





Quantitative Risk Assessment







Risk Analysis training course NIHE, Hanoi, Vietnam, 4-5 May 2016



Definition

- QRA involves estimating both the probability that an unwanted outcome will occur and the magnitude of the effect of this occurrence;
- Mathematical modelling process

Steps to model your system

- □ 7 related steps:
 - Define the risk to be assessed
 - Outline the model framework
 - Identify the inputs and acquire the data
 - Adopt an iterative approach to the model
 - Undertake sensitivity testing
 - Ensure clarity by producing a transparent model description
 - Obtain feedback

The RA components in a OIE quantitative framework

- Release of agent (from the animal/farm) EVENTS OUTSIDE END RECEPTOR
 - Starts by using herd and within-herd prevalence data.
 - Looks at pathways in which the infected commodities are not detected under the surveillance system components (sanitary measures; diagnostic tests)
- Exposure to agent (exposure of susceptible animal in importing country)
 - Conditional probability
 - The exposure end-point is usually taken as the point of entry into the importing country, not exposure of a susceptible animal which is harder to estimate.
 - Looks at pathways in which the infected animals (or animal products) are not detected under the sanitary measure (usually diagnostic tests).

Establishment, spread and other consequences

- Is considerably harder to model
- Involves dose-response relationship and modelling spread of outbreak, and social, financial and animal welfare costs.

Possible risk management questions

Annual probability of importing at least one infected animal

- Annual number of infected units imported
- Number of years between entries

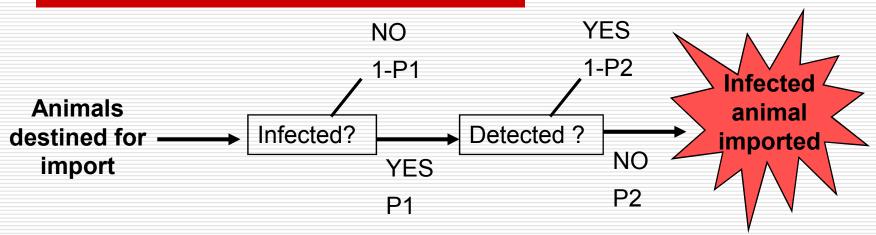
Steps to model your system

- □ 7 related steps:
 - Define the risk to be assessed
 - Outline the model framework
 - Identify the inputs and acquire the data
 - Adopt an iterative approach to the model
 - Undertake sensitivity testing
 - Ensure clarity by producing a transparent model description
 - Obtain feedback

Modelling risk pathways

- Scenario tree modelling graphical method to systematically model the process of disease through a chain of events (i.e. risk events; surveillance system components; release – exposure - consequence)
 - Traces the probabilities that a unit of interest (i.e. animal, herd, flock...) will yield either a positive or a negative outcome
 - Estimate the probability of a given decision
 - Inclusion of factors affecting Pr (infection) and Pr (detection)

Example: simple import RA (1)

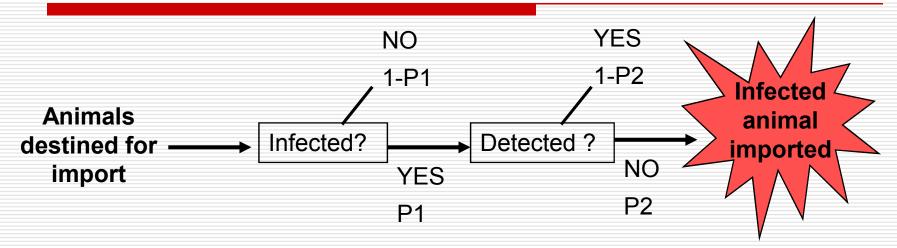


- □ P1 = probability that animals are infected
- 1-P1 = probability that animals are not infected

Infected animals may or may not be detected by a preexport test

- P2 = probability that infected animals are not detected
- □ 1-P2 = probability that infected animals are detected

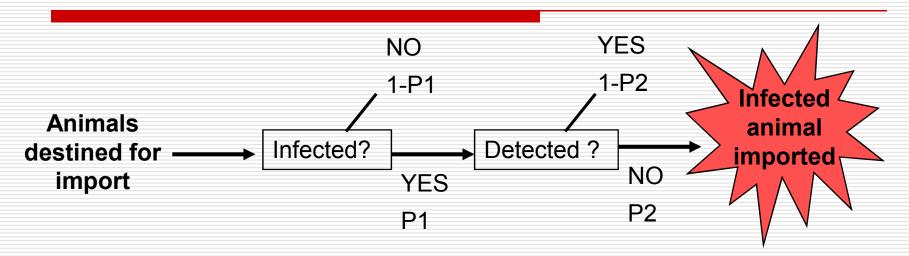
Example: simple import RA (2)



R = probability that an animal destined for import is infected and gets through testing

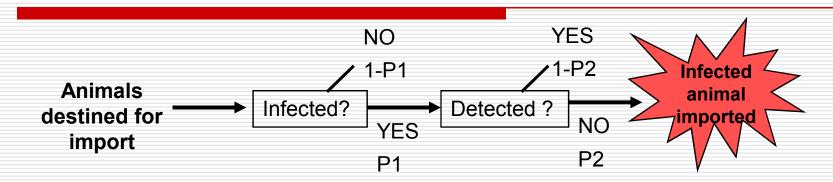
$$R = P1 \times P2$$

Exercise 1



□ An official from the regulatory authorities in the importing country needs to know the probability (R₁) that any animal entering his country is infected. **How to calculate this probability?**

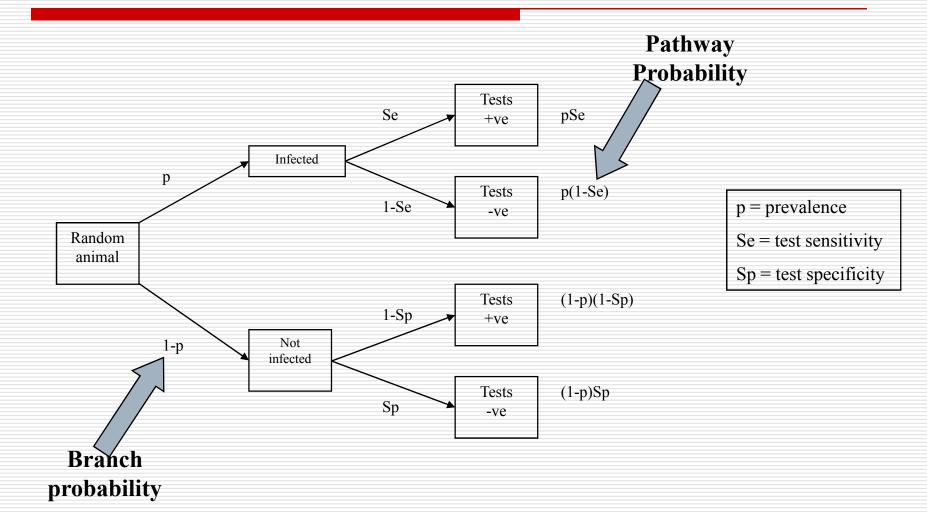
Exercise 1



- \square R₁= proportion of animals entering that are infected
 - Proportion of all animals destined for import that eventually enter the country:
 - ☐ Infected animals not detected by the test P1P2
 - □ Animals not infected (1-P1)
 - \square Animals entering = **P1P2** + (1-**P1**)

$$R_1 = \frac{P1P2}{P1P2 + (1-P1)}$$

Estimating "desired" risk



Steps to model your system

- □ 7 related steps:
 - Define the risk to be assessed
 - Outline the model framework
 - Identify the inputs and acquire the data
 - Adopt an iterative approach to the model
 - Undertake sensitivity testing
 - Ensure clarity by producing a transparent model description
 - Obtain feedback

Identify inputs and acquire data

- For the variables associated with each steps of your pathway you need to associate numbers (from collected data)
- Ex: probability that a random animal is infected with a specific pathogen = prevalence of disease (vary from region, breed, age...)
- You can break down the model pathway into smaller sections to show how inputs and outputs are related to each other

Simulation Inputs

- Definition: values from a risky variable which contribute to the models parameters
- Determine type of information this will determine different approches to develop a distribution based on the source information
 - Source
 - Empirical data
 - Expert opinion
 - Amount
 - Abundant
 - □ Few
 - Complete absence
 - Representativeness
- Decide if variability from uncertainty must be separated
- Document assumptions

Steps to model your system

- □ 7 related steps:
 - Define the risk to be assessed
 - Outline the model framework
 - Identify the inputs and acquire the data
 - Adopt an iterative approach to the model
 - Undertake sensitivity testing
 - Ensure clarity by producing a transparent model description
 - Obtain feedback

Iterative approach

- Your model framework will not be complete from the first outline.
- Many cycles of change, followed by discussions, may occur and in this way an iterative modelling approach is adopted

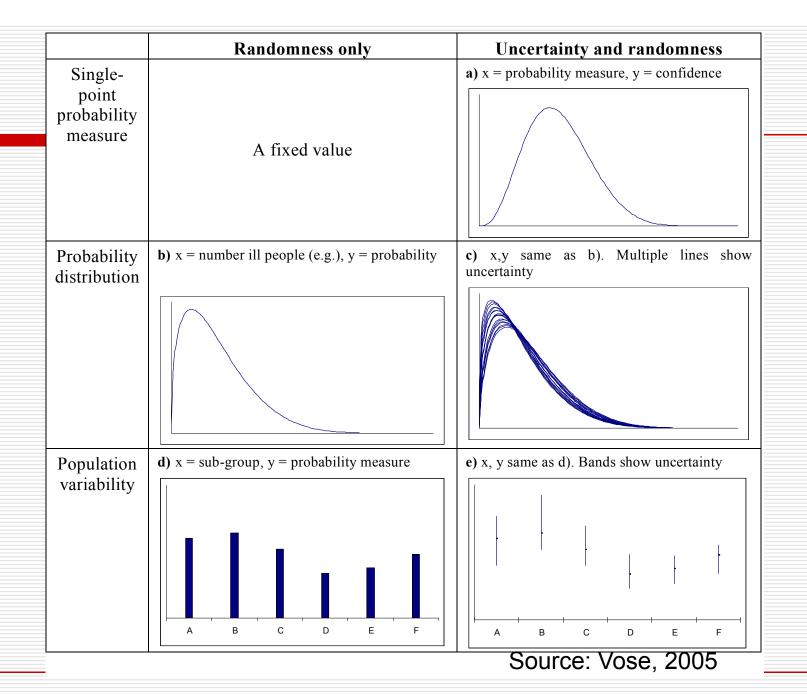
Determinist vs Probabilistic RA

- Deterministic
 - Single (point) values are assigned to each variables
 - Normally a best guess, based on literature or expert opinion.
 - Output is a single value
 - Sensitivity analysis by "what-if" scenario

- Probabilistic
 - Also called stochastic
 - Uses probability distributions to describe variables
 - Range of possible values that a variable can have (e.g. possible weights for cows; prevalence levels within a country)
 - Output is a distribution

Types of data distributions used in risk analysis

- Frequency distributions describing variability between individual data
- Probability distributions describing randomness
- ☐ *Uncertainty distributions* describing <u>assessors</u> uncertainty about some model parameter
 - When we are taking a random sample from a population a frequency distribution is used as a probability distribution
 - When uncertainty about the parameters of the frequency and probability distributions is present we use uncertainty distributions to describe that uncertainty.

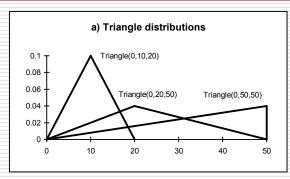


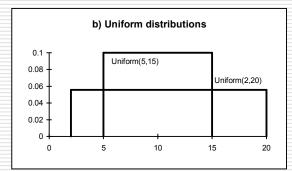
Probability distributions

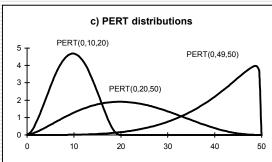
- Non-parametric
 distributions useful for
 empirical data and for
 modeling expert
 opinion:
 - Uniform
 - PERT
 - Triangle
 - Cumulative
 - General
 - Discrete

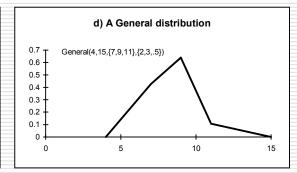
- Parametricdistributions also usedin literature:
 - Binomial
 - NegBin
 - Beta
 - Poisson
 - Gamma
 - Hypergeometric

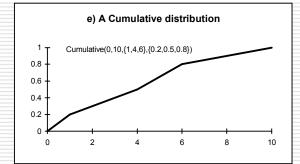
Probability distributions

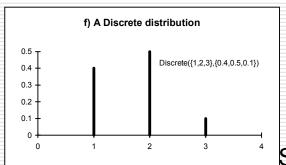












Source: Vose, 2005

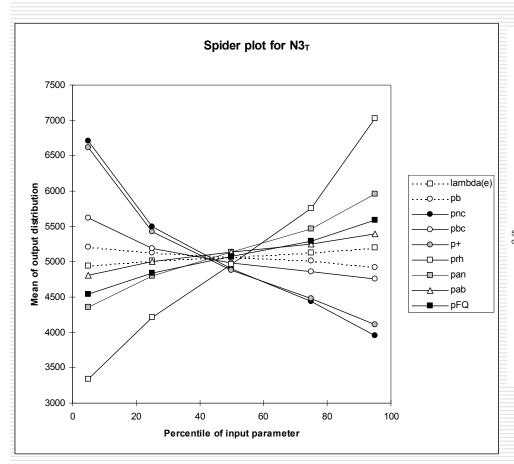
Variability and uncertainty of Simulation Inputs

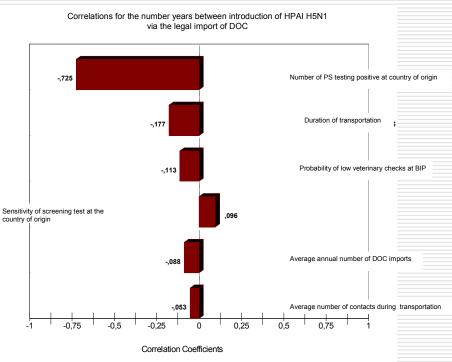
- "Variability is a phenomenon in the physical world to be measured, analysed and where appropriate explained. By contrast, uncertainty is an aspect of knowledge." (Sir David Cox)
- Variability is a function of the system and is the effect of chance; cannot be reduced <u>aleatory uncertainty</u>
- Uncertainty is a function of the assessor; can be reduced <u>epistemic uncertainty</u>
- Both represent the unpredictability of the system total uncertainty

Analysing the uncertainty

- We can use tornado plots (and other graphing techniques) to determine key uncertainties in our model – parameters that are influential.
- This helps focus on future data needs
- Doesn't tell us sensitivity to:
 - Model assumptions
 - Inappropriateness of data

Showing key uncertainties - Tornado charts Spider plot





Steps to model your system

- 7 related steps:
 - Define the risk to be assessed
 - Outline the model framework
 - Identify the inputs and acquire the data
 - Adopt an iterative approach to the model
 - Undertake sensitivity testing
 - Ensure clarity by producing a transparent model description
 - Obtain feedback

Sensitivity testing

- ☐ Your outputs will be constructed in terms of a number of interrelated variables.
- Ex: probability to import disease will depend on prevalence, Se and Sp or test, number of animals imported...
- You need to determine which variable has the most influence in the output = Sensitivity testing
- Highlight the key points in your analysis

Sensitivity analysis – key aspects

- A filtering technique
- Worst case scenario used to filter out whether worth worrying about
- ☐ Best case scenario as preliminary filter on risk management options
 - Risk manager can discount options where most optimistic benefits does not justify its cost.
 - Avoid exaggerating the extreme scenario beyond what is feasible.

☐ What-if scenarios

Where quantified uncertainty about parameter, a value is used that gives the required extreme, like 1 percentile, or 99 percentile.

Steps to model your system

- 7 related steps:
 - Define the risk to be assessed
 - Outline the model framework
 - Identify the inputs and acquire the data
 - Adopt an iterative approach to the model
 - Undertake sensitivity testing
 - Ensure clarity by producing a transparent model description
 - Obtain feedback

Ensure clarity / Feedback

- □ Transparency
- Referenced
- Circulation of the model for discussion and constructive criticism

Evaluation of risk model – key questions:

- 1. Does the risk assessment answer clearly to the defined risk question?
- 2. Is the response useful?
- 3. What are the data gaps?
- 4. What modeling techniques (statistics, distributions, methods of calculation) did they use?
- 5. Did they report key assumptions of the risk pathway parameters?
- **6.** Did they perform a sensitivity analysis? How?
- 7. How would you rate the analysis? Would you be confident to use it to make decisions? Were they hiding anything?







Quantitative Risk Assessment

Report available at: http://www.ifpri.org/sites/default/files/publications/hpaiwp21 ethiopia.pdf







Risk Analysis training course NIHE, Hanoi, Vietnam, 4-5 May 2016

