

Predator-Prey Lecture 10



PHYS 246 class 10
Fall 2025
J Noronha-Hostler

<https://jnoronhahostler.github.io/IntroductionToComputationalPhysics/intro.html>

Announcements

- I'll need to leave early this week and next due to some family constraints.
- I changed some notation's on today's assignment earlier today, please use the current version!
-

Final Guidance

- Teams up to 4 people
- Small coding project
- Presentations ~5 minutes
- Graded on:
 - Presentation skills: Intro, methodology, results, conclusions
 - Working code
 - Shared work responsibilities
- Date/Time: Wednesday 17 December at 1:30 PM
- Location: 32 psychology

Fox $F(t)$ vs bunnies $B(t)$

Dynamics over time

- Prey reproduce: each prey produces $\frac{\Delta B(t)}{B(t)} = \alpha \Delta T$ offspring

$$\frac{dB(t)}{dt} = \alpha B(t)$$

- Predators die: each prey dies as $\frac{\Delta F(t)}{F(t)} = -\gamma \Delta t$

$$\frac{dF(t)}{dt} = -\gamma F(t)$$

- Predators eat prey:

- Rate Predators run into prey: $F(t)B(t)$
- Prey gets eaten: $\Delta B(t) = -\beta F(t)B(t)$
- Predators eats prey, not hungry, will reproduce: $\Delta F(t) = \delta F(t)B(t)$

Fox $F(t)$ vs bunnies $B(t)$

Dynamics over time

- Prey reproduce: each prey produces $\frac{\Delta B(t)}{B(t)} = \alpha \Delta T$ offspring $\frac{dB(t)}{dt} = \alpha B(t)$
- Predators die: each prey dies as $\frac{\Delta F(t)}{F(t)} = -\gamma \Delta t$
$$\frac{dF(t)}{dt} = -\gamma F(t)$$
- Predators eat prey:
 - Rate Predators run into prey: $F(t)B(t)$
 - Prey gets eaten: $\frac{dB(t)}{dt} = -\beta F(t)B(t)$
 - Predators eats prey, not hungry, will reproduce: $\frac{dF(t)}{dt} = \delta F(t)B(t)$

Combine all equations

Entire ecosystem

- Change of prey: $\frac{dB(t)}{dt} = \alpha B(t) - \beta F(t)B(t)$
- Change of predators: $\frac{dF(t)}{dt} = \delta F(t)B(t) - \gamma F(t)$
- Requires initial conditions: $F_0 = F(0), \quad B_0 = B(0)$
- Defined by set of parameters: $\{\alpha, \beta, \delta, \gamma\}$
- Called Lotka-Volterra equations

Stationary points

Places when total numbers don't change?

Original equation

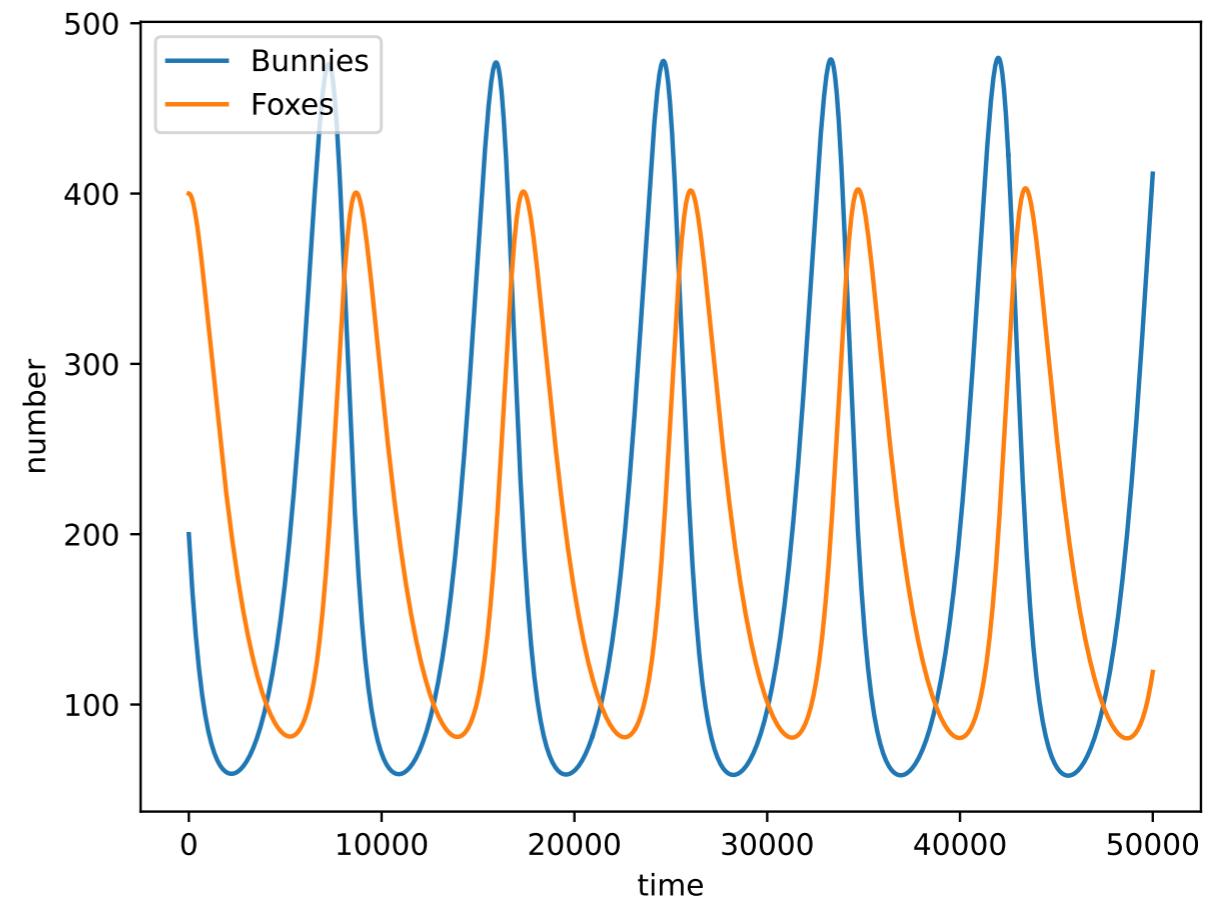
$$\frac{dB(t)}{dt} = \alpha B(t) - \beta F(t)B(t)$$

Euler method

$$B(t + \Delta t) \simeq B(t) + \left(\frac{dB}{dt} \right) \Delta t$$

Plug in original equation

$$B(t + \Delta t) \simeq B(t) + (\alpha B(t) - \beta F(t)B(t))\Delta t$$



Stochastic Method

Fractions of bunnies!



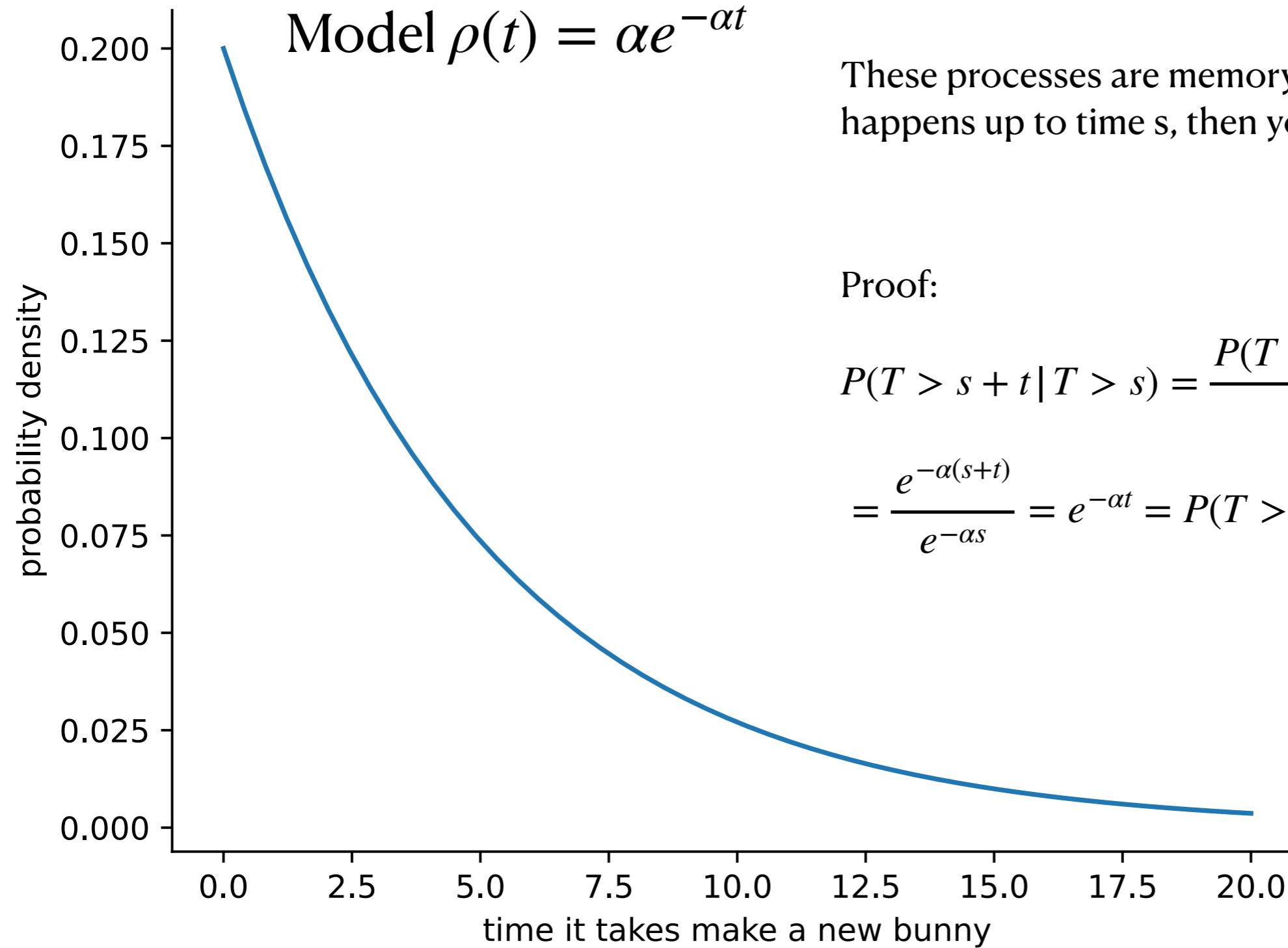
Lotka-Volterra equations are averaged 'mean-field' equations. They are easy to solve and simulate but average over a bunch of detail.

B and F are floats, they really represent sort of an average number. So you can have 36.232 bunnies in expectation.

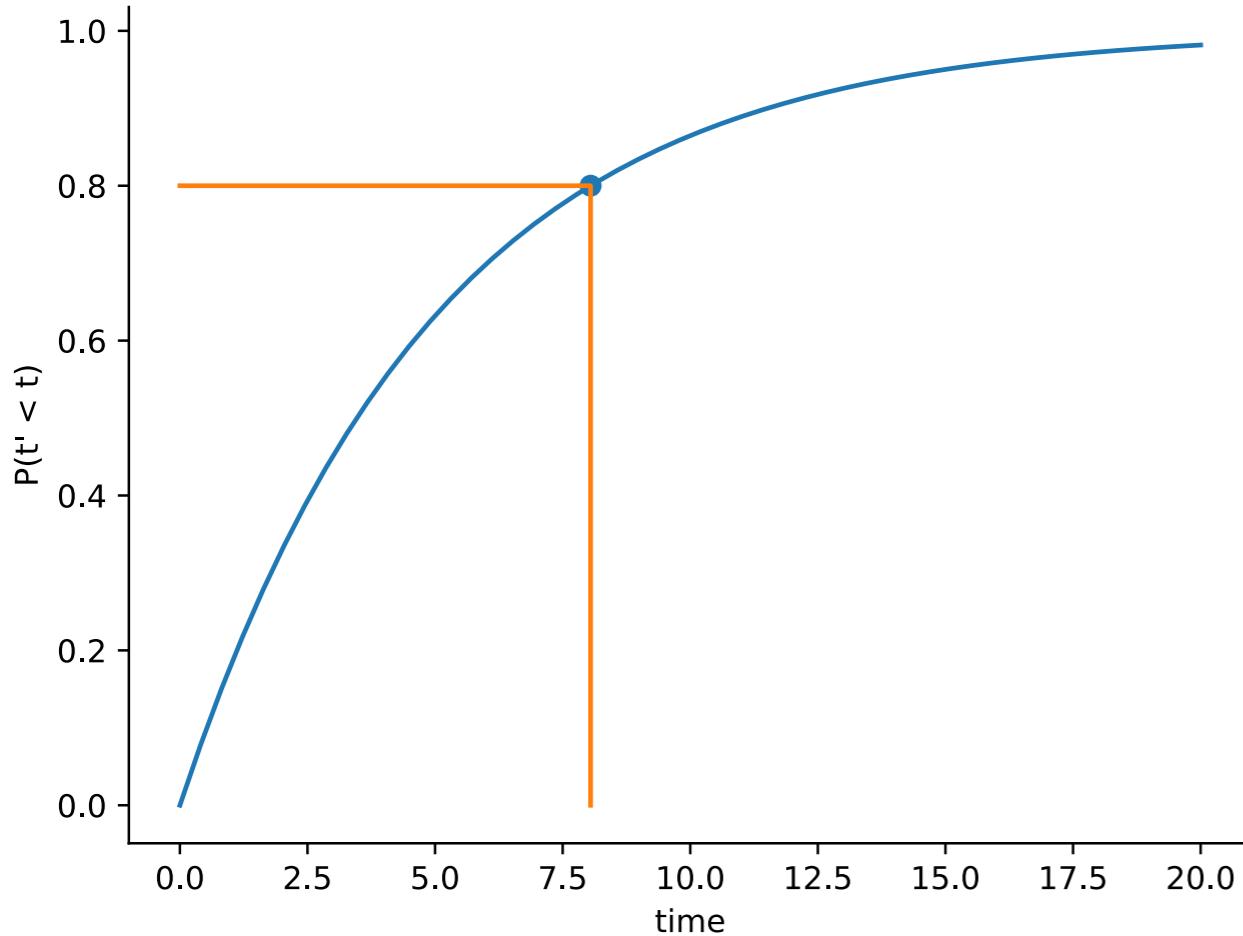
Reality has fluctuations! We can incorporate those by doing a harder calculation..

Exponential Distribution

Average time behavior (Poisson Process)



Sampling from an exponential distribution



Find Δt ,
time to make first bunny!

$$\int_0^t \alpha e^{-\alpha t'} dt' = 1 - e^{-\alpha t}$$

Generate random number χ

$$t' = -\ln(1 - \chi)/\alpha = -\ln(\chi)/\alpha$$

is sampled from the exponential.

Continuous time Monte Carlos

You have a bunch of Poisson processes with rate r_i .

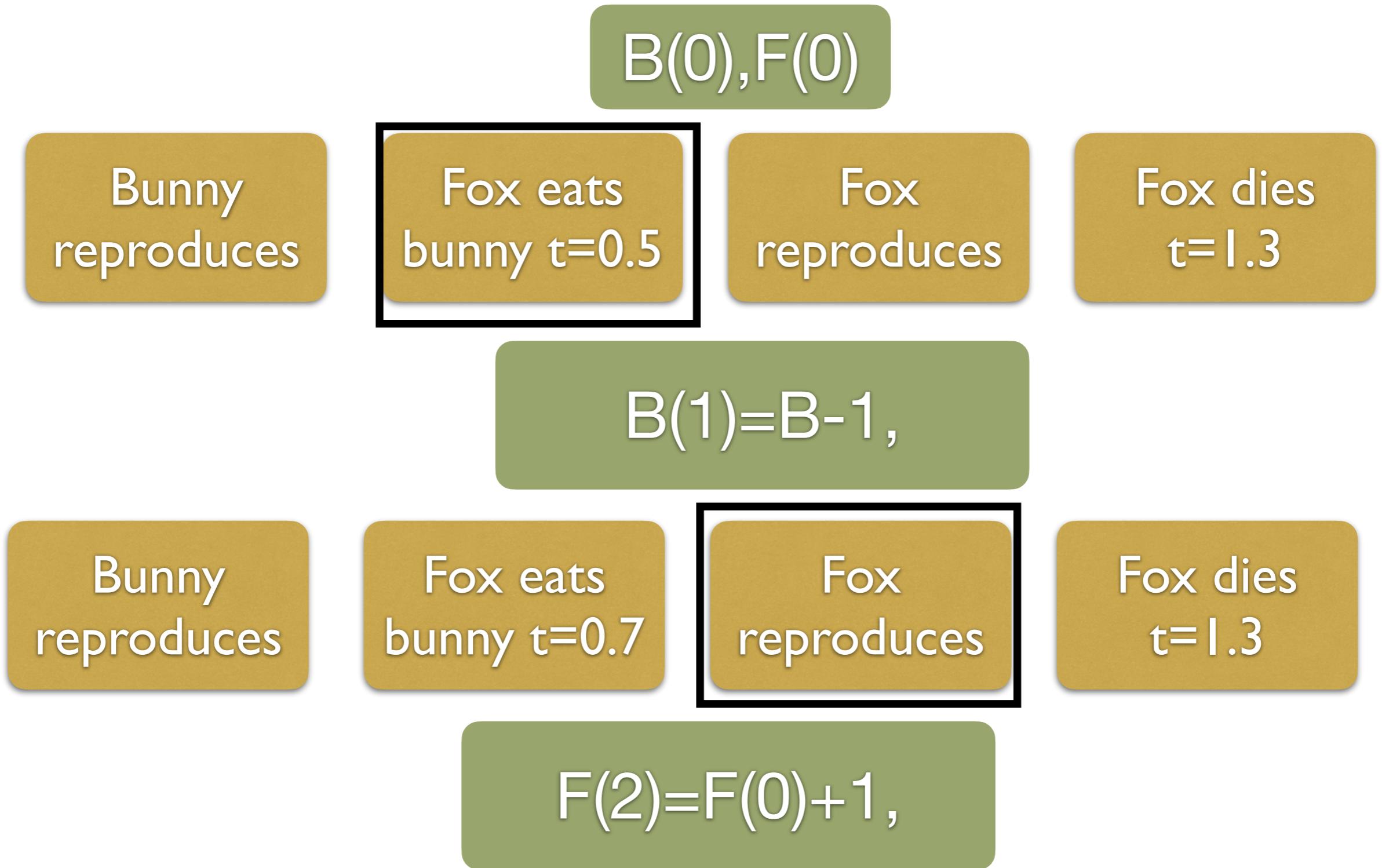
As long as nothing changes, each would happen at time

$$t_i = -\ln(\chi_i)/r_i$$

Algorithm:

- * the action that would happen first happens first.
- * Then the rates change
- * Memoryless -> recompute all rates and find the first action again

Continuous time Monte Carlos



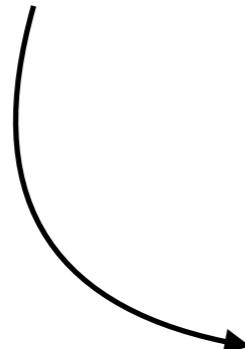
Hints for doing a Monte Carlo

```
t1 = -np.log(np.random.rand())/(alpha*B)
t2 = -np.log(np.random.rand())/(gamma*F)

if t1 < t2: # and t1 < t3 and t1 < t4
    B+=1
elif t2 < t3: # and t2 < t4
    F-=1|
```

Instead of this

Chooses which event happens first



```
'changes = [(1,0), (0,-1)]
rates = np.array([alpha*B, gamma*F])
ts = -np.log(np.random.rand(2))/rates
choose = np.argmax(ts)
B+=changes[choose][0]
F+=changes[choose][1]
time += ts[choose]
```

Try this