

Modelling for prostate cancer screening to assess the cost-effectiveness

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About me

- Health economist who recently defended my PhD thesis (2022)
- "Prostate cancer testing in Sweden: the interplay between cost and effectiveness"
- Research interests: health economic evaluations of cancer screening, diagnosis and treatment
- Consultant at the Stockholm Centre for Health Economics (StoCHE), Region Stockholm
- PhD Health Economics, MPH Health Economics, MSc Applied Economics, BSc Statistics
- 9-year work experience in the consulting and pharmaceutical industry



The way health economists do things: trial-based vs. lifetime Modelling approach: Markov vs. individual-based simulations Per-protocol vs. intention-to-treat: implications for modelling Sensitivity analyses: one-way vs. probabilistic



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Types of economic evaluation

Economic evaluation	Health consequences		
CUA Cost-utility analysis	QALYs (quality-adjusted life years) Generic or disease specific outcome measures ("utilities")		
CEA Cost-effectiveness analysis	A single, common effect that may differ in magnitude; e.g. case detected		
CBA Cost-benefit analysis	Translate effects into monetary benefits; e.g. translate disability days avoided, life-years or QALYs gained		

Advantage of using QALYs

Simultaneously capture gains from

- reduced morbidity
- reduced mortality

Integrate these into a single measure



Perspective Direct		Indirect		
Healthcare	Costs for testing, diagnosis and managing the disease	N/A		
Societal	 Direct healthcare Non-healthcare, e.g. transportation, social services, informal care* 	 Productivity losses Morbidity: short- and long-term sick leave, early retirement Premature mortality 		

Primary: Healthcare

Societal perspective may discriminate against those who are not in the labour market: children, disabled, unemployed, elderly

* Informal care is debatable. It is sometimes considered as indirect costs.

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Cost-utility analysis

Comparisons



Incremental cost-effectiveness ratio (ICER)

$$\text{ICER} = \frac{\Delta C}{\Delta E} = \frac{C_1 - C_0}{E_1 - E_0}$$

Cost-effectiveness threshold (willingness-to-pay)

Category	Costs per QALY gained		
Low	<100,000 SEK		
Moderate	100,000 (incl.) – 500,000 SEK		
High	500,000 (incl.) – 1,000,000 SEK		
Very high	≥1,000,000 SEK		

Trial-based vs. Lifetime



- Easy to compare screening strategies with no screening or current testing practice
- Straightforward to calculate the health consequences and costs
- Focus on shorter-term effects rather than effects after the trial period

Trial-based vs. Lifetime



- Captures the long-term health consequences and costs
- More applicable to health interventions that will affect survival
- Encounter challenges in how to capture the long-term effects
 - Lack of evidence
 - Difficult to model

Research question



MRI: magnetic resonance imaging; PI-RADS: Prostate Imaging Reporting and Data System; PSA: prostate-specific antigen; SBx: systematic biopsy; TBx: MRI-guided targeted biopsy

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Research question



Aim

No Screenin (Diagnos

through

To assess the cost-effectiveness of MRI-based screening **using the Stockholm3 reflex test** for prostate cancer in Sweden.



MRI: magnetic resonance imaging; PI-RADS: Prostate Imaging Reporting and Data System; PSA: prostate-specific antigen; SBx: systematic biopsy; TBx: MRI-guided targeted biopsy

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.1%



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Per-protocol vs. intention-to-treat: implications for modelling

Sensitivity analyses: one-way vs. probabilistic

Overview of health economic models

- 1. Aggregate, branch-based, discrete time: Decision tree
- 2. Aggregate, state-based, discrete time, clock-forward: Markov
- 3. Aggregate, state-based, continuous time, clock-forward: Markov
- 4. Aggregate, state-based, continuous time, clock-reset: semi-Markov
- 5. Individual, state-based + attribute-based, continuous time, mixed time scales: Microsimulation or Discrete Event Simulation (DES)

Credit: Mark Clements' talk Model taxonomy for HTA on 2022-06-14

Why microsimulation?

Using individual-level data to address individual heterogeneity

- Available longitudinal data
- Ability to account for individual heterogeneity within the population of interest

State transition intensities depend on a patient's history

- Incorporate the memory of events occurring for simulated individuals in the model e.g. time since disease onset, the occurrence of previous events, or time-varying response to treatment
- Also supports individualised screening interventions

Flexible distributions for event times

Bring evidence from specific RCTs together with data from other sources

Difficulties in using microsimulation

- Complexity
- "Black box", less transparent
- Data availability
- High computational workload (for multi-core simulations in R and C++, see our microsimulation package on CRAN)
- Monte Carlo uncertainty for individual-based simulations

Microsimulation model

The Prostata microsimulation model

- Simulated individual life histories from disease onset, disease progression through to deaths
- Included a longitudinal PSA submodel (same as FHCRC), T-stage, M-stage and ISUP Grade Group
- Validated and well calibrated for Sweden and ERSPC
- Supports reliable health economic evaluations of cancer screening
- Open access to the model (github.com/mclements/prostate)





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Per-protocol vs. Intention-to-treat

	Intention-to-treat	Per-protocol		
Principal	As randomised, ignoring the actual test or treatment received	As randomised, conditional on protocol compliance		
Scope	Effectiveness of the test or treatment	Efficacy of the test or treatment under ideal circumstances (compliance)		
Strengths	Randomisation – protected from bias due to imbalance of baseline characteristics	Proof of the diagnostic or therapeutic concept; more direct for modelling		
Limitations	 Missing data: imputation required Generalisability: difference in compliance between the study and real-life 	 Power: may be reduced Violation of randomisation: selection bias May over-estimate the effect and does not represent the real-life situation 		

Primary choice in modelling for prostate cancer screening: per-protocol

- Issues with generalisability of ITT
- Uncertainties in screening compliance (participation) and biopsy compliance: captured as parameters in the model
- Address reduced power: model-based multiple imputation (assuming missing at random given other covariates)
- Example: two-arm diagnostic trial (standard biopsy vs MRI-first and then biopsy), with poor biopsy compliance in the standard arm

Differences in the test performance using PP and ITT

Test characteristics	PP	95% CI (PP)	ITT	95% CI (ITT)
Pr(MRI+ PSA+, GG=0, MRI+TBx/SBx)	0.148	(0.126, 0.192)	0.150	(0.129, 0.195)
Pr(MRI+ PSA≥1.5, S3M≥15%, GG=0, MRI+TBx/SBx)	0.167	(0.124, 0.224)	0.175	(0.131, 0.233)
Pr(MRI+ PSA≥2, S3M≥15%, GG=0, MRI+TBx/SBx)	0.164	(0.119, 0.226)	0.174	(0.127, 0.236)



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Differences in the results using PP and ITT



HSV: health state value; ITT: intention-to-treat; MRI: magnetic resonance imaging; PSA: prostate-specific antigen; QALY: quality-adjust life-year; S3M: Stockholm3 test; TBx: targeted biopsy; TBx/SBx: the combined targeted and systematic biopsy; WHO: World Health Organisation



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Both one-way and probabilistic sensitivity analyses are important

One-way: how a system is sensitive to key parameters

- Costs, utilities, discount rate
- Screening: age, screening interval, test threshold, trial evidence using PP or ITT, biopsy procedures, etc.

Probabilistic: addresses joint uncertainties

- Parametric bootstrap or Bayesian analysis
- May assume parameter independence
- Cost-effectiveness plane, cost-effectiveness acceptability curve (CEAC), etc.
- Screening: test performance, costs, utilities

- Both are required by the Swedish guidelines for assessing cost-effectiveness
- Not necessarily required by guidelines from other countries

Mathematically, for CEAC

- $E_{\theta}(Y \mid \theta)$ vs. $Y(E_{\theta}(\theta))$ where θ are the parameters (as a distribution) and Y could be costs or effects. Different countries accept one or both of these approaches
- CEAC_k = Pr (Strategy k is cost effective | θ , τ), where τ is the cost-effectiveness threshold (willingness-to-pay threshold)

Cost-effectiveness acceptability curve (CEAC)



At a nominal threshold of €47,218 per QALY gained

 Blue: 70% probability of being cost-effective

At a nominal threshold of €83,000 per QALY gained

 Blue & red: equal probability of being cost-effective

At a nominal threshold of €94,446 per QALY gained

• Red: more than 50% probability of being cost-effective compared with other strategies



- Health economics is an important component for translating epidemiological findings into policy
- Health economics needs good modelling which is potentially a nice fit for biostatisticians ©
 - Long-term predictions outside of the observed data
 - Modelling for the disease natural history (e.g. using joint models)

