

# Integro-differential–Equation Models for Infectious Disease

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# Introduction

- ▶ Many ecological and epidemiological systems are better modeled in continuous time instead of discrete time.
- ▶ There are many continuous-time frameworks: reaction–diffusion integral equations, integro-differential equations.
- ▶ **Diffusion** is a **local** operator, *i.e.*, individuals can only influence their immediate neighbors and may also present some difficulty in matching with experimental data
- ▶ Integrals have been used instead to model **non-local** spatial processes giving rise to **integral** and **integro-differential equations**.
- ▶ These models have all the advantages of the discrete-time **integrodifference equations**: more flexibility in using dispersal data, predicting faster wave speeds, and the possibility for accelerating waves.

## Spatial epidemic models

- ▶ Reaction–diffusion: Noble (1974); Bailey (1975); Murray, Staley, & Brown (1986)

$$\begin{aligned}\partial_t S &= -\beta(I + \alpha \nabla^2 I)S + D_S \nabla^2 S, \\ \partial_t I &= \beta(I + \alpha \nabla^2 I)S - \gamma I + D_I \nabla^2 I, \\ \partial_t R &= \gamma I + D_R \nabla^2 R\end{aligned}$$

- ▶ Integral equations: Thieme (1977, 1979); Diekmann (1978, 1979); Rass & Radcliffe (1984, 1986, 2003)

$$\begin{aligned}\dot{S} &= -\lambda S, \\ i(t, 0, x) &= \lambda S, \\ i(t, \tau, x) &= i(t - \tau, 0, x), \\ \lambda(t, x) &= \int_0^\infty \int_\Omega i(t, \tau, \xi) A(\tau, x, \xi) \, d\xi d\tau\end{aligned}$$

- ▶ Integro-differential equations: Kendall (1957, 1965) and Mollison (1972); Medlock & Kot (2003)
- ▶ Integrodifference equations: Allen & Ernest (2002)

$$S_{t+1} = \int_{\Omega} [S_t(y) + (\mu + \gamma)I_t(y) - \beta I_t(y)S_t(y)] k(x - y)dy,$$

$$I_{t+1} = \int_{\Omega} [(1 - \mu - \gamma)I_t(y) + \beta I_t(y)S_t(y)] k(x - y)dy$$

# Outline

- ▶ Two integro-differential-equation models from epidemiology
  - ▶ The SI model: Brief review
  - ▶ Distributed Contacts model
  - ▶ Distributed Infectives model
  - ▶ The movement process: Linear integro-differential equations
  - ▶ Wave speed
  - ▶ Wave shape
- ▶ Conclusions

## SI model: Brief review

- ▶ We divide the population into two groups:
  - ▶ Susceptible individuals,  $S(t)$
  - ▶ Infective individuals,  $I(t)$



- ▶ Assumptions
  - ▶ Population size is large and constant,  $S(t) + I(t) = K$
  - ▶ No birth, death, immigration, or emigration
  - ▶ No recovery or latency
  - ▶ Homogeneous mixing
  - ▶ Infection rate is proportional to the number of infectives, *i.e.*,  
 $\lambda = \beta I$



- ▶ A pair of ordinary differential equations describes this model (Ross, 1911):

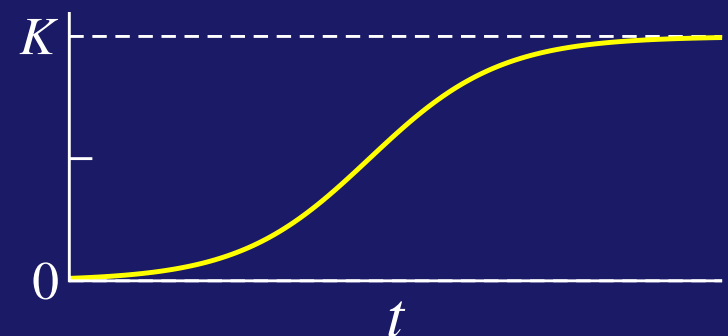
$$\begin{aligned} S'(t) &= -\beta I(t)S(t), \\ I'(t) &= \beta I(t)S(t). \end{aligned}$$

- ▶ With constant population size,  $K = S(t) + I(t)$ , this reduces to

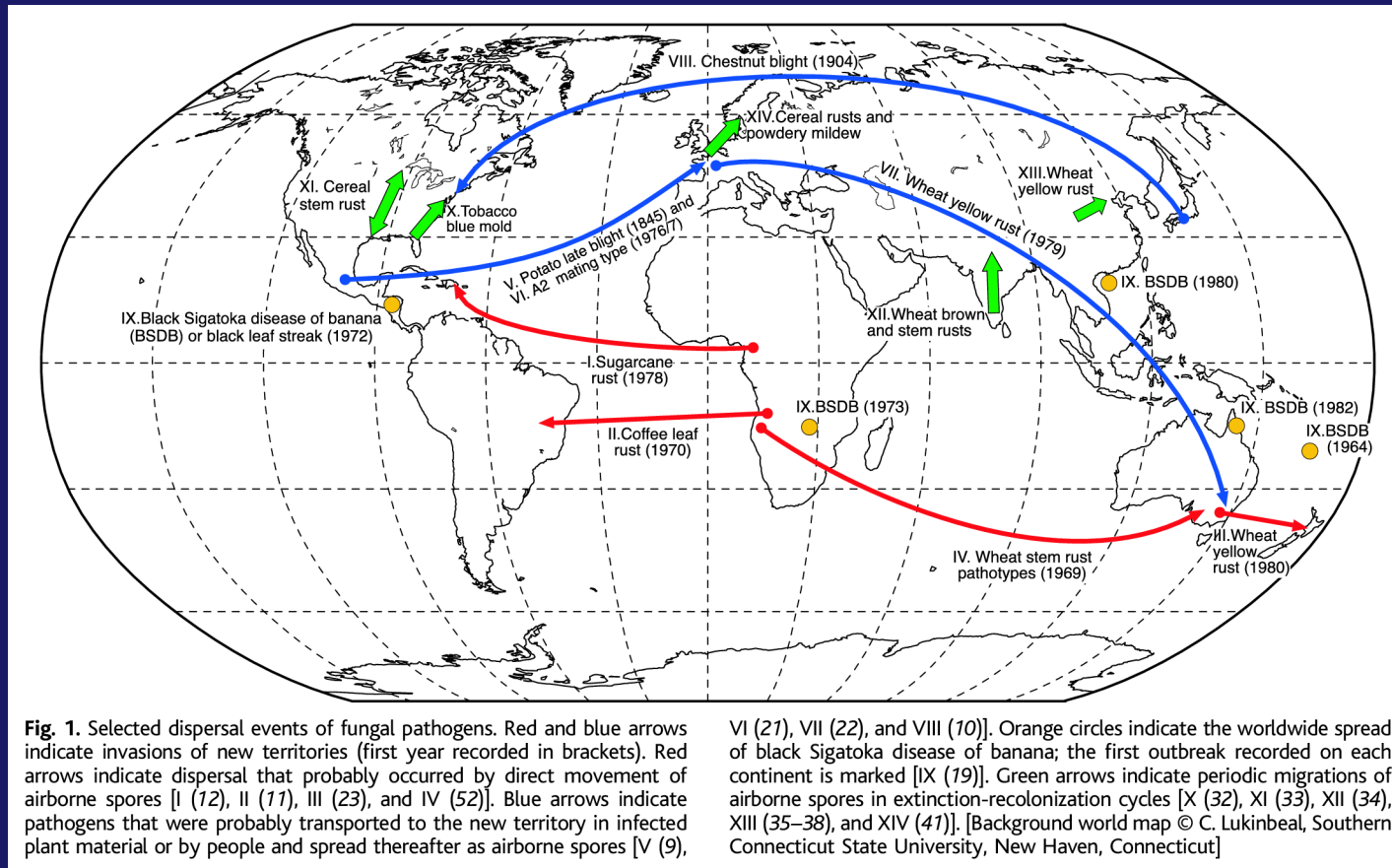
$$I'(t) = \beta I(t)(K - I(t)).$$

- ▶ The solution is the logistic curve,

$$I(t) = \frac{I(0)K}{I(0) + (K - I(0))e^{-\beta K t}}.$$



# Spatial spread: Two different non-local mechanisms



From Brown, J.K.M. & Hovmøller, M.S., *Science*, **297** 537.



From <http://www.hort.uconn.edu/ipm/veg/htms/potblpic.htm>.



## Distributed Contacts (DC)

- ▶ The governing equations (Kendall, 1957, 1965; Mollison, 1972):

$$\partial_t S = -\beta \left( \int_{\Omega} k(x-y) I(y, t) \, dy \right) S,$$

$$\partial_t I = \beta \left( \int_{\Omega} k(x-y) I(y, t) \, dy \right) S.$$

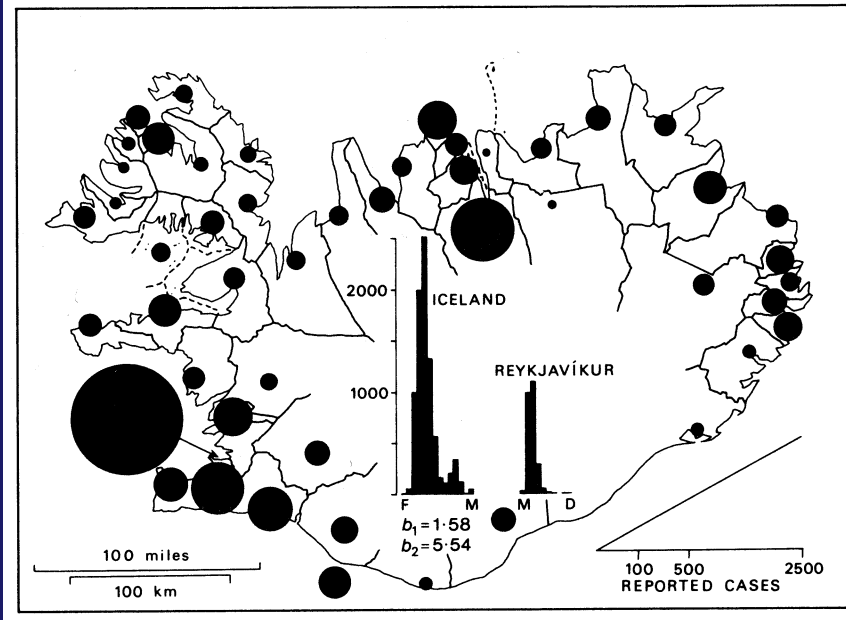
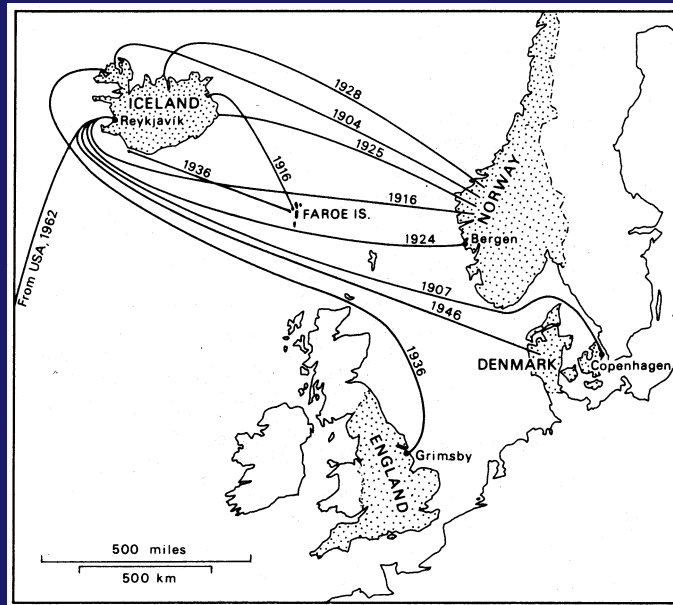
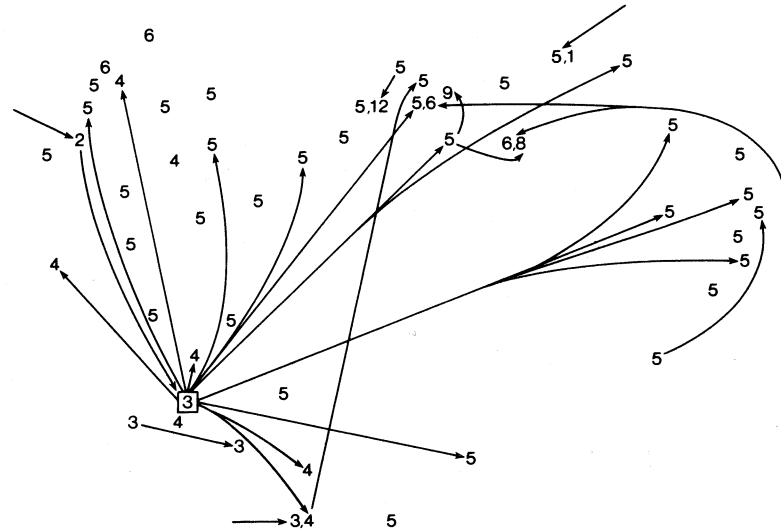
- ▶  $\beta$  is the infection rate
- ▶  $k(u)$ , the kernel, is the contact distribution
- ▶ Assumptions:  $K = S + I$  is constant and  $\Omega = \mathbb{R}$ , give

$$\partial_t I = \beta \left( \int_{\mathbb{R}} k(x-y) I(y, t) \, dy \right) (K - I).$$



From <http://www.phls.wales.nhs.uk/measlg.htm>.

FEBRUARY 1936 - MARCH 1937



From Cliff, A.D., Haggett, P., Ord, J.K. & Versey, G.R., *Spatial Diffusion*, 1981.

## Distributed Infectives (DI)

- ▶ The governing equations (Fedotov, 2000; Medlock & Kot, 2003; Lutscher *et al.*, 2004):

$$\partial_t S = -\beta IS - DS + D \int_{\Omega} k(x-y) S(y, t) \, dy,$$

$$\partial_t I = \beta IS - DI + D \int_{\Omega} k(x-y) I(y, t) \, dy.$$

- ▶  $\beta$  is the infection rate
- ▶  $D$  is the dispersal rate
- ▶  $k(u)$ , the kernel, is the dispersal distribution
- ▶ Assumptions:  $K = S + I$  is constant and  $\Omega = \mathbb{R}$ , give

$$\partial_t I = \beta I(K - I) - DI + D \int_{\mathbb{R}} k(x-y) I(y, t) \, dy.$$

## The movement process: Linear integro-differential equations

- ▶  $N(x, t)$  is the density of a population at position  $x$  and time  $t$
- ▶ At rate  $D$ , individuals move to a new location **instantaneously**
- ▶  $k(x - y)$  is the proportion of individuals moving from  $y$  to  $x$

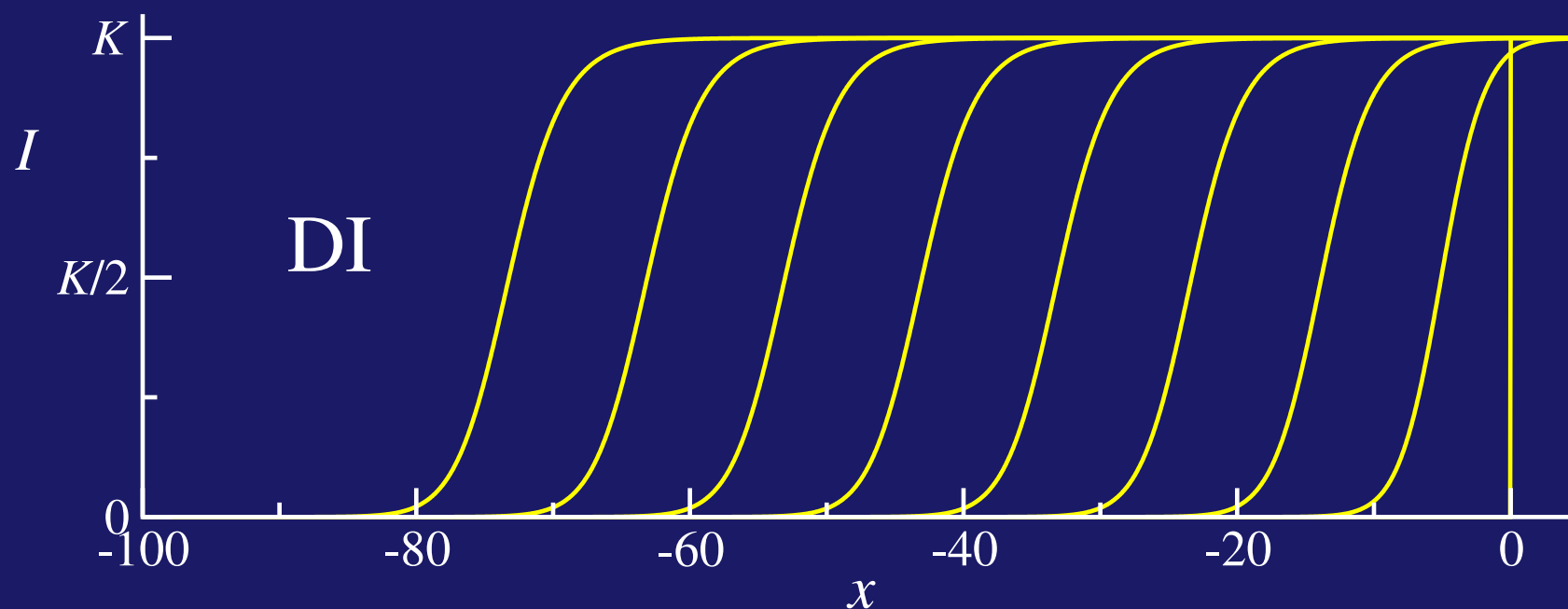
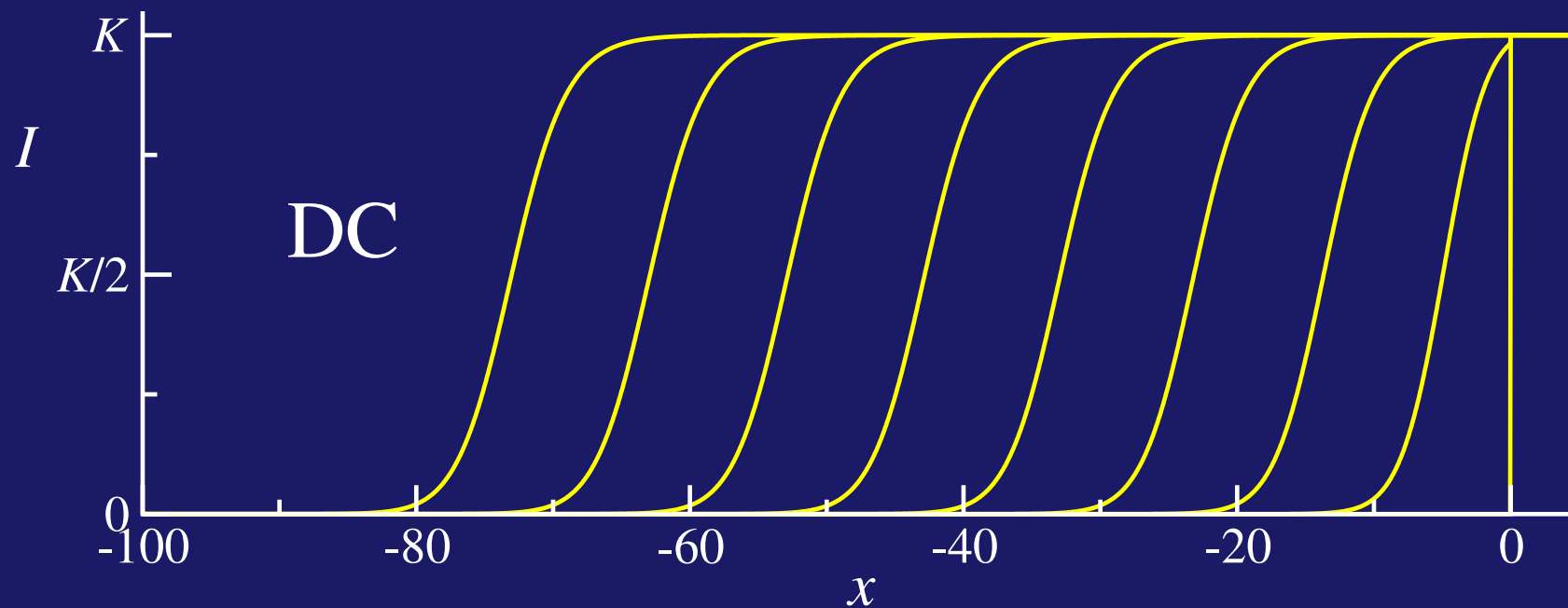
$$\partial_t N = D \int_{\mathbb{R}} k(x - y) N(y, t) \, dy - DN$$

# Position jump process or kangaroo process (Othmer *et al.*, 1988)

- ▶ An individual starts at  $x = 0$  at  $t = 0$
- ▶ He waits an exponentially-distributed time (with parameter  $D$ )...
- ▶ ...and then jumps to a new position  $y$  that is governed by the distribution  $k(x - y)$

$$\partial_t N = D \int_{\mathbb{R}} k(x - y) N(y, t) \, dy - DN$$

# Both models have traveling wave solutions



## Wave speed

### ► Distributed Contacts

$$\partial_t I = \beta \left( \int_{\mathbb{R}} k(x-y) I(y, t) \, dy \right) (K - I)$$

Using the traveling-wave coordinate,  $z = x + ct$ ,

$$cI' = \beta \left( \int_{\mathbb{R}} k(z-y) I(y) \, dy \right) (K - I).$$

Linearizing about  $I = 0$ ,

$$cI' = \beta K \left( \int_{\mathbb{R}} k(z-y) I(y) \, dy \right).$$

Assuming  $I = Ae^{-\theta z}$ ,

$$c\theta = -\beta K M(\theta),$$

where

$$M(\theta) = \int_{\mathbb{R}} k(u) e^{\theta u} \, du.$$

For a constant speed traveling wave,  $k(x) \leq A e^{-\gamma|x|}$ , which implies  $M(\theta)$  exists for  $|\theta| < \gamma$ .

The critical speeds are

$$c_R = -\beta K \sup_{\theta > 0} \frac{M(\theta)}{\theta}, \quad c_L = -\beta K \inf_{\theta < 0} \frac{M(\theta)}{\theta}.$$

Parametrically,

$$\begin{aligned} c^* &= -\beta K M'(\theta^*), \\ M(\theta^*) &= \theta^* M'(\theta^*). \end{aligned}$$

Aronson (1977) showed this is the asymptotic speed.



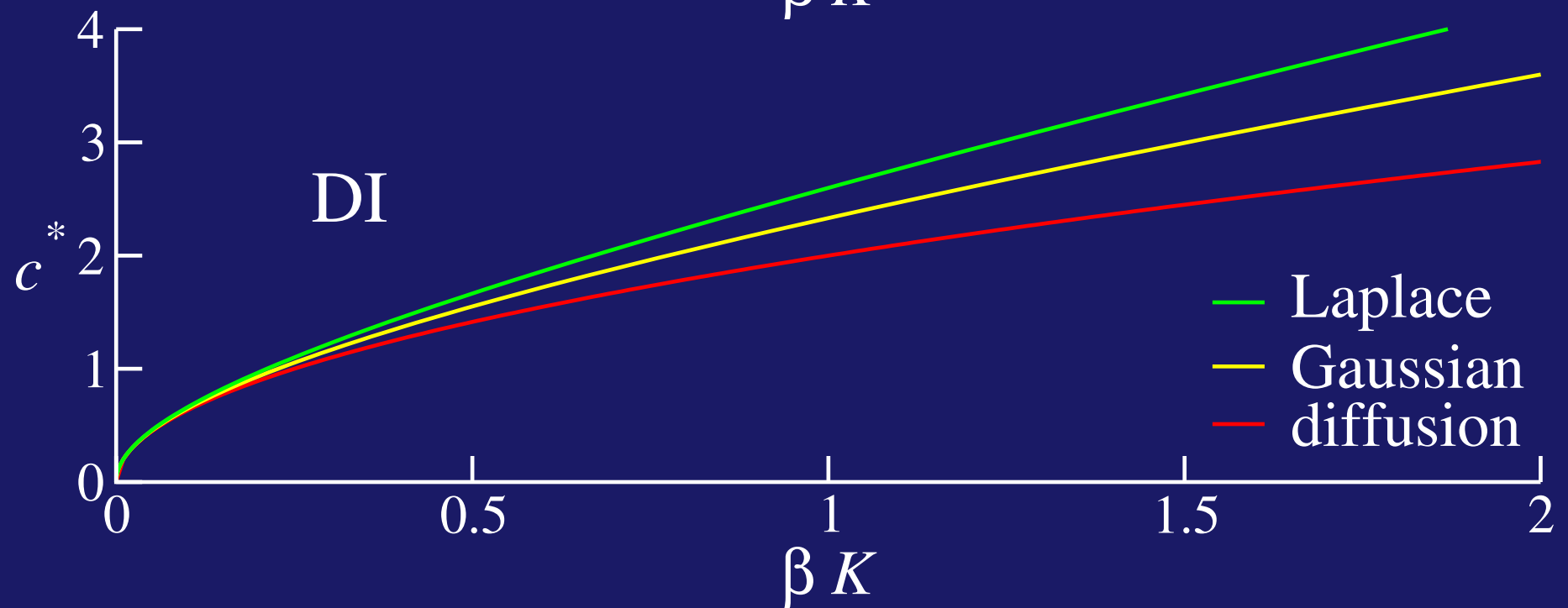
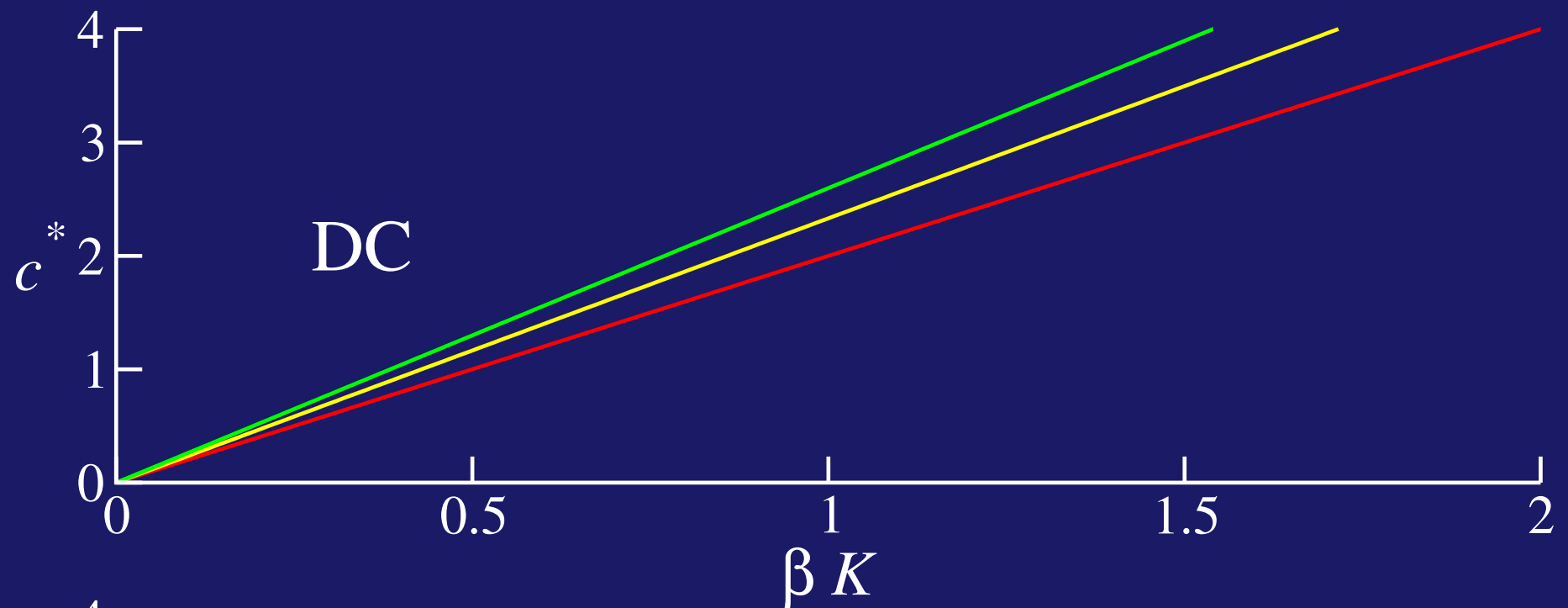
## ► Distributed Infectives

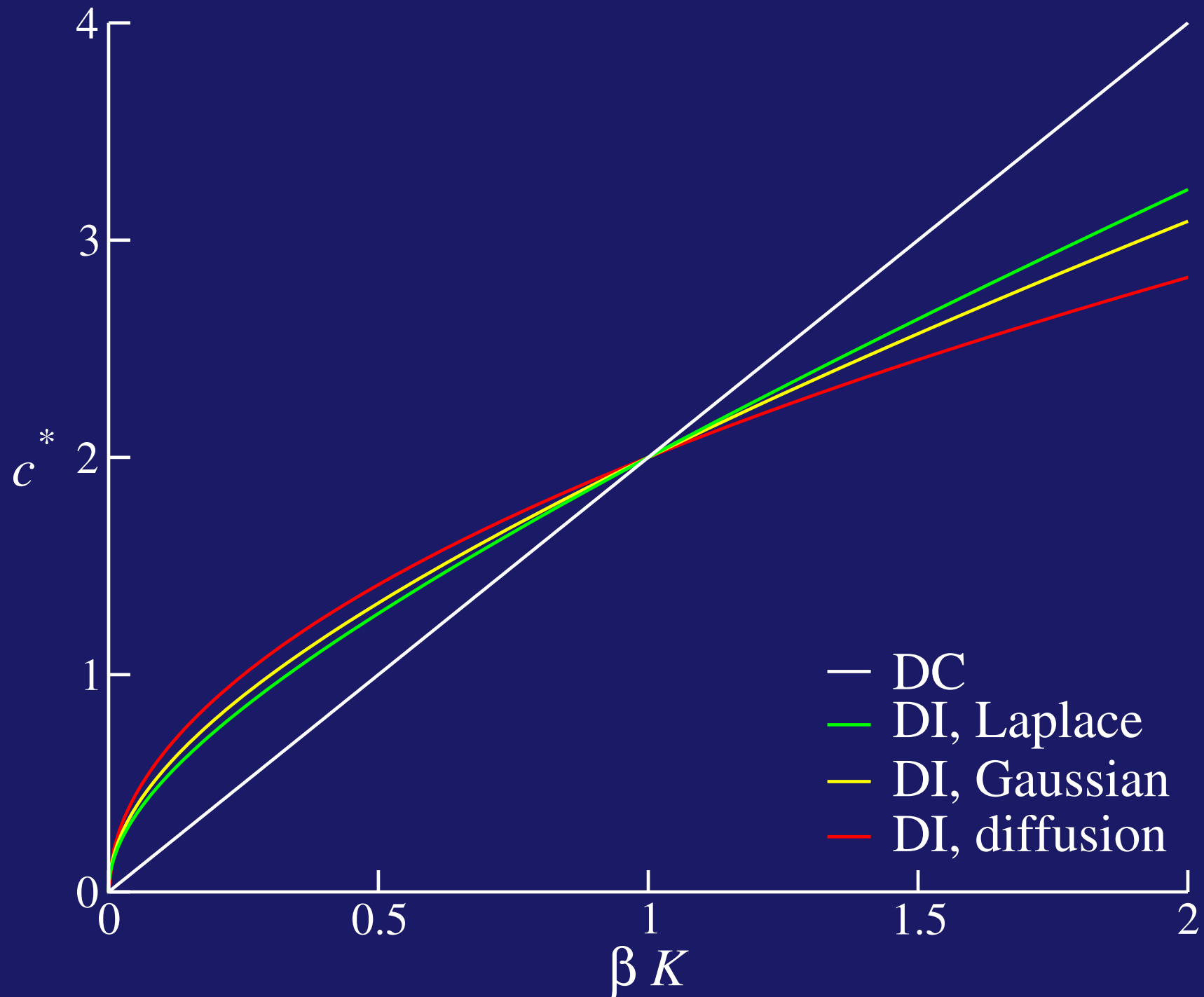
$$\partial_t I = \beta I(K - I) - DI + D \int_{\mathbb{R}} k(x - y) I(y, t) \, dy$$

The same procedure gives

$$\begin{aligned} c^* &= -DM'(\theta^*), \\ \beta K &= D[1 - M(\theta^*) + \theta^* M'(\theta^*)]. \end{aligned}$$

Lutscher *et al.* (2004) showed this is the asymptotic speed.





## A perturbation scheme for the wave shape

The key to the perturbation scheme is expanding the convolution

$$k * I = \int_{\mathbb{R}} k(y) I(z - y) \, dz$$

for large  $c$ .

Let  $\xi = z/c$  and  $h(\xi) = I(z)$ . Then

$$\begin{aligned} k * I &= \int_{\mathbb{R}} k(y) I(z - y) \, dy \\ &= \int_{\mathbb{R}} k(y) h\left(\xi - \frac{y}{c}\right) \, dy \\ &= \int_{\mathbb{R}} k(y) \left[ h(\xi) - \frac{y}{c} h'(\xi) + O\left(\frac{1}{c^2}\right) \right] \, dy \\ &= h(\xi) - \frac{M'(0)}{c} h'(\xi) + O\left(\frac{1}{c^2}\right). \end{aligned}$$

## Distributed Infectives

$$cI' = \beta I(K - I) - DI + D \int_{\mathbb{R}} k(y)I(z - y) \, dy$$

with  $I(-\infty) = 0$ ,  $I(+\infty) = K$ ,  $I(0) = K/2$ .

Letting  $\xi = z/c$  and  $h(\xi) = I(z)$  gives

$$h' = \beta h(K - h) - D \left[ \frac{M'(0)}{c} h' + O\left(\frac{1}{c^2}\right) \right].$$

Expanding  $h = h_0 + \frac{1}{c}h_1 + O\left(\frac{1}{c^2}\right)$ ,

$$h'_0 = \beta h_0(K - h_0),$$

$$h'_1 = \beta h_1(K - 2h_0) - DM'(0)h'_0,$$

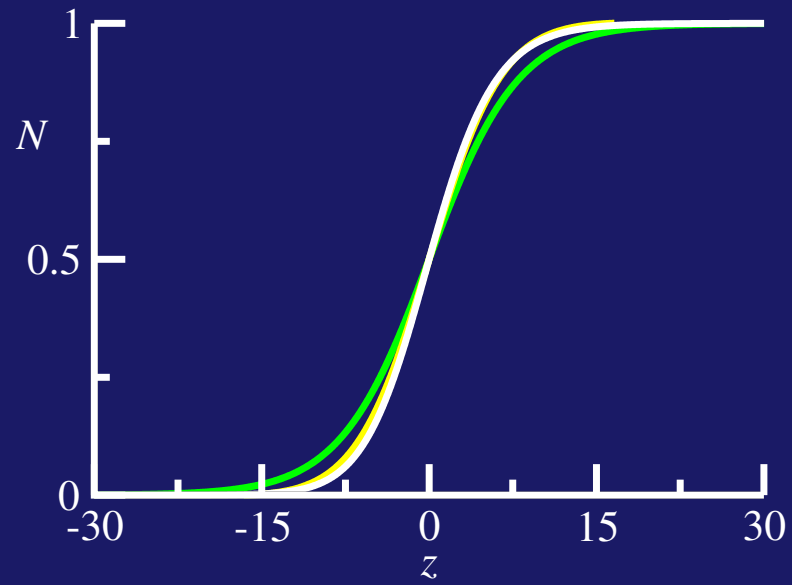
with  $h_0(-\infty) = 0$ ,  $h_0(+\infty) = K$ ,  $h_0(0) = K/2$ ,  
and  $h_1(-\infty) = h_1(+\infty) = h_1(0) = 0$ .

Then

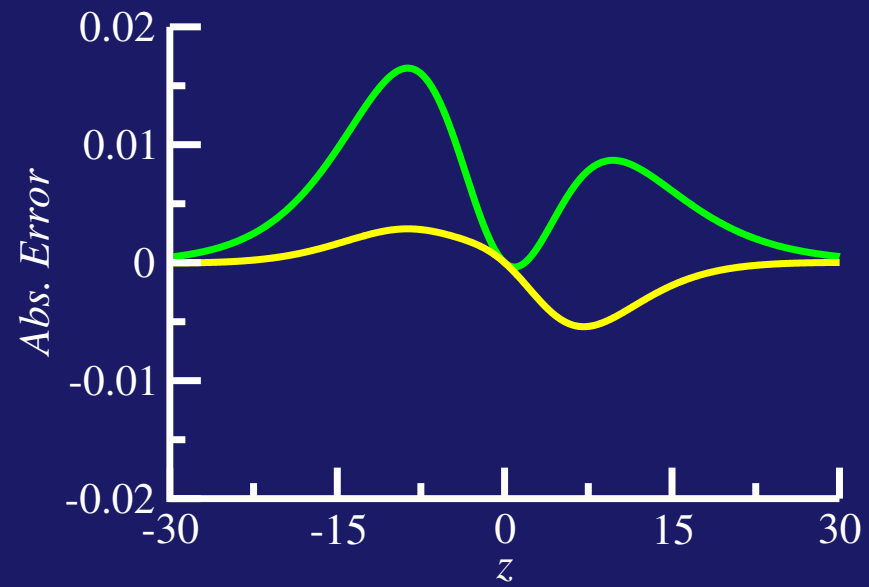
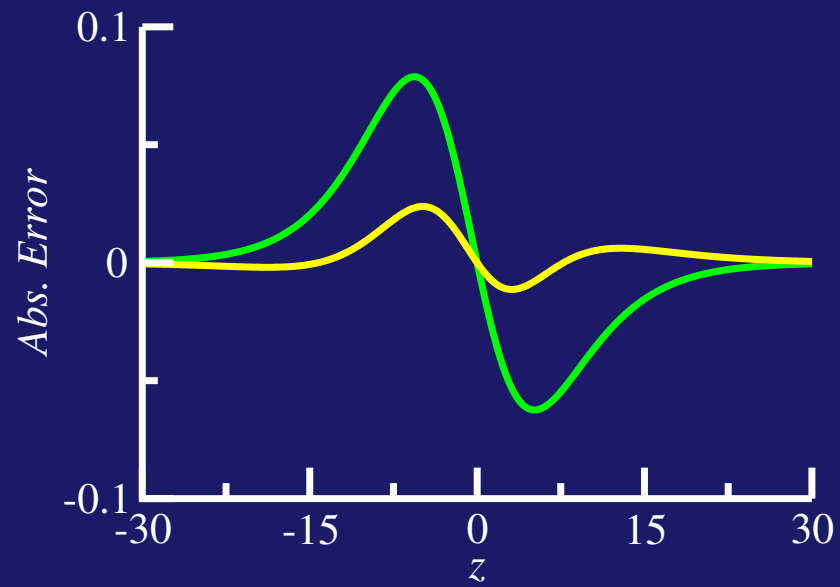
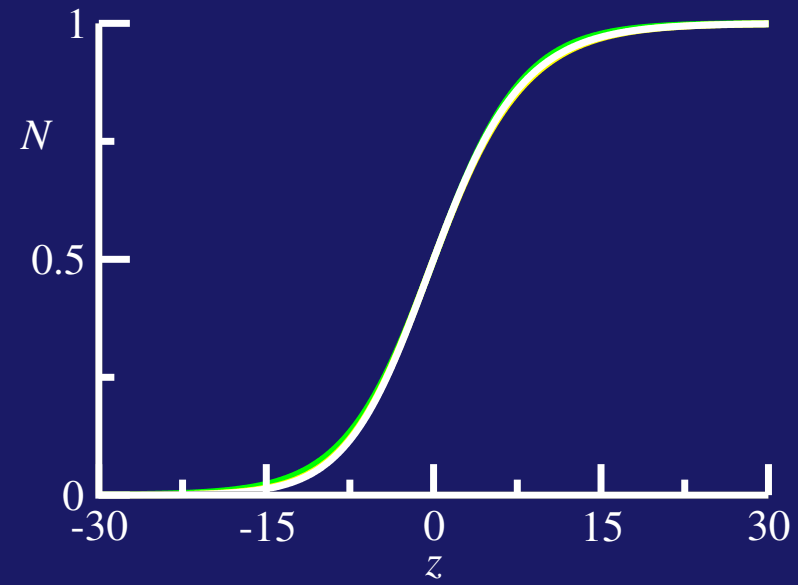
$$h_0(\xi) = \frac{Ke^{\beta K\xi}}{1 + e^{\beta K\xi}},$$

$$h_1(\xi) = D\beta K^2 M'(0) \frac{\xi e^{\beta K\xi}}{(1 + e^{\beta K\xi})^2}.$$

Exponential kernel



Laplace kernel



## Asymptotic speed

For an accelerating invasion  $k(x) > Ae^{-\gamma|x|}$ , we can compute the wave asymptotically.

- ▶ Starting with the linearized DC equation

$$\partial_t I = \beta K \int_{\mathbb{R}} k(x-y) I(y, t) dy, \quad I(x, 0) = I_0 \delta(x),$$

taking the Fourier transform

$$\partial_t \hat{I} = \beta K \hat{k}(\omega) \hat{I}, \quad \hat{I}(\omega) = I_0,$$

gives

$$\hat{I}(\omega, t) = I_0 e^{\beta K \hat{k}(\omega) t}.$$



Inverting gives

$$I(x, t) = \frac{I_0}{2\pi} \int_{\mathbb{R}} e^{-i\omega x} e^{\beta K \hat{k}(\omega)t} d\omega \approx I_0 e^{\beta K t} k(x) \quad \text{as } |x| \rightarrow \infty,$$

subject to some technical conditions.

► Likewise for the DI model

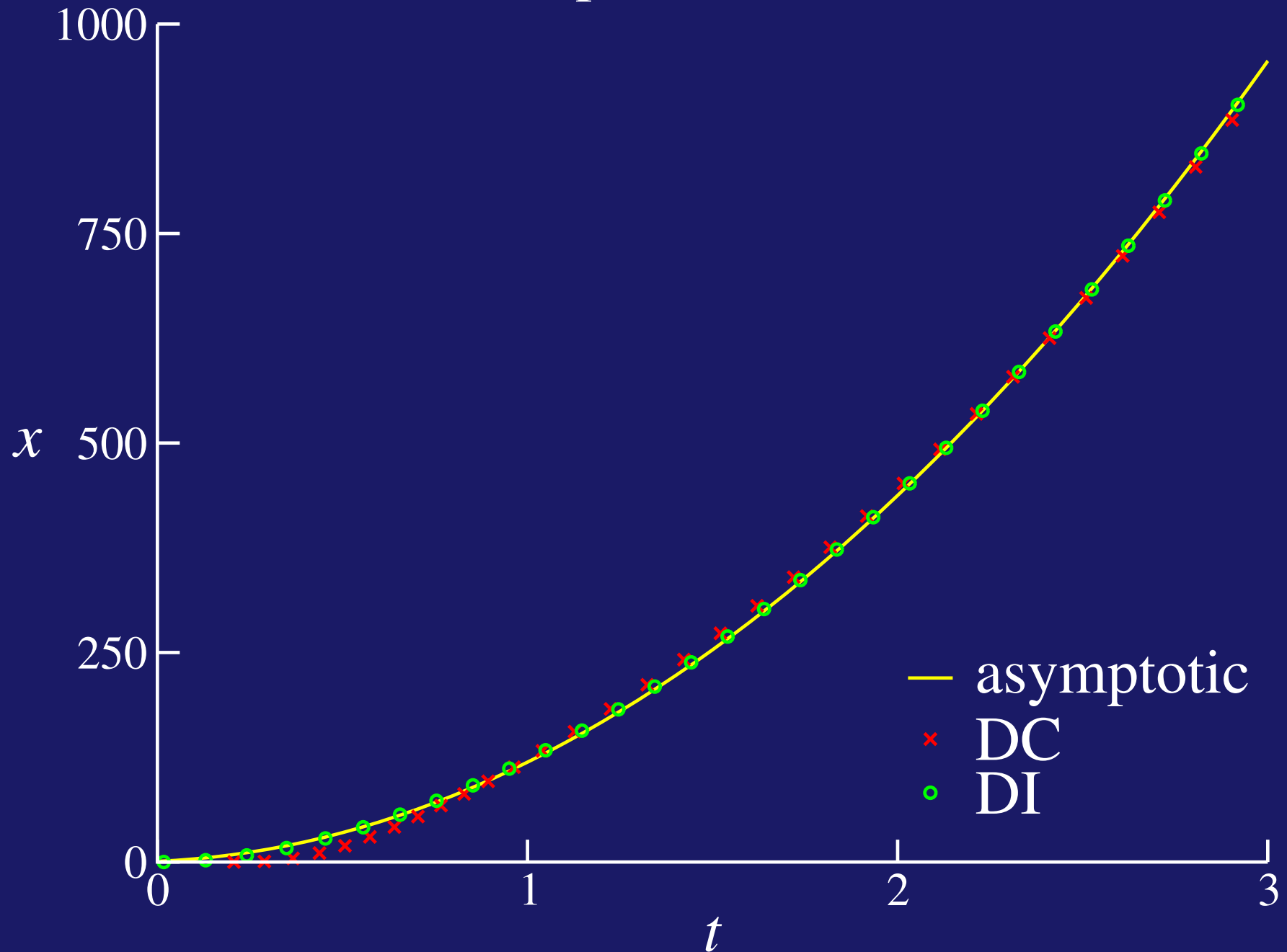
$$I(x, t) \approx I_0 e^{\beta K t} k(x) \quad \text{as } |x| \rightarrow \infty.$$

Setting some threshold value,  $\tilde{I}$ , gives

$$x \approx k^{-1} \left( \frac{\tilde{I}}{I_0} e^{-\beta K t} \right),$$

the position of the threshold as a function of time.

# front position vs. time



## Time-periodic coefficients

$$\partial_t I = \beta(t)I(K(t) - I) - D(t)I + D(t) \int_{\mathbb{R}} k(x - y, t)I(y, t) \, dy$$

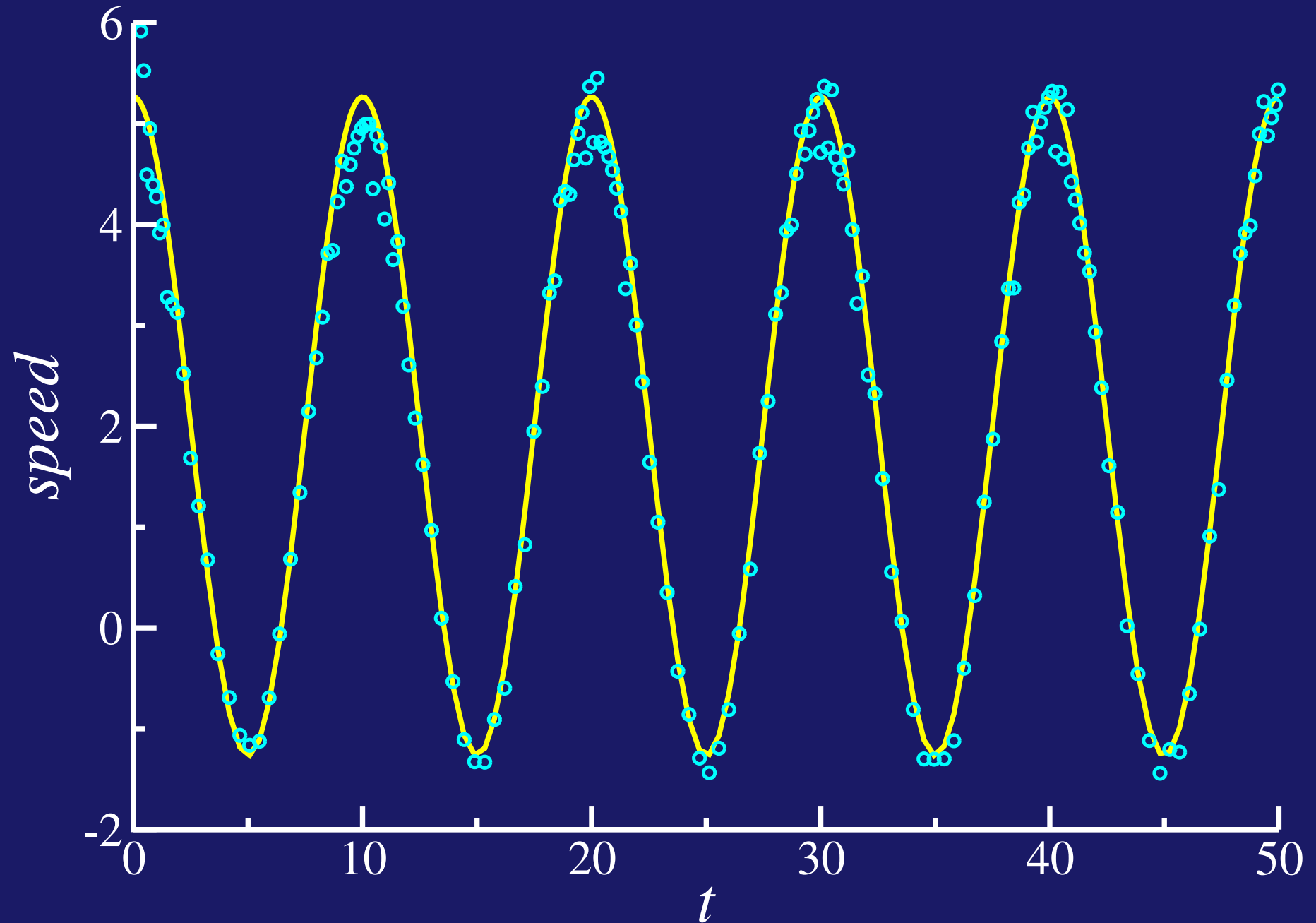
Using the same linearization procedure, the speed is given by

$$\begin{aligned} \overline{\beta K} &= \overline{D} - \overline{DM}(\theta) + \theta \overline{D\partial_\theta M}(\theta), \\ \overline{c} &= -\overline{D\partial_\theta M}(\theta), \end{aligned}$$

where

$$\overline{f} = \frac{1}{T} \int_0^T f(t) \, dt.$$

speed vs. time



## Numerics: IDEs are better than PDEs!

- ▶ Integral dispersal is much more stable than diffusion, which allows much larger time steps
- ▶ In general,

$$\partial_t \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix} = f \left( \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix}, \begin{bmatrix} k_1 * u_1 & \dots & k_m * u_1 \\ \vdots & & \vdots \\ k_1 * u_n & \dots & k_m * u_n \end{bmatrix} \right).$$

- ▶ Trapezoid rule and Simpson's rule are  $O(N^2)$ .
- ▶ FFT is  $O(N \log N)$  (Andersen, 1991)

$$\mathcal{F}(k_i * u_j) = \mathcal{F}(k_i) \mathcal{F}(u_j).$$

## ▶ Algorithm

- ▶ Compute  $\mathcal{F}(k_i)$  at the beginning.
- ▶ Each time step, compute  $\mathcal{F}(u_j)$  and  $\mathcal{F}^{-1} [\mathcal{F}(k_i)\mathcal{F}(u_j)]$ .
- ▶ Step in time with Runge–Kutta scheme.

## Conclusions

- ▶ There are more flexible frameworks than reaction–diffusion for continuous-time models.
- ▶ Integro-differential equations can be used with empirical dispersal data, give faster speeds than r–d, and can give accelerating waves.
- ▶ Traveling wave speed is sensitive to the form of the model; care should be used.
- ▶ Perturbation techniques can be used to compute approximate solutions to the IDEs.

## Problems

- ▶ Instead of space,  $x$  represents some other characteristic.
- ▶ Estimating kernels from dispersal data.
- ▶ The inverse problem: Given a traveling wave, what is the kernel?