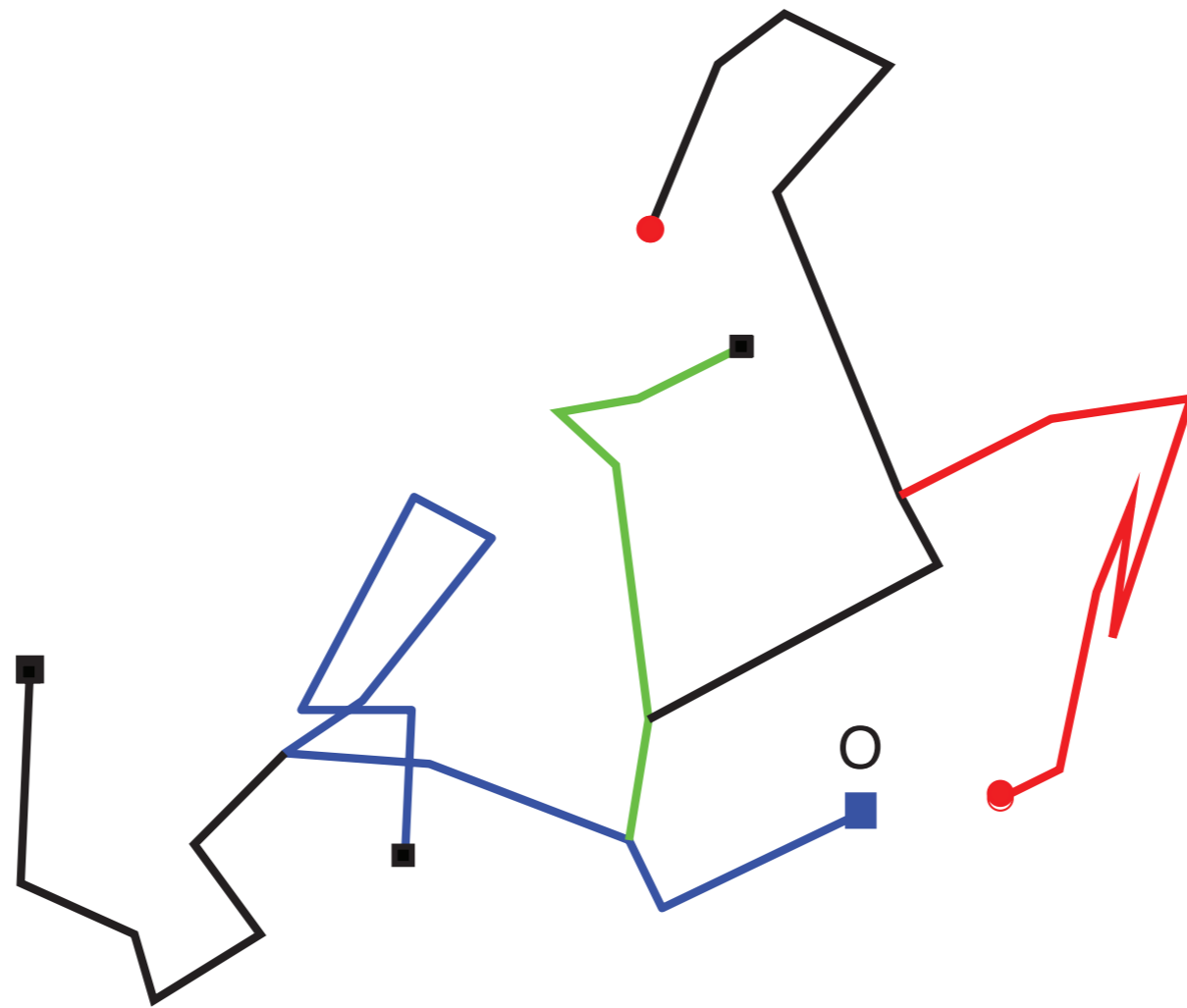


Spatial extent of an outbreak in animal epidemics

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SIR model for epidemics

Three species : susceptibles (S), infected (I), recovered (R)

$$\frac{dS}{dt} = -\beta I S$$

- mean field fully connected model

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

- β rate of infection transmission

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

- γ rate at which an infected recovers

$$I(t) + S(t) + R(t) = N$$

N being the total population

Outbreak of an epidemic

Initial condition : $I(0) = 1, S(0) = N - 1 \approx N, R(0) = 0$

$$\frac{dS}{dt} = -\beta I S$$

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

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

$$t \approx 0, S \approx N$$



Outbreak regime

$$\frac{dI}{dt} \simeq (\beta N - \gamma) I$$

Reproduction rate: $R_0 = \frac{\beta N}{\gamma}$

Deterministic and stochastic models

SIR is a deterministic model. In the outbreak fluctuations are important

- Stochastic process: Galton-Watson (mean field)
- each infected individual transmits the disease at rate $N\beta$
- each infected individual recovers at rate γ

Reproduction rate:

$$R_0 = \frac{\beta N}{\gamma}$$

- $R_0 < 1$ epidemics extinction
- $R_0 > 1$ epidemics invasion
- $R_0 = 1$ critical case

How far in space can an epidemic spread?

Problem 1: How to model the space?

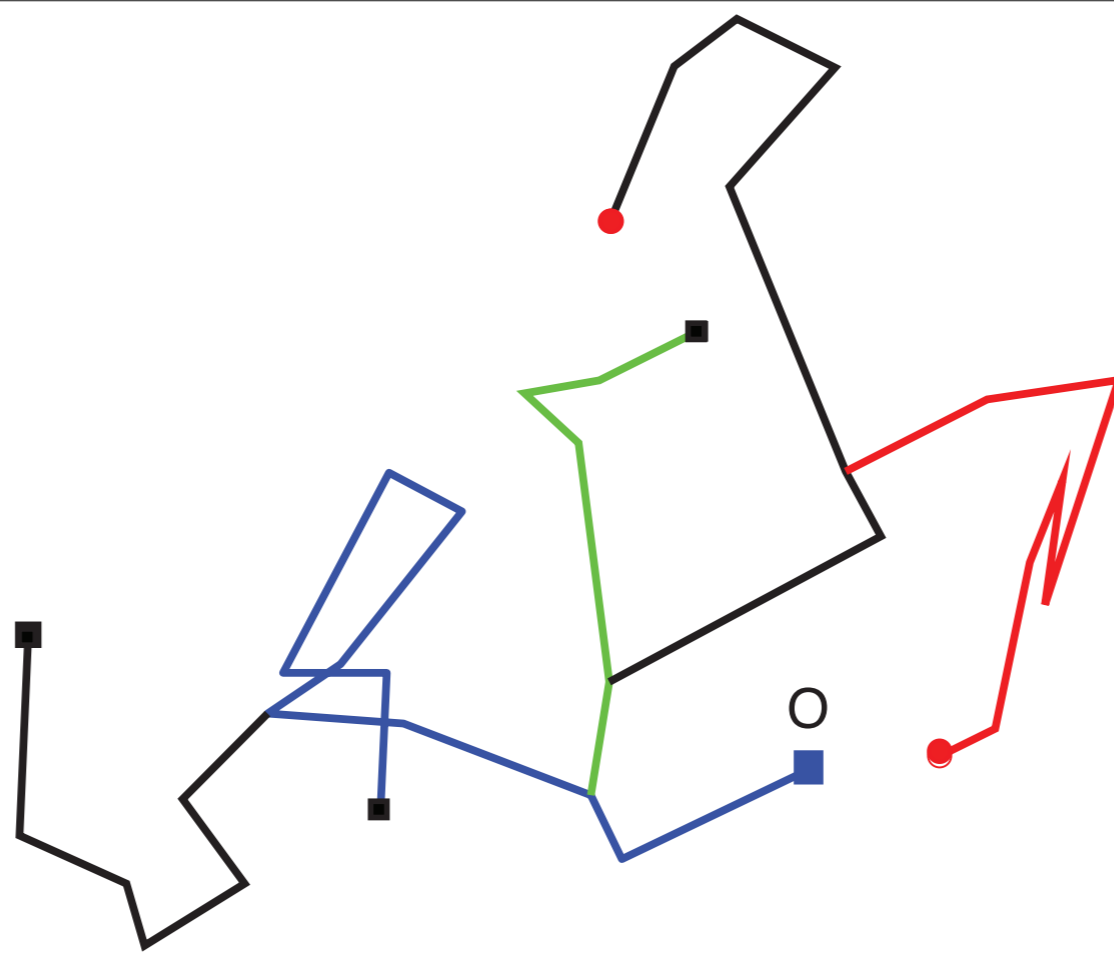
The population is uniformly distributed

At time $t = 0$ an infected individual appears

... and moves in space

Brownian motion is the paradigm of animal migration

while human beings take the plane (even when they are sick)



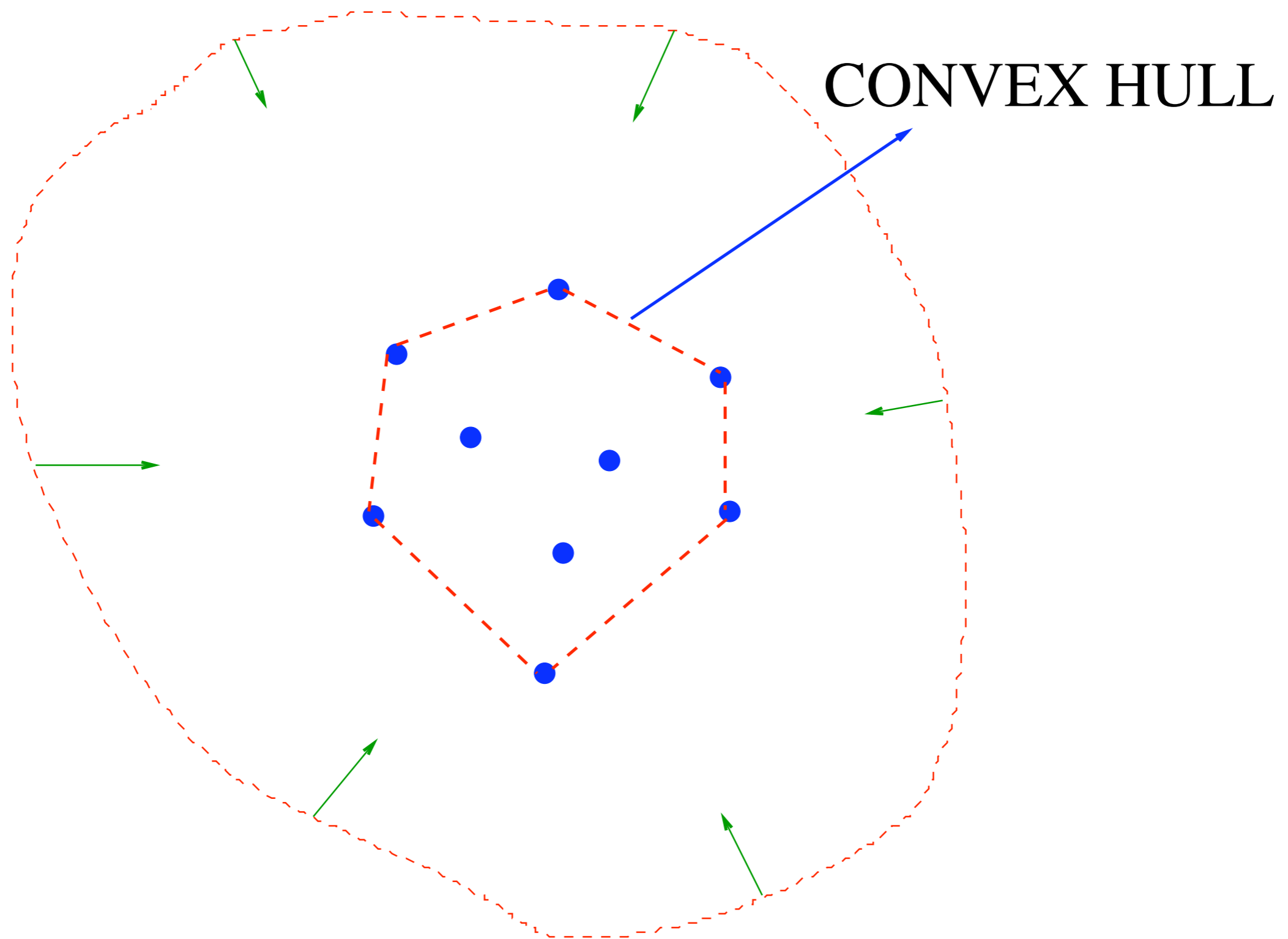
The good candidate: Brownian process with branching and death

In dt , each infected can:

- recovers with probability γdt
- infects with probability $\beta N dt = \gamma R_0 dt$
- otherwise, it diffuses (D diffusion const.)

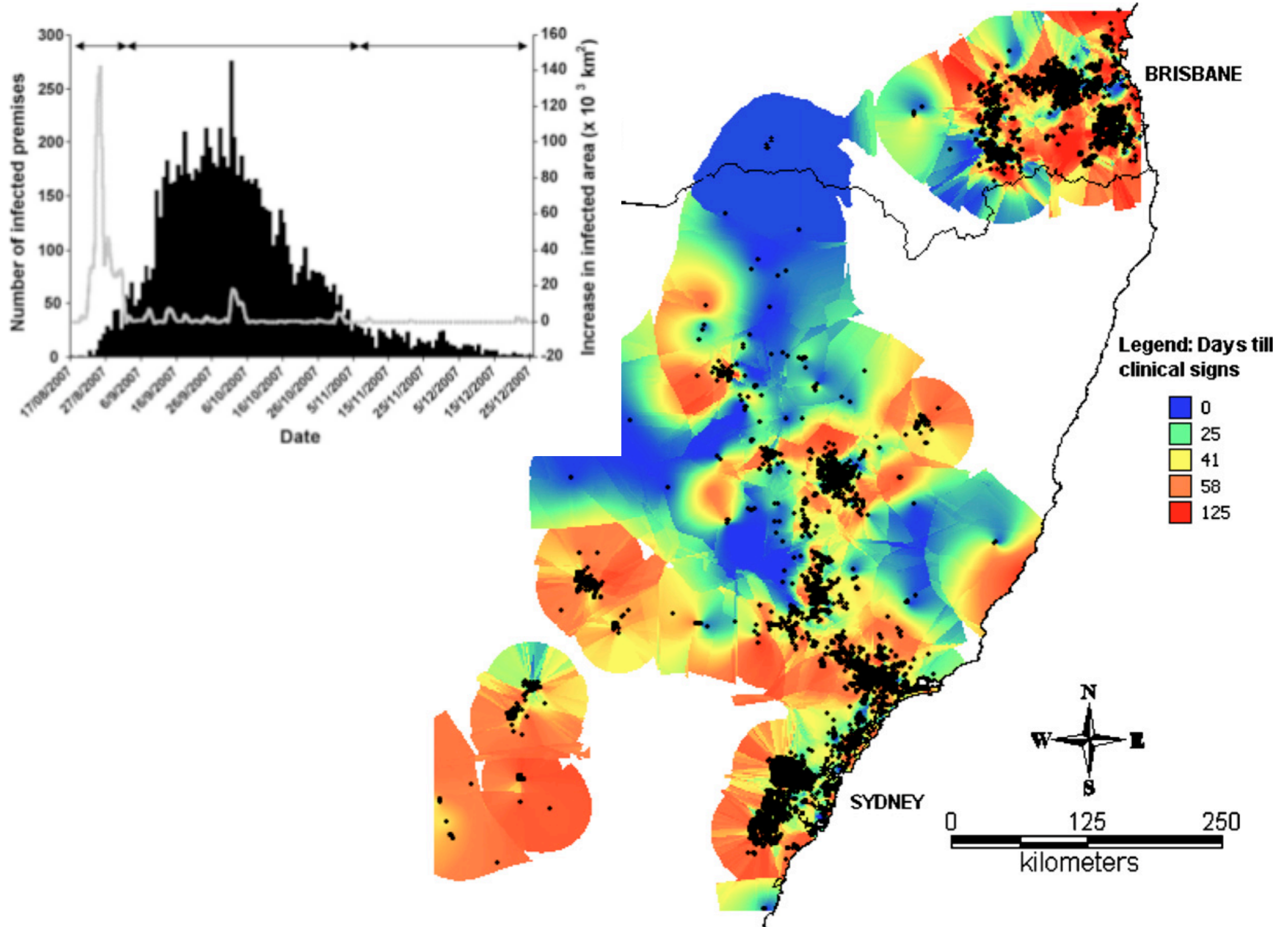
$$\text{Diffusion } \Delta x \sim \sqrt{2Ddt}, \langle \Delta x \rangle = 0, \langle \Delta x^2 \rangle = \sqrt{2Ddt}$$

Problem 2: How to quantify the area that needs to be quarantined?



Algorithms: Graham Scan ($N \log(N)$)

Real applications

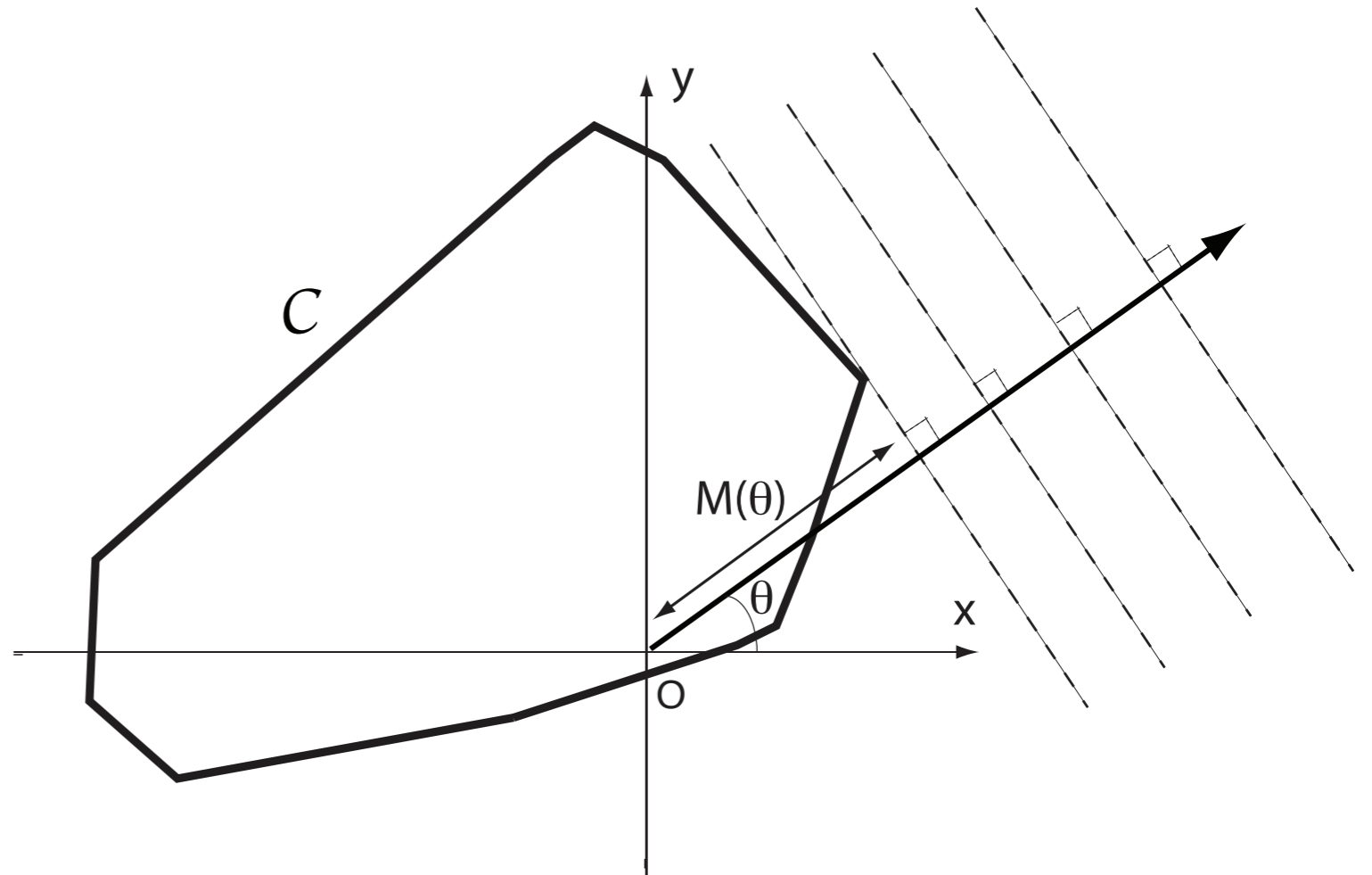


How to compute the convex hull of Branching processes?

Cauchy formulas

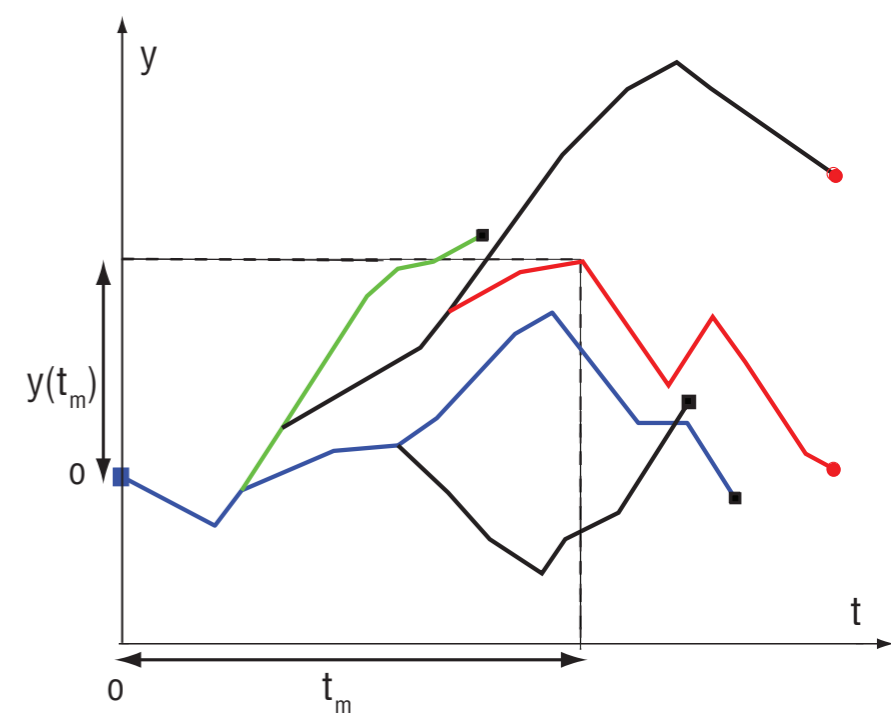
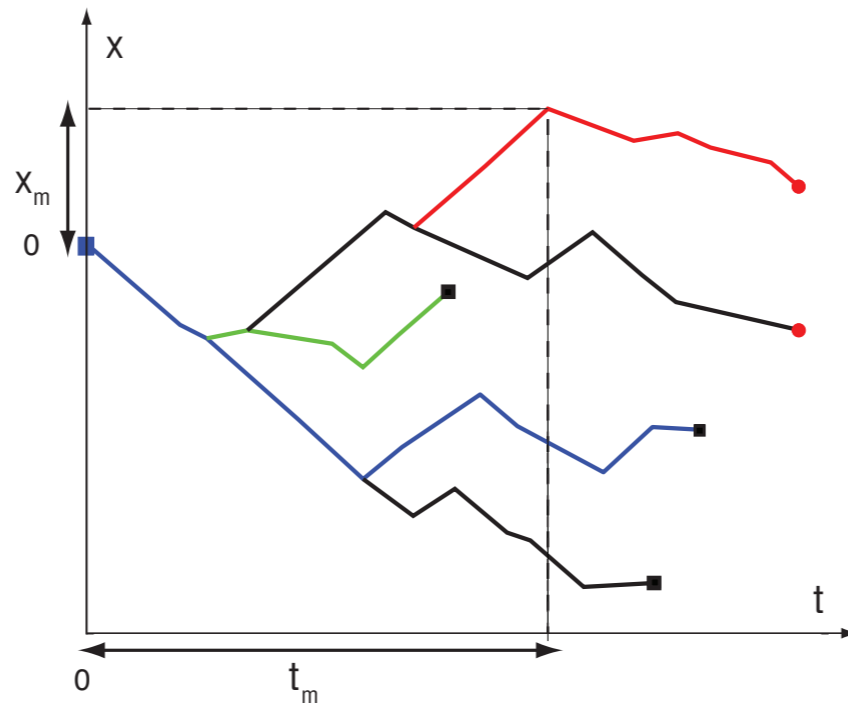
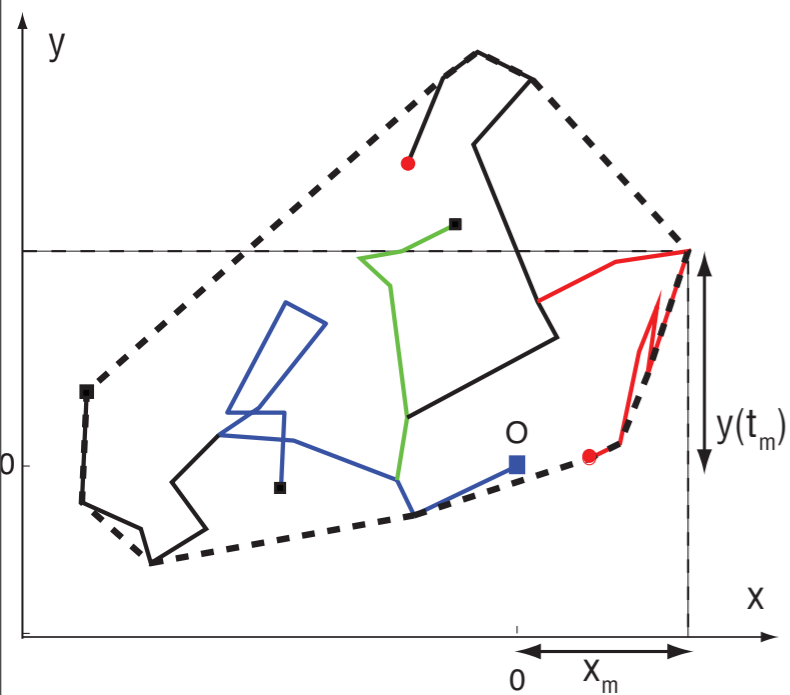
$$L = \int_0^{2\pi} M(\theta) d\theta$$

$$A = \frac{1}{2} \int_0^{2\pi} [M^2(\theta) - (M'(\theta))^2] d\theta$$



Support Function

$$M(\theta) = \max_{0 \leq \tau \leq t} [x_\tau \cos \theta + y_\tau \sin \theta]$$



- $x_m(t)$ x -maximum up to time t
- t_m time *location* of the maximum

$$M(0) = x_{\tau=t_m} = x_m(t)$$

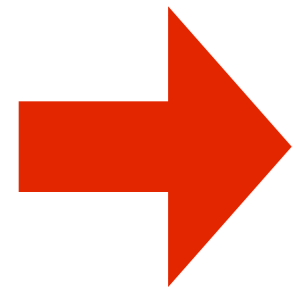
$$M'(\theta = 0) = -x_{t_m} \sin \theta + y_{t_m} \cos \theta \Big|_{\theta=0} = y_{t_m}$$

Dimensional reduction

$$L = \int_0^{2\pi} M(\theta) d\theta$$

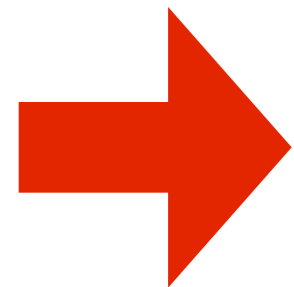
$$A = \frac{1}{2} \int_0^{2\pi} [M^2(\theta) - (M'(\theta))^2] d\theta$$

If the process is rotationally invariant any average is independent of θ



$$\langle L(t) \rangle = 2\pi \langle M(0) \rangle$$

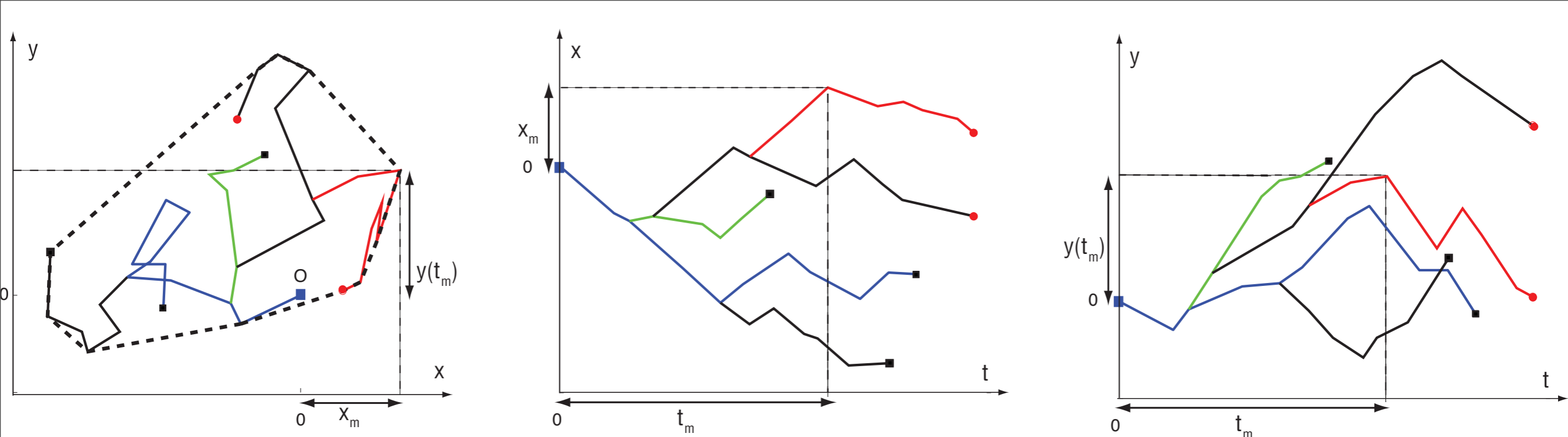
$$\langle A(t) \rangle = \pi [\langle M^2(0) \rangle - \langle M'(0)^2 \rangle]$$



$$\langle L(t) \rangle = 2\pi \langle x_m(t) \rangle$$

$$\langle A(t) \rangle = \pi [\langle x_m^2(t) \rangle - \langle y^2(t_m) \rangle]$$

This relation is valid ONLY in average



$$\langle L(t) \rangle = 2\pi \langle x_m(t) \rangle$$

$$\langle A(t) \rangle = \pi \left[\langle x_m^2(t) \rangle - \langle y^2(t_m) \rangle \right]$$

consider a 1d branching process evolving in $(0, t)$

- x_m is the global maximum
- t_m is the location of the maximum
- $\langle y^2(t_m) \rangle = \dots = 2D t_m$

Backward Fokker Planck equation

$$Q_t(x_m) = \text{Proba}[\text{global max up to } t < x_m]$$

$$Q_{t+dt}(x_m) = \gamma dt + R_0 \gamma dt Q_t^2(x_m) + [1 - \gamma(R_0 + 1)] dt \langle Q_t(x_m - \Delta x) \rangle$$

- $\langle Q_t(x_m - \Delta x) \rangle = Q_t(x_m) - \langle \Delta x \rangle Q_t'(x_m) + \langle \frac{\Delta x^2}{2} \rangle Q_t''(x_m) + \dots$
- $\langle \Delta x \rangle = 0$
- $\langle \Delta x^2 \rangle = 2Ddt$

$$\langle Q_t(x_m - \Delta x) \rangle = Q_t(x_m) + Ddt \partial_x^2 Q_t(x_m) + \dots$$

$$\frac{\partial}{\partial t} Q = D \frac{\partial^2}{\partial x_m^2} Q - \gamma(R_0 + 1)Q + \gamma R_0 Q^2 + \gamma$$

- initial condition $Q_{t=0}(x_m) = \theta(x_m)$
- boundary condition $Q_t(x_m < 0) = 0$
- boundary condition $Q_t(x_m \rightarrow \infty) = 1$

$$\langle L(t) \rangle = 2\pi \int_0^\infty [1 - Q_t(x_m)] dx_m.$$

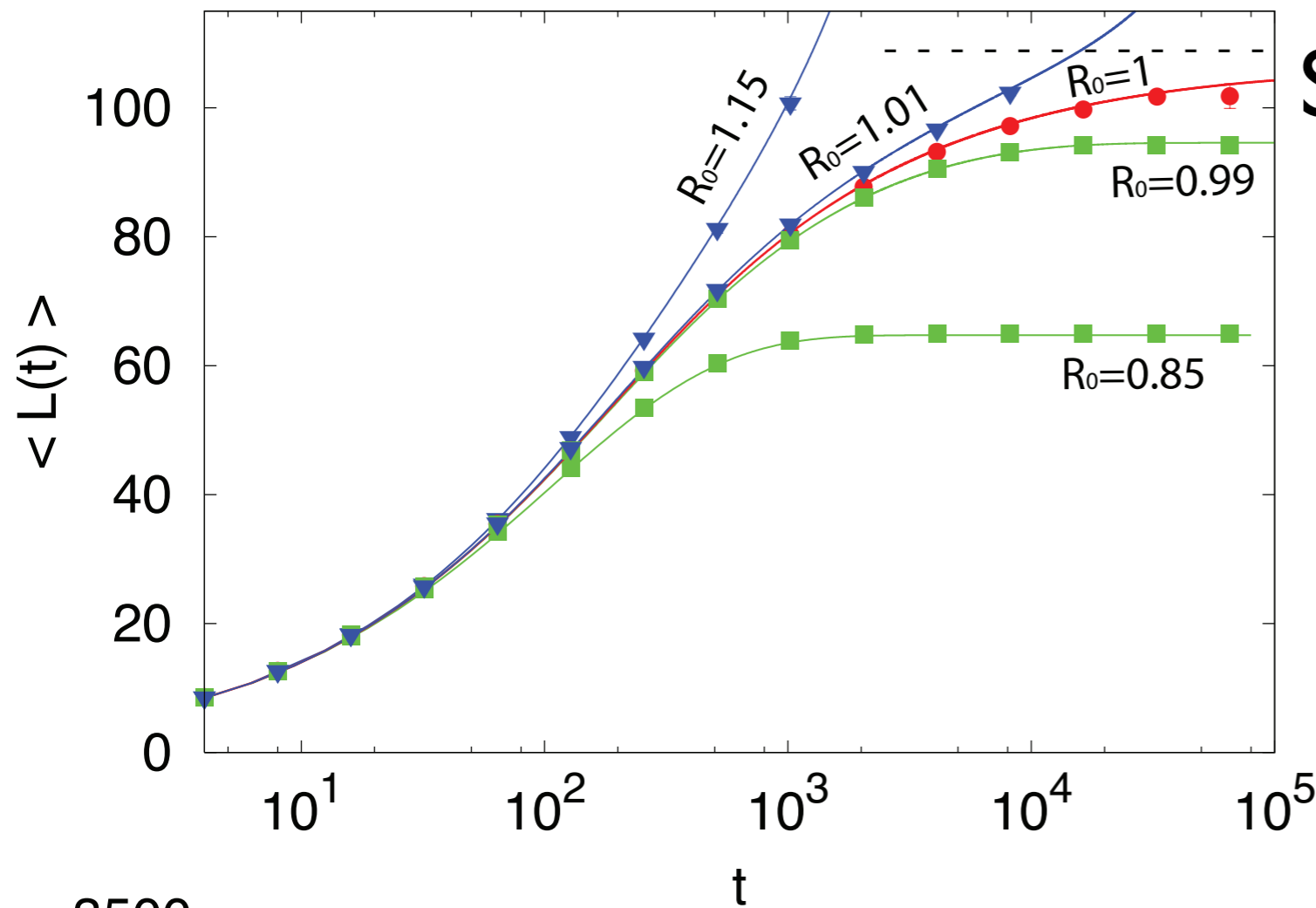
Similar calculations allows to express the mean area as:

$$\langle A(t) \rangle = \pi \int_0^\infty dx_m [2x_m(1 - Q_t(x_m)) - T_t(x_m)]$$

Where the evolution of $T_t(x_m)$ is governed by:

$$\frac{\partial}{\partial t} T_t + \partial_x Q_t(x_m) = \left[D \frac{\partial^2}{\partial x_m^2} + 2\gamma R_0 Q_t - \gamma (R_0 + 1) \right] T_t,$$

**Both PDE can be integrated numerically and solved
in some asymptotic limit**

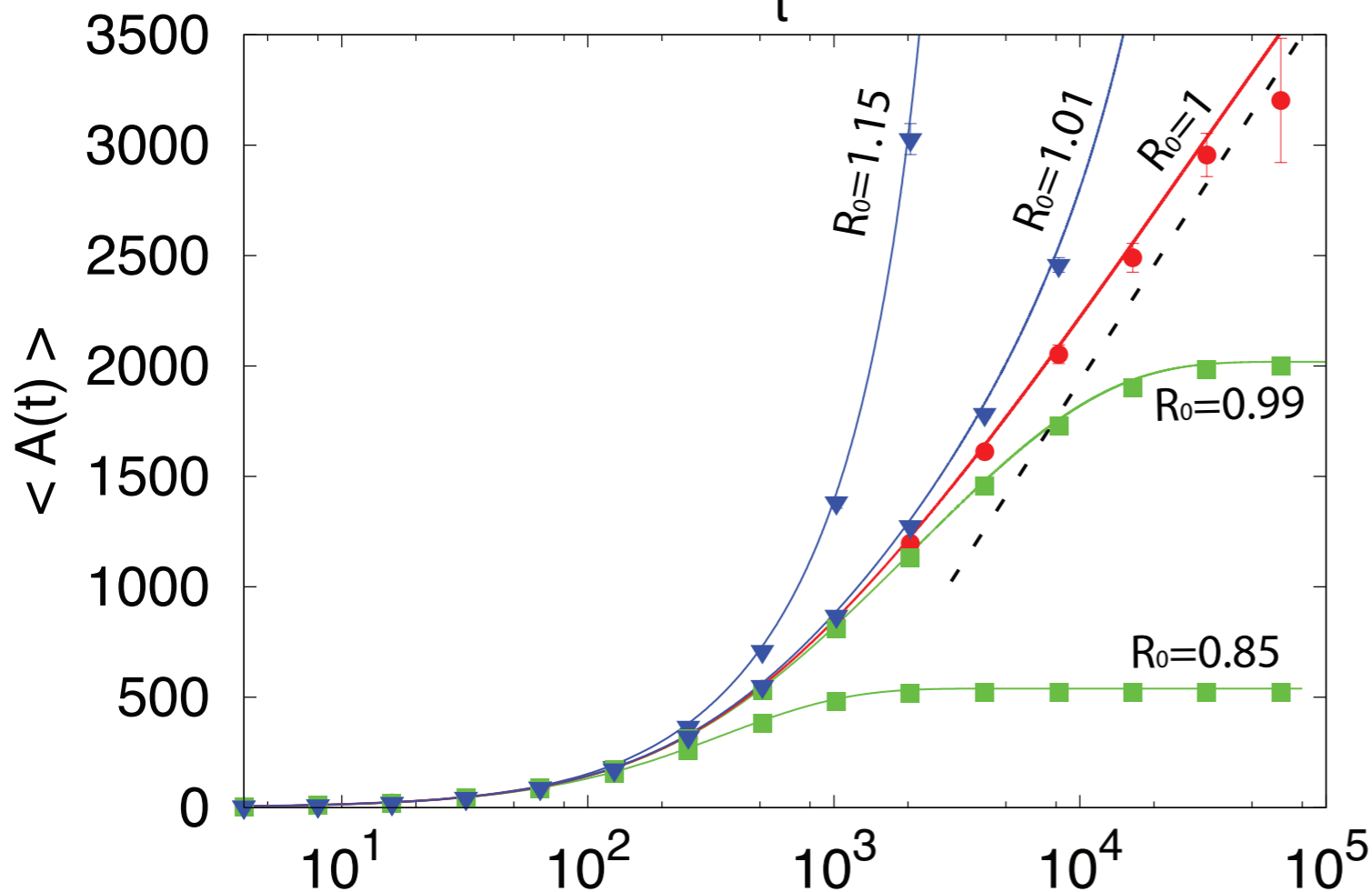


Solid lines: our predictions

Blue lines: super-critical

Red lines: critical

Green lines: sub-critical



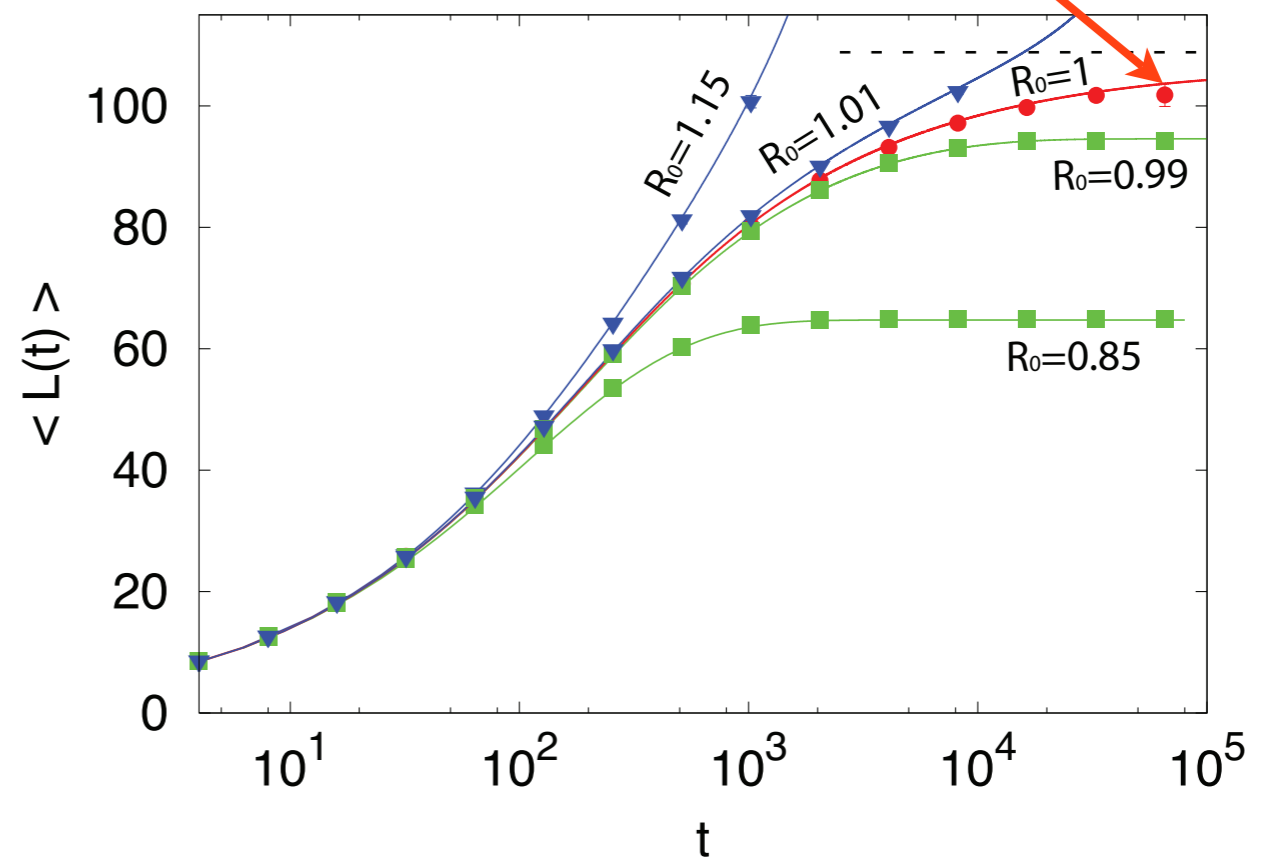
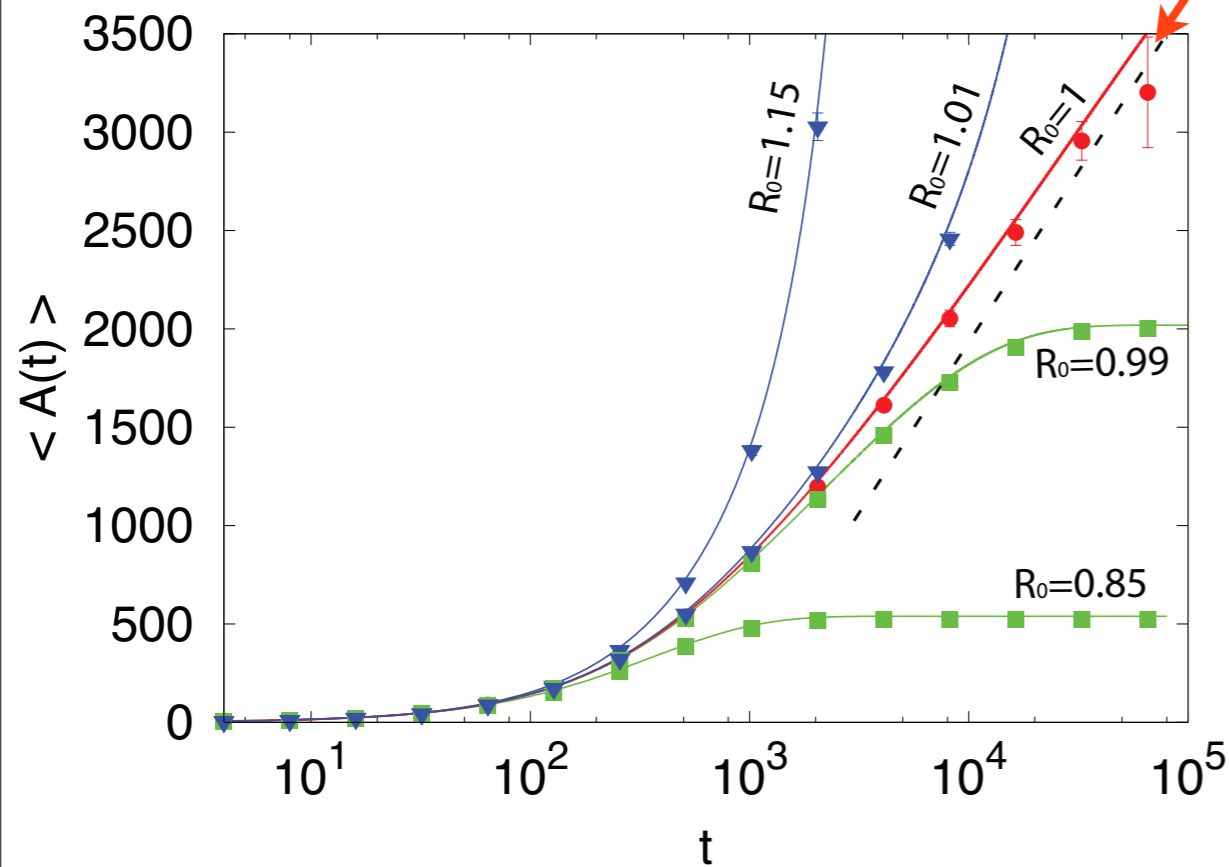
Symbols: Monte Carlo simulations

Dashed lines: analytical asymptotic results

The critical case

$$\langle L(t \rightarrow \infty) \rangle = 2\pi \sqrt{\frac{6D}{\gamma}} + \mathcal{O}(t^{-1/2})$$

$$\langle A(t \rightarrow \infty) \rangle = \frac{24\pi D}{5\gamma} \ln t + \mathcal{O}(1)$$

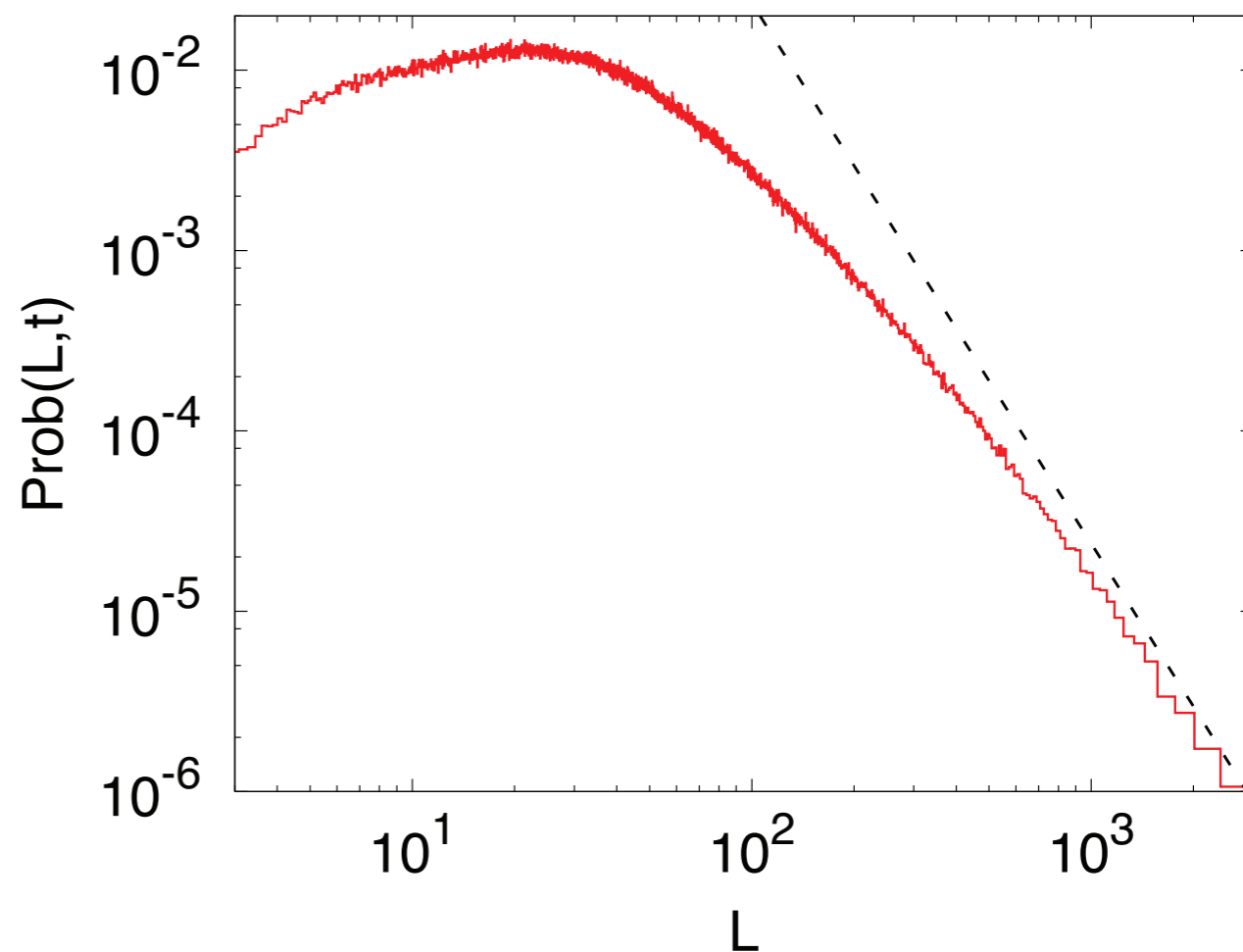
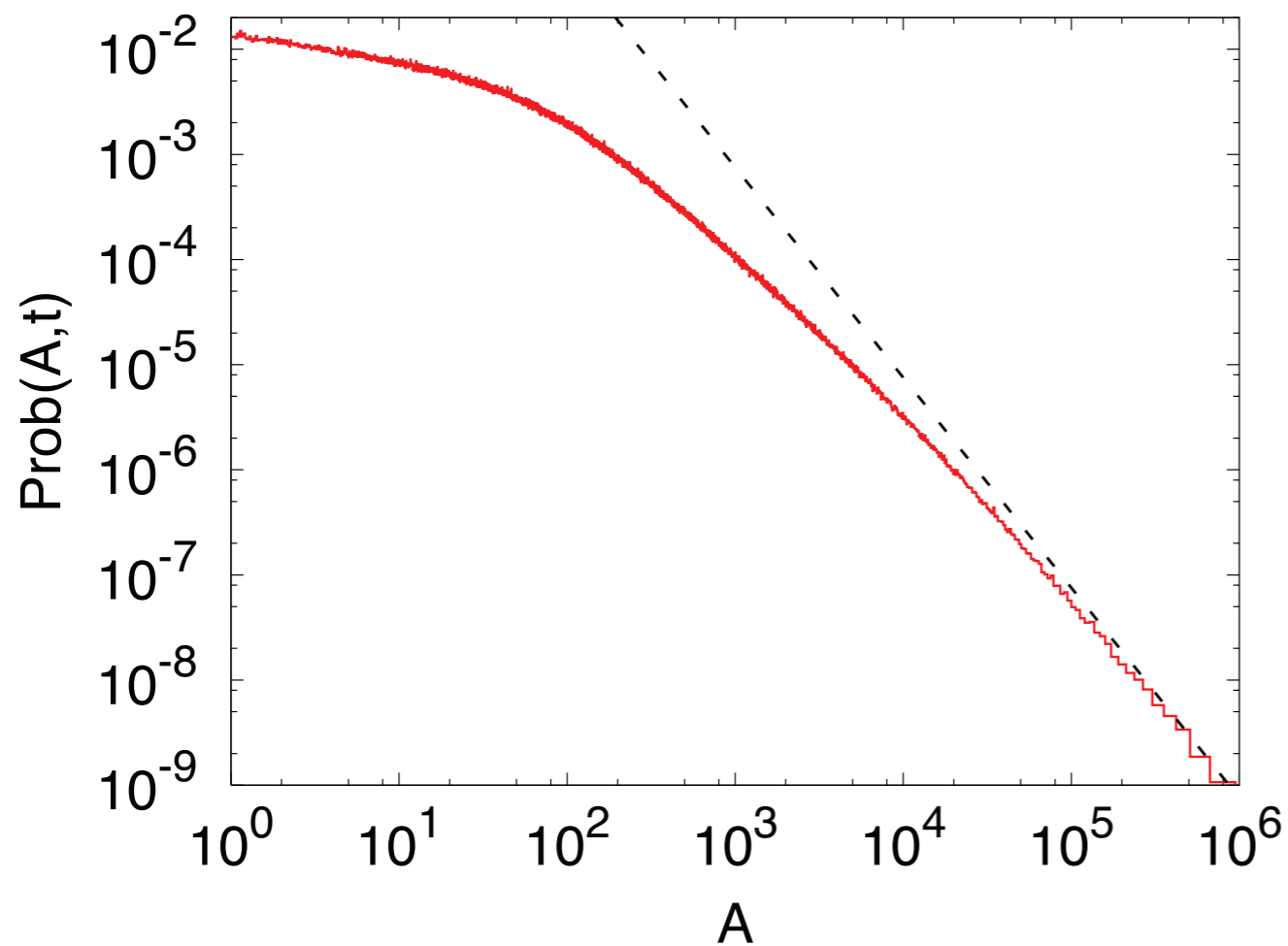


When $t \rightarrow \infty$ the perimeter remains finite, but the area diverges!

How it is possible ? ... Fluctuations

$$\text{Prob}(A) \xrightarrow[A \rightarrow \infty]{t = \infty} \frac{24\pi D}{5\gamma} A^{-2}$$

$$\text{Prob}(L) \xrightarrow[L \rightarrow \infty]{t = \infty} L^{-3}$$



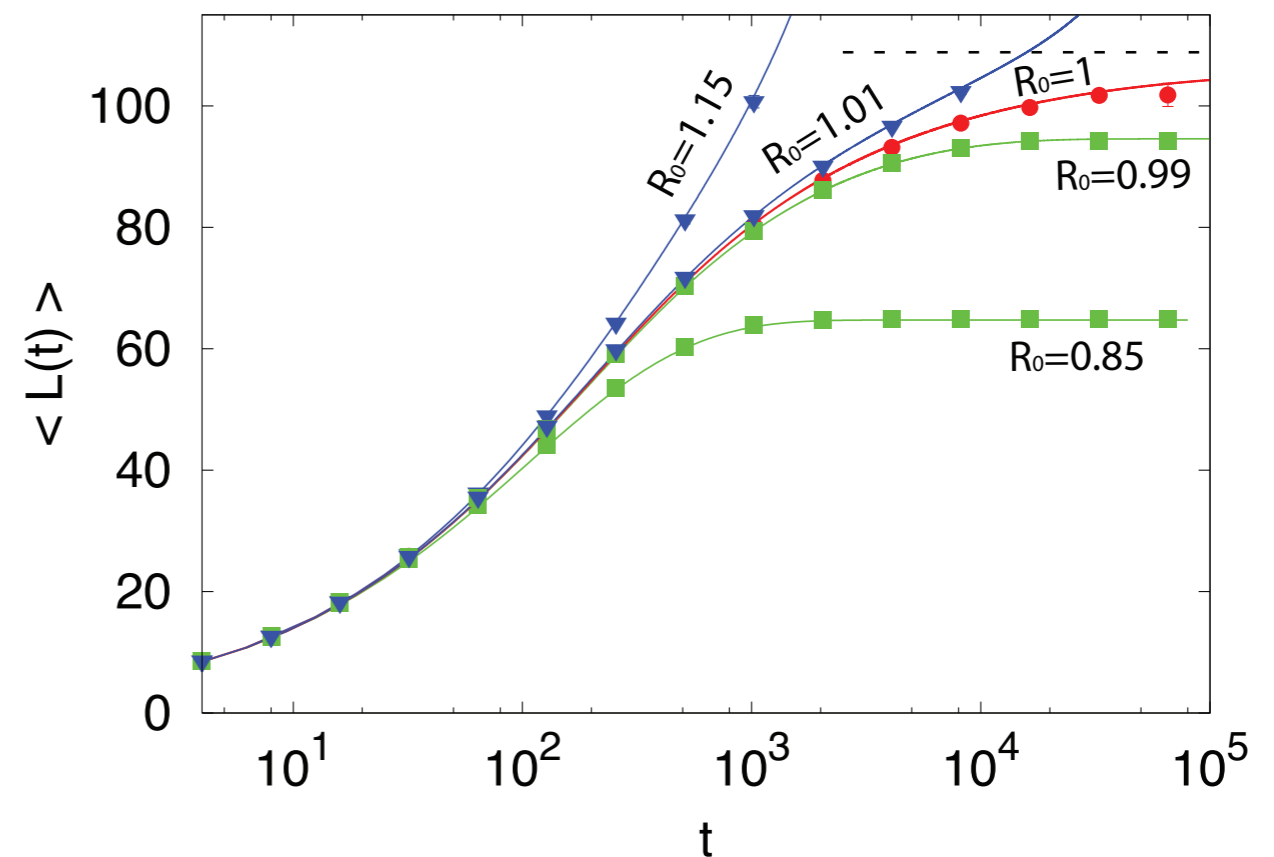
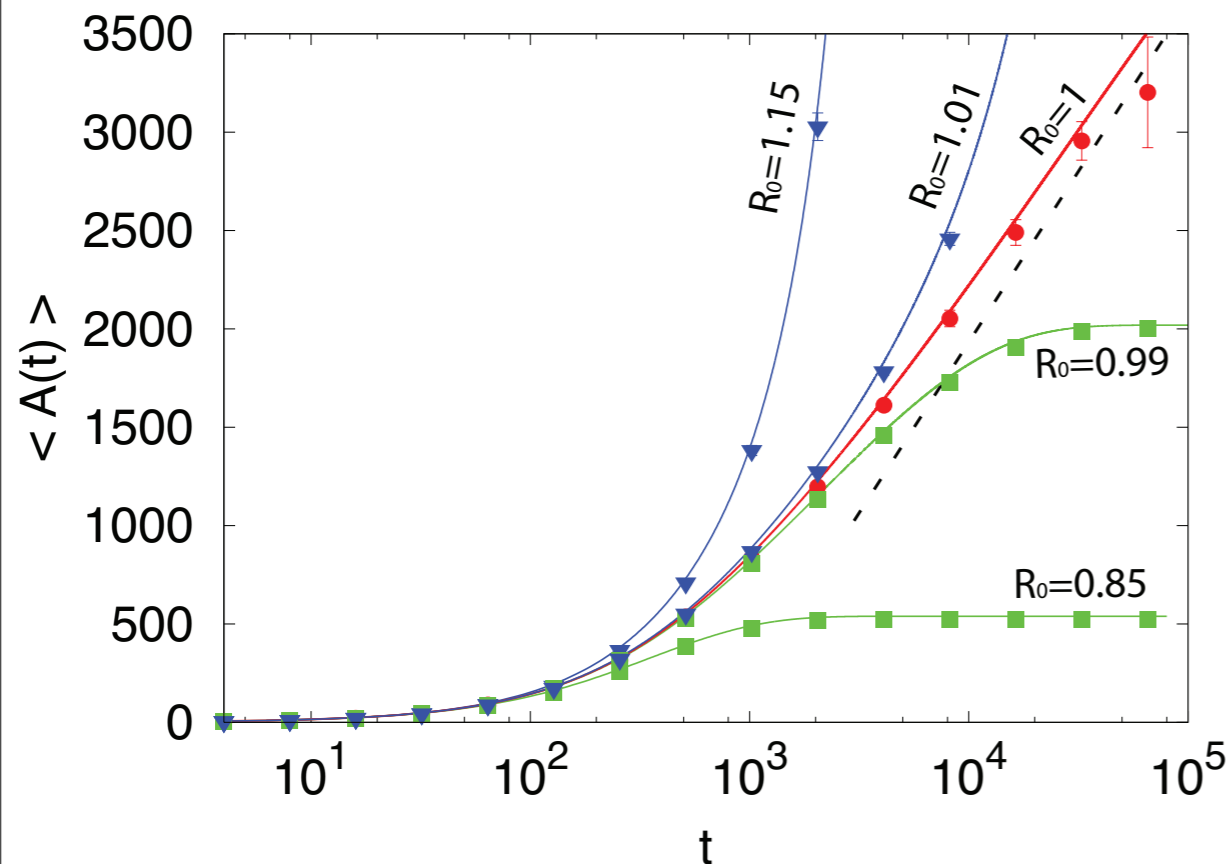
Out of criticality

When $R_0 \neq 1$, characteristic time $t^* \sim |R_0 - 1|^{-1}$.

For times $t < t^*$ the epidemic behaves as in the critical regime.

In the *subcritical* regime, for $t > t^*$ the epidemic goes to extinction.

In the *supercritical* regime, with probability $1 - 1/R_0$ epidemic explodes.

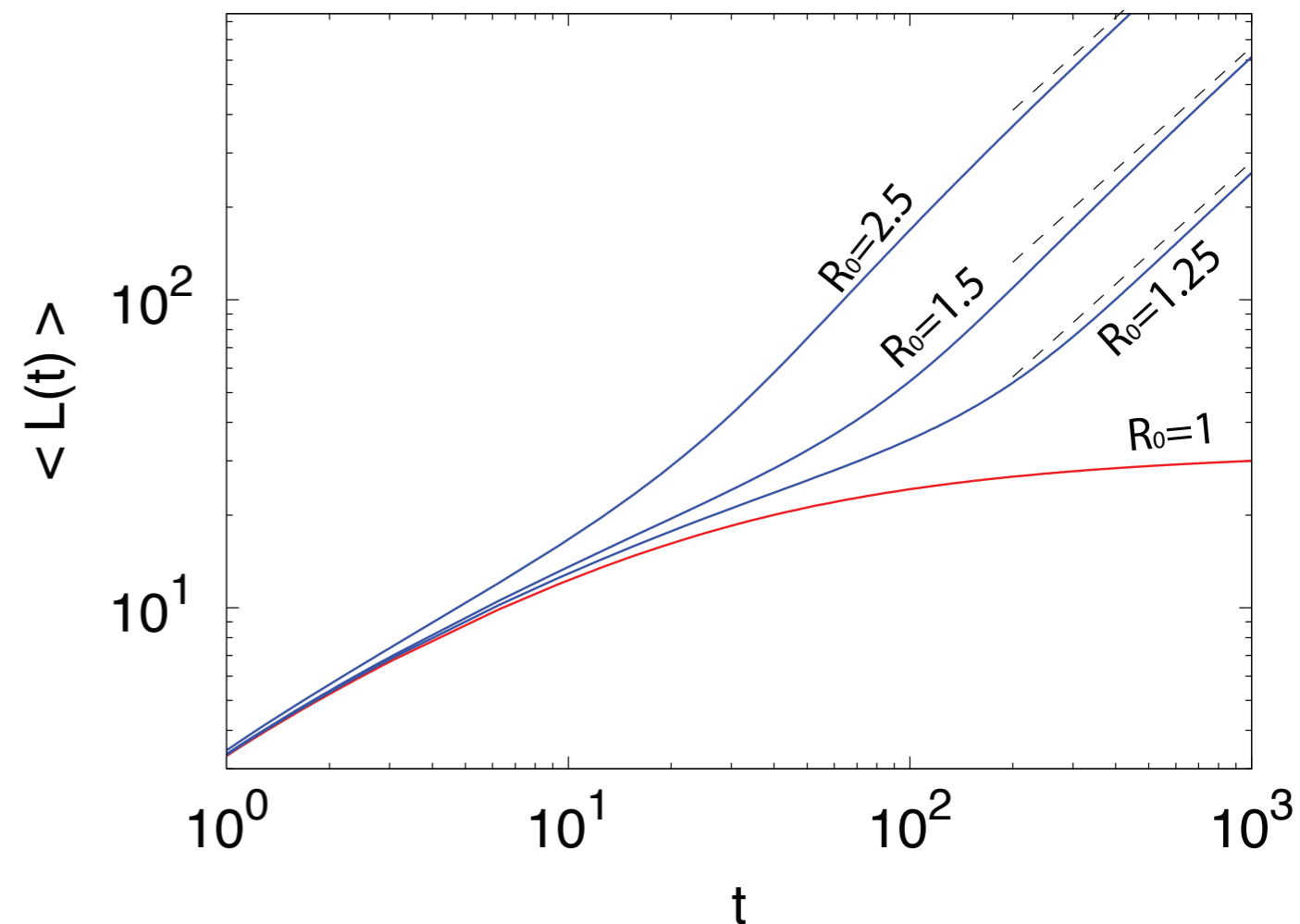
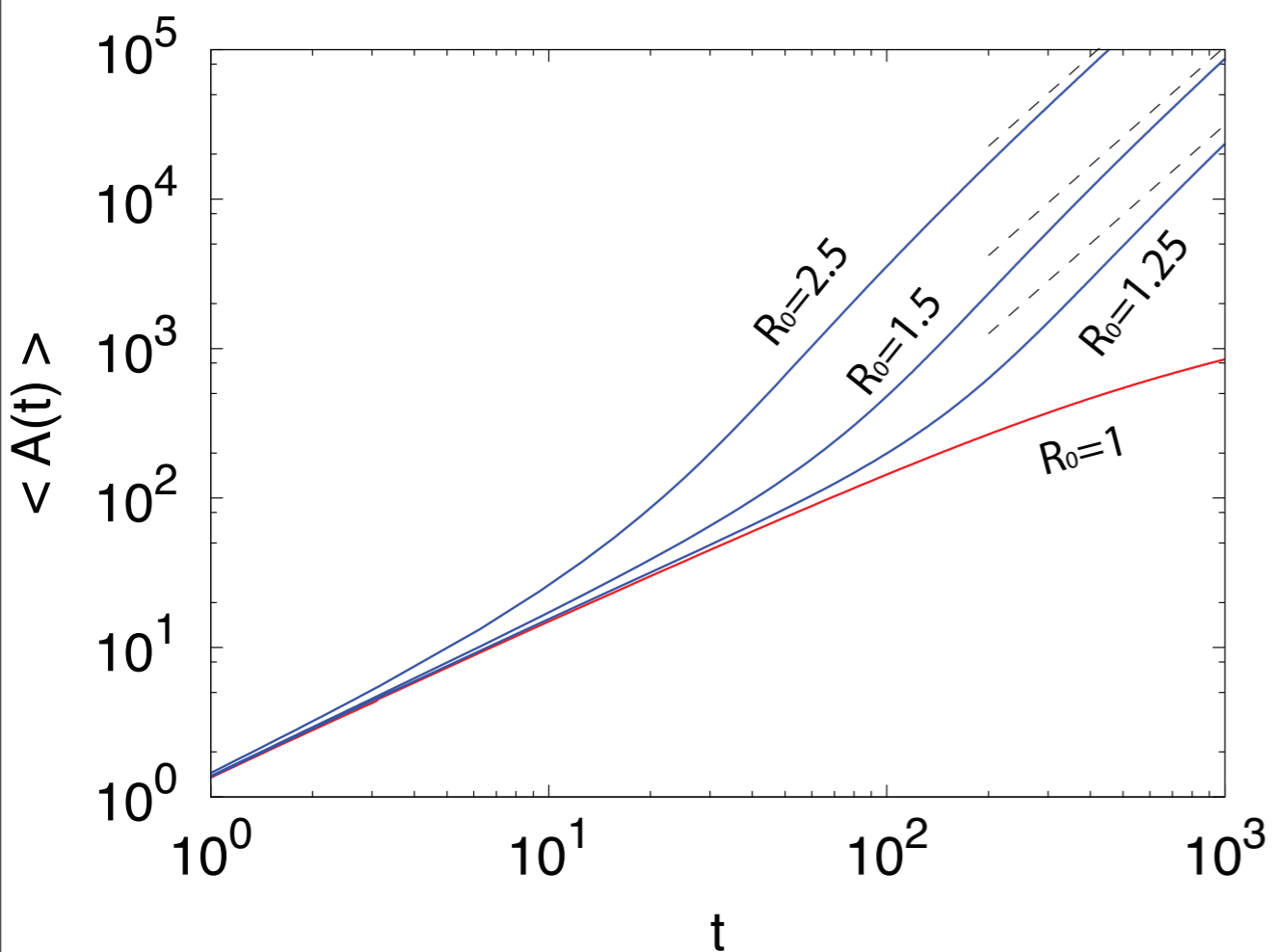


Supercritical

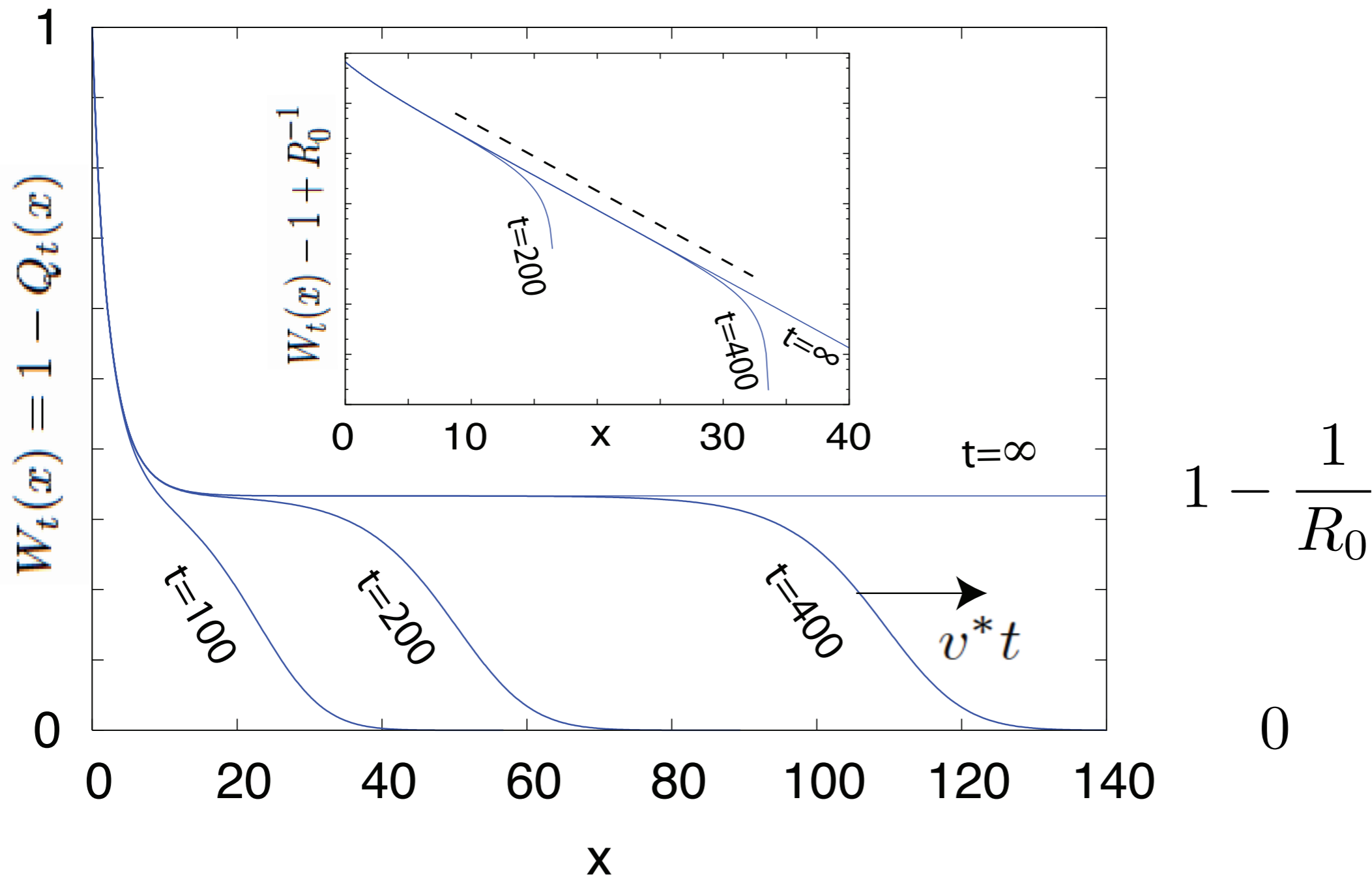
$$\langle L(t \gg t^*) \rangle = 4\pi \left(1 - \frac{1}{R_0}\right) \sqrt{D \gamma (R_0 - 1)} t$$

$$\langle A(t \gg t^*) \rangle = 4\pi \left(1 - \frac{1}{R_0}\right) D \gamma (R_0 - 1) t^2$$

$$t^* \sim |R_0 - 1|^{-1}$$



$$\frac{\partial}{\partial t} W = D \frac{\partial^2}{\partial x_m^2} W + \gamma(R_0 - 1)W - \gamma R_0 W^2$$



Traveling front solution

Conclusions:

- Branching Brownian motion with death as a model for the spatial extent of animal epidemics
- Using Cauchy Formulas we can map the convex hull problem in the extreme statistic of the 1-dimensional process
- Backward F-P equations for the extreme distributions
- Critical case has very large fluctuations
- Super Critical case: traveling front solution