

Modelling for Informing Policy Development: Constraints and Challenges

Dirk U. Pfeiffer Professor of Veterinary Epidemiology

Outline

- context
- policy development
- modelling
- validity und uncertainty
- examples
- conclusions



Thinking about the Future

- need to be able to handle
 - o uncertainty
 - \circ complexity
 - o multiple plausible futures
- approaches
 - o narratives
 - o group processes
 - Delphi -> consensus-based response
 - Foresight -> focus on deliberations



Climate Change

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Scenarios Modelled by Intergovernmental Panel on Climate Change (IPCC)





Projected Temperature Change for 2090-99 (relative to 1980-1999)



Source: IPCC 2007

Model Predictions of Multi-Model Global Averages of Surface Warming (relative to 1980–1999) 90% confidence intervals



Costs of Animal Disease Outbreaks



Source: Bio Economic Research Associates paper on the *Economic Risks Associated with an Influenza Pandemic*, presented to the United States Senate Committee on Foreign Relations, November 9, 2005

Global Risk of AI Pandemic







From maps.maplecroft.net





Disease Control Policy

- disease control policy informed by understanding of cause-effect relationships

 experience and judgment
 scientific explanations / predictions
 - risks and uncertainty
- success of disease control programmes influenced by commitment of stakeholders towards programme goals

 perceptions in relation to validity and relevance of cause-effect relationships



Disease Control Policy cont.

- 'knowledge-based' society requires extensive consultation and debate about approaches, progress and outcomes
 - o effective communication becomes integral component of disease control programme
 - o participatory approaches



Policy Development, Implementation and Impact



Policy Development and Communication

- communicate risks and uncertainty in relation to likelihood of possible outcomes of policy implementation to stakeholders
 - generally recognised reduction of public trust in science
 - o influenced by risk perception, education and effectiveness of risk communication strategy









Asia's avian flu is spreading rapidly, increasing the risk that it will mutate into something far deadlier

With quakes and tsunamis, no one's got time for bird flu

The fight against the disease is unlikely to attract much attention from the resource-strapped Indonesian authorities until the flu becomes a mass killer









Possible Roles of Scientists in Policy Development

value consensus AND low uncertainty

 \circ no policy connection

pure scientist

o policy connection

- science arbiter
- no value consensus OR high uncertainty

 \circ need to reduce scope of choice

issue advocate

 \circ no need to reduce scope of choice

honest broker of policy alternatives



Modelling in Policy Development Process

- risk assessment
 - o quantify risks
 - o identify key risk factors (explanation)
- risk management
 - guidance with respect to policy choices (prediction)



Modelling

- reflection of our understanding of 'real world'
- explain or predict effects
- qualitative and quantitative
- focus on quantitative models
- needed where mental simulation not able to represent multiple causal links within system

 limit usually reached with 3 variables and 6 transitions from one state to another (Klein 1998)



Models as Representation of Reality



Quantitative Models

- used to generate information or knowledge from data
- "All models are wrong but some are useful" (Box 1979)
- "... the establishment that a model accurately represents the 'actual processes occurring in a real system' is not even a theoretical possibility." (Oreskes et al 1994)
- need to decide whether model good enough rather than true



data- and knowledge-driven models

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Data-driven Modelling

- uses statistical approaches to derive quantitative relationships from datasets
- usually used for explanation
 - to generate knowledge in relation to causeeffect relationships
- can be predictive



Knowledge-driven Modelling

 based on existing understanding of biological relationships within system

o or hypotheses in this respect

- strength and weakness -> ability to represent dynamics of complex biological systems
 - o particularly important for infectious disease
 - propagation of infection inherently time-dependent
 - number of new infections at particular time depends on number of infectious and susceptible individuals at preceding points in time
- possible to identify key factors within system



Knowledge-driven Modelling cont.

- may have 'emerging properties' resulting from interactions between multiple effects represented in model
- can be used to test impact of changes in system
- includes knowledge derived from data-driven models as well as expert opinion



Explanation and Prediction of Effects

- explanation
 - to understand biological mechanisms that lead to occurrence of outcome
 - o particular strength of data-driven models
- prediction / forecasting
 - knowledge about mechanism may also be of interest
 - \circ to understand what will happen in future
 - medium to long term
 - tactical usage during an outbreak
 - o with knowledge-driven models
 - possible to simulate different control scenarios



Modelling Outcomes

- effect estimates
 - \circ risks
 - \circ economic impact
 - \circ comparative assessments
- uncertainty



Uncertainty and Validity

 need to be considered when using outputs from modelling activities for informing policy development

o uncertainty (precision, random error)
o validity (bias, systematic error)

 influenced by assumptions, data quality and knowledge about biological system



Uncertainty

- uncertainty is the condition of all human life (after John Maynard Keynes)
- more than one outcome consistent with our expectations
 - expectations influenced by knowledge and lack thereof, societal values and interests
 - science and technology needed to clarify expectations and facilitate desired outcomes



Uncertainty cont.

 can increase with advancing knowledge due to complexity of systems

 therefore ignorance is bliss because it is accompanied by lack of uncertainty



Validity of Data-driven Models

- internal validity -> factors resulting in incorrect inferences in relation to presence of relationships within dataset
- external validity -> ability to extrapolate



Validity of Knowledge-driven Models

- complex to assess validity of outputs
 - typically based on quantitative relationships derived in very different studies, or from expert opinion
- usually assessed by comparing model behaviour with observed 'real world' system behaviour
- due to lack of suitable 'real world' data often necessary to consider plausibility of quantitative outcomes resulting from varying input parameters

Science, Models and Complexity





Problems with Models of Natural Processes

- errors in characterisation of processes modelled
 - \circ use of averages
 - \circ scaling up
 - o substituting lab measurements for nature
 - o substituting mathematics for nature

o assuming linear relationships

- omissions of important processes
- Iack of knowledge about initial conditions
- Influences from outside modelled system

Examples (from UK)

- BSE
- FMD
- Bovine TB



WHATEVER HAPPENED TO MAD COW DISEASE?

In 1996, BSE—or mad cow disease—spread food hysteria across the Continent, as 10 people died from a new form of Creutzfeldt-Jakob disease that was linked to contaminated beef. At one point, epidemiologists from Imperial College, London, predicted that vCJD could kill millions. So far, 139 have died, just a handful of them outside the U.K. The crisis is one of the most potent examples of how science can get risk wrong. A timeline:

1986

Bovine spongiform encephalopathy (BSE), an in-

fection that riddles the brain full of holes, is identified in cattle at a farm in West Sussex, England.

1987 The British government announces that large numbers of cattle are infected with BSE. The probable cause: feed that con-

TIME, JULY 28, 2003

tained meat or bone of sheep infected with scrapie.

1988 The government orders the slaughter of all BSEinfected cattle, above, and bans meat and bone meal in cattle feed.

1989 Britain bans human consumption of cattle brains,

spinal cords and other body parts that some scientists suspect could affect human health.

1990 Donald Acheson, the Chief Medical Officer, assures the public that beef is safe to eat. To prove the point, Agriculture Minister John Gummer publicly shares a hamburger with his daughter, left.



tion peaks: three cows in every 1,000 have the disease.

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UK FMD 2001



UK FMD 2001: Consequential Reality and Theory of Models







Bovine TB and Wild Badgers in UK

- risk of introduction of BTB to cattle herd
 - o cause-effect relationships
 - between-herd transmission vs. wildlife reservoirs
 - $_{\odot}$ 20 years political and scientific debate
 - in GB three government committee reports conclude
 - strong indications for significant role of badgers
 - badger population control methods considered inhumane



Bovine TB and Wild Badgers *cont*.

- UK badger culling trial to assess impact of badger TB on cattle TB, and effectiveness of badger population reduction (1998-2006 -> £40Mill)
 - 10 clusters with three treatment groups of 100km² each
 - \circ preliminary finding in mid 2003
 - reactive treatment increases cattle TB reactor rate by 27% (95% CI -2.4%-65%; Donnelly *et al* 2003)
 - interpreted as 'likely' cause-effect relationship -> UK government stopped reactive trial activities
 - was decision justified given scientific evidence?
 - significant sources of sampling error and bias



Conclusions

- predicting the future is not possible, but models can be a mechanism for making informed decisions
- modelling outcomes always associated with varying degrees of uncertainty and validity, which are often poorly communicated or understood



Conclusions cont.

- scientists need to be conscious about their role in policy development process

 issue advocate vs. honest broker
- ultimately, decision to use models for informing policy making will have to be based on opinion and judgment

