

R Foundations Workshop DAY 1 | 9 April 2025



R-HTA in LMICs

Increasing accessibility to R for HTA in LMICs

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Modelling Basics: Decision Tree

HTA in LMICs R Foundations Workshop DAY 1
(9 April 2025)

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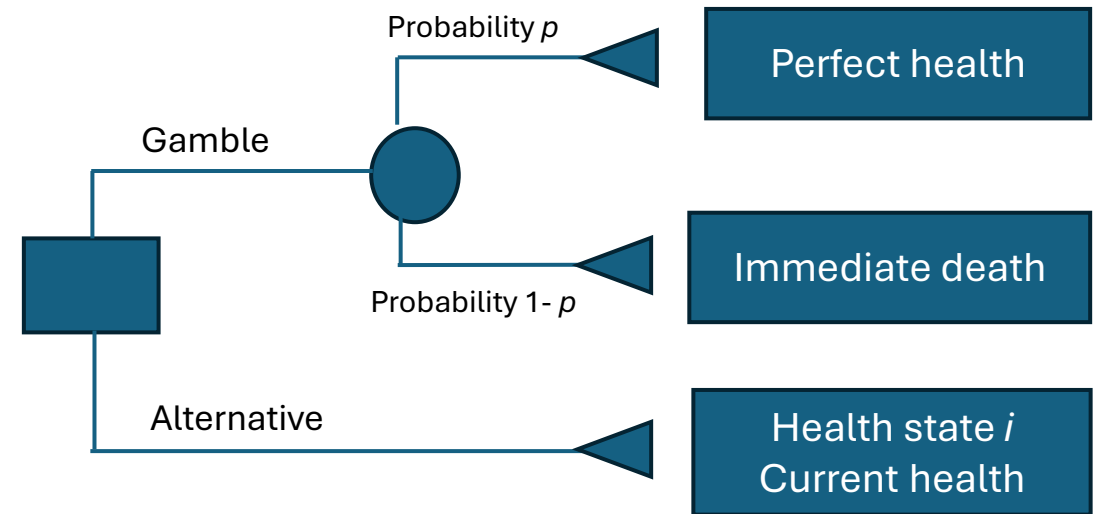
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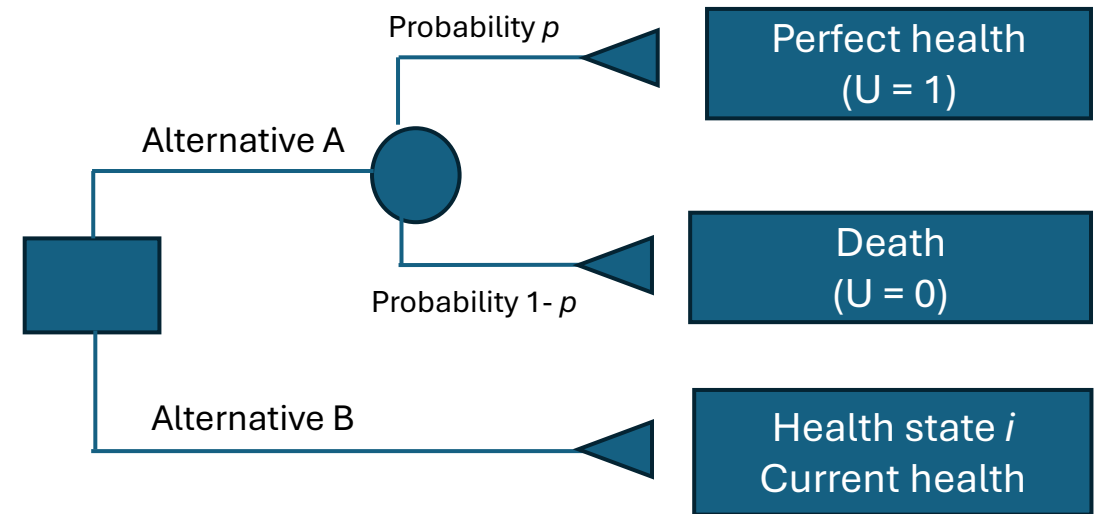
Standard Gamble theory

- The standard gamble is **a method that has its theoretical basis in the von Neumann–Morgenstern axioms of expected utility theory.**
- It aims to measure the 'disutility' of a health state by observing the willingness to accept a certain risk of death in order to avoid the state.
- Typical standard gamble framework, a respondent is asked to consider a choice between two alternatives:
 - **Alternative A:** the person would live with a particular health problem (the one for which the valuation is needed) with certainty, for the remainder of his or her life.
 - **Alternative B (Gamble)** is usually characterised as a risky treatment, with two possible outcomes: life in a state of optimal health, with probability p , or immediate death, with probability $(1-p)$.
 - The measurement objective for the standard gamble is to identify the probability of optimal health, p , at which the respondent is 'indifferent' between alternatives A and B, in other words, the point at which the two alternatives seem equally attractive.



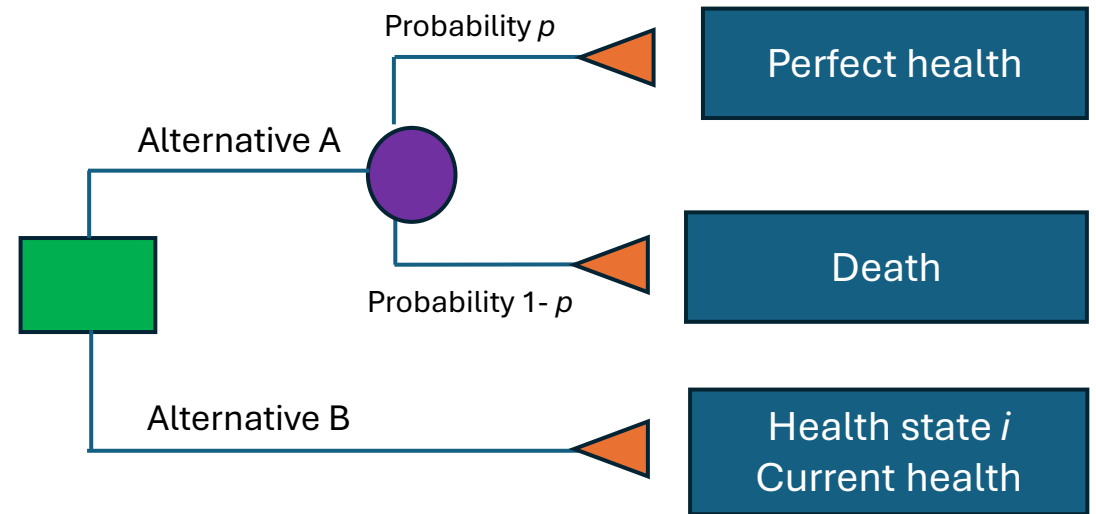
Standard gamble theory

- Once this 'indifference' point is identified, a health state valuation for the particular health problem of interest is equal to p .
- The logic of this inference derives from setting the utility of optimal health to 1.0 and that of death to 0 and assuming that at the point of indifference, the respondent considers the expected utility of alternatives A and B to be the same.
- In mathematical terms, the equality is stated as $p \times U(\text{optimal}) + (1-p) \times U(\text{death}) = U(\text{health outcome})$
- $p \times 1 + (1-p) \times 0 = U(\text{health outcome})$, which simplifies to $p = U(\text{health outcome})$.



Decision Tree: basics

- Simplest form of decision model
- Short time horizon: e.g. 6-12 months
- Perspective: e.g. health system/provider, societal etc
- Pathways
 - Are routes through the decision tree
 - These are a sequence of mutually exclusive events
- Probabilities
 - Show likelihood of a particular event occurring at a chance node
 - Move from left to right
 - 1st probability: shows probability of an event.
 - 2nd probability: conditional or depends on whether an earlier event occurred or did not occur.
 - Pathway joint probability: obtained from multiplying probabilities along a pathway

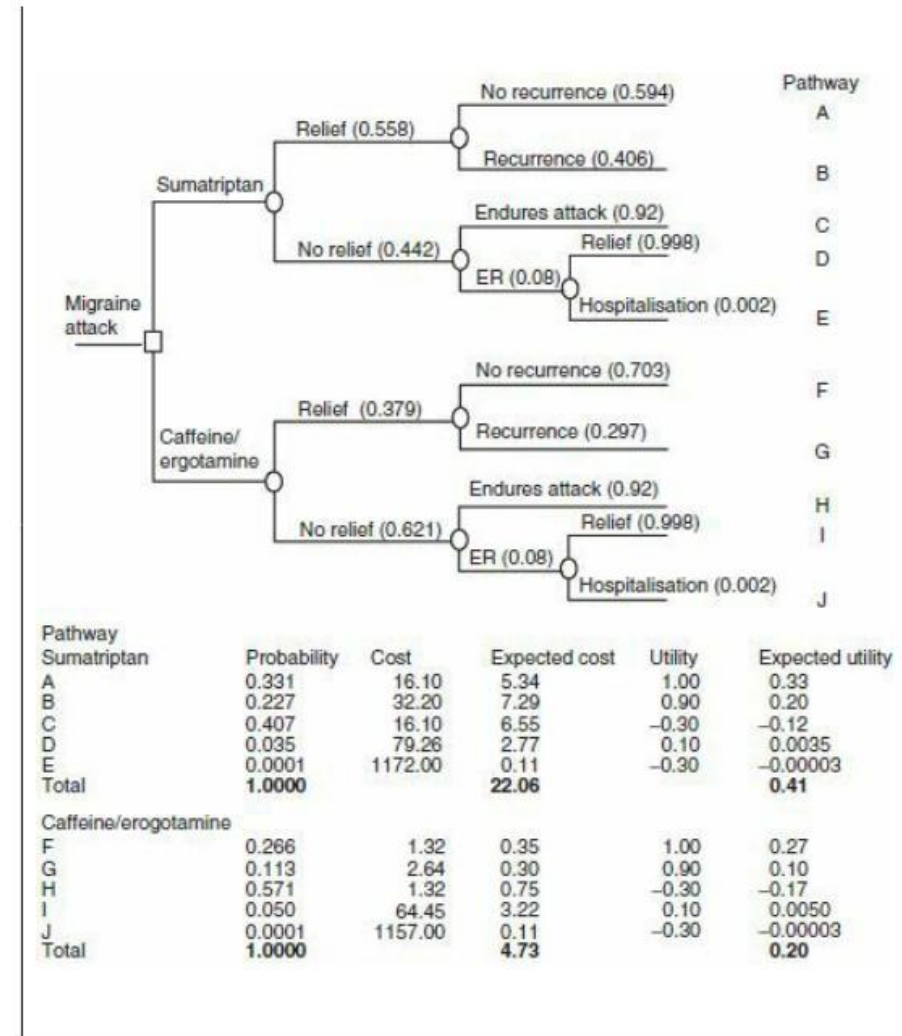


Decision nodes:

- **Square decision Node** (At start of tree):
 - Show decision point between alternative options.
- **Circular chance node** (branches coming out of the node):
 - Show a point where 2 or more alternative events for the patient pathway are possible.
- **Terminal nodes (Leaf node)**:
 - Contains records that do not pass through any further decisions

Decision Tree Probabilities

- **Probabilities in decision analysis:** a number indicating a likelihood of an event occurring in future.
- **Joint probability:**
 - The likelihood of 2 events taking place at the same time.
 - Notation = $P(A \text{ and } B)$
- **Conditional probability:**
 - Probability of event A, given that event B is known to have taken place.
 - Notation = $P(A|B)$
- **Independence:**
 - Events A and B are independent if the probability of event A, $P(A)$, is the same as the probability of $P(A|B)$.
 - When the events are independent $P(A \text{ and } B) = P(A) \times P(B)$.
- Relation of joint and conditional probabilities is shown here: $P(A \text{ and } B) = P(A|B) \times P(B)$

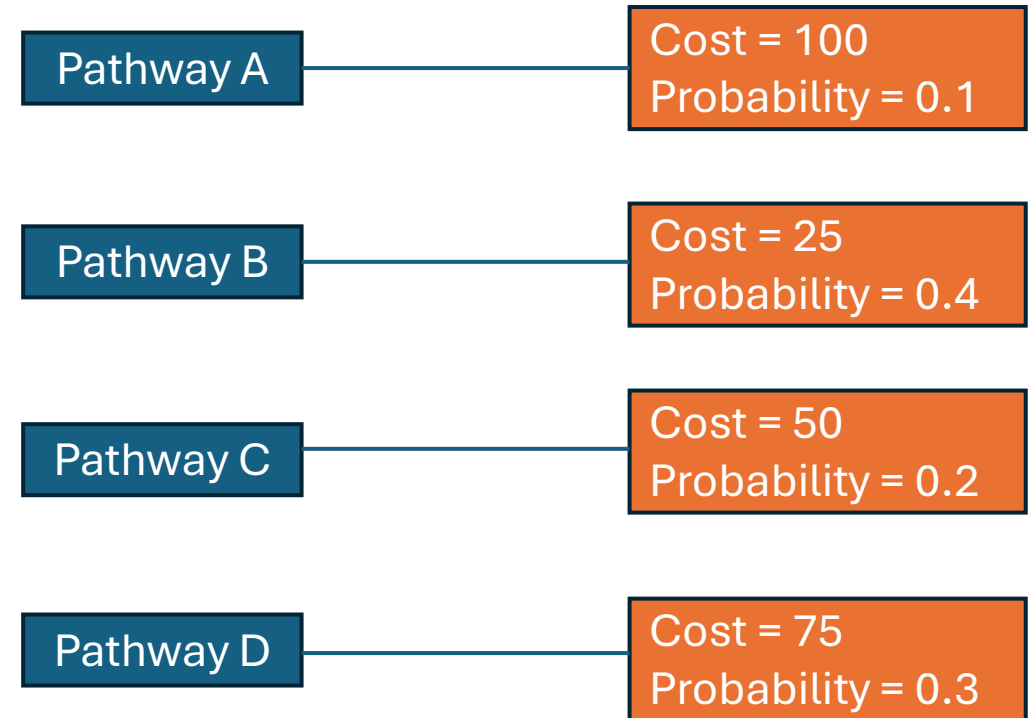


Expected values using costs

For a given option, the likelihood of each prognosis can be quantified using probabilities and their cost implications or health outcomes.

Health outcomes or Payoffs

- Costs would typically be one form of payoff but, on the effects side, a range of outcomes may be defined depending on the type of study.
- Increasingly, quality-adjusted life-years (QALYs) would be one of the payoffs in a decision model for cost-effectiveness analysis, which may or may not be based on utilities elicited using the standard gamble.
- The principle of identifying a preferred option on the basis of a decision analytic model is on the basis of expected values.

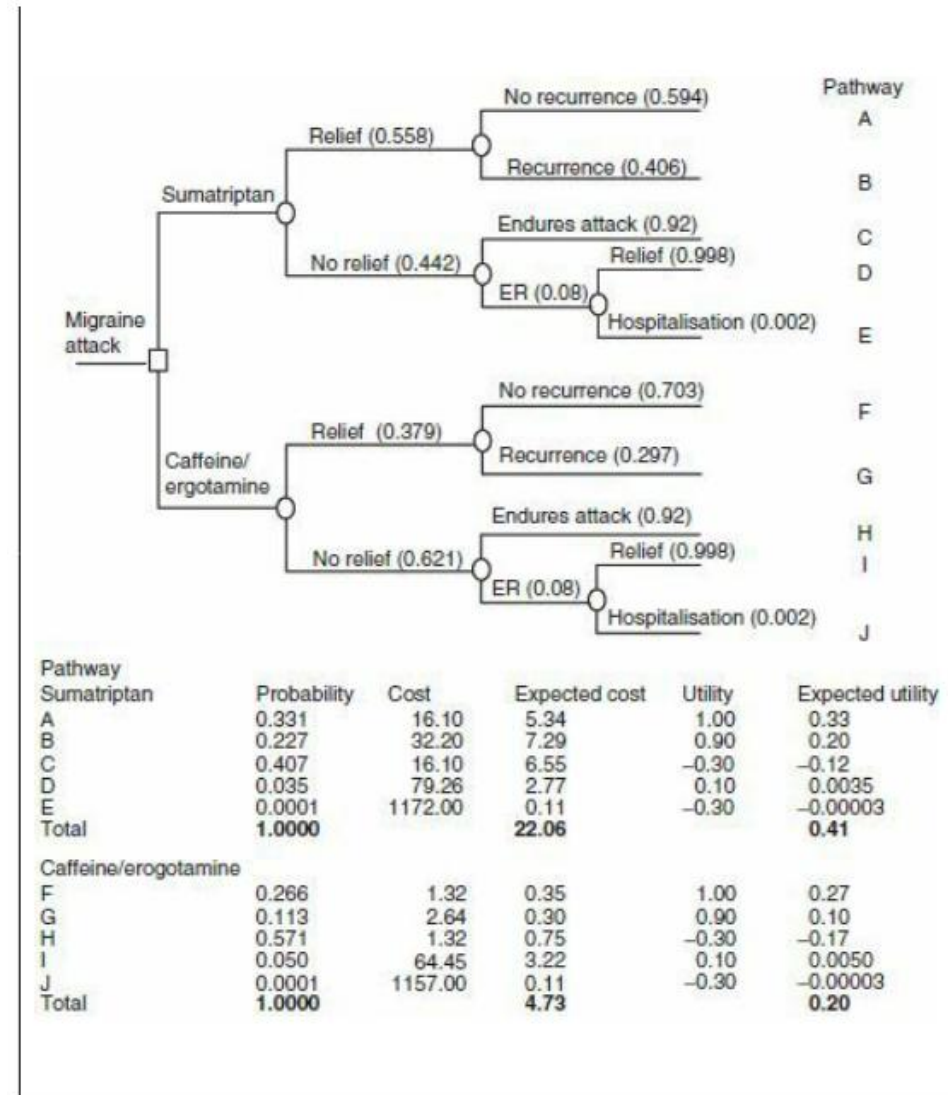


Expected costs = Path A + Path B + Path C + Path D
= $(100 \times 0.1) + (25 \times 0.4) + (50 \times 0.2) + (75 \times 0.3)$
= 52.50

Calculations

- Pathway A = Prob. of Relief_A x Prob. of NoRecurrence_A
 $= 0.558 \times 0.594$
 $= 0.331$
- Probabilities must equal to 1
 - Pathway probabilities (A+B+C+D+E) = 1
- Expected costs = Probability_B x Cost_B
 $= 0.227 \times 32.30$
 $= 7.33$
- Expected Utility = Probability_C x Utility_C
 $= 0.407 \times -0.30$
 $= -0.12$

Note: Utilities may be provided in the data (If not, you will find in the published literature)



(Source: Briggs *et al.*, 2006, pg 46 (Box2.3))

Decision Tree in R: Exercise

Q1. Set the sample size to 300.

Q2. Change the tree model 'type' to 2 or 5. What happens to the structure of the tree?

Q3. Change extra to 101. How many patients are in palliative care (Hint: numerical number)?

Q4. Remove the neat layout of the decision tree (Hint: opposite of TRUE)

Q5: Change the title of the decision tree to 'Decision Tree for Breast Cancer Treatment'

THANK YOU!



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