Introduction to Mathematical Models of Infectious Disease in Livestock Lecture 3: Some model examples Andrea Doeschl-Wilson

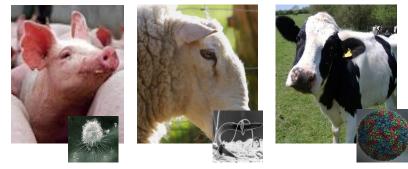


### **Purpose of this lecture**

- Provide a brief demonstration of existing mathematical models of infectious disease dynamics in livestock
  - Chosen subjectively
  - Not a comprehensive review!
- Get some insight into:
  - How biological concepts and mathematical approaches are incorporated in models
  - How the diverse models contribute to infectious disease research & policy



### **Overview**



#### Models of within host infection dynamics:

1. PRRS virus infection in pigs

Micro-parasite infection; based on immunological principles

2. Gastro-intestinal parasite infection in sheep

Macro-parasite infection; based on resource allocation theory

#### **Epidemiological models:**

3. Gastro-intestinal parasite infection in sheep

Example for using models to inform breeding strategies

4. Foot and mouth disease

Example for using models as decision making tool

# Mathematical models of infection dynamics

• Distinguish between two broad categories

(1) Within host models

Model interactions between pathogen and host response

(2) Epidemiological models

- Model disease spread between hosts / farms
- They require different knowledge, use different data & answer different kind of questions
  - But use similar mathematical tools
- Both models can be combined into an immuno-epidemiological model

#### Within-host infection models

- Relatively few models for animal diseases
- 2 contrasting examples to demonstrate diversity in approach and scope:
  - 1. Modelling PRRS virus infection in pigs
    - Mathematical representation of immune response to micro-parasite infection
  - 2. Modelling nematode infection in sheep
    - Mathematical representation of resource allocation theory for macro-parasitic infections





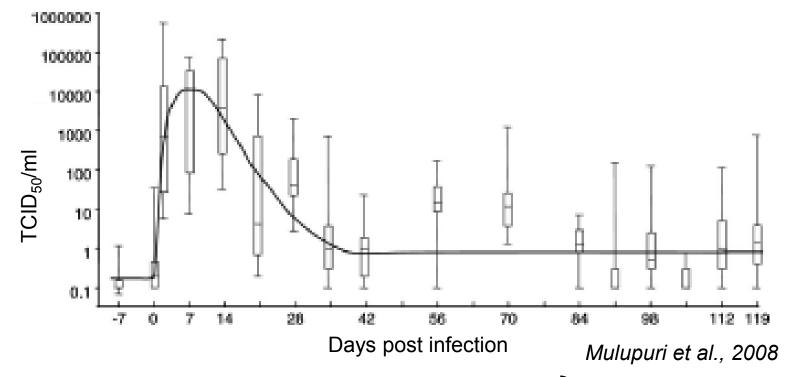
## Within-host model of PRRS virus infections in pigs

The Porcine Reproductive and Respiratory Syndrome (PRRS)

- Endemic viral disease worldwide
  - Infectious agent: RNA-virus PRRSV
  - causes dramatic losses to pig industry
- Symptoms:
  - Reproductive failure in mature pigs
  - Respiratory problems, fever, weight loss, death in growing pigs
- Target cells: subpopulation of macrophages in lung and other tissues
- Vast amount of research, but no efficient control measures



#### Atypical & highly diverse virus load profile

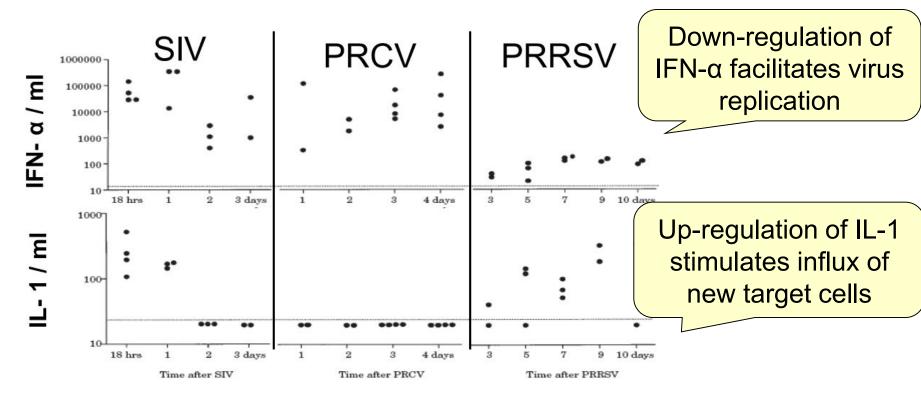


- peak levels at 7-14 days post infection
- acute phase lasts approx. 4 weeks
- long-term persistence at low levels
- large variation between hosts

Atypical profile for virus infections!

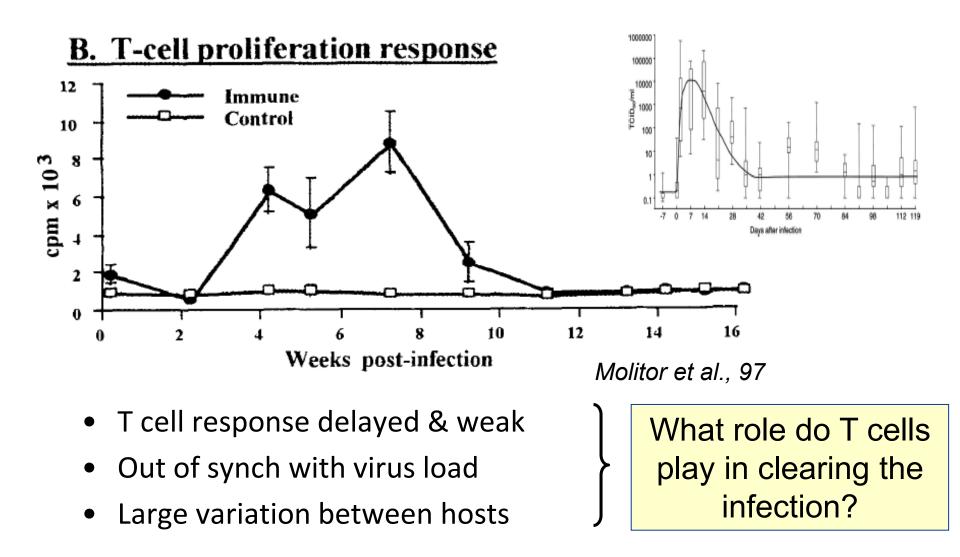
#### Weak innate immune response

- Lack of typical cytokine expressions
- Virus seems to manipulate innate response to its favour

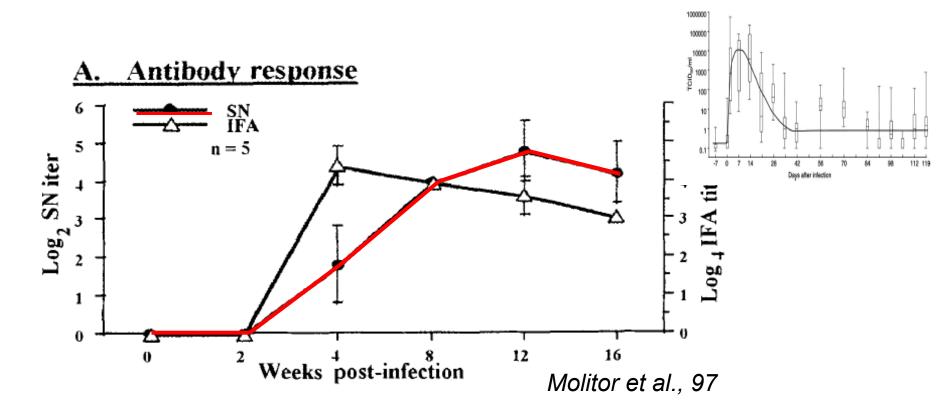


Van Reeth et al., 1999

#### Weak & delayed adaptive response



#### Atypical neutralizing antibody response



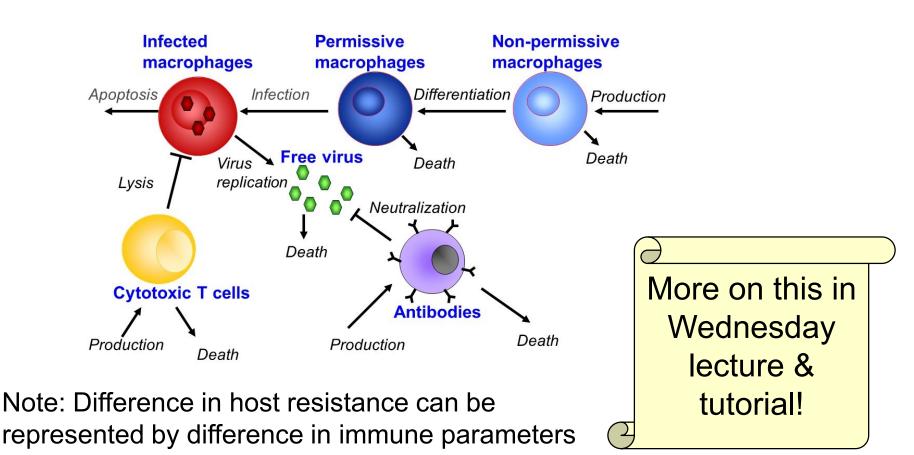
- Neutralizing antibodies appear late
- Antibody levels remain high
- Large variation between hosts

What role do neutral. antibodies play in clearing the virus?

# Questions addressed by the mathematical model

- What causes the observed diversity in PRRS viraemia profiles?
- Which biological processes are responsible for viral clearance?
  - What role does the adaptive immune response play?

# Modelling approach: A mechanistic model of virus and immune system dynamics



### Within-host model of gastro-intestinal parasite infections in sheep

# Gastro-intestinal parasite infections in sheep

- Endemic; major problem for sheep production: slow growth -> economic loss
- Conventional control strategies no longer work (anthelmintic resistance)
- Strong evidence for large influence of diet & host genetics on parasite burden
- Breeding for resistance a possible solution?
- But not clear if breeding for resistance would indeed lead to faster growth:
  - Estimates of genetic correlation between parasite burden & growth in field studies range between -0.8 to 0.4





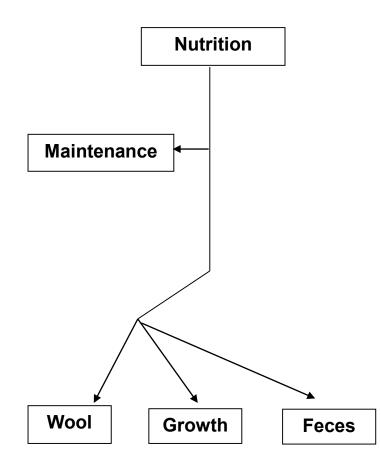
#### Model objectives:

• To investigate what causes the conflicting estimates from field studies

Hypotheses: Field estimates depend on

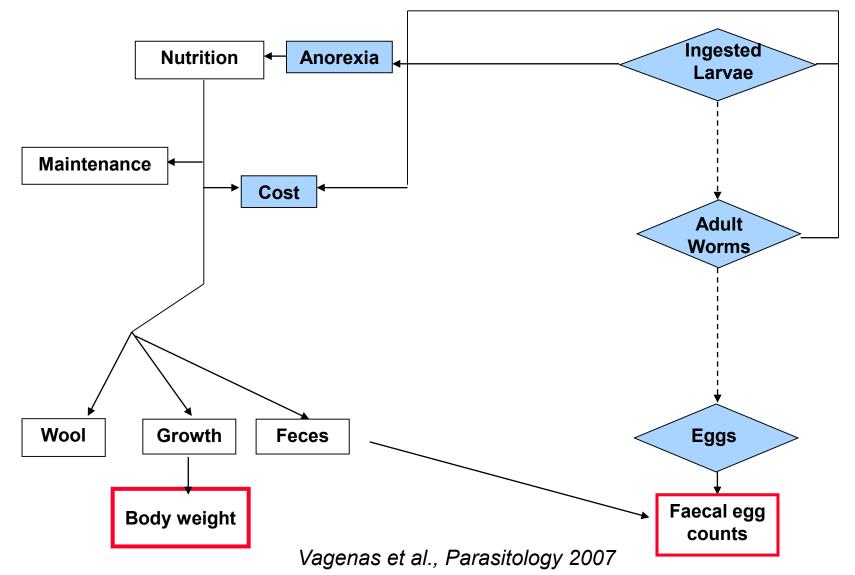
- Parasite challenge
- Breed
- Nutritional environment
- Timing of measurements
- Ultimately: to determine if / under what conditions selective breeding is a viable alternative to drug administration

## Base model: Nutrient allocation in healthy animal

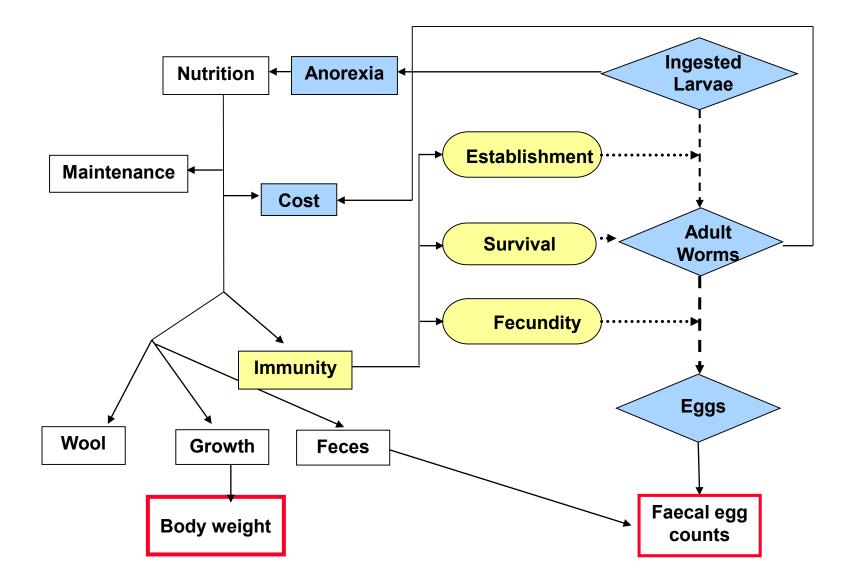


Vagenas et al., Parasitology 2007

## Influence of parasite on nutrient allocation and performance



#### Influence of host immune response



#### Mathematical representation

Hybrid mechanistic model of nutrient allocation:

- Mass balance equations for nutrient flow
- Parameterization based on empirical estimates for energy and protein costs associated with diverse biological processes
  - Assume production and health traits change proportional to nutrients allocated to these
- Within host-dynamics represented by large system of nonlinear equations (discrete time step, deterministic)
- Expand to a population model: Host genetic variation represented by normal distributions in key model parameters (e.g. rate of parasite establishment) (stochastic)

#### **Simulation experiment**

Simulate growth and immune response for a population of lambs

Immune challenge: Trickle infection with 3000 nematode larvae

Simulate 2 x 2 factorial experiment:

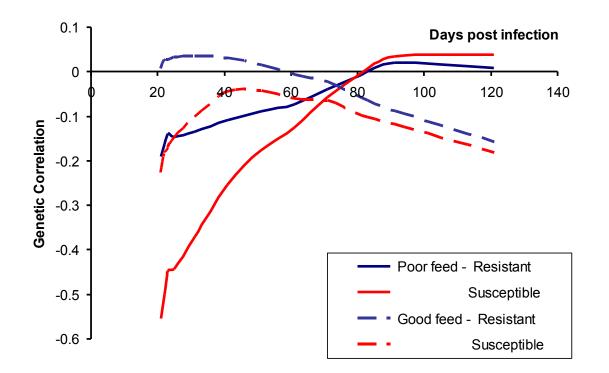
- 2 breeds for resistance (different population means for immune parameters): Susceptible / Resistant breed
- 2 diets (ad libitum access): Good / poor quality grass





# Impact of host genetics & diet on genetic correlations

Genetic correlations between body weight and faecal egg counts (log)



Doeschl-Wilson et al., GSE 2008

#### Summary of model findings

Model can explain ambiguous field study estimates:

- Traits and relationships change drastically over time
  - Repeated measurements required
- Strong breed x diet interactions
  - Trends are consistent with field observations
  - Model provides insight of underlying mechanisms
- Strong positive correlations between growth and health could only be obtained if growth and resistance mechanisms are controlled by common genes
- Genetic selection for disease resistance is most beneficial for susceptible breeds in poor quality diet

#### Extension to epidemiological models

- Both models produce predictions for pathogen burden (viral load or faecal egg counts) over time
- They thus lends themselves to expansion to immuno-epidemiological models





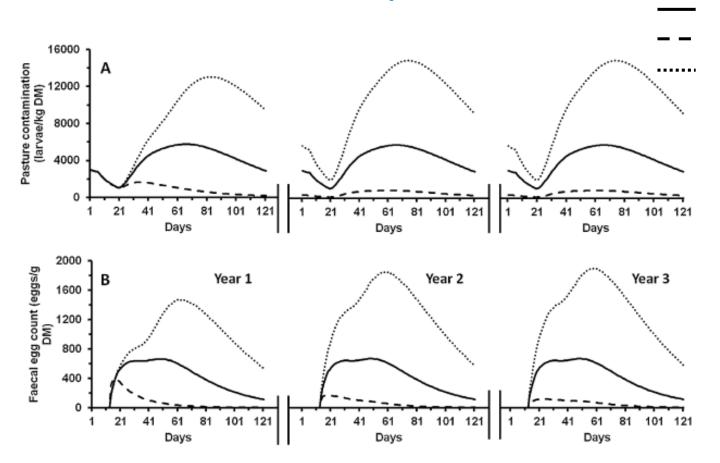
### Epidemiological model of gastro-intestinal parasite infections in sheep



Epidemiological model for gastro-

- Extension of the within-host GIP model by a 'pasture module'
  - Define pasture characteristics (size, volume of grass available, initial contamination)
  - Calculate pasture contamination over time: based on parasite eggs excreted (FEC) by infected lambs, natural parasite life-cycle on grass & removal of larvae by grazing
- Aim: to explore epidemiological consequences of resistance and grazing management

#### Model predictions



Model predicts that biggest benefit of anthelmintics treatment and grazing is to be expected for susceptible sheep → Promotes targeted selective treatment

Laurenson et al., 2012.

Mixed group

Resistant group

Susceptible group

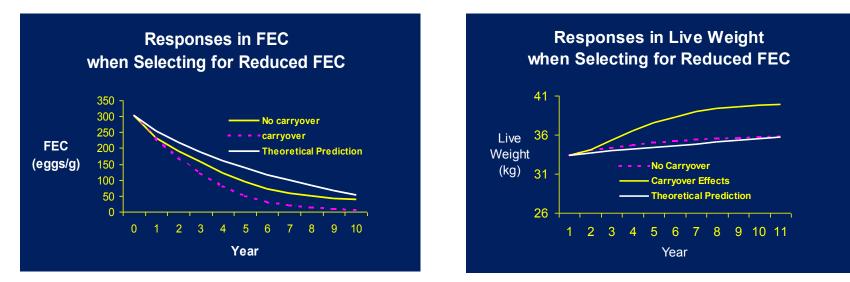
### Model application: Predicting response to selection

- Classical quantitative genetics formula for calculating response to selection ignores epidemics
- Positive epidemiological feedback:

Select for reduced faecal egg counts (FEC): => decrease pasture larval contamination => decrease larval challenge for others => **lower FEC in ALL sheep at same pasture** => greater productivity in ALL sheep

• Resistant sheep protect all sheep

### Model application: Predicting Response to Selection



- Theoretical prediction: based on quantitative genetics theory (ignores epidemics)
- No carryover: include epidemic, but ignore long-term benefits
- Carryover: include epidemic & long-term benefits

Bishop & Stear 1997

# Epidemiological models as decision making tools

#### Foot & Mouth Disease (FMD)



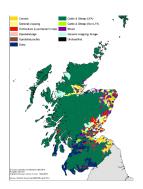
- 2001 FMD crisis in UK:
  - Led to the killing of over 10 million sheep & cattle
  - Cost US\$16bn
- Problem: Rapid transmission between wide range of livestock species
- Infection is rarely fatal, but causes severe reduction in growth rate and in milk production (dairy cattle)
- Strong economic impact: export ban of milk and meat, and movement restrictions in affected farms



#### Epidemiological models & Policy decisions

- Several control options available:
  - Culling, vaccination (with resulting export ban), prolonged movement & export restrictions …
- Main policy aim: achieve disease-free status asap
- Trade-off: minimize time vs minimize disturbance
  - Difficult to achieve optimal balance without a quantitative predictive framework
- "Scientific policy approach": Appointment of Prof. Roy Anderson, leading epidemiological modeler
- 3 epidemiological models for FMD were developed to inform policy decisions

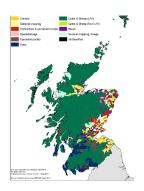
#### Model 1: InterSpread



- Large, complex, very flexible stochastic simulation model
  - Predicts spread of infection between farms influenced by many mechanisms
  - Most accurate representation of reality (amongst the 3 models)
- Accurate spatial representation
  - seeded with known location of all farms and their number / types of livestock
- Difficult to parameterize, very slow simulation times
  - Requires 'expert opinion' or guess for parameter values
  - Difficult to validate
  - Restricted exploration possible

Morris et al., 2001; Keeling 2005

### Model 2: Cambridge-Edinburgh model



- Stochastic simulation model
- Takes spatial structure of farms into account
  - same initialization as InterSpread
- Less explicit representation of temporal aspects
- More simple, transparent transmission mechanisms
  - Fewer parameters, easier parameterization
  - Still slow simulation times  $\rightarrow$  restricted exploration

Keeling et al., 2001; Keeling 2005

#### Model 3: Imperial model

#### • Deterministic model

- Only possible to predict average outcomes, not outcomes with low probabilities
- Simplistic representation of the spatial structure
  - Cannot distinguish between high / low risk areas
- Easy to parameterize, fast simulation times
  - Allowed for extensive exploration of a variety of scenarios (delay in reporting, diverse vaccination / culling strategies)

Whilesmith et al. 2003; Keeling 2005

#### Model predictions

- Models focused on different aspects, depending on the model type:
  - InterSpread: Identify high risk areas by comparing shortterm model predictions with observed nr of cases
  - Cambridge-Edinburgh: vaccination / culling strategies
  - Imperial: compare a wide range of control options
- Models overlapped in their main predictions:
  - Successful control of FMD requires rigorous application of culling (or vaccination) on a wide scale

#### **Stakeholder Reactions**

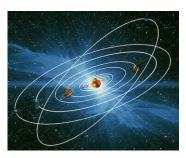
- Policy makers: application of stringent culling
  UK reverted to FMD free status within a few months
- Farmers & Veterinarians:





### Resolving the conflict (Keeling 2005): The issue of scale

- Optimum approach & control strategy depends on the scale:
- Individual farm level / local scale:
  - Veterinary judgement is most accurate / suitable
  - Less stringent control measure is optimal
- National level / global scale:
  - Mathematical model best suited to weigh pros & cons
  - More stringent control measure is optimal



#### Lessons learnt

- Epidemiological models can help decision making when faced with complex problems
- There is not one best model: Different modelling approaches can provide different insights
- Epidemic models can cause friction between modellers / veterinarians / farmers
- All epidemic models over-simplify and lack crucial aspects:
  - Failure to represent within-farm dynamics
  - Failure to capture individual, spatial or temporal heterogeneity
  - Failure to include economic aspects ...

#### Summary



- Mathematical models have proved useful for:
  - Providing explanations for conflicting experimental or field observations
  - Predicting outcome of infection / control strategies
- But the application of mathematical models to livestock diseases is still in its infancy
  - Lack of appropriate data for model parameterization& validation
  - Lack of base models to build upon & inference techniques

#### Further reading

- Doeschl-Wilson, A. B. "The role of mathematical models of host–pathogen interactions for livestock health and production–a review." *animal* 5.06 (2011): 895-910.
- Doeschl-Wilson, Andrea B., et al. "Unravelling the relationship between animal growth and immune response during micro-parasitic infections." *PLoS One* 4.10 (2009): e7508.
- Bishop, S. C., and M. J. Stear. "Modelling responses to selection for resistance to gastro-intestinal parasites in sheep." *Animal Science* 64.03 (1997): 469-478.
- Vagenas, D., S. C. Bishop, and I. Kyriazakis. "A model to account for the consequences of host nutrition on the outcome of gastrointestinal parasitism in sheep: logic and concepts." *Parasitology* 134.09 (2007): 1263-1277.
- Doeschl-Wilson, et al. "Exploring the assumptions underlying genetic variation in host nematode resistance." *Genet. Sel. Evol* 40 (2008): 241-264. Laurenson et al. "Exploration of the epidemiological consequences of resistance to gastro-intestinal parasitism and grazing management of sheep through a mathematical model." *Veterinary parasitology* 189.2 (2012): 238-249.
- Wilesmith, J. W., et al. "Spatio-temporal epidemiology of foot-and-mouth disease in two counties of Great Britain in 2001." *Preventive veterinary medicine* 61.3 (2003): 157-170.
- Keeling, Matt J. "Models of foot-and-mouth disease." *Proceedings of the Royal Society of London B: Biological Sciences* 272.1569 (2005): 1195-1202
- Keeling, Matt J., et al. "Dynamics of the 2001 UK foot and mouth epidemic: stochastic dispersal in a heterogeneous landscape." *Science* 294.5543 (2001): 813-817.