

EPIB 678-001 Advanced health decision modelling

DRAFT Course syllabus, Winter 2026, McGill University

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Course description

Developing decision-analytic models to synthesize evidence and simulate the outcomes of strategies in public health and medicine, including health technology assessment and model-based economic evaluation. Methodologies include cohort state transition modelling, microsimulation, discrete event simulation, Bayesian model calibration, probabilistic sensitivity analysis.

sis, value of information analysis, and equity-informative modelling frameworks. Applications include disease screening, prevention, and treatment, public health strategy, and prioritization of health research.

Learning objectives

By the end of the course, students will be able to:

- Develop decision analytic models to inform health policy and clinical decisions
- Understand and critically appraise model-based decision analyses
- Create transparent reports of model-based decision analyses that conform to open science principles

Instructor

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Class meetings & office hours

Class meetings will be held 11:35am – 12:55am on Tuesdays and Thursdays at 2001 McGill College Ave in room #1203. Attendance is expected.

Office hours will be held **TBD**. Students can take advantage of office hours to discuss assignments, projects, and readings or ask general career/research questions. All students are encouraged to attend. Most discussions take place openly so that other students can listen in or join in. Students who would like a private discussion should let the instructor know.

Target audience

The course is oriented towards towards PhD students and advanced Masters students interested in applying model-based analyses of health policies, decisions, or technologies in research. Modeling skills developed in the course will also be useful for students interested in industry or government roles related to economic evaluation, policy modeling, health technology assessment, or health economics and outcomes research.

Prerequisites

The ideal preparation of this course includes (1) at least an introductory course in R or a similar programming language, (2) a course in probability, (3) a course in statistics. A course on economic evaluation such as PPHS 528 is not necessary but would provide helpful context. Prior experience programming in R is ideal, but students with solid experience in programming fundamentals (e.g., nested loops, functions, data structures) should be able to learn the R syntax and environment as they go. Students who lack a prerequisite other than programming may need to spend extra time on some course components but should be able to succeed in the course. Students without programming, or without training in either statistics or probability, should not take the course. To help make an informed decision on how well-prepared you are for the course, you may [attempt to complete assignment 0 early \(available on Github\)](#).

Readings

Readings drawn from various sources are assigned before most class sessions as indicated in the schedule below. PDFs of the readings are available on MyCourses. Some readings are tutorials with accompanying code, which you are encouraged to download and run. To get full credit for class engagement, students should come to class prepared to discuss the readings.

If >20 pages are assigned for a given class, students may skim some of the material, particularly readings that are marked as ‘secondary.’ But students should be sufficiently familiar with each reading to quickly find relevant material as they work on assignments and their project.

Laptops and software

Students should bring a personal laptop to class. Please let the instructor know right away if this is a challenge so we can find a solution. Before the first session, students should install and configure the necessary software (R, RStudio, Quarto, git) [using the ‘Setting up your software environment’ instructions from session 1](#).

Approach to learning

The course provides hands-on experience applying course concepts early and often. The course has two phases with some overlap:

- **The methods phase (15 sessions)** focuses on understanding methods and workflows and building programming skills. Readings are primarily pedagogical in nature, including tutorials with accompanied R code you can examine and run. Some class sessions will include running sample code on your own laptop. Six programming assignments will provide hands-on experience in coding a model-based decision analysis.

- **The application phase (11 sessions)** focuses on the use of decision analytic modeling to inform health policy and clinical practice. Students will be assigned a diverse set of applied modeling analyses as readings and hear from a variety of decision-analytic modellers in class sessions. Students will also analyze and present an existing open-source modeling study. For the course project, students will develop their own model-based decision analysis.

Assignments & evaluation

Class engagement (8%)

Students are expected to attend class on time and prepared to discuss assigned readings. Students should actively engage in discussions, ask questions, and follow along with programming examples on their own laptops.

R programming assignments (48%)

A series of time-intensive programming assignments will provide hands-on experience developing decision analytic modeling analyses. For each assignment, students will download a Zip file that contains a Quarto document from the [class Github site](#). The Quarto document includes questions, instructions, and starter code. Students will complete the assignments by completing the code and answering questions. Once complete, students will render their Quarto document to an HTML file, which they upload to [MyCourses](#) for grading.

0. Simpler models (8%)
 1. Decision trees, cost-effectiveness (8%)
 2. Cohort models (8%)
 3. Simulation (8%)
 4. Uncertainty analysis (8%)
 5. Calibration, validation, and value of information (8%)

Out of fairness to all classmates, late assignments will be penalized 20% per day unless an exception is made more than 2 days before the deadline or under exceptional circumstances. All assignments will be available at least 10 days before their deadline. Students should:

- Start early
- Read the whole assignment before you begin and feel free to go out of order
- Spend some time on it each day for 3 or more days (instead of completing it in one sitting)
- Work alone first, then discuss with classmates and/or the instructor (but you must write your own code and responses)

- Come to office hours to discuss

Presentation of open-source modeling analysis (8%)

Students will find a published applied decision-analytic modeling analysis that includes code. Students will read the paper, analyze the approach to coding, attempt to run the model on their own machine, and give a brief in-class presentation about their experience.

Course project (36%)

The default course project entails developing a decision-analytic model and use it to inform a decision from health policy or clinical practice. Students are encouraged to analyze a decision problem related to their own research and/or engage with domain experts outside the class when applicable. Other project topics, such as developing or assessing a novel method or using modeling methods for an application that is not a decision analysis, may be approved by the instructor on a case-by-case basis. Students are strongly encouraged to come to office hours to discuss their project with the instructor.

1. Proposal (3%)
2. Class presentation (8%)
3. Written report (with code) (25%)

Logistics

All readings are available to enrolled students in [MyCourses](#). Assignments are due at 11:59pm EST on the date indicated in the schedule below, to be uploaded on MyCourses. Grades, feedback, and assignment solution will also be provided within MyCourses. All other course materials are available on [Github](#), including the files needed to complete assignments.

Academic integrity

McGill University values academic integrity. Therefore, all students must understand the meaning and consequences of cheating, plagiarism and other academic offences under the *Code of Student Conduct and Disciplinary Procedures* ([see here for more details](#)). Students found guilty of such academic offences will be given a failing grade to this course.

Language of submission

In accordance with McGill University's *Charter of Students' Rights*, students have the right to submit any written work that is to be graded in either English or French.

Students who enroll during the add/drop period

Students who should watch the lecture recordings on MyCourses for any course they missed and should submit Assignment 0 within one week of registering for the course.

Course schedule

i The guest lectures shown here are from Fall 2025. Schedule will be updated to reflect the Winter 2026 guest lectures, and some topics may be re-ordered.

Date	Topics, Readings/tasks to be completed before class
6 Jan Tu	1. Course introduction <ul style="list-style-type: none">• Why do decision analysis• Analytic perspective• Using R and RStudio• Assignment 0 available (simpler models) <p>(1) Follow the ‘Setting up your software enviroment’ instructions to make sure you have an up-to-date local version of R and Rstudio desktop on your laptop.</p> <p>(2) Read through the Hello, Quarto tutorial for RStudio</p>
8 Jan Th	2. Economic evaluation <ul style="list-style-type: none">• Decision analytic frameworks• Quantifying health and costs• Incremental analysis• Assignment 1 available (decision trees, cost-effectiveness) <p>Ch. 1 ‘Introduction to health economic evaluation’ from Bayesian Methods in Health Economics (2012) by Gianluca Baio</p>

Date	Topics, Readings/tasks to be completed before class
13 Jan Tu	3. Probability, decision trees <ul style="list-style-type: none"> • Conditional probability • Distributions, expectation, variance • Decision trees • Assignment 0 due Wednesday <p>Ch 8 ‘Decision tree models’ from R for Health Technology Assessment August 2025 edition</p>
15 Jan Th	4. Cohort models I <ul style="list-style-type: none"> • Discrete time state-transition models • Markov model solution • Time-varying transitions • Assignment 2 available (cohort models) <p>Alarid Escudero et. al. 2022. An introductory tutorial on cohort state-transition models in R using a cost-effectiveness analysis example</p>
20 Jan Tu	5. Cohort models II <ul style="list-style-type: none"> • Capturing health-economic outcomes • Discrete time error corrections • Capturing epidemiological outcomes • Differential equation models • Assignment 1 due Friday <p>Alarid Escudero et. al. 2022. A tutorial on time-dependent cohort state-transition models in R using a cost-effectiveness analysis example</p>
22 Jan Th	6. Workflow and good coding practices <ul style="list-style-type: none"> • Open science practices • An R project workflow • Version control • Open-source model presentation assigned <p>Alarid-Escudero et. al. 2019. A Need for Change! A Coding Framework for Improving Transparency in Decision Modeling</p>

Date	Topics, Readings/tasks to be completed before class
27 Jan Tu	7. Simulation I <ul style="list-style-type: none"> • Monte Carlo simulation • Simulating decision trees • Simulating state transition models • Assignment 3 available (simulation) <p>Krijkamp et. al. 2018. Microsimulation modeling for health decision sciences Using R: A tutorial</p>
29 Jan Th	8. Simulation II <ul style="list-style-type: none"> • Discrete event simulation • Simulating resource constraint • Assignment 2 due Wednesday <p>Lopez-Mendez et. al. 2025. A Tutorial on Discrete Event Simulation Models in R Using a Cost-Effectiveness Analysis Example</p>
3 Feb Tu	9. Simulation III <ul style="list-style-type: none"> • Efficient coding strategies • Efficiency improvement for simulations • Agent based models • Network model models <p>Chhatwal & He 2015. Economic evaluations with agent-based modelling: An introduction.</p>
5 Feb Th	10. Designing & parameterizing an analysis <ul style="list-style-type: none"> • Perspective, time horizons, comparators • Deciding on scope & structure • Literature-based inputs • Patient-level data inputs • Project proposal assigned <p>Primary: Roberts et. al. 2012. Conceptualizing a model: a report of the ISPOR-SMDM Modeling Good Research Practices Task Force</p> <p>Secondary: Ch 8 (pg 209 - 236) 'Finding and summarizing the evidence' from Decision Making in Health and Medicine (2014) by Hunink et. al.</p>

Date	Topics, Readings/tasks to be completed before class
10 Feb Tu	11. Uncertainty analysis <ul style="list-style-type: none"> • Probabilistic sensitivity analysis • Credible intervals • Univariate sensitivity analysis • Scenario analysis • Assignment 3 due Friday • Assignment 4 available (Sensitivity analysis) <p>Briggs et. al. 2012. Model parameter estimation and uncertainty analysis: ISPOR-SMDM modeling good practices task force report</p>
12 Feb Th	12. Statistical models, causal inference, and decision-analytic modeling <ul style="list-style-type: none"> • Time-to-event models • Competing events • Temporal extrapolation <p>Primary: Kühne et. al. (2022). Causal evidence in health decision making: methodological approaches of causal inference and health decision science Supplemental: Williams et. al. (2016). Cost-effectiveness analysis in R using a multi-state modeling survival analysis framework: A tutorial</p>
17 Feb Tu	13. Calibration & validation <ul style="list-style-type: none"> • Bayesian model calibration • Model validation <p>(readings may change) Primary: Menzies et. al. 2017. Bayesian methods for calibrating health policy models: A tutorial. Supplemental: Eddy et. al. 2012. Model transparency and validation: ISPOR-SMDM modeling good practices task force report</p>
19 Feb Th	14. Guest lecture (ZOOM ONLY): Adapting a model of cervical carcinogenesis to self-identified Black women to evaluate racial disparities in the United States (Jenny Spencer PhD, University of Texas at Austin) <ul style="list-style-type: none"> • Office hours on Zoom only <p>Primary manuscript, Supplemental manuscript</p>

Date	Topics, Readings/tasks to be completed before class
24 Feb Tu	15. Guest lecture: Two microsimulation-based decision analyses: Needle and syringe programs for people who inject drugs (Jihoon Lim, PhD, McGill) and medications to improve cardiorenal outcomes in patients with diabetes (Ethan McNally, McGill) <ul style="list-style-type: none"> • Assignment 4 due Friday
26 Feb Th	Needle syringe programs manuscript , Code 16. Value of information analysis (ZOOM ONLY) <ul style="list-style-type: none"> • For research prioritization and as sensitivity analysis • VOI measures (EVPI, EVPPI, EVSI) • Challenges and limitation <ul style="list-style-type: none"> – Assignment 5 available (calibration, validation, VoI) • Office hours on Zoom only
10 Mar Tu	Fenwick et. al. 2020. Value of information analysis for research decisions—an introduction: Report 1 of the ISPOR Value of Information Analysis Emerging Good Practices Task Force <i>Winter reading break March 2 to March 6, inclusive</i> 17. Equity and distributional considerations <ul style="list-style-type: none"> • Evaluation frameworks • Estimating heterogeneity • Project proposal due Friday
12 Mar Th	Killedar et. al. 2023, Modelled distributional cost-effectiveness analysis of childhood obesity interventions: A demonstration 18. Student presentations of open-source modeling analyses
17 Mar Tu	19. Guest lecture: Estimating Treatment-Switching Bias in a Randomized Clinical Trial of Ovarian Cancer Treatment: Combining Causal Inference with Decision-Analytic Modeling (Uwe Siebert, MD and Felicitas Kuehne, MSc, UMIT Triol) Manuscript
19 Mar Th	20. Guest lecture: Choose your recorded guest lecture Students choose a guest lecture recording from an earlier year to watch.

Date	Topics, Readings/tasks to be completed before class
24 Mar Tu	21. Guest lecture: Dynamic compartmental modelling to inform influenza dynamics and vaccine policies in the United States by Kyu Lee, PhD (University of Washington School of Pharmacy) Manuscript , Code , Supplemental manuscript
	<ul style="list-style-type: none"> • Assignment 5 due Friday
26 Mar Th	22 Guest lecture: Modeling the Balance of Benefits and Harms of Cervical Cancer Screening with Cytology and Human Papillomavirus Testing by Talía Malagón, PhD (McGill University) Manuscript , Code on MyCourses
31 Mar Tu	23. Guest lecture: “Estimating the total incidence of type 1 diabetes in children and adolescents aged 0–19 years from 1990 to 2050: a global simulation-based analysis” by Zachary J. Ward, MPH PhD (Harvard T.H. Chan School of Public Health) Manuscript
2 Apr Th	24. Guest lecture: “Integrating decision modeling and machine learning to inform treatment stratification” by John Giardina, PhD (Mass General Research Institute / Harvard University) Manuscript
7 Apr Tu	25. Guest lecture: “Costs, quality-adjusted life years, and value-of-information of different thresholds for the initiation of invasive ventilation in hypoxemic respiratory failure” by Christopher Yarnell, MD (University of Toronto / University Health Network) Manuscript , Code
9 Apr Th	26. Final presentations and wrap-up lecture
17 Apr Fr	Final project report + code due on MyCourses (no class)