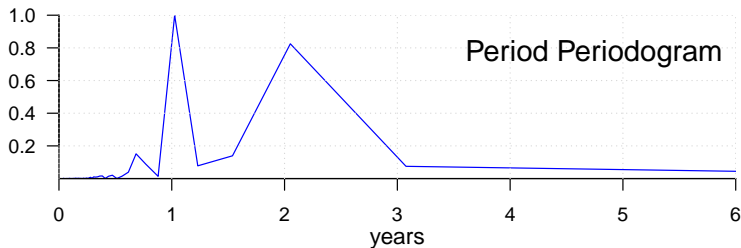
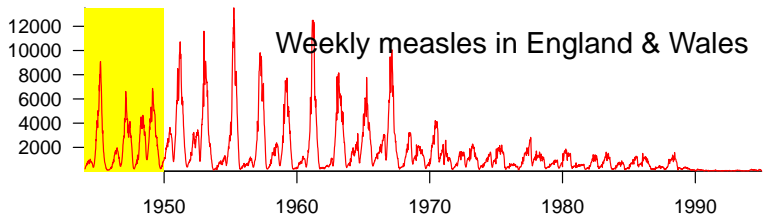


13 Mechanistic Modelling of Recurrent Epidemics

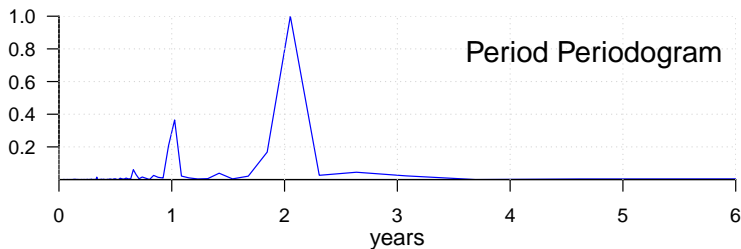
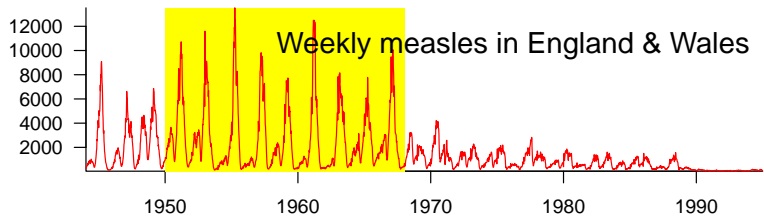
14 Mechanistic Modelling of Recurrent Epidemics II

# Mechanistic Modelling of Recurrent Epidemics

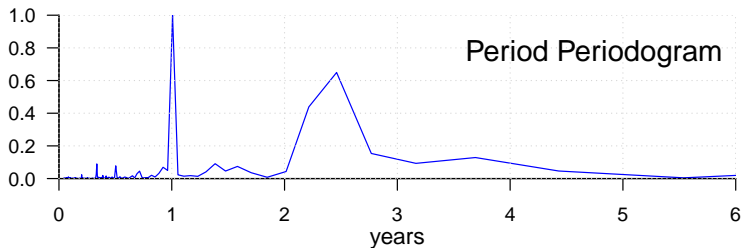
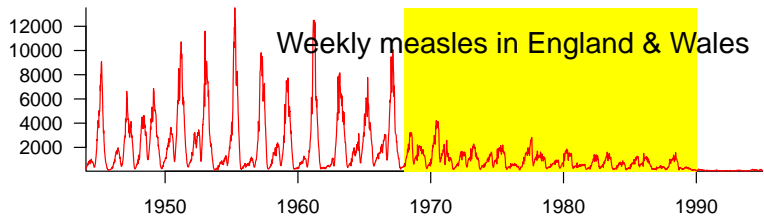
# What causes changes in frequency content over time?



# What causes changes in frequency content over time?

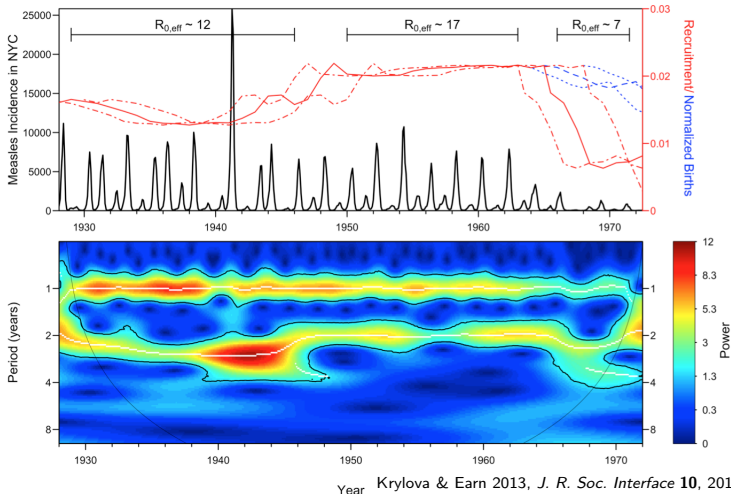


# What causes changes in frequency content over time?



# What causes changes in frequency content over time?

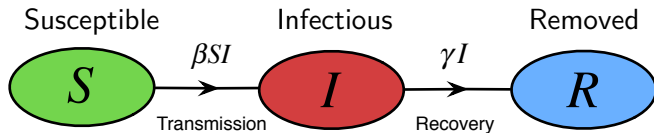
## Measles in New York City



# Mechanistic Epidemic Modelling: Principles

- Consider the biological mechanisms involved in disease transmission and spread
- Model mechanisms and infer their effects
- Start as simple as possible!
- Rule out simple models by comparing results with observed time series of incidence or mortality
- Add complexity one step at a time, so key mechanisms can be identified
- Ideally converge on simplest possible model that can explain observed patterns

# The SIR model: Flow Chart and Parameters



$$\frac{dS}{dt} = -\beta SI$$

$$\frac{dI}{dt} = \beta SI - \gamma I$$

$$\frac{dR}{dt} = \gamma I$$

## ■ Parameters:

- Transmission rate  $\beta$
- Recovery rate  $\gamma$   
(or Removal rate)



# The SIR model: Derived Parameters



$$\frac{dS}{dt} = -\beta SI$$

$$\frac{dI}{dt} = \beta SI - \gamma I$$

$$\frac{dR}{dt} = \gamma I$$

## ■ Derived Parameters:

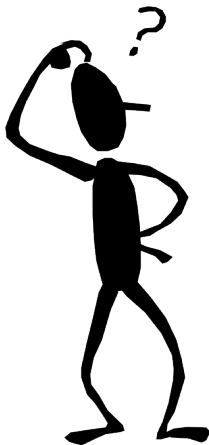
- Initial growth rate  $\beta - \gamma$
- Mean infectious period  $\frac{1}{\gamma}$
- Basic Reproduction Number

$$\mathcal{R}_0 = \frac{\beta}{\gamma}$$

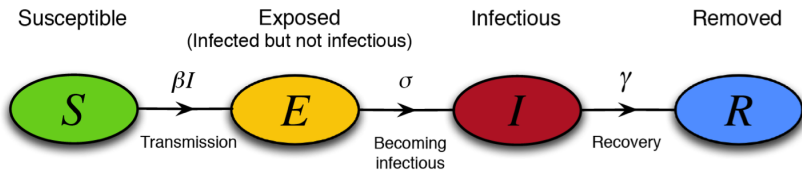
# Basic SIR Model: Important Results

- Epidemic occurs if and only if  $\mathcal{R}_0 > 1$
- Exact solution for phase portrait
- Single epidemic, then disease disappears
- Exact formula for final size as a function of  $\mathcal{R}_0$
  
- Cannot explain diseases that persist
- Cannot explain recurrent cycles of epidemics

# What are we missing?



# SEIR Model: flow chart

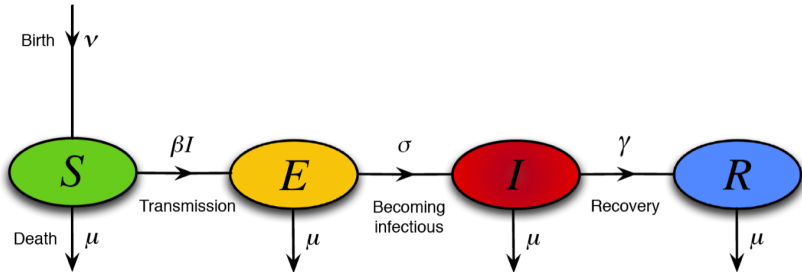


- Introduces only one new parameter ( $\sigma$ )
- Mean latent period ( $1/\sigma$ ) can often be estimated
- But... effect of inclusion of exposed class usually small

What are we **still** missing?



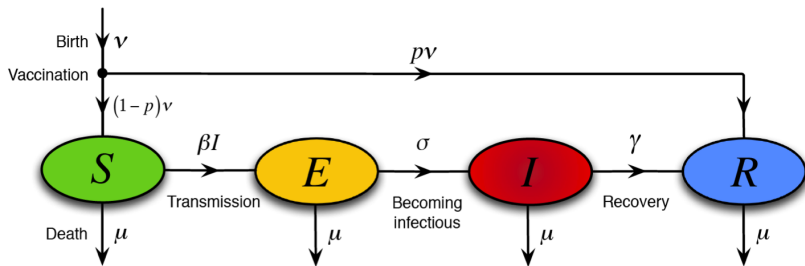
# SEIR Model with vital dynamics: flow chart



New Parameters:

- Birth rate ( $\nu$  for natality)
- Death rate ( $\mu$  for mortality)
- Mean latent period ( $1/\sigma$ )

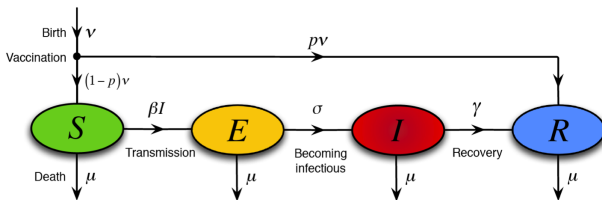
# SEIR with vital dynamics and vaccination: flow chart



New Parameters:

- Birth rate ( $\nu$  for natality)
- Death rate ( $\mu$  for mortality)
- Mean latent period ( $1/\sigma$ )
- Proportion vaccinated ( $p$ )

## SEIR with vital dynamics and vaccination: Equations



$$\frac{dS}{dt} = \nu(1 - p) - \beta SI - \mu S$$

$$\frac{dE}{dt} = \beta SI - \sigma E - \mu E$$

$$\frac{dI}{dt} = \sigma E - \gamma I - \mu I$$

$$\frac{dR}{dt} = \nu p + \gamma I - \mu R$$

- Birth rate ( $\nu$  for natality)
- Death rate ( $\mu$  for mortality)
- Proportion vaccinated ( $p$ )
- Transmission rate ( $\beta$ )
- Mean latent period ( $1/\sigma$ )
- Mean infectious period ( $1/\gamma$ )





Mathematics  
and Statistics

$$\int_M d\omega = \int_{\partial M} \omega$$

## Mathematics 4MB3/6MB3 Mathematical Biology

Instructor: David Earn

Lecture 13

Mechanistic Modelling of Recurrent Epidemics

Monday 5 February 2018

# Announcements

- **Assignment 2:** Due TODAY!  
Do the [Group contribution survey](#) TODAY.
- **Assignment 3:** Will be posted tonight or tomorrow.  
Due Wednesday 28 February 2018, 11:30am.
- **Midterm test:**
  - *Date:* Thursday 8 March 2018
  - *Time:* 7:00pm to 9:00pm
  - *Location:* BSB-B154

# SEIR with vital dynamics and vaccination: Analysis

- $\mathcal{R}_0$  ?
  - Biological derivation: (assuming  $\nu = \mu$  and  $p = 0$ )
$$\mathcal{R}_0 = \beta \times \frac{\sigma}{\sigma + \mu} \times \frac{1}{\gamma + \mu} \simeq \frac{\beta}{\gamma} \quad \because \frac{1}{\mu} \gg \max\left(\frac{1}{\sigma}, \frac{1}{\gamma}\right)$$
  - Mathematical derivation:  
 $\mathcal{R}_0 = 1$  is stability boundary
- Final size ? Not well defined (because of continuous source of new susceptibles).
- Equilibria ?
  - Disease Free Equilibrium (DFE)
  - Endemic Equilibrium (EE)
  - That's all folks.
- Periodic solutions ? No.
- What else ? Chaos?

# SEIR with vital dynamics and vaccination: Results

- $\exists$  Endemic Equilibrium  $\iff \mathcal{R}_0(1 - p) > 1$ 
  - EE is GAS in this case.
  - DFE is GAS otherwise.
- Eradication  $\iff p > 1 - \frac{1}{\mathcal{R}_0}$  (herd immunity)
  - Smallpox:  $\mathcal{R}_0 \sim 4 \implies p_{\text{crit}} \sim 75\%$
  - Measles:  $\mathcal{R}_0 \sim 20 \implies p_{\text{crit}} \sim 95\%$
- Explains persistence of diseases (via births)
- No periodic solutions  $\stackrel{?}{\implies}$  no recurrent epidemics
- GAS equilibrium  $\implies$  no periodic solutions and no chaos
- Equilibrium approached by *damped oscillations*  
 $\implies$  recurrent epidemics
- But observed epidemic patterns show *undamped* oscillations...

# What are we **STILL** missing?



# Demographic Stochasticity

- Differential equations describe the expected behaviour in the limit that the population size goes to infinity
- How do dynamics differ in finite populations?
- Re-cast the **SEIR model** as a stochastic process (**Continuous time Markov process**)
- Proving anything about stochastic epidemic models is difficult, but we can easily simulate them and learn a lot
- Standard algorithm for creating realizations of a stochastic epidemic model attributed to Daniel T. Gillespie

Gillespie 1976, *J. Comp. Phys.* 22, 403–434

- Rather than rates of change of compartment sizes consider event rates for transitions between disease states
- Finite number of individuals
- Assume event rates depend only on current state of population

# Gillespie Algorithm

- Let  $a_1, a_2, \dots$ , be the rates at which the various processes occur, *e.g.*,
  - $a_1 =$  birth rate,
  - $a_2 =$  rate of going from susceptible to exposed,
  - $a_3 =$  the rate of going from infectious to removed (recovering),
  - *etc.*
- Let  $a_0$  be the overall event rate, *i.e.*,  $a_0 = \sum_i a_i$  (so average time between events  $= 1/a_0$ ).
- Assume time spent in any state is exponentially distributed (transitions between states are “Poisson processes”)
- $\therefore$  Probability next event occurs in  $(t, t + dt)$  is  $a_0 e^{-a_0 t} dt$
- Let  $u = 1 - e^{-a_0 t}$ . Then  $u \in [0, 1]$  and  $du = a_0 e^{-a_0 t} dt \implies u$  is uniformly distributed in  $[0, 1]$ .
- $\therefore$  Get time  $t$  to next event by sampling  $u$  from uniform distribution in  $[0, 1]$  and setting  $t = \frac{1}{a_0} \ln \frac{1}{1-u}$ .

# Gillespie Algorithm continued

- We now know the time  $t$  of the next event, but we must still determine what type of event occurs at time  $t$ .
- Probability of event of type  $i$  is  $\frac{a_i}{a_0}$
- $\therefore$  Can easily determine type of event by sampling a point from a uniform distribution on  $[0, a_0]$ :
  - Event is type  $i$  if the uniform deviate lies in the  $i$ th interval in the following list:

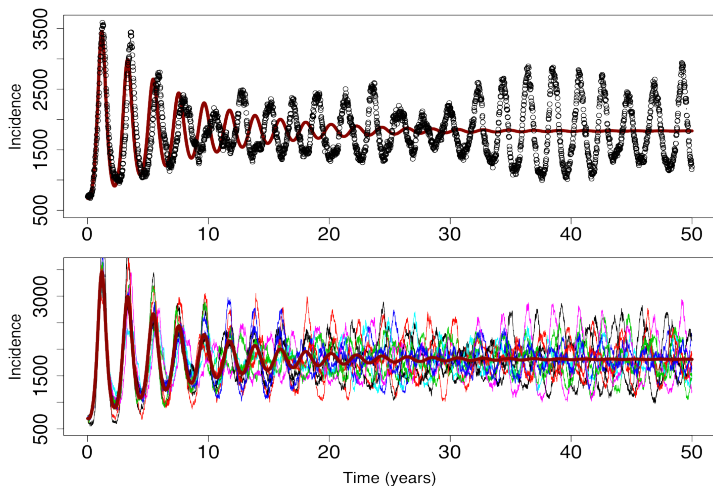
$$[0, a_1), [a_1, a_1 + a_2), \dots, [a_1 + \dots + a_{i-1}, a_1 + \dots + a_i), \dots$$

- How do realizations of this process differ from the solution of the deterministic (differential equation) model?



# Gillespie Simulations: Results for Measles Parameters

$\mathcal{R}_0 = 17$ ,  $T_{\text{lat}} = 8$  days,  $T_{\text{inf}} = 5$  days,  $\nu = \mu = 0.02/\text{year}$ ,  $N = 5,000,000$



Earn 2009, *IAS/Park City Mathematics Series* **14**, 151–186

# Effects of Demographic Stochasticity

- Sustains transient behaviour (oscillations do not damp out) (Bartlett 1950's)
- Explains undamped oscillations at a single period
- But, unable to explain changes in interepidemic period, or irregularity

# What are we **STILL** missing?





Mathematics  
and Statistics

$$\int_M d\omega = \int_{\partial M} \omega$$

## Mathematics 4MB3/6MB3 Mathematical Biology

Instructor: David Earn

Lecture 14

Mechanistic Modelling of Recurrent Epidemics II

Wednesday 7 February 2018

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# Contact rates are higher during school terms!

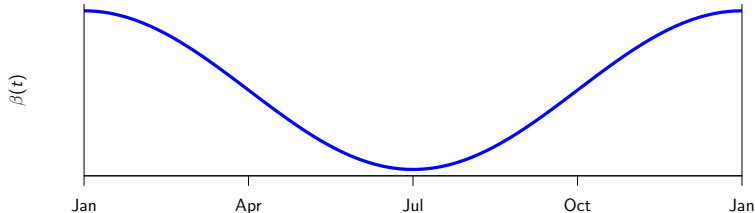


# Sinusoidal SEIR Model

- Transmission rate  $\beta$  is not constant:  
high during school terms, low in summer
- For simplicity, model as a sine wave:

$$\beta(t) = \langle \beta \rangle (1 + \alpha \cos 2\pi t)$$

- $\langle \beta \rangle$  = mean transmission rate
- $\alpha$  = amplitude of seasonal variation in contact rate



# Is this change significant?

- We now have a forced nonlinear system
- Forcing frequency can resonate with the natural timescales of the disease (e.g., damping period)
- Very rich dynamical system...  
(analogy: forced pendulum)



# Sinusoidal SEIR Model: Numerical Results

- Stable cycles of various lengths (annual, biennial, 3-year, . . . )
- Multiple co-existing stable cycles
- Chaotic dynamics
- Lots of work on this model in 1980s and 1990s

Smith HL, 1983, *J. Math. Biol.* **17**, 163–177

Schwartz IB, Smith HL, 1983, *J. Math. Biol.* **18**, 233–253

Aron JL, Schwartz IB, 1984, *J. theor. Biol.* **110**, 665–679

Olsen LF, Schaffer WM, 1990, *Science* **249**, 499–504

. . .

# Sinusoidal SEIR Model: Rigorous Results

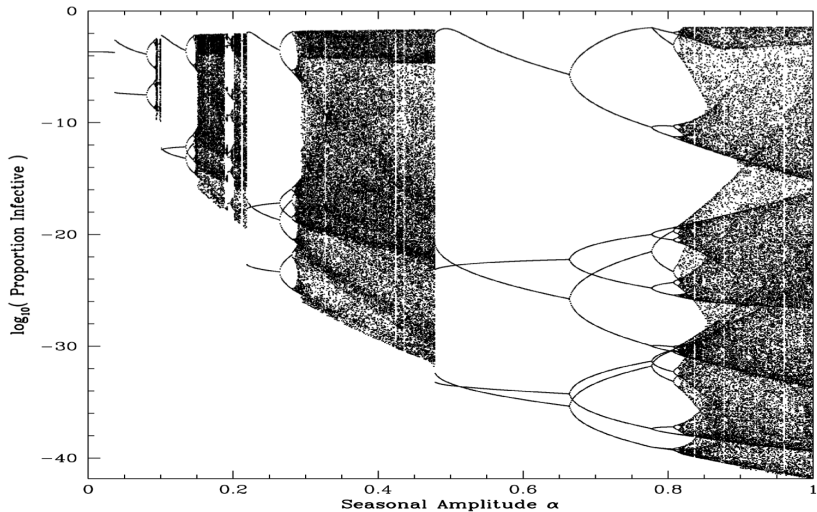
- There exist parameter values such that infinitely many stable cycles co-exist

Schwartz IB, Smith HL, 1983, *J. Math. Biol.* **18**, 233–253

- There exist chaotic repellers (in a modified SEIR model)

Glendinning P, Perry LP, 1997, *J. Math. Biol.* **35**, 359–373

# Measles Bifurcation Diagram (Sinusoidal SEIR model)



Earn (2009) *IAS/Park City Mathematics Series* 14, 151–186

# Does Sinusoidal SEIR Model Explain Measles Dynamics?

SEIR model with sinusoidal forcing:

- Produces recurrent undamped epidemics of all frequencies observed in measles time series.
- Produces chaos, which can explain irregular behaviour and transitions from one type of cycle to another
  - If correct, this implies these transitions are *unpredictable*.
- BUT... the model also predicts **rapid extinction** of the virus (not persistence).

What are we STILL missing?



# Is Age Structure Important?

- Real system is not homogeneously mixed
- Contact structure is age-dependent
- Schenzle (1984) argued for creating a Realistically Age-Structured (RAS) SEIR model
  - 21 age classes (0–1, 1–2, ..., 19–20, > 20)
  - SEIR compartments for each class
  - Different contact rates between all these age classes

$$\beta(t) \quad \longrightarrow \quad \begin{pmatrix} \beta_{1,1}(t) & \beta_{1,2}(t) & \cdots & \beta_{1,21}(t) \\ \beta_{2,1}(t) & \beta_{2,2}(t) & \cdots & \beta_{2,21}(t) \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{21,1}(t) & \beta_{21,2}(t) & \cdots & \beta_{21,21}(t) \end{pmatrix}$$

Schenzle D (1984) *IMA Journal of Mathematics Applied in Medicine and Biology* 1, 169–191

- Lots of work on RAS models since Schenzle (1984)

# RAS SEIR model: Results for Measles

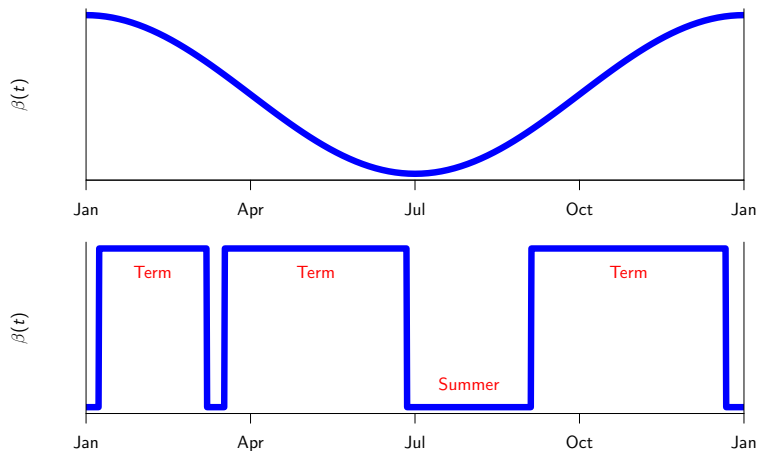
- Persistent biennial cycle
- Matches biennial cycle in data extremely well
- And we need only 84 ODEs and fewer than 500 new parameters!
- Can get an even better fit by adding spatial structure with 6000 ODEs and only 1500 new parameters!
- **Woohoo!** Time to celebrate.
- hmmm... maybe not...
- In fact, age structure is a **RED HERRING!**
- Critical ingredient of **RAS model** is...

Contact rates are higher during school terms!

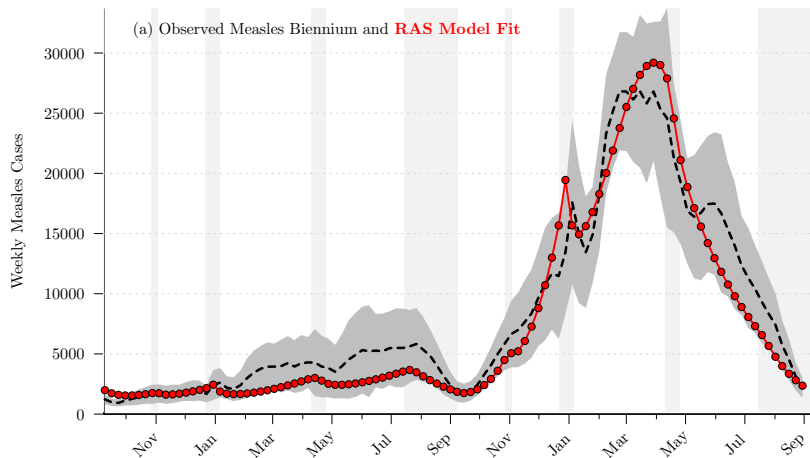




# Sinusoidal forcing vs Term-time forcing

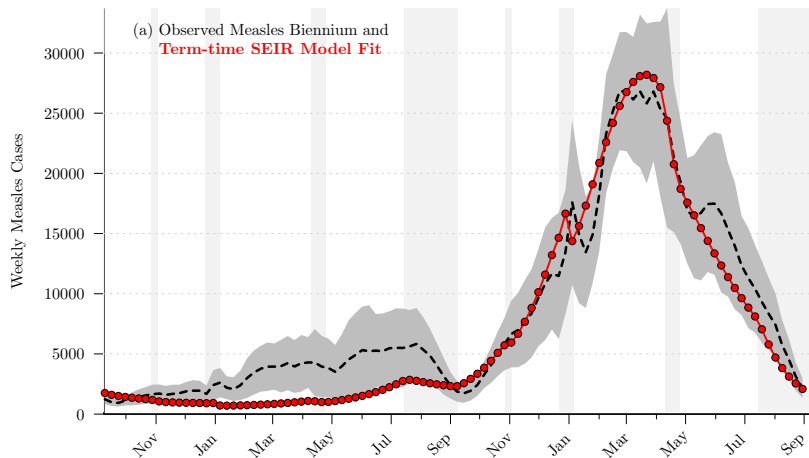


# RAS model fit to measles in England and Wales



He & Earn (2016) *J. R. Soc. Interface* **13**, 20160156

# Term-time SEIR model fit to measles in England and Wales



He & Earn (2016) *J. R. Soc. Interface* **13**, 20160156