

1. Time-dependence (cont.)

2. Uncertainty

Ivy Tsai
20231031

Agenda 20231031

- 1st hour: quick review and R package – time-varying values
- 2nd hour: video and self-study
- 3rd hour: discussion – uncertainty/heterogeneity

Homework for 20231107:

Exercise 3.5 Hip Fracture **(MUST)**

Videos **(if no time to read Briggs and Edlin)**

Exercise 3.5 Hip Fracture

- Page 67-75

<https://www.herc.ox.ac.uk/downloads/decision-modelling-for-health-economic-evaluation>

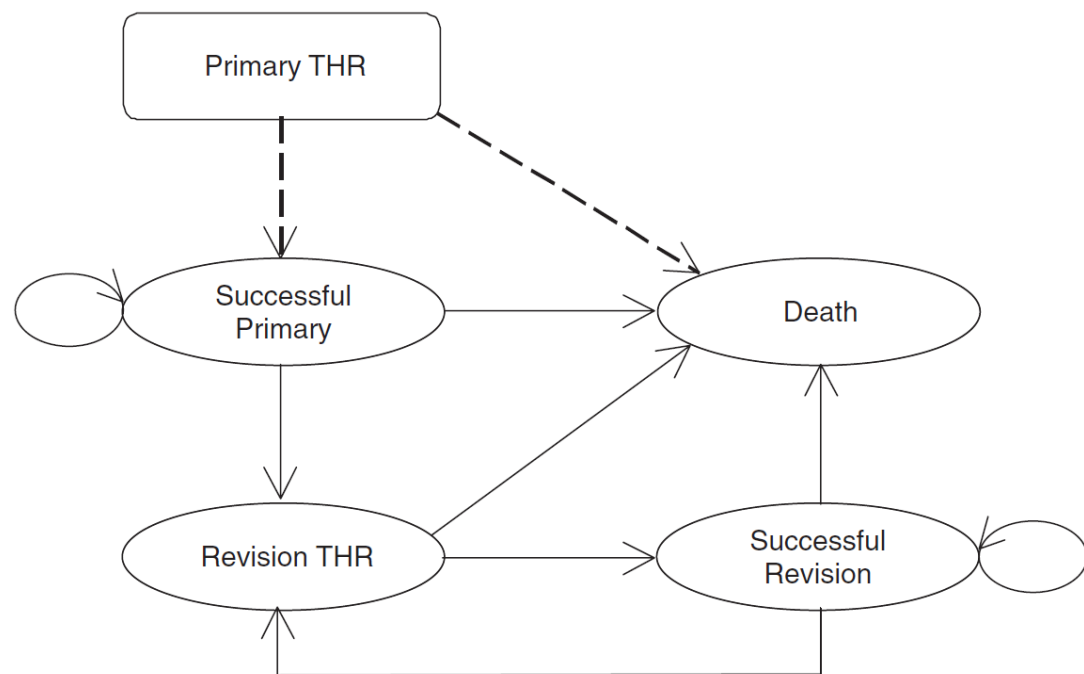
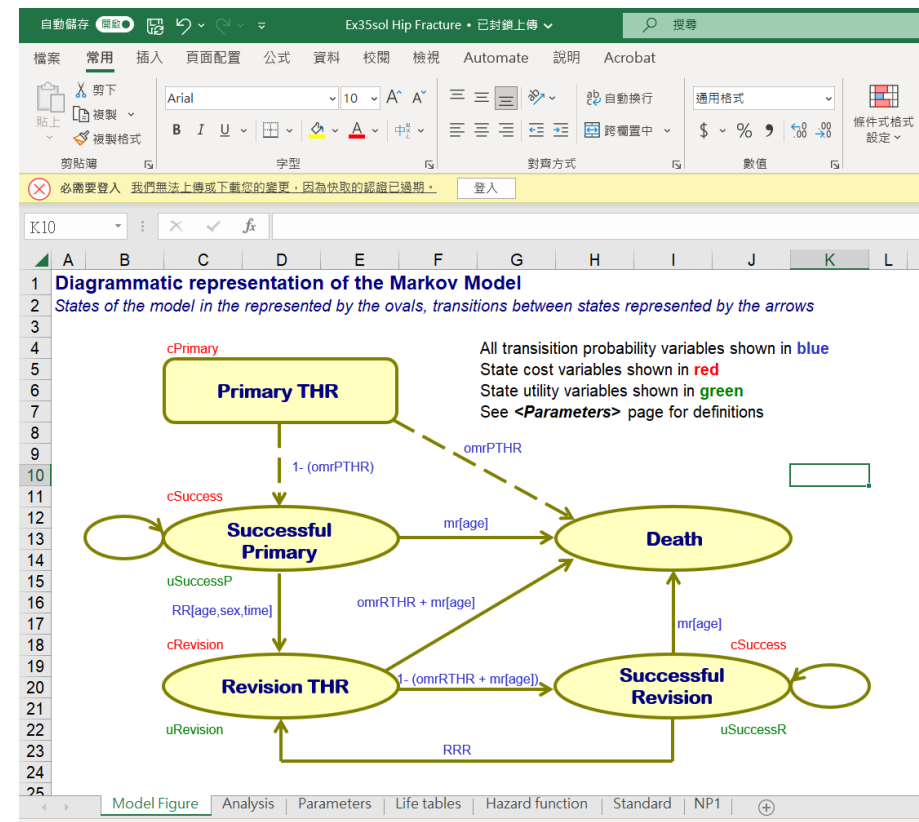
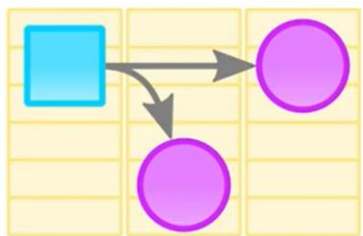


Fig. 3.7 Diagram showing the structure of the model (Briggs et al. 2004, Fig. 1).





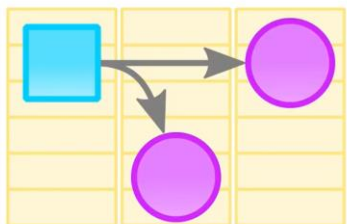
Decision analytical modelling
in health economics

15 March 2021

Spotlight on: Beta distribution

Spotlight on: Beta distribution

<https://www.youtube.com/watch?v=i8KvU-ODqgA>



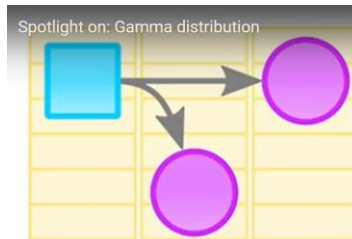
Decision analytical modelling
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18 June 2021

Spotlight on: Lognormal distribution

Spotlight on: Lognormal distribution

<https://www.youtube.com/watch?v=qxQ2GC1uYjs>



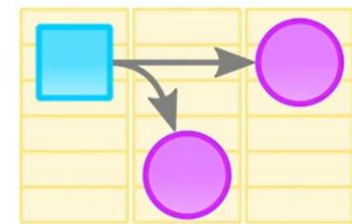
Decision analytical modelling
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18 June 2021

Spotlight on: Gamma distribution

Spotlight on: Gamma distribution

<https://www.youtube.com/watch?v=-pIIVcwq6NI>



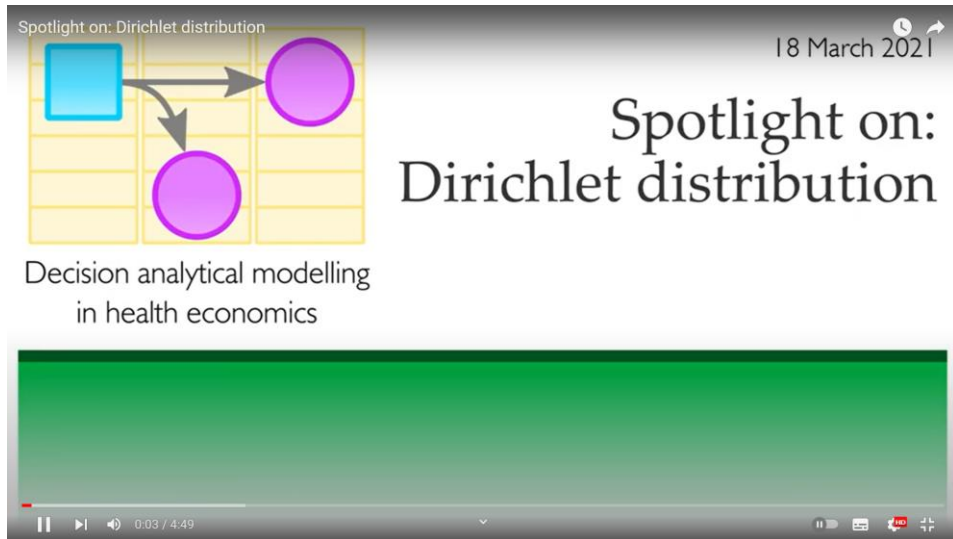
Decision analytical modelling
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18 March 2021

Spotlight on: Chi-square distribution

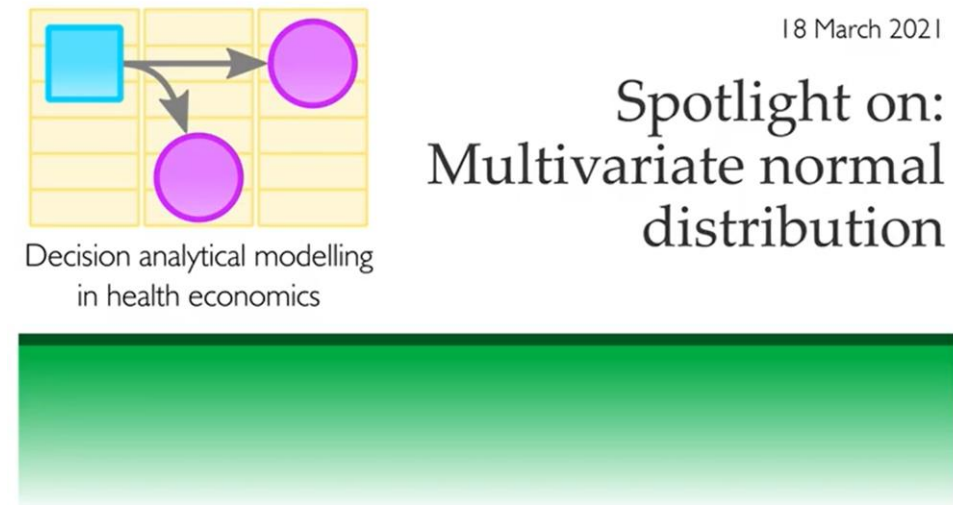
Spotlight on: Chi-square distribution

<https://www.youtube.com/watch?v=rtv7elZT-OQ>



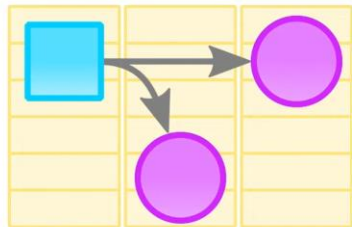
Spotlight on: Dirichlet distribution

<https://www.youtube.com/watch?v=eOOzVyBDHjc>



Spotlight on: Multivariate normal distribution

<https://www.youtube.com/watch?v=X3sbaqJ2tcY>



23 February 2021

Sensitivity analyses

Decision analytical modelling
in health economics

How do changes in the inputs affect the outputs?

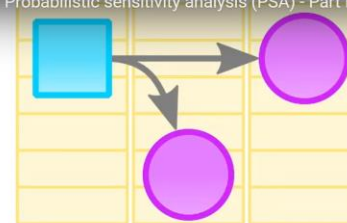


Sensitivity analyses in cost-effectiveness modelling

<https://www.youtube.com/watch?v=1txk9FUslbw&list=PLIT7NqYN7YT3LNYOoTEFchRXX7crLfYa5&index=10>

Probabilistic sensitivity analysis (PSA) - Part I (Introduction)

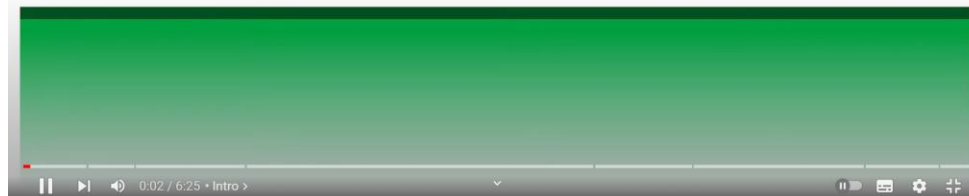
23 February 2021



Probabilistic sensitivity analysis

Decision analytical modelling
in health economics

Part I – Introduction

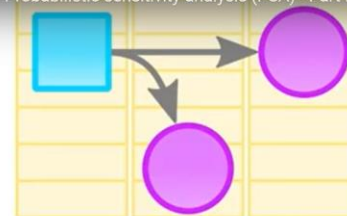


Probabilistic sensitivity analysis (PSA) - Part I (Introduction)

<https://www.youtube.com/watch?v=Mnftc8V0pU4>

Probabilistic sensitivity analysis (PSA) - Part II (Analysing and presenting results)

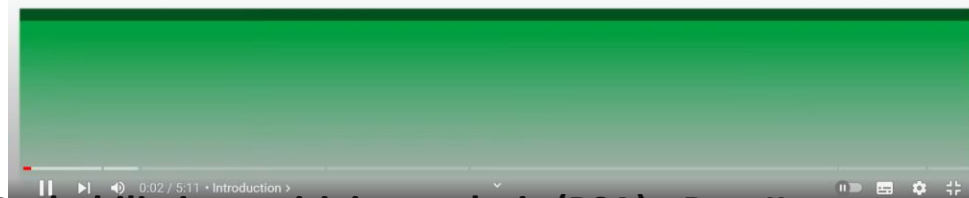
26 February 2021



Probabilistic sensitivity analysis

Decision analytical modelling
in health economics

Part II – Analysing and presenting results



Probabilistic sensitivity analysis (PSA) - Part II (Analysing and presenting results)

<https://www.youtube.com/watch?v=cTEceyHOqQQ>

Uncertainty in decision model

- York Materials



1. Define the elements of decision problem

Population:

Patients with advanced EC who *failed 1st line platinum-based chemotherapy*.

2nd line treatment:

Lenvatinib + pembrolizumab (LP) regimen

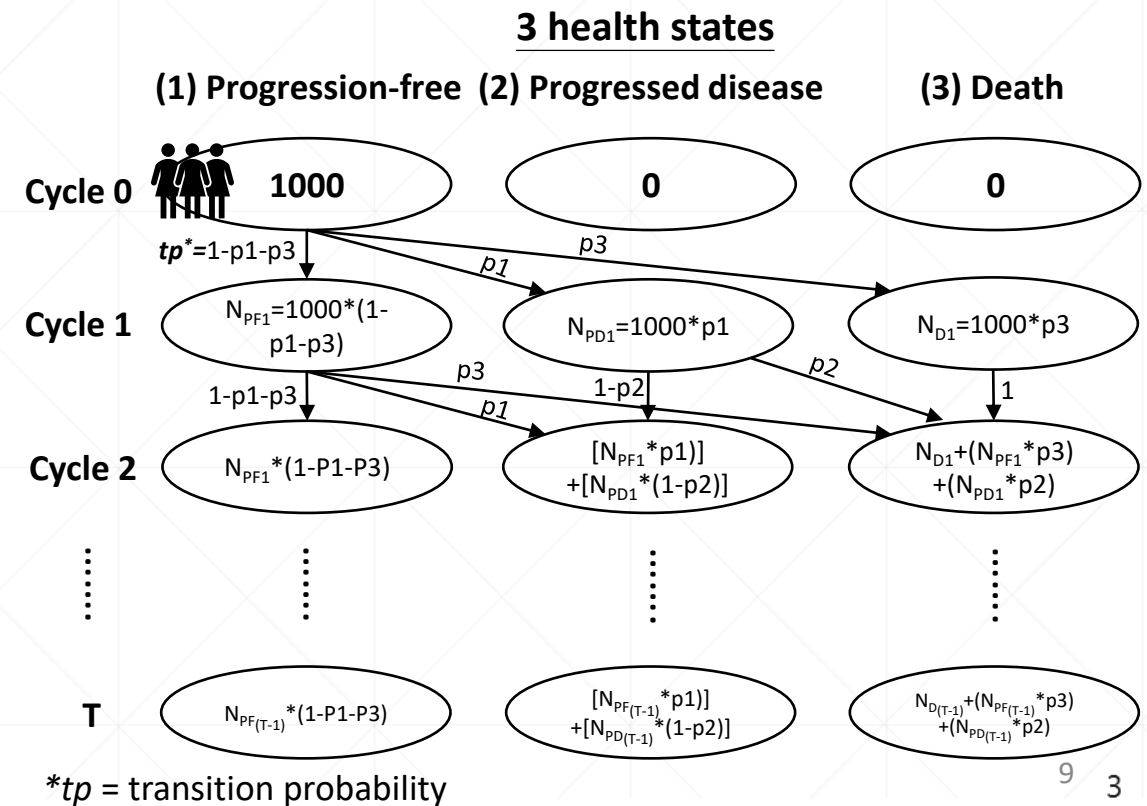
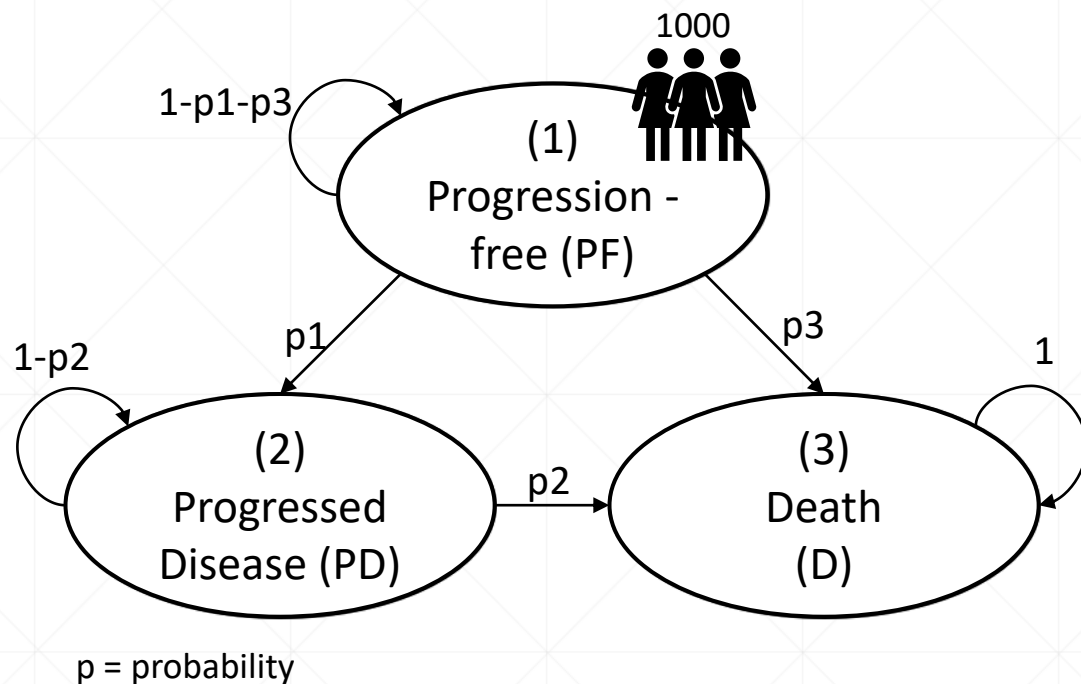
Chemotherapy (doxorubicin or paclitaxel)

Progressed from 2nd line
Supportive care

- **Cycle length:** 3 weeks
- **Time horizon (t):** 20 years
- **Discount rate (r):** 3% per year
- **Outcome variables:**
 - **Effectiveness (E_I and E_C)**
→ Life years (Lys), quality-adjusted life years (QALYs = Life year × quality of life).
 - **Directed medical costs (C_I and C_C)**
→ Medical cost reimbursed by NHI and NHI listing price.

2. Analytical Model building

- **Decision analytical model:** Cohort-based Markov model
- **Disease model:** 3 health states
 - (1) Progression-free (PF): Disease stable and keep current treatment.
 - (2) Progressed disease (PD): Disease progressed and shift to supportive care.
 - (3) Death (D)

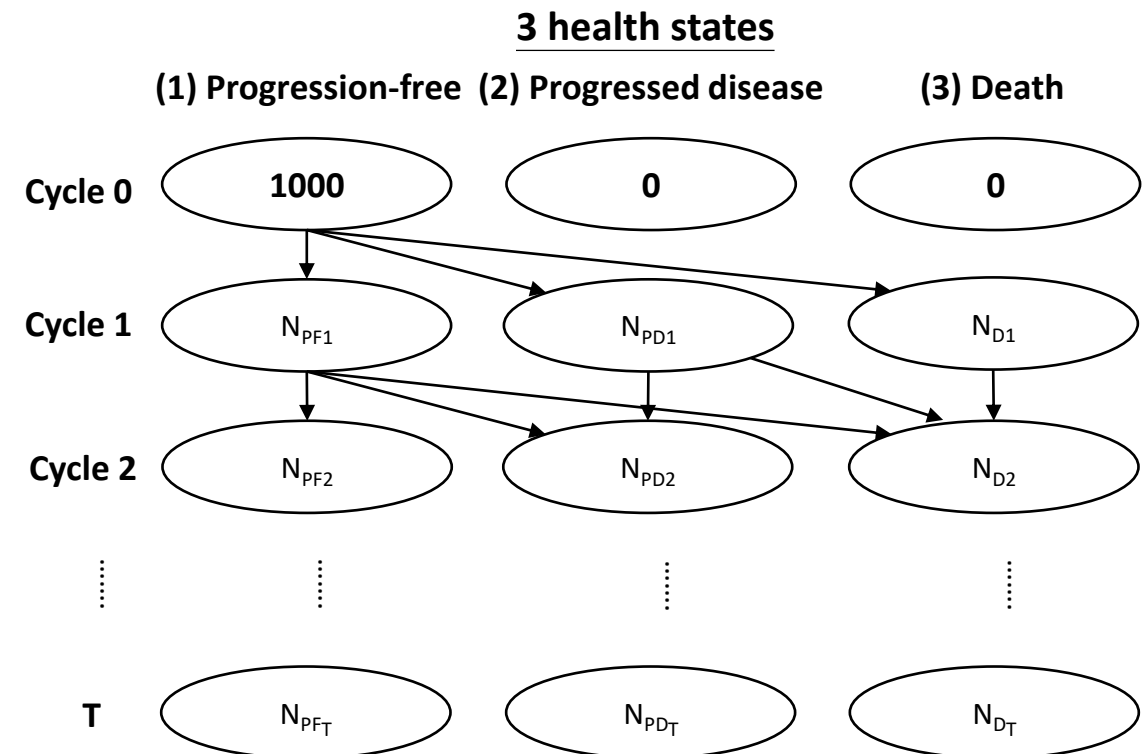


2. Analytical Model building

- **Decision analytical model:** Cohort-based Markov model
- **Cumulated outputs:** E_t , E_c , C_t , and C_c

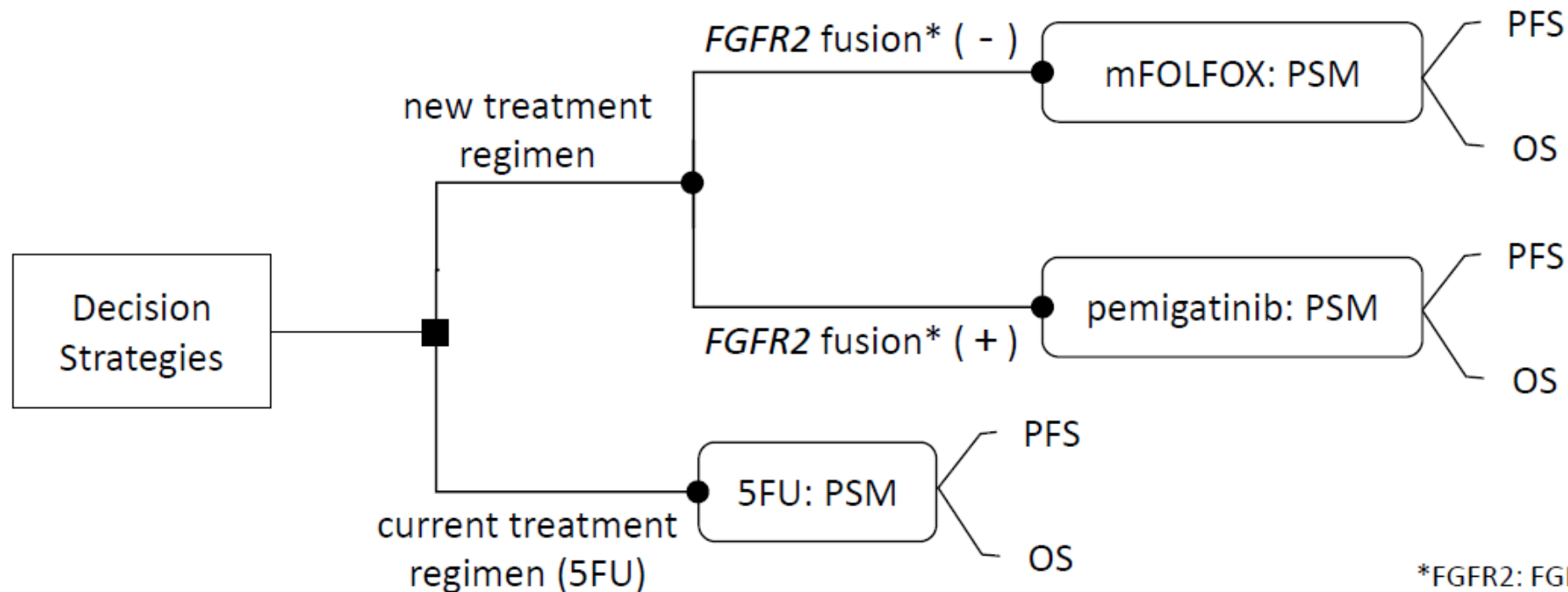
$$E = \sum_0^T \frac{1}{(1+r)^t} E_t \quad C = \sum_0^T \frac{1}{(1+r)^t} C_t$$

Cycle	Effectiveness (E) (Unit: LYs, QALYs*)	Cost (C) (Unit: NTD)
0	$E_0 = E_{PF0}$	$C_0 = C_{PF0}$
1	$E_1 = E_{PF1} + E_{PD1}$	$C_1 = C_{PF1} + C_{PD1} + C_{D1}$
2	$E_2 = E_{PF2} + E_{PD2}$	$C_2 = C_{PF2} + C_{PD2} + C_{D2}$
⋮	⋮	⋮
T	$E_T = E_{PFT} + E_{PDT}$	$C_T = C_{PFT} + C_{PDT} + C_{DT}$



*LYs = life years, QALYs = quality-adjusted life years.

- **Target population:** advanced ICC patients who failed their 1st line treatment
- **Treatment regimens:** 2nd line treatment for advanced ICC patients
 - Intervention: new treatment regimen (mFOLFOX and pemigatinib)
 - Comparator: current treatment regimen (5FU)



**FGFR2*: *FGFR2* gene fusion/rearrangement

- **Decision analytical model**

- Partitioned survival analysis (PartSA) model

- **Model structure**

- 3 Health states
 1. **Progression free (PF)**
 - Clinical performance: stable
 - Treatment: 2nd line new or current regimen
 2. **Progressed disease (PD)**
 - Clinical performance: tumor progressed
 - Treatment: supportive care
 3. **Death**
- Cycle length: 1 month
- Time horizon: 5 years
- Discounting rate: 3% (CDE, 2014)

- **Perspective**

- National Health Insurance Administration, Taiwan

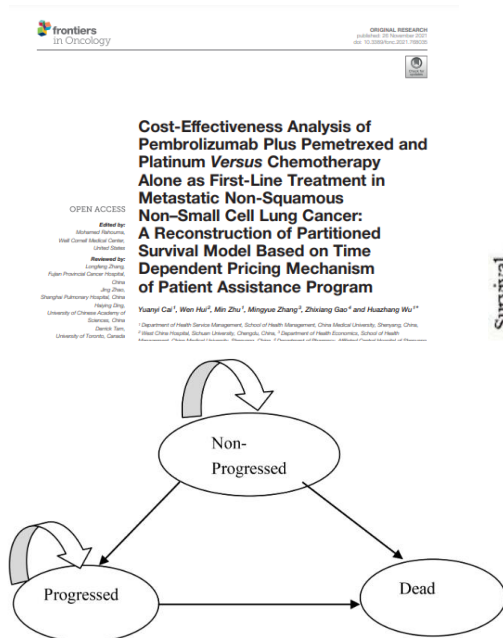
- **Outcomes**

- Life years, quality-adjusted life years (QALYs), direct medical costs, cost-effectiveness results

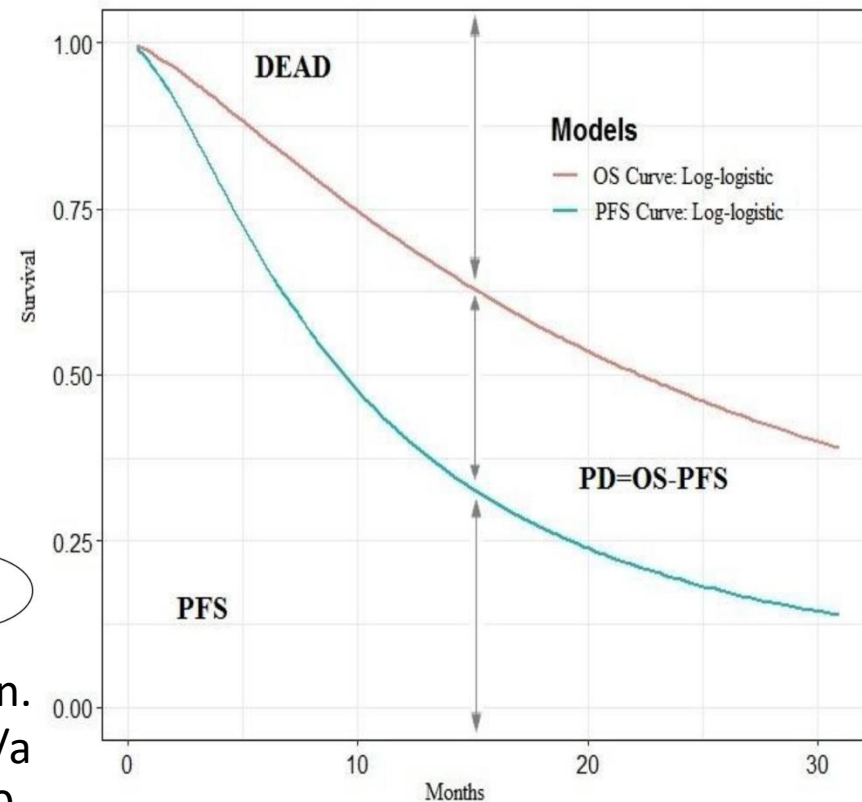
- **Parameters**

- Proportion of ICC patients with FGFR2 gene fusion
 - Taiwan: 7.7% (Chiang et al., 2021)
- Willingness-to-pay (WTP, λ) = 3 times GDP per capita
- Clinical efficacy
- Utility
- Direct medical costs

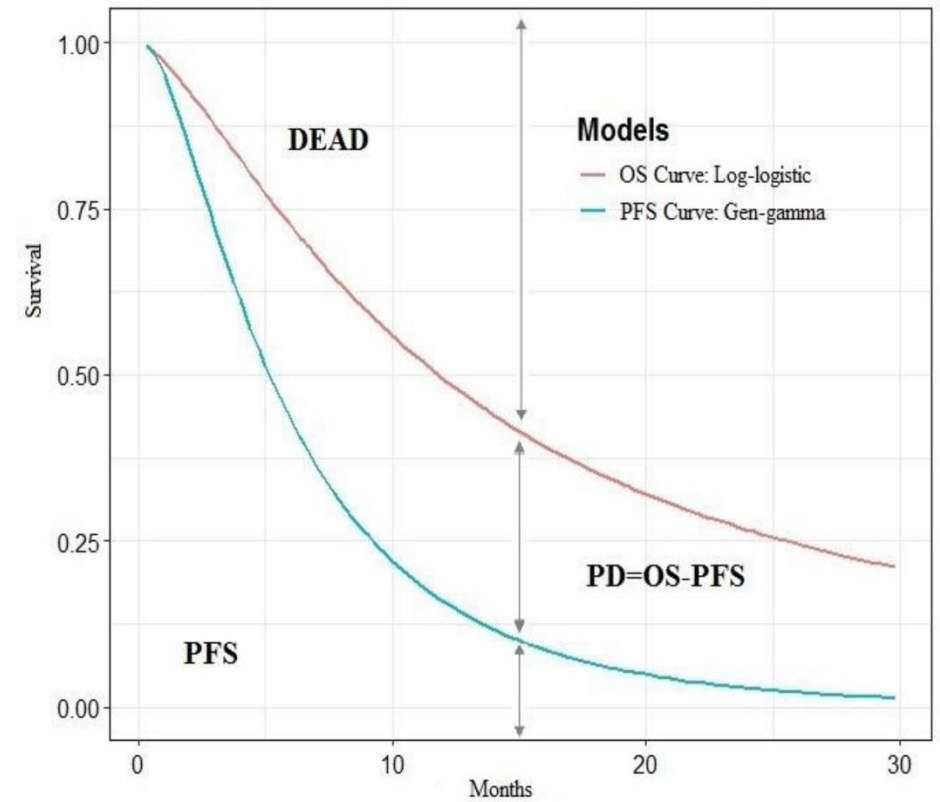
Partitioned Survival Model



<https://www.frontiersin.org/journals/oncology/articles/10.3389/fonc.2020.01768/full>



Intervention group



Comparator group

Partitioned Survival Model



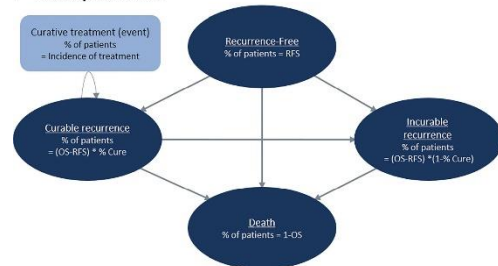
Cost-Effectiveness of Adjuvant Immunotherapy With Cytokine-Induced Killer Cell for Hepatocellular Carcinoma Based on a Randomized Controlled Trial and Real-World Data

Jeong-Yeon Cho^{1†}, Sun-Hong Kwon^{1†}, Eun-Kyung Lee², Jeong-Hoon Lee^{3*} and Hye-Lin Kim^{4*}

¹ School of Pharmacy, Sungkyunkwan University, Suwon, South Korea, ² Department of Internal Medicine and Liver Research Institute, Seoul National University College of Medicine, Seoul, South Korea, ³ College of Pharmacy, Seoul National University, Seoul, South Korea, ⁴ College of Pharmacy, Seoul National University, Seoul, South Korea

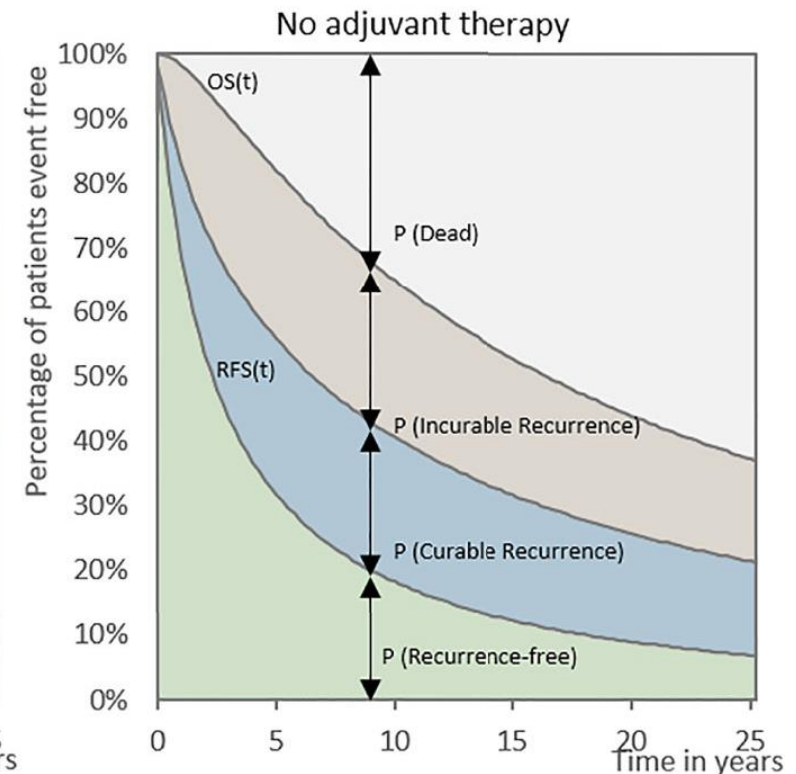
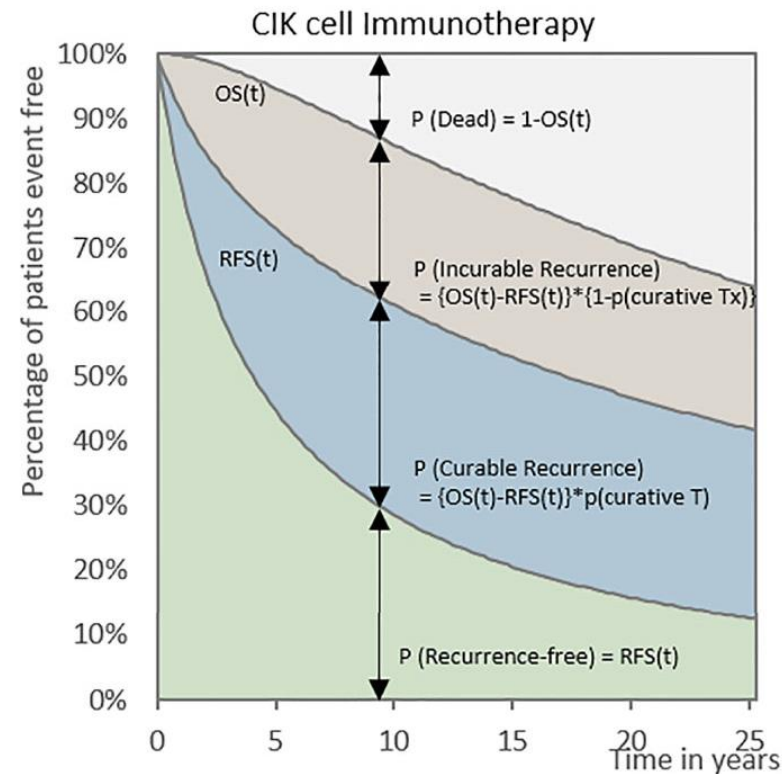
Background: Studies using data from randomized controlled trials (RCTs) and real-world data (RWD) have suggested that adjuvant cytokine-induced killer (CIK) cell

A Conceptual model



<https://www.frontiersin.org/journals/oncology/articles/10.3389/fonc.2021.728740/full>

B Partitioned survival model



● Incremental cost effectiveness ratio (ICER)

The cost per unit of the health outcome/effect.

Function:
$$\text{ICER} = \frac{C_2 - C_1}{E_2 - E_1} = \frac{\Delta C}{\Delta E}$$

● Net monetary benefit (NMB)

Function:
$$\text{NMB} = \lambda \times \Delta E - \Delta C$$

C_1 : the cost under the comparator.

C_2 : the cost under the intervention of interest.

E_1 : the effectiveness under the comparator.

E_2 : the effectiveness under the intervention of interest.

λ : threshold, willingness to pay (WTP)

✓ Decision criteria

- $\text{ICER} < \lambda$
- $\text{NMB} > 0$

$\lambda = 3$ times of GDP per capita in Taiwan (2021)

(Edlin et al., 2015; Gray et al., 2010)

● Base-case: Cost-effectiveness outcomes in 5 years

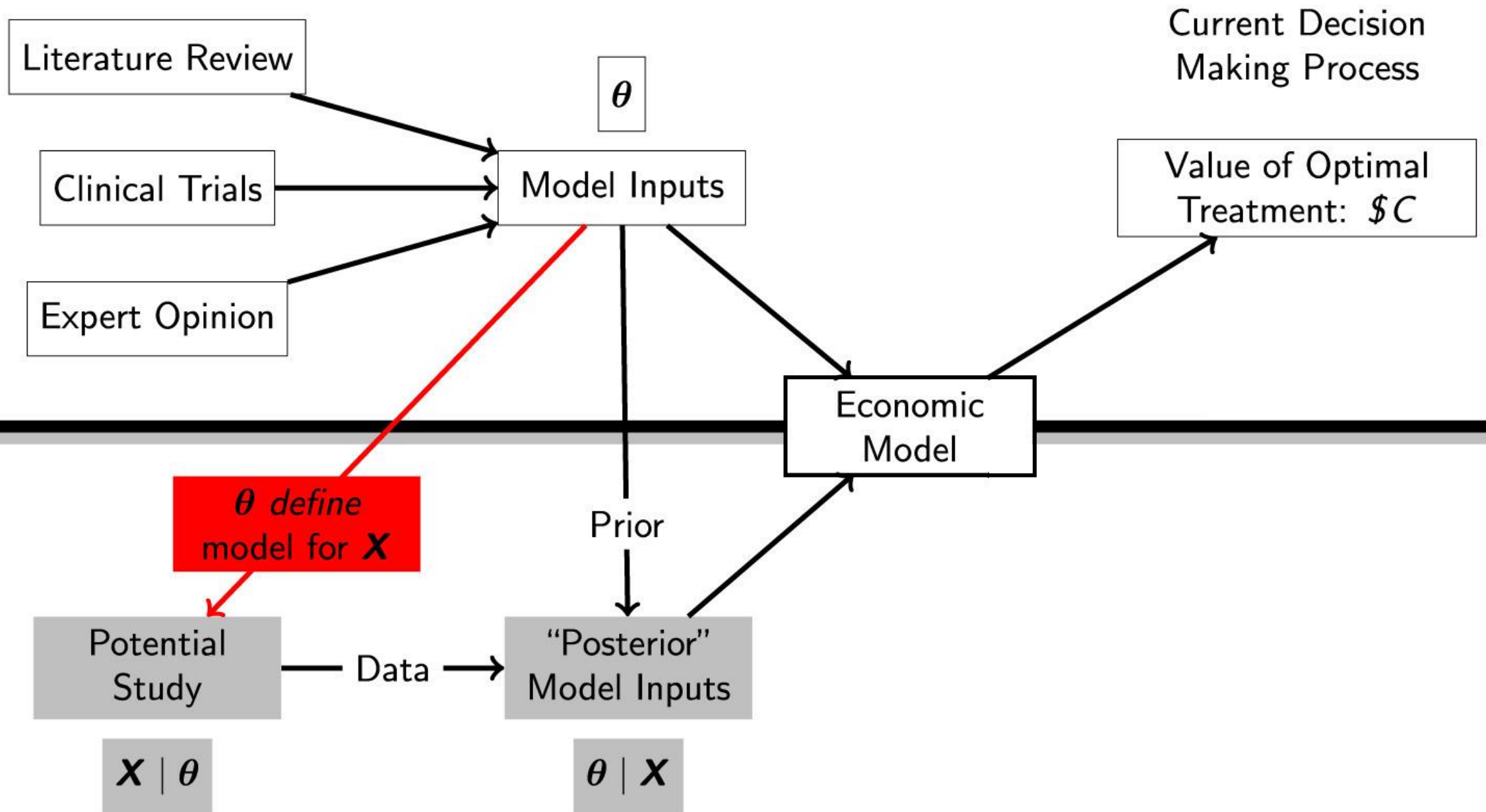
	Current regimen (5FU)	New regimen (mFOLFOX/pemigatinib)	Incremental change
Cost	524,472	984,168	459,697
Total cost of PF state	369,229	795,614	
• Genetic test cost	0	30,000	
• Medication costs (PF)	63,430	387,176	
• Non-medication cost (PF)	305,799	378,437	
Total cost of PD state	155,243	188,555	
Life years			
Progression-free	0.36	0.48	0.12
Overall	0.67	0.86	0.19
Quality-adjusted life years			
Progression-free	0.26	0.35	0.09
Overall	0.47	0.61	0.13
Incremental cost per QALY (ICER)	WTP (λ) = 3 times GDP (NT\$2,889,684)		3,411,098
NMB			-70,269

LYs: life years, QALYs: quality-adjusted life years, ICER: incremental cost-effectiveness ratio, NMB: net monetary benefit

Findings:

- ICER (NT\$ 3,411,098) > WTP (NT\$2,889,684)
- NMB (NT\$ -70,269) < 0

NOT cost-effectiveness in base-case analysis!



Q1: When we make decisions based on point estimates in cost-effectiveness analysis, what potential issues or concerns are there?

- **Why** and **what** do we (have to) be worried or concerned about the “evidence-based” decision?

□ **Why?**

□ **What?**

Not sure about the evidence (point estimates of CEA and NMB/NHB):

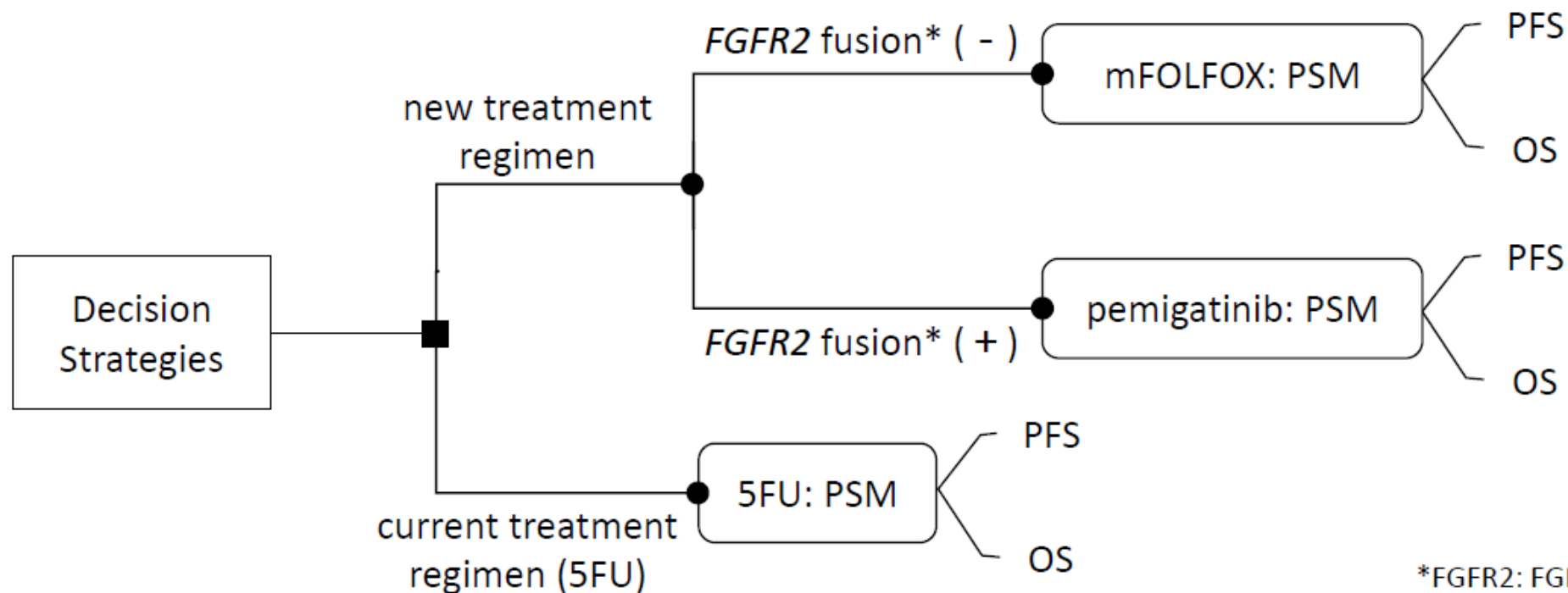
What is uncertain about cost-effectiveness analysis?

The repercussions of a
wrong decision,
founded on
inadequate evidence.

- Decisions should not be based on little or poor quality evidence
- Always a chance that the wrong adoption decision is made, resulting in health benefit and resources forgone



- **Target population:** advanced ICC patients who failed their 1st line treatment
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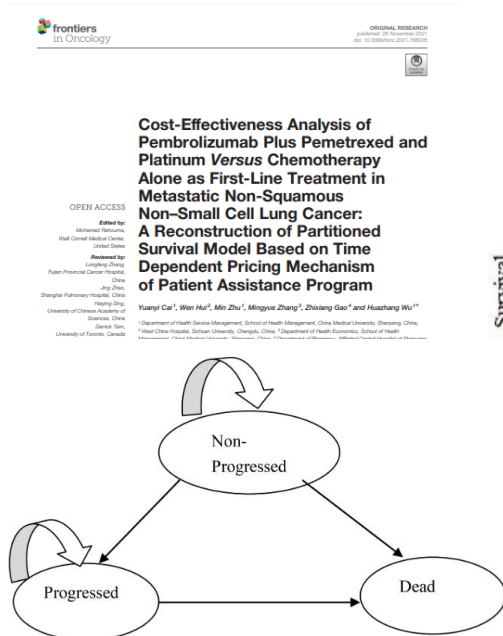
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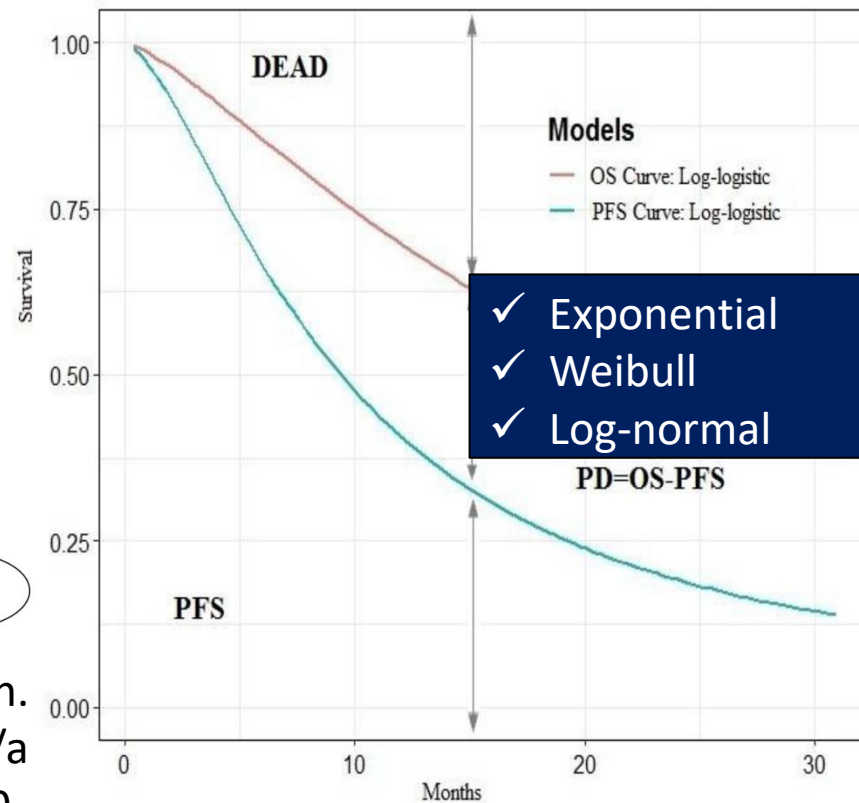
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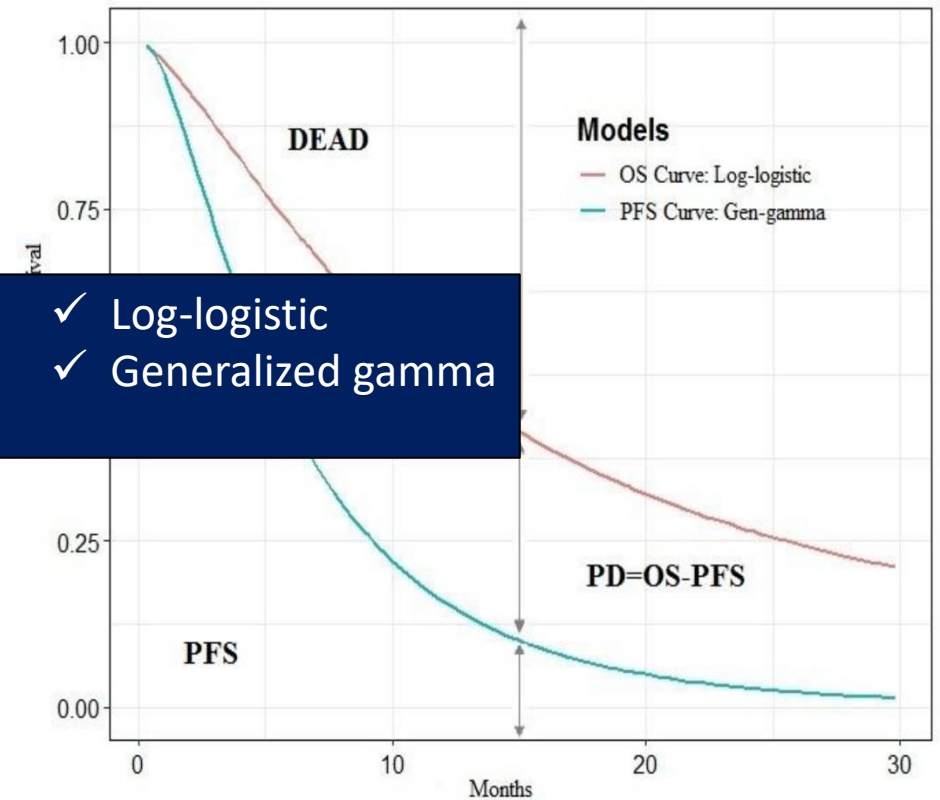
Partitioned Survival Model



<https://www.frontiersin.org/journals/oncology/articles/10.3389/fonc.2021.768035/full>



Intervention group

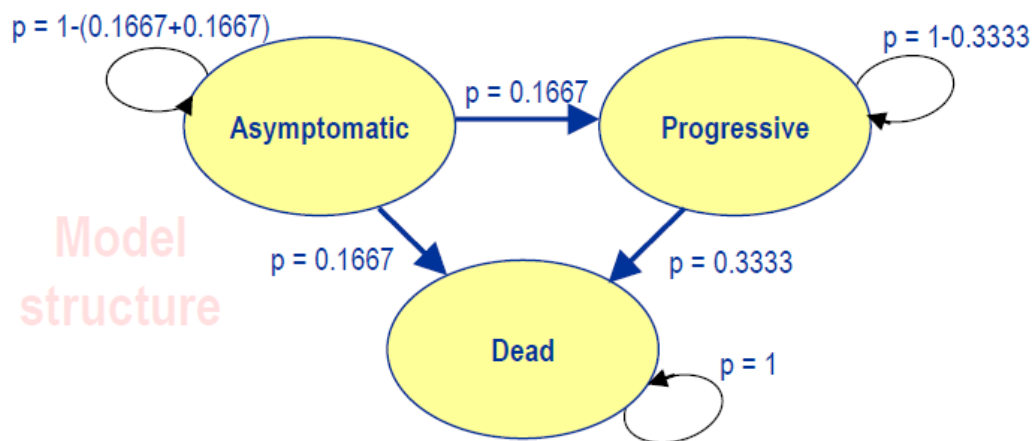


Comparator group

✓ Exponential
✓ Weibull
✓ Log-normal

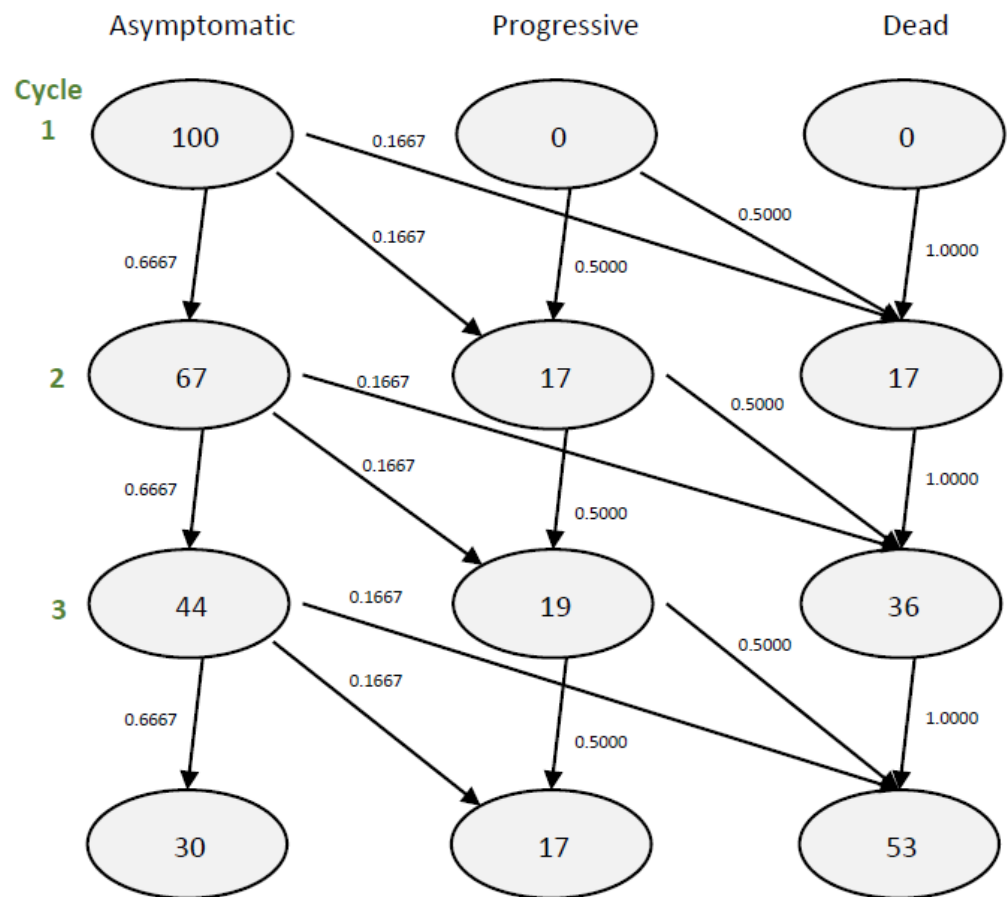
✓ Log-logistic
✓ Generalized gamma

Outline of a simple Markov model



Current treatment	Asymptomatic	Progressive	Dead	Cost	QALY
Asymptomatic	0.6667	0.1667	0.1667	£150	0.9
Progressive		0.5000	0.5000	£325	0.78
Dead			1.0000		

Parameter values



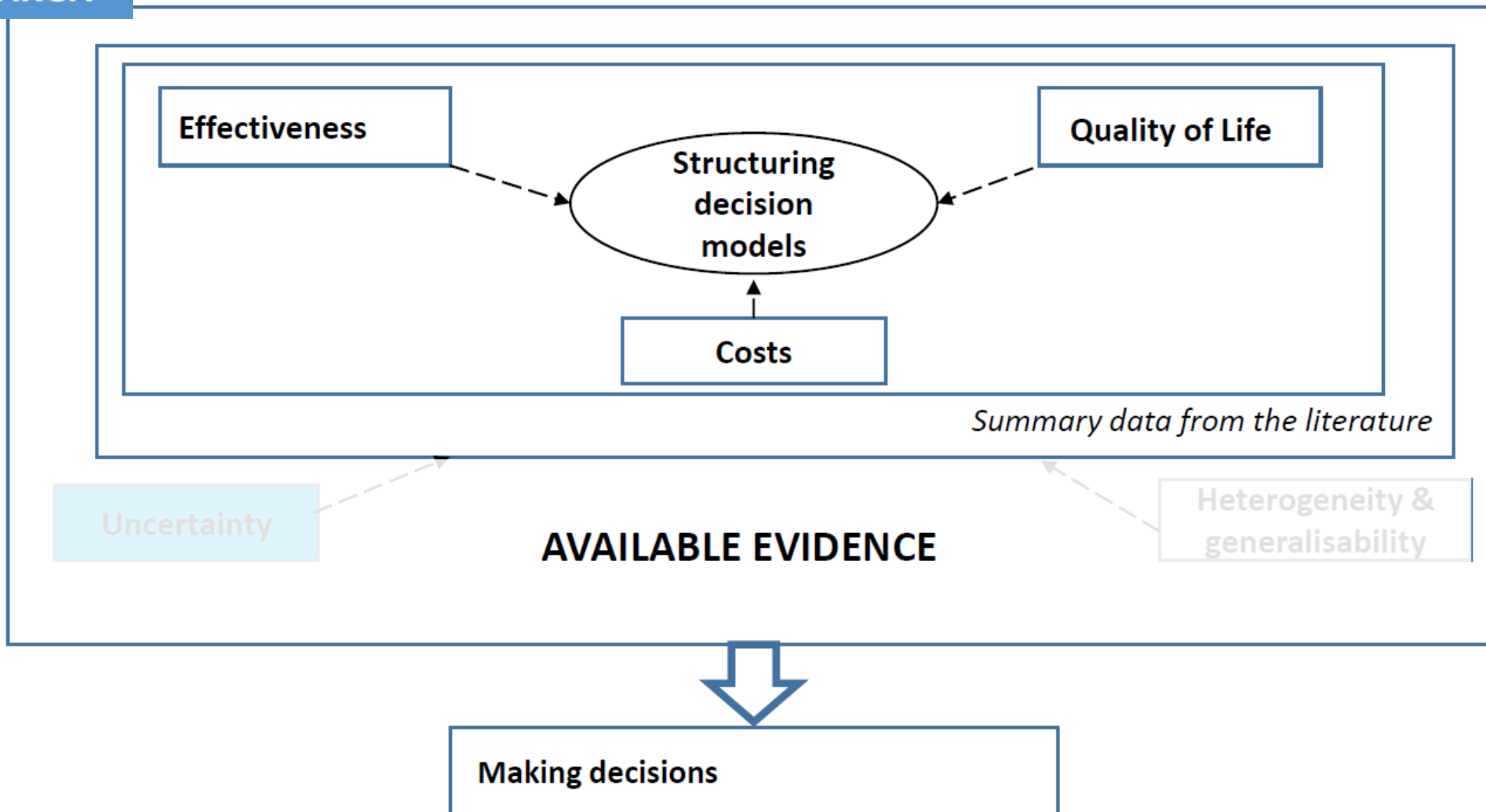
Cohort simulation

What is uncertain about cost-effectiveness analysis?

The repercussions of a
wrong decision,
founded on
inadequate evidence.

- Decisions should not be based on little or poor quality evidence
- Always a chance that the wrong adoption decision is made, resulting in health benefit and resources forgone
- Different possible values for the parameters
 - Lack of knowledge about the parameter values
 - Different outcomes in different populations
- Structural uncertainty
 - Choice of health states, choice of modelling approach
- Distinguish between
 - Uncertainty, variability, heterogeneity and policy choices

Course structure – where are we up to?



Q2. What are bias, variability, heterogeneity, uncertainty?

- (Briggs: **Box 2.1.**; page 80-82)

Box 2.1. Key concept in understanding uncertainty and heterogeneity in decision models for cost-effectiveness analysis

First and second order uncertainty

1st order

- Distribution of outcomes in population
≈ Sample variance
- Standard deviation in a mean value
 - Range of outcomes in sample
- Incorporate in CEA by simulating and recording pathway of individual patients through a model
- Large number of patients required to estimate mean and standard deviation
- Must repeatedly sample large numbers of patients to estimate uncertainty in mean and standard error

2nd order

- Distribution of sample mean outcome
≈ Variance of sample mean
- Standard error of mean
 - Range of population mean values supported by the sample outcomes
- Incorporate in CEA by simulating and recording pathway of cohort through a model
- One cohort provides estimate of mean but no information on standard deviation
- Large number of cohorts entered into models to estimate uncertainty in mean and standard error

1st order uncertainty – screen share example

- Simulate individual patients progress through model
- Random numbers to determine occurrence of chance events
- Markov trace generated for multiple individuals to get mean costs and QALYs

TRANSITION MATRIX <i>From</i>	<i>To</i>		
	Asymptomatic	Progressive	Dead
Asymptomatic	0.6667	0.1667	0.1667
Progressive		0.6666	0.3333
Dead			1.0000



	Asymptomatic	Progressive	Dead
Asymptomatic	1, 2, 3, 4	5	6
Progressive		1, 2, 3, 4	5, 6
Dead			1, 2, 3, 4, 5, 6

Uncertain decisions

- 1st order uncertainty and variation within groups of patients not the focus of CEA
 - Decision must be made for group as a whole
 - Variability cannot be reduced
 - Computationally time consuming when combined with 2nd order uncertainty
- 2nd order uncertainty is the focus of CEA
 - Informs questions about likelihood of making wrong decision, and likelihood of new information changing the optimal decision
- Structural uncertainty
 - Lack of knowledge about most appropriate model structure
 - Different modelling approaches provide different estimates of mean costs and QALYs
 - Contributes to uncertainty in mean outcomes

Heterogeneity – multiple decisions

- ‘Baseline’ characteristics ‘explain’ a proportion of overall variability between patients (e.g. age, sex)
- Can condition decision on these characteristics, and recommend different options in different groups
- To incorporate in CEA, generate mean parameter values per sub-group population
 - Variability within sub-group will remain
 - Need to present results by sub-group (defined by patient characteristics)

Policy choices and value judgements

- The authority taking the decision may set some parameter values
 - For example, the discount rate applied to costs and health outcomes
- The values are relevant for particular decision
 - E.g. NICE specifies 3.5% for costs and health outcomes
 - Sensitivity analysis of 1.5% per annum
- Different decision makers may have different values
 - Heterogeneity in value of parameter between decision makers
 - Choice taken by an individual decision maker is not uncertain

‘5.1.2 There is considerable debate about the most **appropriate methods** to use for some aspects of health technology assessment. This **uncertainty relates to choices that are essentially value judgements**; for example, whose preferences to use for valuation of health outcomes.... The reference case specifies the methods considered by the Institute to be the most appropriate for the Appraisal Committee’s purpose and consistent with an NHS objective of maximising health gain from limited resources.’

NICE Guide to the methods of technology appraisal 2013

An analogy with regression(Briggs, page 83)

$$Y = \alpha + \sum_{j=1}^p \beta_j X_j + \varepsilon$$

Q3: what should we focus on in CEA analysis?

- Parameter uncertainty
- Heterogeneity
- Structural
- Variability
- Policy choice

Types of uncertainty - Summary

Need to address	Not main focus of CEA

Q4: how to handle heterogeneity?

Data	Probabilities from binomial data	Probabilities from time-to-event data	Continuous variable like cost and utility
Model			
$E[Y]$			
$\text{Var}(Y)$			
Distribution			
Notes			