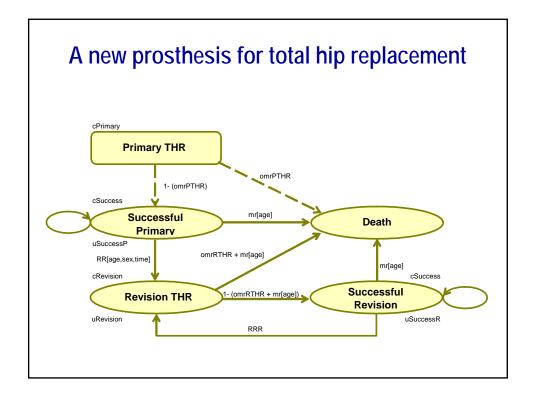
### **Probabilistic Sensitivity Analysis**

Dispelling the myths...

Andrew Briggs, DPhil *University of Oxford* 

#### Some common myths surrounding PSA...

- But Claxton tells us 'inference is irrelevant' which means that we don't have to worry about uncertainty? Right?
- But PSA is complicated, we can handle uncertainty using standard sensitivity analysis methods? Right?
- But there are so many potential distributions, PSA is just an arbitrary exercise? Right?, WY
- PSA is all very well for simple models, but for my all-singing all-dancing discrete-event micro-simulation Markov-decision-process with policy relevant individual patient prediction module PSA is computationally expensive? Right?

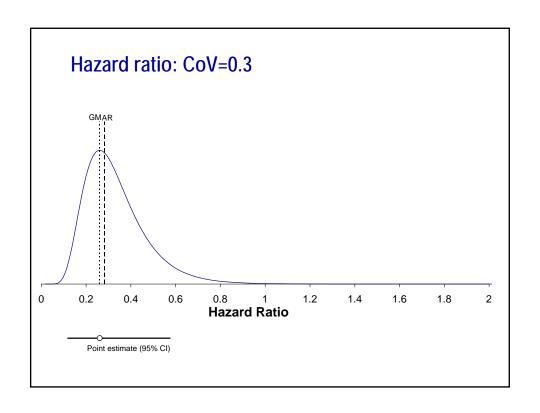


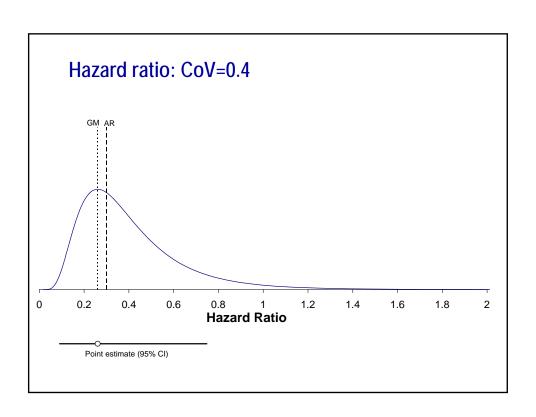
# But Claxton tells us 'inference is irrelevant' which means that we don't have to worry about uncertainty?

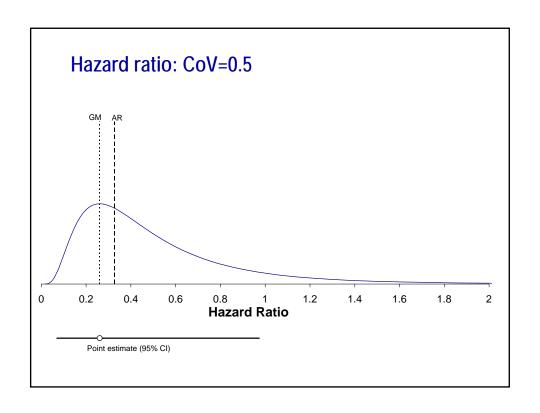
Even if this were true...

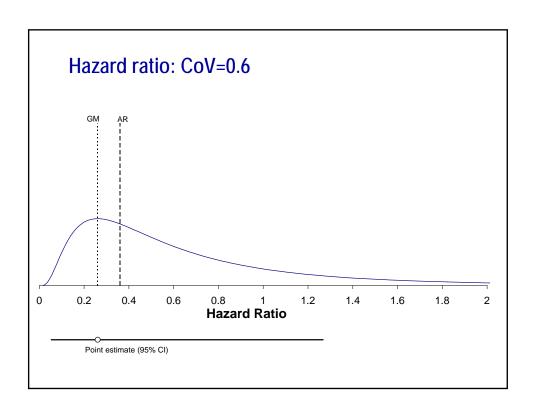
- Input parameters may be skewed and standard reporting may not focus on mean
- · Most decision models are not linear
- Uncertainty in input parameters affects the Expectation of the output parameters of interest

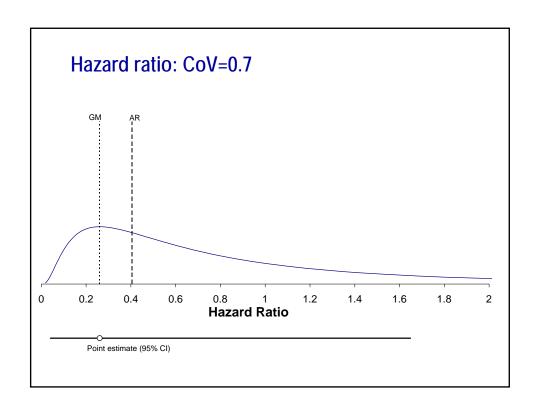
Example: Hazard ratio of risk of revision in hip model

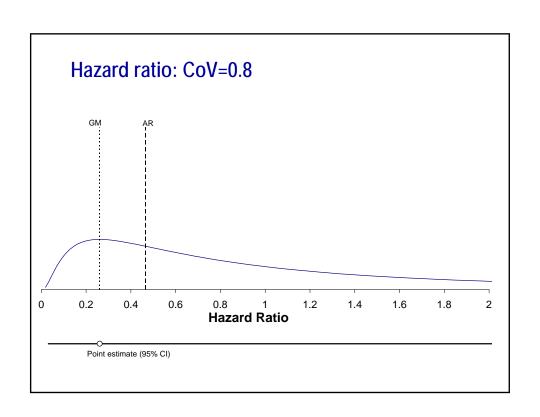


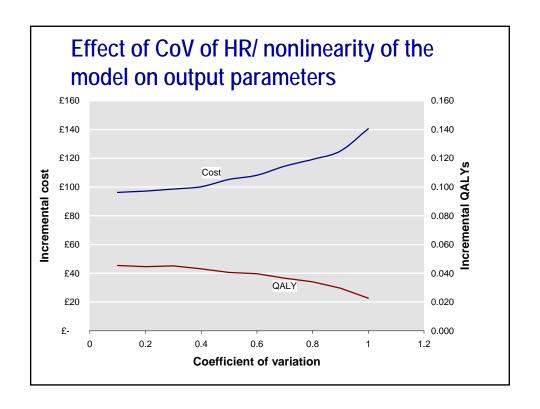






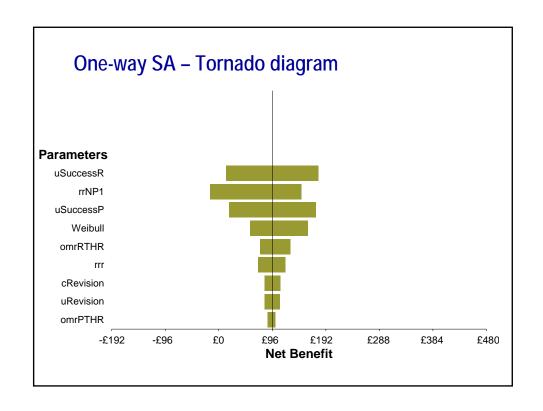


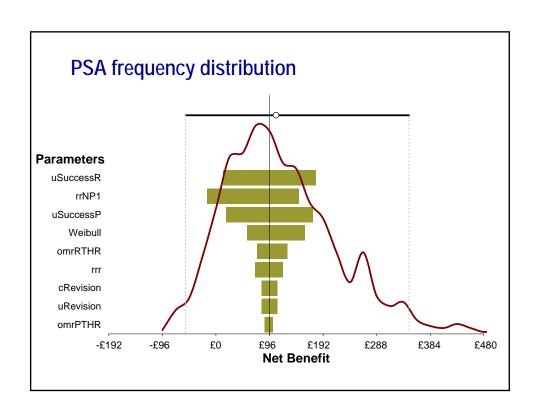


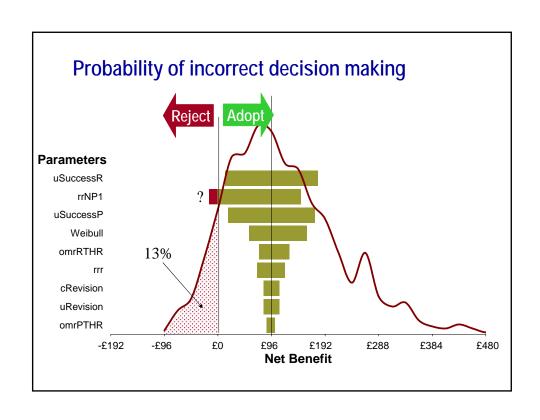


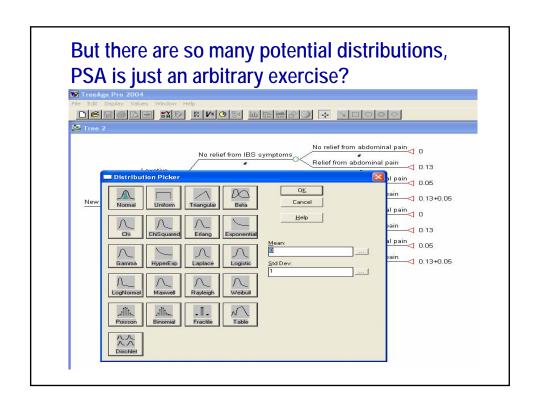
## But PSA is complicated, we can handle uncertainty using standard sensitivity analysis methods?

- One-way sensitivity analysis can under-represent decision uncertainty
- Multi-way analysis is cumbersome and doesn't say anything sensible about how likely different combinations of parameters are
- Both effectively ignore correlation









#### Choice of distribution is not arbitrary...

- A normal distribution is always a candidate
  - Central Limit Theorem
- Otherwise small number of candidates:
  - Type of parameter (logical restrictions)
  - Data being used to inform parameter estimation
  - Method of estimation

### For example: Probability parameters Logical restriction: zero-one interval

Type of data: binomial

Estimation: univariate proportion

- Use Beta distribution

· Type of data: binomial

Estimation: multivariate logistic regression

- Use multivariate normality on log odds scale

 Type of data: time to event Estimation: survival analysis

- Use multivariate normality on log hazards scale

PSA is all very well for simple models, but for my all-singing all-dancing discrete-event microsimulation Markov-decision-process with policy relevant individual patient prediction module, PSA is computationally expensive?

True, but...

- The ends justify the means
- Uncertainty is important, therefore PSA is necessary

So...

- Buy a faster computer
- Leave it running over the weekend/semester/until 2006 (haven't you heard of Occam's Razor?)

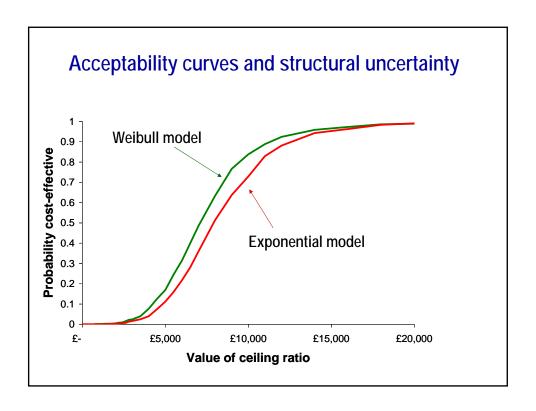
#### Nevertheless, some important caveats...

Garbage in / Garbage out

- If we want to make statistical statements about outcomes, we had better be careful how we characterise our inputs
- Your VOI analysis is only as good as the PSA on which it is based!

Parameter uncertainty is not the only uncertainty!

- Structural uncertainty may be just as important
- Continuing role for traditional sensitivity analysis
- Need to see PSA for all scenarios?



#### **Concluding comments**

- · Decision uncertainty is important
- PSA can characterise parameter uncertainty to inform decision uncertainty
  - Less arbitrary than traditional SA
  - Process of choosing distributions should encourage more careful consideration of parameter uncertainty
- Continuing role for traditional SA for structural uncertainty
  - In addition to PSA