationship 28 between the number of individuals vaccinated and prevention of the disease of interest.²⁸ 29 30 It can often be difficult to obtain accurate parameter estimates for infectious disease

31 models since researchers frequently must rely on observational studies or surveillance

- 32 data.²⁸ Parameter values derived from surveillance data may be biased because the
- 33 proportion of cases detected is often low and varies considerably between different
- 34 diseases, even for infectious diseases that are reportable as part of Public Health

surveillance requirements.¹⁴⁹ Severity of the infectious disease impacts detection. For
 example, infection with pertussis may be asymptomatic, associated with mild symptoms,
 or severe coughing or even death.¹⁵⁰ Thus, surveillance systems that rely upon passive
 reporting often overestimate disease severity, morbidity, and mortality, while
 underestimating the true incidence of infection in the population.^{28,150}

7 Uncertainty in estimates of vaccine effectiveness may arise from differences between 8 data obtained from RCTs compared to that obtained from large observational studies. In 9 RCTs, the force of infection does not change and leads to an underestimate of the true 10 population-based effectiveness of the vaccine because these studies do not account for 11 indirect effects of vaccination (i.e., impact of community immunity). On the other hand, 12 large population-based observational studies of vaccine effectiveness do account for 13 indirect effects, but they are limited by the potential for selection bias and unmeasured confounding.^{28,151} Selection bias may result from systematic differences in sampling of 14 15 individuals who are vaccinated compared to those who are not, resulting in a distribution 16 of exposures and outcomes that is no longer representative of the source population. 17 Confounding occurs when all or part of the apparent association between the exposure 18 (the vaccination program) and outcome (e.g., hospitalizations averted, deaths averted) 19 is, in fact, accounted for by other variables that affect the outcome and are not 20 themselves affected by exposure. Examples of these factors could include level of 21 access to healthcare services, socioeconomic status, and prevalence of natural 22 immunity.¹⁵¹ Researchers should note that observational studies of vaccine effectiveness 23 are difficult to conduct, and usually cannot be undertaken prior to a vaccine being 24 licensed. Also, effectiveness of vaccination programs for preventing disease in both 25 vaccinated and unvaccinated individuals at the population-level (community immunity) is 26 dependent upon vaccine coverage and dose completion rates. Uncertainty in these 27 parameters should be accounted for. When considering observational studies of vaccine 28 effectiveness for inclusion in economic evaluations, researchers may consider referring 29 to published guidelines for assessing evidence from comparative effectiveness studies.^{152,153} 30

31

32 Uncertainty in parameters related to transmission of infection between individuals should

33 be reflected in an uncertainty analysis. These parameters include contact patterns

34 between individuals, as well as other behaviours that may influence disease prevention

1	and control. Researchers should account for any differences in these parameters
2	between groups. For example, in diseases where asymptomatic or mildly symptomatic
3	individuals can transmit infection to others, these individuals are less likely to modify
4	their behaviours to reduce transmission compared to individuals whose symptoms are
5	more severe. ²⁸
6	
7	In some cases parameter values are estimated using models, which could be considered
8	sub-models of the primary decision-analytic model. ¹⁴⁴ For example, a predictive model
9	may be required to establish the relationship between immune biomarkers that are
10	vaccine CoPs, and the incidence of clinically apparent disease. In this case, uncertainty
11	in the values related to CoPs as well as uncertainty in the methods used to model the
12	relationship between CoPs and clinical disease need to be accounted for.
13	
14	When calibration is used to estimate model parameters, uncertainty in the estimates
15	derived from the calibration process should be explored. ¹⁴⁵ As Taylor et al. have
16	demonstrated in their cost-effectiveness analysis of HPV vaccine, failing to account for
17	uncertainty related to calibrated parameters in the model underestimates the true extent
18	of uncertainty in the cost-effectiveness estimates.61
19	
20	Structural Uncertainty
21	
22	Structural uncertainty relates to the choice of model structure. When constructing models
23	for economic evaluation of vaccination programs, researchers need to ensure that model
24	structure accounts for factors related to transmission of infection between individuals,
25	including the role of population subgroups that may be at high risk for transmitting or
26	acquiring infection, the natural history of the disease being modeled, as well as the direct
27	and indirect effects of the vaccination program. ^{28,46}
28	
29	Structural uncertainty related to transmission of infection can be related to any of the
30	following factors: 1) mode of transmission; 2) the relationship between severity of
31	symptoms and transmissibility (i.e., whether asymptomatic or minimally symptomatic
32	individuals can transmit infection); 3) mixing and contact patterns of individuals within
33	populations; and 4) behavioural changes in response to disease outbreaks. ^{28,154,155}
34	Researchers should test alternate assumptions related to these factors in all applicable

1 situations to ensure that uncertainty related to transmission has been adequately

2 examined.

3

For certain infectious diseases, population subgroups may be epidemiologically important to disease transmission, or to risk of infection in the overall population. These subgroups may intersect with equity-relevant subgroups, which are discussed in Chapter 14 on Equity. Examples include persons with hepatitis A, men who have sex with men and injection and non-injection drug users— all of whom are at high risk for transmitting and acquiring infection.¹⁵⁶ Model assumptions of the role of high-risk groups with respect to transmission should be tested to understand the degree of uncertainty they contribute.

12 Uncertainty about the natural history of an infectious disease often centers on whether it 13 leads to latent infection or natural immunity. This is the case with certain high-risk strains 14 of HPV. Decision analytic models, where different assumptions on latency and natural 15 immunity to these HPV strains were employed, demonstrated that results are highly 16 sensitive to these assumptions.¹⁵⁷ Accordingly, researchers should account for such 17 uncertainties in the structure of a model. In this HPV example, both susceptible-18 infectious-susceptible (SIS) and susceptible-infectious-removed (immune) (SIR) 19 frameworks could be examined to assess how results vary with alternate model 20 structures.²⁸

21

Important aspects of structural uncertainty related to vaccination programs include the
 timing of vaccine doses, the duration of protection provided by the vaccination, and any
 indirect effects that may result from the vaccine.^{28,86} When applicable, the use of vaccine
 booster doses should also be evaluated.

26

27 In situations where there is uncertainty about whether the protection provided by a 28 vaccine wanes, researchers should test different plausible assumptions related to 29 duration of protection. These assumptions, where possible, should be based on 30 immunologic evidence on the relationship between immune correlates of long-term protection and occurrence of clinical disease in the post-vaccination period.^{86,158} 31 32 Epidemiologic data related to disease outbreaks, when available, might also be useful in 33 modelling duration of protection conferred by vaccines as has been demonstrated with the examples of childhood mumps vaccination,¹⁵⁹ and whole-cell and acellular pertussis 34

vaccination.¹² Examples of some methods used to predict duration of effect include
linear functions, logarithmic functions, and exponential functions. Constant functions are
used in models that assume no waning of protection.^{31,86}

4

5 One of the critical decisions faced by researchers conducting economic evaluations of 6 vaccination programs is whether to model the indirect effects of vaccines using dynamic 7 models. It has been argued that if the only indirect effect of a vaccination program is 8 community immunity, then the only uncertainty in results based on a static model (which 9 does not account for these effects) is on how much more favourable the results of the 10 economic evaluation would be towards the vaccination program being evaluated.⁹ This 11 uncertainty, however, is only acceptable in situations where a static model has 12 demonstrated that a vaccination program is cost-effective; in cases where the 13 vaccination program has not been found to be cost-effective, it is problematic since a 14 dynamic model could either confirm the lack of cost-effectiveness or produce a result 15 that demonstrates that the vaccination program is indeed cost-effective. 16 17 In cases where there are indirect effects associated with a vaccine such as age-shifting 18 of disease or serotype replacement, the decision to use a static model instead of a 19 dynamic model could result in a greater degree of uncertainty. Dynamic models are

20 required to account for uncertainties on age-shifting of the disease in economic

21 evaluations of vaccination programs. Researchers should consider various scenarios

22 related to changing epidemiology of disease after the introduction of a vaccination

program to assess the effects of age-shifting on the results of an economic evaluation.

24 The effects of serotype replacement should also be explored using dynamic models.

25

Decisions on how to address uncertainty related to the different dimensions of vaccine effectiveness (prevention of clinical disease, severity of clinical disease, infection, and infectiousness) may be complicated in cases where these effects are not well understood. For example, it is thought that meningococcal B vaccines do not provide community immunity by preventing transmission of the bacteria between individuals, but data on the true extent of the effectiveness of this vaccine is limited.¹⁶⁰ Researchers should use different model structures to explore this type of uncertainty where relevant.

1 Structural uncertainty influences results of economic evaluations at least to the same extent as parameter uncertainty, and often to a greater extent.^{46,145} It is particularly 2 3 important to explore structural uncertainty in dynamic models with uncertainty analysis 4 because of their non-linear effects, which can lead to variable model behaviour.²⁸ 5 Scenario analysis should be used to explore structural uncertainties in models. This 6 technique involves testing alternate model scenarios that are underpinned by different 7 plausible structural assumptions. Results from each scenario analysis should be 8 presented individually. Model averaging can then be used to summarize results from all 9 of the alternate model scenarios that were tested. When averaging results from the 10 scenarios analyses, weightings for each model based on the model's predictive ability 11 according to available data (e.g., measures of fit) should be used.^{161,162} When weightings 12 cannot be derived from data, researchers' judgment and expert opinion may be required. 13 14 Parameterization is an emerging method for addressing structural uncertainty. It involves 15 adding parameters to a model that are assumed to be the sources of structural 16 uncertainty and assigning them a single, often extreme value so that in some cases they 17 may be completely excluded from the model whereas in other cases they are an 18 important component of the model. This method allows structural uncertainty to be 19 internalized in the model, and could be used to inform any decision about future 20 research to resolve these uncertainties.^{145,161} Justification should be provided for any 21 structural uncertainties that have not been addressed.¹⁴⁵ 22 23 24 Methodological Uncertainty 25 26 Methodological uncertainty relates to the different methods that can be used to conduct 27 an economic evaluation. When conducting economic evaluations of vaccination 28 programs, important methodological choices that researchers must consider include the type of analysis, perspective, discounting approach and rate, and time horizon.46,144 29 30 31 Because vaccination programs often prevent diseases that could result in catastrophic 32 consequences (e.g., meningococcal B vaccination could prevent death or permanent 33 neurological sequelae resulting from meningitis due to meningococcal type B bacteria), 34 they produce health-related benefits as well as non-health-related benefits such as

1 improvements in education or lifetime productivity. As such, some authors have argued 2 that CBAs should be considered in the economic evaluation of vaccination programs in 3 order to account for their full range of benefits.⁸ In practice, however, uncertainty related 4 to the type of evaluation conducted is rarely examined.⁴⁶ In principle, the non-health 5 costs of vaccination programs could be captured in a CUA if a broader perspective (e.g., 6 societal perspective) for the analysis is adopted, but capturing non-health benefits may 7 be more challenging.^{8,144} Discrete choice experiments are an increasingly popular option 8 for capturing relevant trade-offs for non-health benefits of interventions for either CBA or 9 CUA. Accordingly, these guidelines recommend conducting two reference case 10 analyses: one from the publicly funded health system perspective and another from the 11 societal perspective.

12

13 Economic evaluations of vaccination programs are particularly sensitive to the 14 discounting strategy, use of annuitization, and time horizon chosen for the analysis since 15 costs related to the initiation of a vaccination program are incurred when the program is 16 introduced while the full benefits of the program often takes a much longer period of time, sometimes many years or decades, to realize.9,46 As such, researchers should 17 18 examine how varying the discounting approach, as well as time horizon of the analysis 19 affects the results. When the time horizon of the analysis is very long (e.g., multiple 20 decades), results of the economic evaluation should be reported for a range of time 21 horizons to ensure that decision-makers are considering costs and outcomes that are 22 relevant to the decision problem they are addressing.

23

To thoroughly explore many aspects of methodological uncertainty, multiple different
models are ideally required, which is often practically difficult. As such, researchers
should transparently collaborate with other groups addressing similar decision problems,
whenever possible, so that the extent of uncertainty related to methodological choices
can be explored.
Value-of-information

30 If a value-of-information analysis is undertaken, summarize the value of additional

31 information using the expected value of perfect parameter information and the

32 population expected value of perfect parameter information. See CADTH Guidelines for

33 further guidance.

1 14. Equity

•		
2	14.1	Researchers and decision-makers should work together to establish which equity
3		dimensions and goals should be included in the economic evaluation of the
4		vaccination program being considered. Equity should be considered in the
5		context of NACI's Ethics, Equity, Feasibility, and Acceptability (EEFA) framework.
6		
7	14.2	Analyses that incorporate relevant equity concerns should accompany the
8		reference case analysis (e.g., distributional cost-effectiveness analysis, extended
9		cost-effectiveness analysis, or other emerging methods), and presented
10		alongside the reference case.
11		
12	The tr	aditional emphasis of economic evaluations of healthcare interventions has been
13	the as	sessment of efficiency. This exercise sits within the larger decision-making
14	frame	work of HTA, which synthesizes and appraises primarily clinical and economic
15	evider	nce related to a new health intervention or technology. However, there is growing
16	recog	nition that ethical and moral questions related to how a technology is appraised
17	and u	sed should be addressed as part of decisions on the adoption of new health
18	techno	ologies. ^{163,164}
19		
20	NACI	has established the Ethics, Equity, Feasibility and Acceptability (EEFA) Framework
21	to sys	tematically consider these factors as part of a multi-criteria approach to vaccine
22	recom	mendations. In this framework, ethics and equity are considered with the feasibility
23	and a	cceptability of a recommendation, alongside a vaccine's clinical effectiveness,
24	immu	nogenicity, safety, and cost-effectiveness. ⁵ Public health ethics is the domain of
25	applie	d ethics relevant to vaccination. It is primarily concerned with the following core
26	ethica	l dimensions: 1) respect for persons and communities; 2) non-maleficence and

- 27 beneficence; 3) trust; and 4) justice.¹⁶⁵
- 28

Equity is considered within the core ethical dimension of justice, and is defined as "the absence of avoidable, unfair, or remedial differences among groups of people, whether those groups are defined socially, economically, demographically or geographically or by means of stratification."^{5,166} Equity in economic evaluations is an approach to distributive justice that concerns judgments about the fairness in distribution of health outcomes and experiences in a population, and it relates to the fair allocation of resources and achievement of health improvements between individuals or groups.¹⁶⁷ There has been
 considerable recent activity and methodological development related to equity in the
 economic evaluation of health technologies.^{164,168-170}

4

5 The distributional consequences related to adoption of a new health technology become 6 particularly important in situations where decision-makers must make trade-offs between 7 attributes of health technologies. For example, health economists frequently flag trade-8 offs between cost-effectiveness and health equity. These situations arise when a 9 technology is cost-effective but increases inequity between groups in a population 10 because some segments of society may benefit from the technology more than others. 11 Alternatively, they arise when a technology is not cost-effective but its adoption would 12 improve equity between groups by reducing disparities in health gains, or when a 13 technology would increase equity between some groups (e.g., income strata) but 14 decrease equity between others (e.g., geographical).¹⁷¹ 15

16 Researchers and decision-makers should work collaboratively to establish which of the 17 following equity goals the vaccination program is aiming to address: 1) improving equity 18 in access to the vaccination program for eligible individuals; 2) improving equity in 19 uptake of the vaccination program in eligible individuals; 3) improving equity in health 20 benefit related to health conditions addressed by the vaccination program; 4) reducing 21 lifetime health inequities between groups by means of the vaccination program; or 5) 22 reducing overall (i.e., health and non-health-related) inequities between groups by 23 means of the vaccination program. Different approaches may be used to conceptualize 24 equity goals (e.g., proportionate universalism, egalitarianism). Researchers may find it 25 helpful to refer to published literature on this topic when conceptualizing equity objectives for inclusion in an economic evaluation.¹⁷²⁻¹⁷⁴ 26

27

When establishing equity goals researchers should consider whether there are key groups of individuals experiencing health inequities and barriers to health that could be reduced or addressed by the vaccination program. Examples of groups that may experience health inequity in Canada include Indigenous Peoples (specifically, First Nations, Inuit, and Métis Peoples for the purposes of these guidelines), individuals of low socioeconomic status, people who are part of ethnic, sexual, or gender minority groups, populations living in certain geographic locations (urban vs. rural vs. remote and isolated), individuals with disabilities, and vulnerable groups such as children, seniors or
 institutionalized persons.^{5,117,170,175,176}

3

Researchers should also consider factors that could lead to differences in health benefits
resulting from the vaccination program between groups experiencing health inequities.
Factors include underlying health conditions, potential for lifetime benefit, health-seeking
behaviours, uptake of the vaccine and the role of community immunity in reducing or
increasing inequities between groups, risk-taking behaviours, different mixing or contact
patterns within groups, and access to culturally safe healthcare.^{6,177}

10

Researchers should be aware that some groups may benefit from the vaccination program, and some groups may not, thus potentially increasing inequities. For example, differential access to an HPV vaccination program can worsen inequity by reducing the rate of cervical cancer in a population who was already at lower risk but have greater access, thereby increasing the difference in outcomes between groups. Interventions that appear to reduce inequities should be examined to investigate how relevant barriers to access faced by the population would be overcome.

18

19 Once researchers have established the equity-relevant outcomes of interest, features of 20 the vaccination program intended to achieve these outcomes should be considered. For 21 instance, if the goal of the program is to improve equity in access to the vaccine for all 22 eligible individuals, then a program that decreases barriers to access should be 23 considered. An example of such a program would be a school-based HPV vaccination 24 program that eliminates barriers for individuals such as the cost of the vaccine doses. 25 and the need for transportation to a clinic or physician's office.¹¹⁷ If the goal of the 26 program, however, is to improve equity in uptake of the vaccine, researchers could 27 consider scenarios in which vaccines are mandatory or that address misinformation 28 about the vaccine. An example of such a program would be a legally mandated school-29 based program for HPV vaccination, with a provision for active opt-out.¹⁷⁸ If the goal is to 30 reduce lifetime health inequity between groups with the vaccination program, a program 31 that is consistent with the principle of vertical equity, which entails treating individuals 32 with different ethically-relevant characteristics differently, should be considered.¹⁷⁹ An 33 example would be a vaccination program aimed at achieving high levels of vaccination 34 coverage among Indigenous Peoples. Indigenous Peoples experience a greater burden

1 of vaccine-preventable diseases than non-Indigenous People in Canada (e.g., cervical 2 cancer, hepatitis A) due to systematic inequities such as poverty, crowded housing 3 conditions, lack of running water, and poor underlying health status, which increase the 4 risk for acquiring these infections.^{180,181} Additionally, Indigenous Peoples living on 5 reserves and in remote communities may also experience inequities in access to 6 treatment when they become ill, increasing their risk of infection-related morbidity and 7 mortality.¹⁸² Researchers must be aware, however, that vaccination programs restricted 8 to certain high-risk groups that are vulnerable or marginalized may serve to further 9 stigmatize those groups. Alternative approaches, such as more universal programs, 10 should be considered. Finally, if the goal is to improve health and non-health equity 11 between groups, researchers could consider vaccination programs that contribute 12 towards improving health as well as economic productivity. Examples of such programs 13 are childhood vaccination programs, which enable children to participate in education, in 14 turn allowing them to become healthy and economically productive adults.¹⁸³ When 15 consideration of equity-relevant outcomes relates to selection and definition of 16 comparator(s) to be included in the analysis, researchers should refer to Chapter 3 on 17 Comparators of these guidelines.

18

When presenting results of economic evaluations by equity-relevant subgroups, researchers should ensure that the criteria for establishing these subgroups has been transparently delineated and justified. A recent review of equity-informative CEAs identified eleven different criteria that have been used to explicitly incorporate equity in a cost-effectiveness framework, with socioeconomic status and race/ ethnicity used most frequently.¹⁸⁴ Distributed (DCEA) and extended CEA (ECEA) frameworks provide guidance and methods for conducting equity-informative CEAs.^{168,185}

26

27 In addition to considering equity-related outcomes associated with vaccination programs, 28 researchers should also consider the distribution of opportunity costs related to the 29 implementation of these programs.¹⁷¹ This redistribution of resources could, for example, 30 result in decreased expenditures on screening programs or non-vaccine preventive 31 measures related to the infection being targeted by vaccination program. Opportunity 32 costs could also fall outside of the health sector, for example, through decreased funding 33 of educational or social programs.¹⁷¹ Although in many cases, it may be difficult to 34 explicitly identify opportunity costs related to implementing vaccination programs, where

possible, researchers should quantify opportunity costs in a manner that is relevant to
decision-makers. In some cases, interventions to improve equity may not carry a net
opportunity cost, since it may be efficient to allocate resources to groups with higher
health burden.

5

6 When relevant, researchers should consider the implications of vaccination programs on 7 intergenerational equity. Vaccination programs that result in externalities have effects on 8 cohorts of individuals other than the cohort that is vaccinated.³⁴ For example, a 9 childhood varicella vaccination program may result in increased cases of herpes zoster in older adults;¹⁸⁶ conversely, an HPV vaccination program may lead to disease 10 eradication for future generations.¹⁸⁷ In both of these examples, the indirect effects on 11 12 cohorts of individuals not intended for the vaccination program should be accounted for 13 with dynamic models used to generate estimates of cost-effectiveness. Researchers 14 should then explicitly consider the equity implications of these results. In the first 15 example, researchers need to qualitatively identify the trade-offs between improved child 16 health and negative health outcomes that may be experienced by older individuals. 17 Quantitatively, the summary costa and outcomes estimated in the analysis capture the 18 trade-offs. In the second example, researchers need to consider how health outcomes 19 that accrue to cohorts far into the future should be valued in present day terms. 20

21 Such valuation of health outcomes in cohorts far into the future is contingent on the 22 discounting strategy employed in the economic evaluation. An equal discount rate 23 results in greater value placed on health outcomes for the present cohort and cohorts 24 close in time to the present, while lesser value is placed on health outcomes for cohorts in the distant future, which some authors argue is an unfair feature of this strategy.³³ Use 25 26 of lower discount rates, however, could result in giving greater weight to health outcomes 27 in cohorts that are likely to have greater income, and access to more health 28 interventions, and as such, more potential to improve health. Given the variable 29 outcomes related to intergenerational effects of vaccination programs that can result 30 from different discounting strategies, researchers should consider and report the 31 intergenerational equity implications of vaccination programs that result in health benefits for cohorts in the distant future. 33, 34, 188 32

1	15.	Reporting	
2	15.1	"The economic evaluation should be reported in a transparent and detailed	
3		manner with enough information to enable the reader or user (e.g., decision-	
4		maker) to critically assess the evaluation. Use a well-structured reporting format."	
5		[CADTH Guideline Statement]	
6			
7	15.2	"A summary of the evaluation written in non-technical language should be	
8		included." [CADTH Guideline Statement]	
9			
10	15.3	"Results of the economic evaluation should be presented in graphical or visual	
11		form, in addition to tabular presentation." [CADTH Guideline Statement]	
12			
13	15.4	"Details and/ or documents describing quality assurance processes and results	
14		for the economic evaluation should be provided. An electronic copy of the model	
15		should be made available for review with accompanying documentation in	
16		adequate detail to facilitate understanding of the model, what it does, and how it	
17		works." [CADTH Guideline Statement]	
18			
19	15.5	"Funding and reporting relationships for the evaluation should be described, and	
20		any conflicts of interest disclosed." [CADTH Guideline Statement]	
21			
22	15.6	Researchers should use NACI's Guidelines for Reporting Economic Evaluations	
23		of Vaccination Programs in Canada, and complete the Impact inventory table for	
24		economic evaluations of vaccination strategies, which is found in Appendix 1.	
25			
26	Repor	ting results of economic evaluations should provide decision-makers with	
27	transp	parent and credible information that enables them to address the decision problem	
28	of inte	rest, and make an optimal funding decision related to the vaccination program	
29	being	considered.	
30	Vaccii	nation-specific reporting considerations should be addressed including the time	
31	horizon of the evaluation, and the mechanisms through which vaccines exert their		
32	effects. In cases where the model time horizon of an economic evaluation spans a long		
33	perioc	l of time, results from various time points over the model time horizon should be	
34	report	ed to ensure that findings of the analysis are relevant to the time horizon being	

1 considered by decision-makers. Since vaccines may exert their effects through various

2 mechanisms (e.g., preventing transmission of infection, preventing infection, preventing

3 disease or decreasing its severity), researchers should report outcomes of vaccination

4 programs not only in terms of QALYs, but also in terms of the number of cases

5 prevented, the number of relevant healthcare utilization units (e.g., hospitalizations)

6 averted, the number of deaths averted, and the number of individuals needed to

7 vaccinate, where applicable. Reporting these metrics in addition to QALYs increases the

8 credibility and transparency of the analysis for decision-makers.

9 Resource use, costs, and outcomes should be reported in a disaggregated fashion for

10 each comparator considered in the analysis and for both reference case analyses. If

analyses are conducted from multiple public payer perspectives (e.g., for multiple

12 provinces/ territories), each should be reported separately.

All assumptions and decision rules used in the analyses should be transparentlyreported.

15

16 Researchers should provide details related to quality assurance processes, and results

17 undertaken as part of the model verification process. A fully executable electronic copy

18 of the model should be made available along with details related to the model's

19 functionality to enable the decision-maker to verify results of the analysis or conduct

20 additional analyses if required.

21

Researchers should disclose all sources of funding for the economic evaluation and state the role of the funder(s) in the identification, design, conduct, and reporting of the analysis. Non-monetary (e.g., in-kind) sources of support should also be disclosed.¹⁸⁹

25

Researchers should disclose all potential conflicts of interest, both financial and nonfinancial. Types of affiliations and interests to disclose include: participation in research,
equity ownership, intellectual property, and any other interest that readers may perceive
as a competing interest (e.g., public statements about the topic).¹⁹⁰

30

31 The "NACI Guidelines for Reporting Economic Evaluations of Vaccination Programs in

32 Canada" on the NACI website provides a standard format for reporting the results of

33 economic evaluations of vaccination programs.¹⁹¹ Researchers should follow the

34 structure outlined in this document when presenting their results.

Appendix 1: Impact Inventory Table

Table 1: Impact inventory table for economic evaluations of vaccination strategies

Area of Impact	Definitions/Examples	Included in Reference Case?		Comments
		Publicly funded health system perspective	Societal perspective	
Health				
Health outcomes	Individual health outcomes for persons inten	ded for vaccinatio	n	
	Mortality			
	Health-related quality of life			
	Safety (i.e., adverse events)			
	Irreversible health impacts not captured by QALYs (e.g., infertility associated with sexually transmitted infections			
	Individual health outcomes for informal care	givers		
	Health-related quality of life			
	Population health outcomes			
	Incidence of disease in vaccinated and unvaccinated individuals			
	Changes in age distribution of individuals who develop infection and disease			

Area of Impact	Definitions/Examples	Included in Reference Case?		Comments
		Publicly funded health system perspective	Societal perspective	
	Changes in infection and disease incidence related to variations of pathogen or other pathogens that replace ones targeted by vaccine			
	Disease eradication			
Health system costs	Healthcare treatment costs			
	Publicly funded healthcare services (e.g., physician visits, diagnostic tests, drug treatment where applicable, hospitalization, formal caregiving, ^a rehabilitation in a facility or at home, ^a home care, ^a long-term care in nursing homes ^a)			
	Future related and unrelated healthcare costs			
	Public health costs			
	Program-related costs (e.g., implementation, delivery and recurrent costs, public health campaigns, health promotion activities, transaction costs, population-based screening, epidemiologic surveillance, contact tracing, investigation and management of outbreaks)			

Area of Impact	Definitions/Examples	Included in Reference Case?		Comments		
		Publicly funded health system perspective	Societal perspective			
	Intervention-related costs (e.g., cost of vaccine doses, distribution such as transportation and cold storage, administration including personnel, wastage and ancillary supplies)					
		N1/A				
Healthcare costs NOT	Drug treatments (in some cases)	N/A				
funded by the health system	Formal caregiver services, ^a rehabilitation in a facility or at home, ^a home care, ^a long- term care in nursing homes ^a (in some cases) Miscellaneous out-of-pocket costs (e.g., non-prescription medications) Ancillary costs (e.g., private insurance copayments, dental care, vision care,	N/A N/A N/A				
	assistive devices, physiotherapy, etc.)					
Non-Health Areas						
Direct out-of-pocket	Transportation costs	N/A				
costs	Accommodation costs	N/A				
Productivity loss	Paid work		·			

Area of Impact	Definitions/Examples	Included in Re	ference Case?	Comments
		Publicly funded health system perspective	Societal perspective	
	Time off work resulting from treatment, illness, disability, or death	N/A		
	Presenteeism	N/A		
	Lifetime productivity consequences of childhood disease			
	Unpaid work			
	Time off work in informal labour market resulting from treatment, illness, disability, or death	N/A		
	Uncompensated household production (e.g., Cooking, cleaning, shopping, raising children, other tasks related to household management)	N/A		
	Informal caregiver productivity			
	Time off work resulting from caring for sick individuals	N/A		
	Caregiver presenteeism	N/A		
	Macroeconomic consequences			
	Labour supply shocks, widespread business closures	N/A		

Area of Impact	Definitions/Examples	Included in Re	ference Case?	Comments
		Publicly funded health system perspective	Societal perspective	
Consumption	Future individual nen medical consumption	N/A		
Consumption	Future individual non-medical consumption	N/A		
	Changes in household consumption	N/A		
	Health impacts of consumption (e.g.,	N/A	_	
	associated with job loss)			
Education	Level of educational achievement as a result of physical health, mental health, and cognition	N/A		
	Costs of special education needs as a result of illness/disability	N/A		
Social services and community services	Social services and community services (e.g., disability support, programs to improve access to vaccination programs for adults)	N/A		
	Child and Youth Services (e.g. awareness programs, family respite, programs to improve access to vaccination programs for children and youth)	N/A		
Environment	Environmental impact of vaccination programs and comparators (e.g.,	N/A		

Area of Impact	Definitions/Examples	Included in Reference Case?		Comments
		Publicly funded health system perspective	Societal perspective	
	manufacturing, distribution, and implementation)			
Other Areas	Consider areas such as legal/criminal or housing when applicable	N/A		

^a Some of these costs may or may not be incurred by the publicly funded health system, depending on the precise nature of these costs and the relevant jurisdiction

1 Appendix 2: Reference Case

2 Specifications

- 3
- 4 Table 2 presents recommendations for the reference case analyses. In situations where
- 5 the analyses do not follow the recommendations presented below, researchers should
- 6 identify any deviations and provide justification based on the decision problem.

7 Table 2: Recommendations for reference case analyses

Section	Guidance
Decision Problem	Specify the details of the vaccination program, setting, perspective, costs, outcomes, time horizon and intended population for the evaluation.
Types of	Conduct a cost-utility analysis (CUA) capturing health outcomes in
Evaluations	terms of quality-adjusted life-years (QALYs).
Study Populations	Identify the population(s) in which the vaccination program will be used, and, when applicable, any populations that might experience externalities resulting from the vaccination program. Conduct stratified analysis where distinct subgroups are identified.
Comparators	Compare all relevant interventions, including other vaccination programs, screening interventions, medical and non-medical preventive interventions, and treatment-based approaches presently used in a Canadian context.
Perspective	Conduct two reference case analyses, one from the publicly funded health system perspective and one from the societal perspective.
Time Horizon	Select a time horizon that is long enough to capture all relevant differences in the future costs and outcomes associated with the interventions being compared.
Discounting	Discount costs and outcomes at a rate of 1.5% per year.
Measurement and Valuation of Health	Identify, measure, and value all relevant health outcomes based on the perspectives of the publicly funded health system and society. Use health preferences that reflect the general Canadian population.
	Obtain health preferences from an indirect method of measurement that is based on a generic classification system.
Resource Use and Costs	Identify, measure, and value all relevant resources and costs based on the perspective of the i) publicly funded health system, and ii) society. Estimate Canadian resources and costs using data that reflect the jurisdiction(s) of interest.
Analysis	Derive expected values of costs and outcomes for both the publicly funded health system perspective analysis and the societal perspective analysis for each intervention through probabilistic analysis, incorporating potential correlation among parameters, whenever possible.

	Where distinct subgroups are identified within the intended population, conduct a stratified analysis and present results for each subgroup.		
	Calculate incremental costs, incremental effectiveness, and incremental cost-effectiveness ratios (ICERs) for both the publicly funded health system and societal perspective analyses. For evaluations with more than two comparators, calculate ICERs sequentially.		
	Address methodological uncertainty by comparing the reference case results to those from a non- reference case analysis.		
Uncertainty	Summarize decision uncertainty, using cost-effectiveness acceptability curves (CEACs) and cost-effectiveness acceptability frontiers (CEAFs), where possible.		
Uncertainty	Use scenario analysis to address structural uncertainty.		
	If a value-of-information analysis is undertaken, summarize the value of additional information using the expected value of perfect parameter information and the population expected value of perfect parameter information.		
	Consider whether there are inequities experienced by specific groups that could be improved by the vaccination program.		
Equity	Equity should be explored using methods such as distributional cost- effectiveness analysis and extended cost-effectiveness analysis. Any additional analyses should accompany the references case analyses when applicable.		

1

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