

Tau-Leaping - Speeding Stochastic Simulation

Applied Numerics in Systems Biology Seminar

Balduin Laubisch

Tau-Leaping Method

- simulate stochastic models faster
- simulate complex stochastic models in feasible time

SSA vs. Tau-Leaping vs. ODE

SSA

$$X_t = X_0 + \sum_j s_{.,j} \mathcal{P} \left(\int_0^t a_j(X_s) ds \right)$$

Tau-Leaping

$$X_{t+\Delta t} \approx X_t + \sum_j s_{.,j} \mathcal{P} (a_j(X_t) \Delta t)$$

ODE

$$\frac{dX_t}{dt} = \sum_j s_{.,j} \cdot a_j(X_t)$$

Tau-Leaping Implementation

```
def tau_leap_simulate(model, T, h):
    # population
    x = [model[0]]
    # propensity functions
    r = model[2]
    # stoichiometric matrix
    s = np.array(model[1])
    # time
    t = [0]
    while (t[-1] < T):
        # calculate reaction rates
        ai = r(x[-1])

        if (t[-1] + h > T):
            # last time step size
            h = T - t[-1]

        # change in x according to current propensity and tau leap condition
        k = np.random.poisson(np.array(ai) * h)
        dx = s.dot(k)

        # update time and population
        t.append(t[-1] + h)
        x.append(np.add(dx, x[-1]))

    return x, t
```

Experimental Setup

Assess Tau-Leaping method on different models and reproduce dynamics

- Birth-death process
- Banana / Boomerang model
- Schlögl model

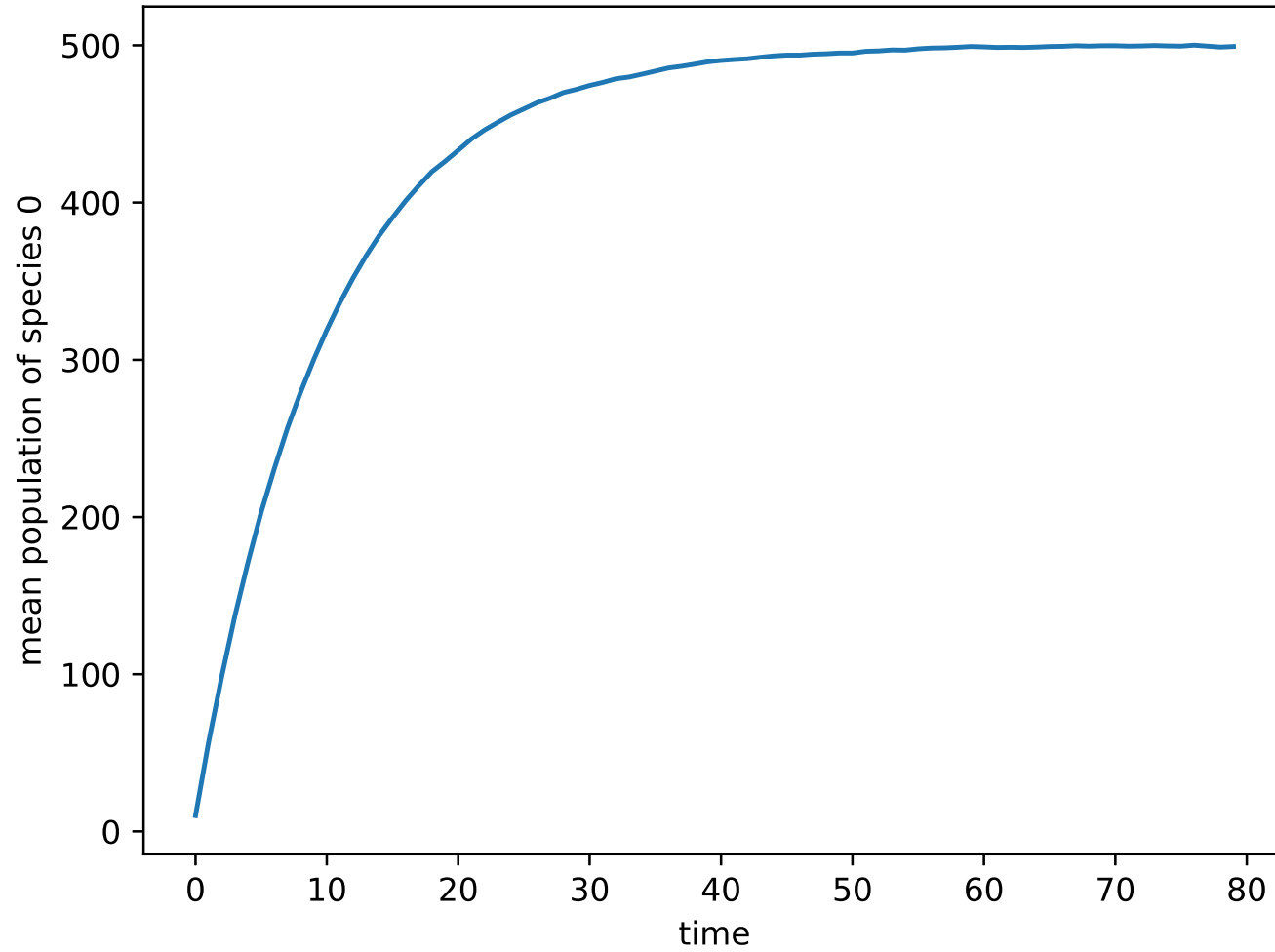
Birth-death process



- $X(0) = 10$
- stationary distribution at $\lambda/\beta = 500$

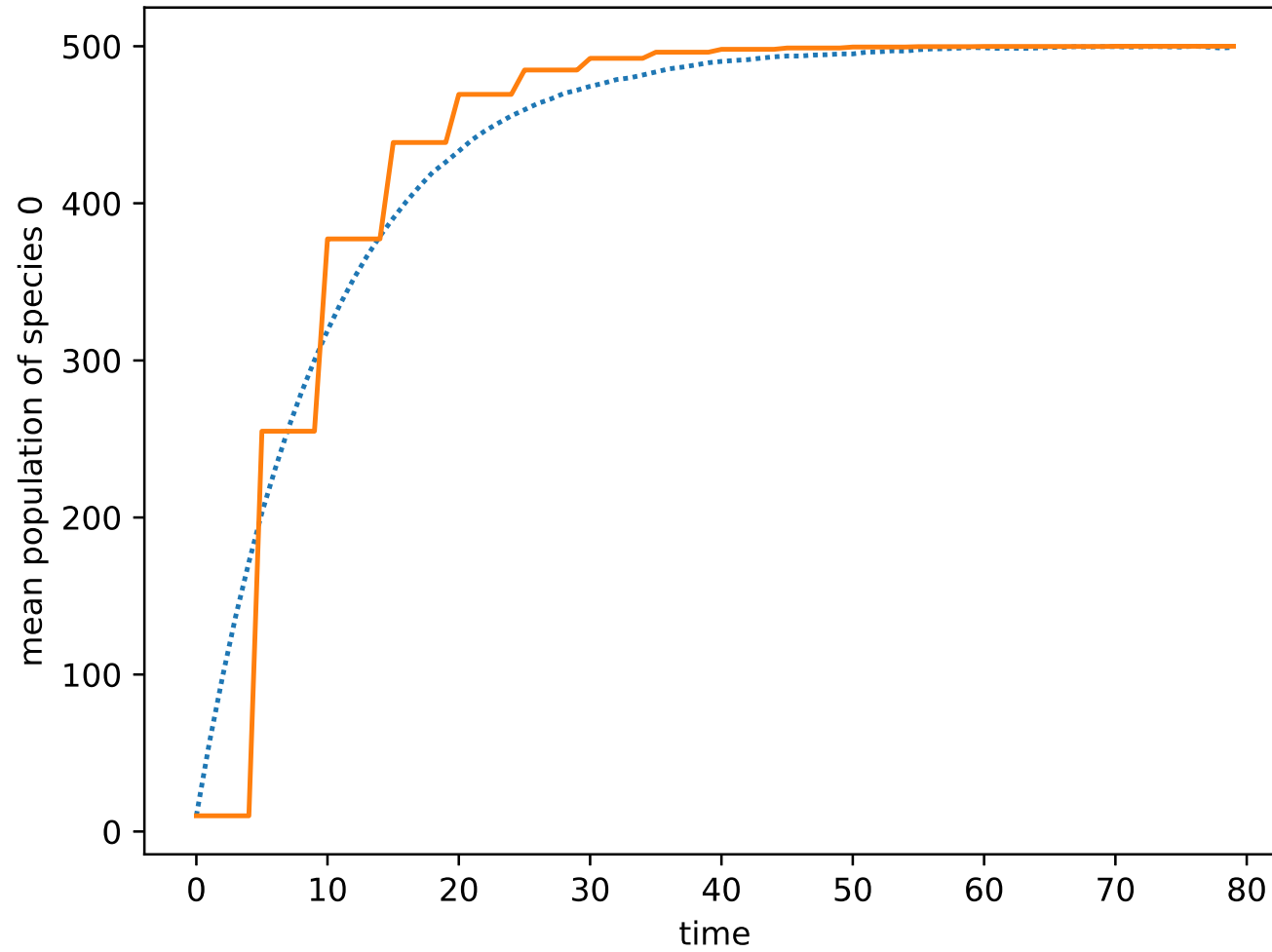
Birth-death stationary distribution

Birth-death Population
SSA ($\tau=0.010657\pm 0.010851$)
1000 Simulations



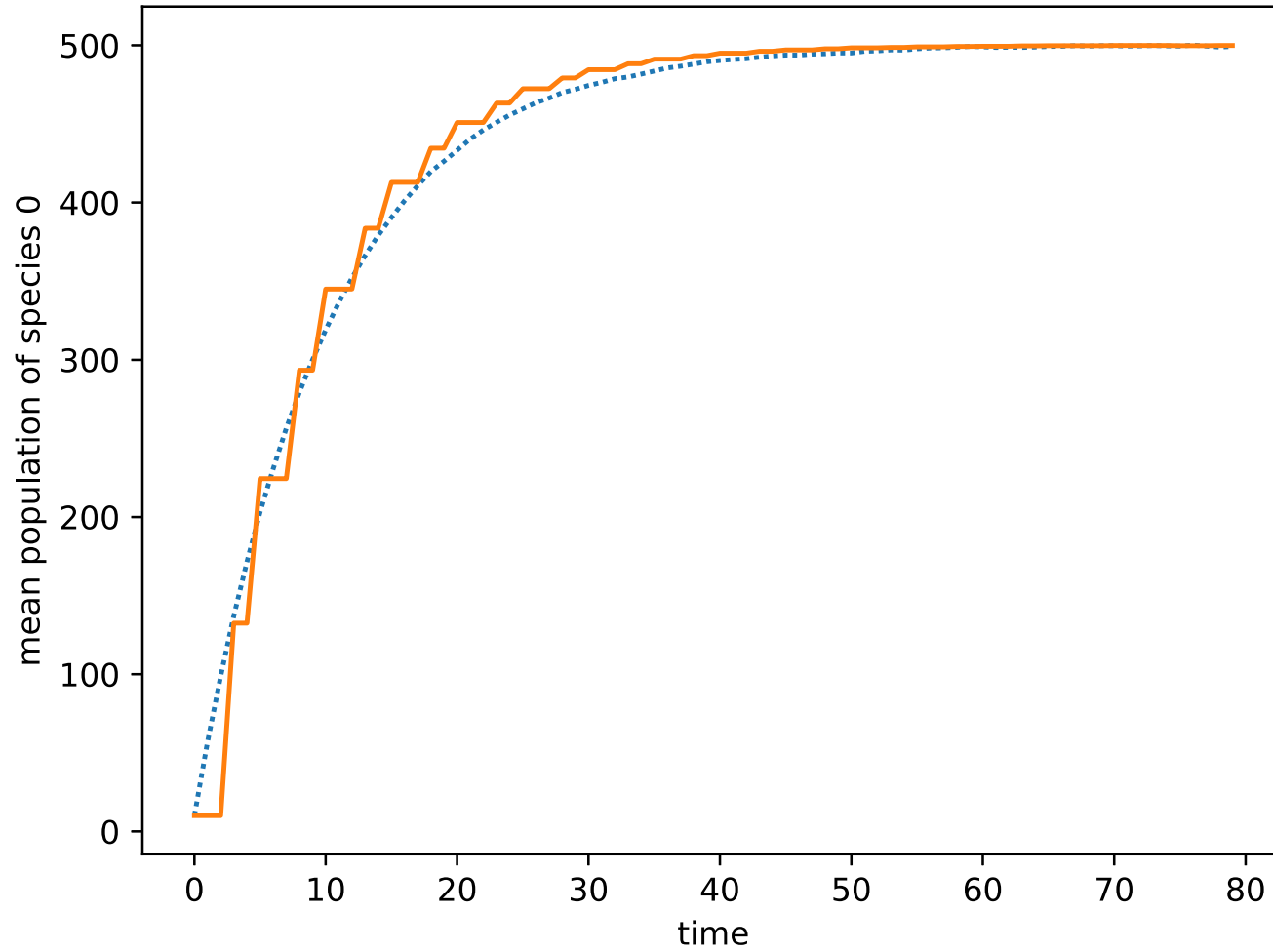
Birth-death stationary distribution

Birth-death Population
Tau-Leap ($h=5.000000$, $\tau=5.000000 \pm 0.000000$)
100000 Simulations



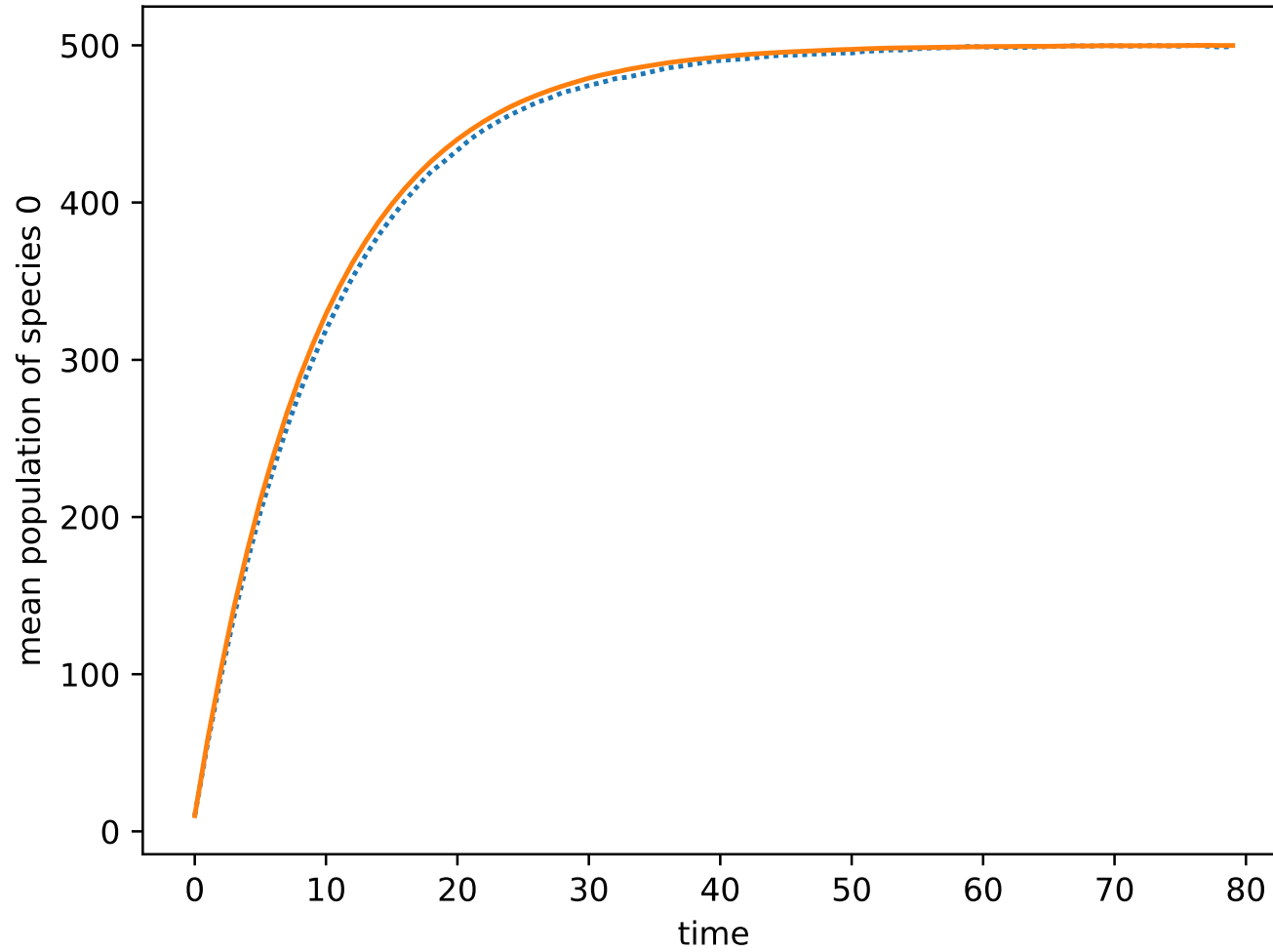
Birth-death stationary distribution

Birth-death Population
Tau-Leap ($h=2.500000$, $\tau=2.500000\pm 0.000000$)
100000 Simulations

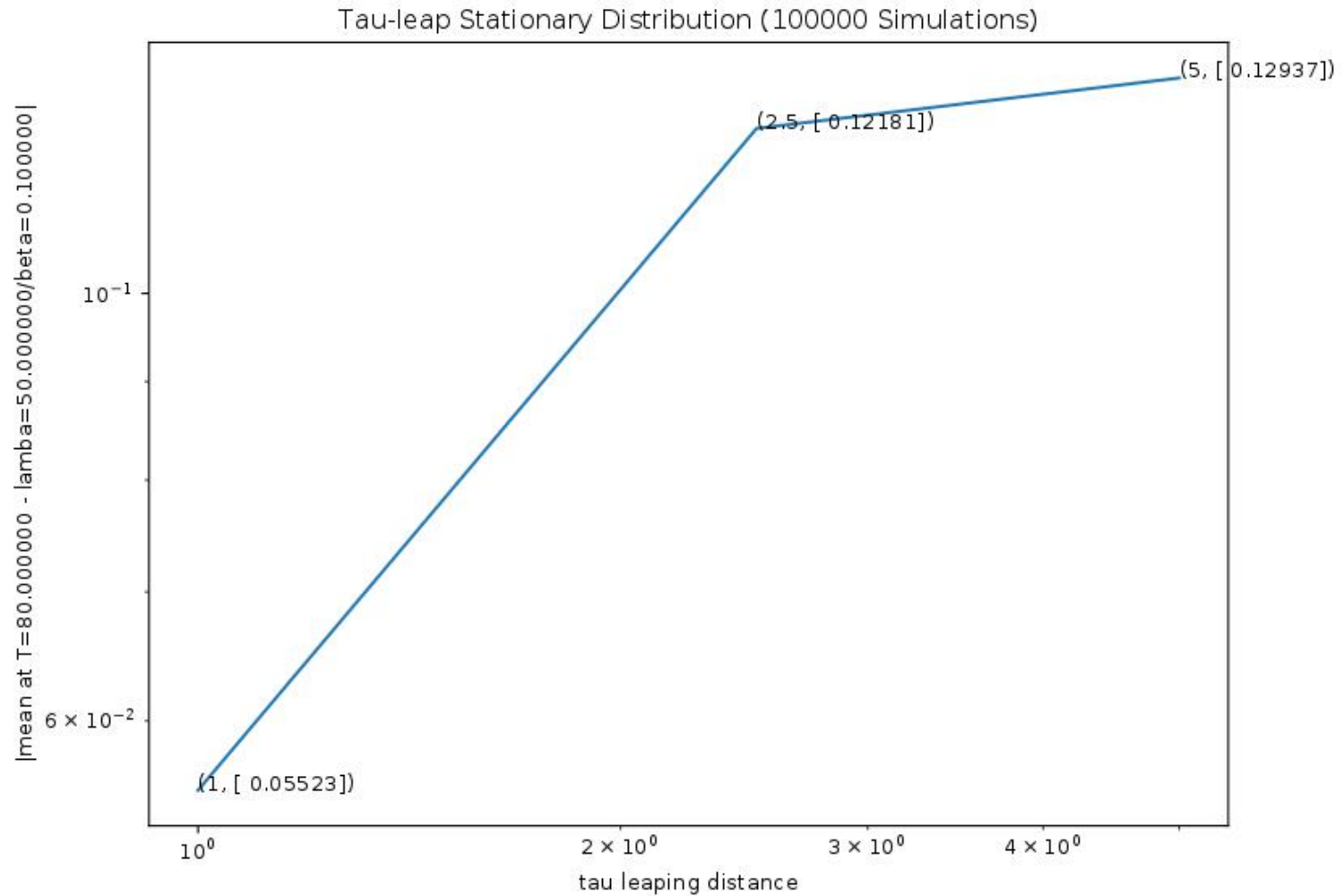


Birth-death stationary distribution

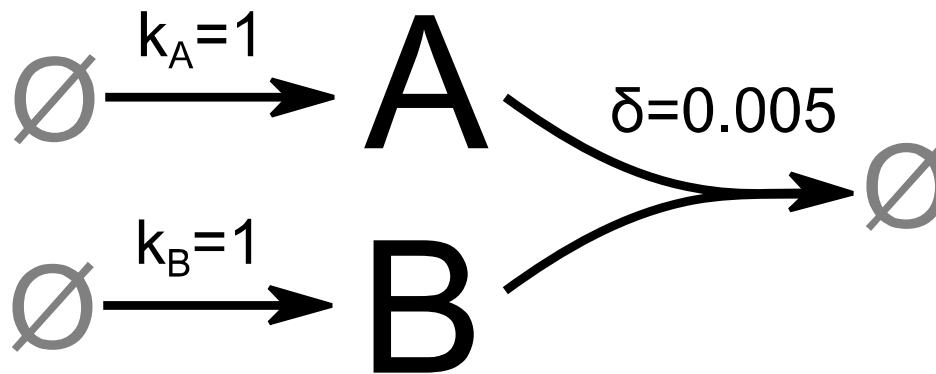
Birth-death Population
Tau-Leap ($h=1.000000$, $\tau=1.000000\pm 0.000000$)
100000 Simulations



Birth-death stationary distribution



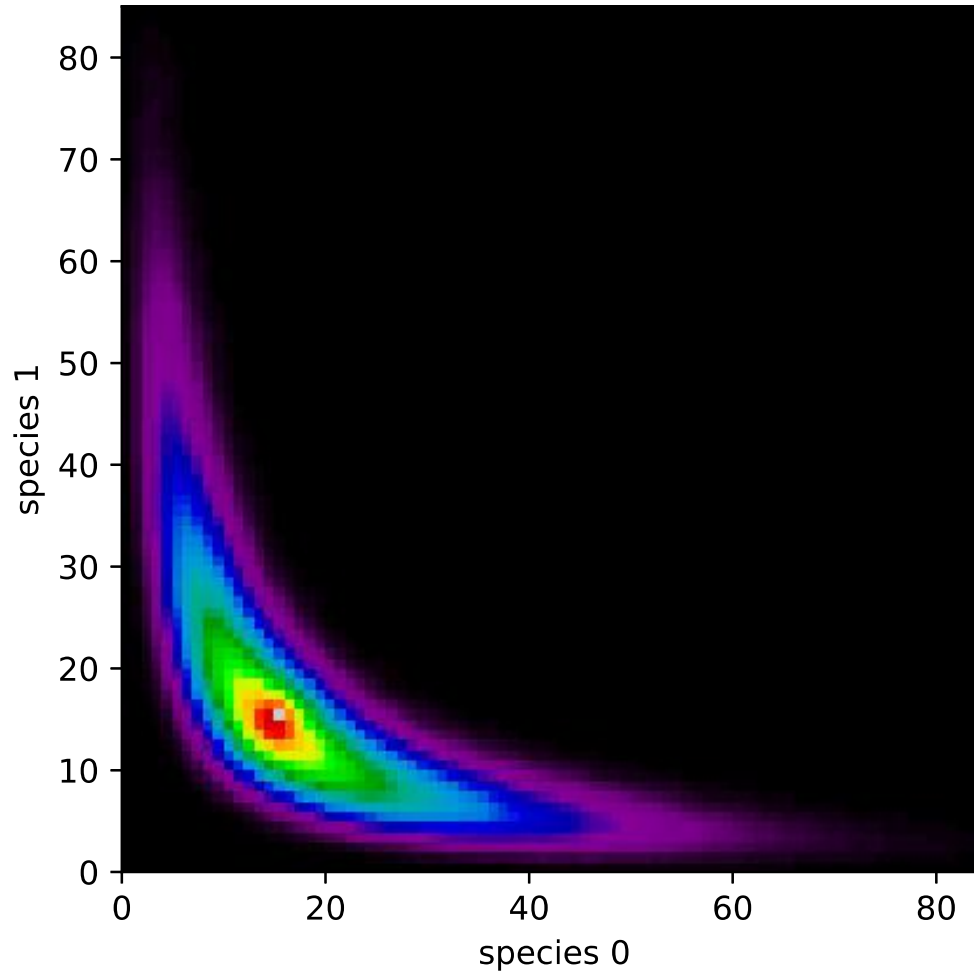
Banana / Boomerang model



- $A(0) = B(0) = 10$
- increasing population for A or B only mutually exclusive

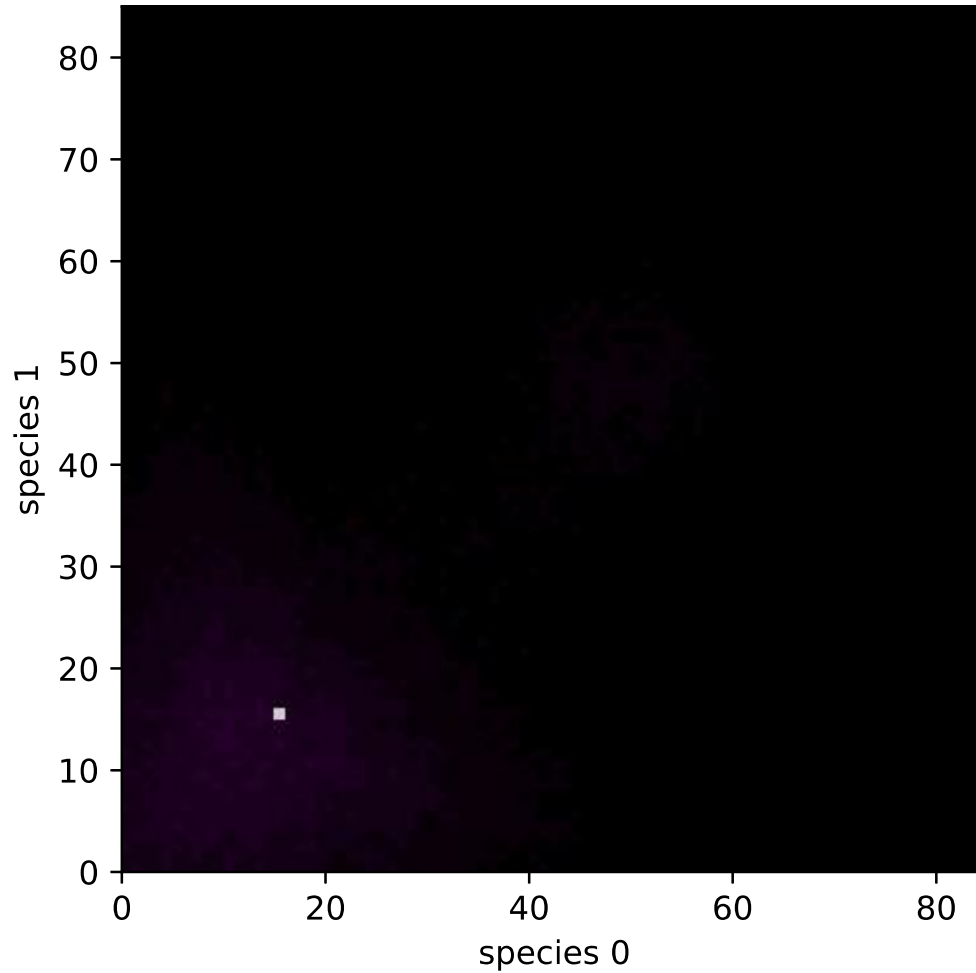
Bending the Banana

Phaseprobability
SSA $\tau=0.334383\pm 0.339502$
Banana (5000 Simulations)

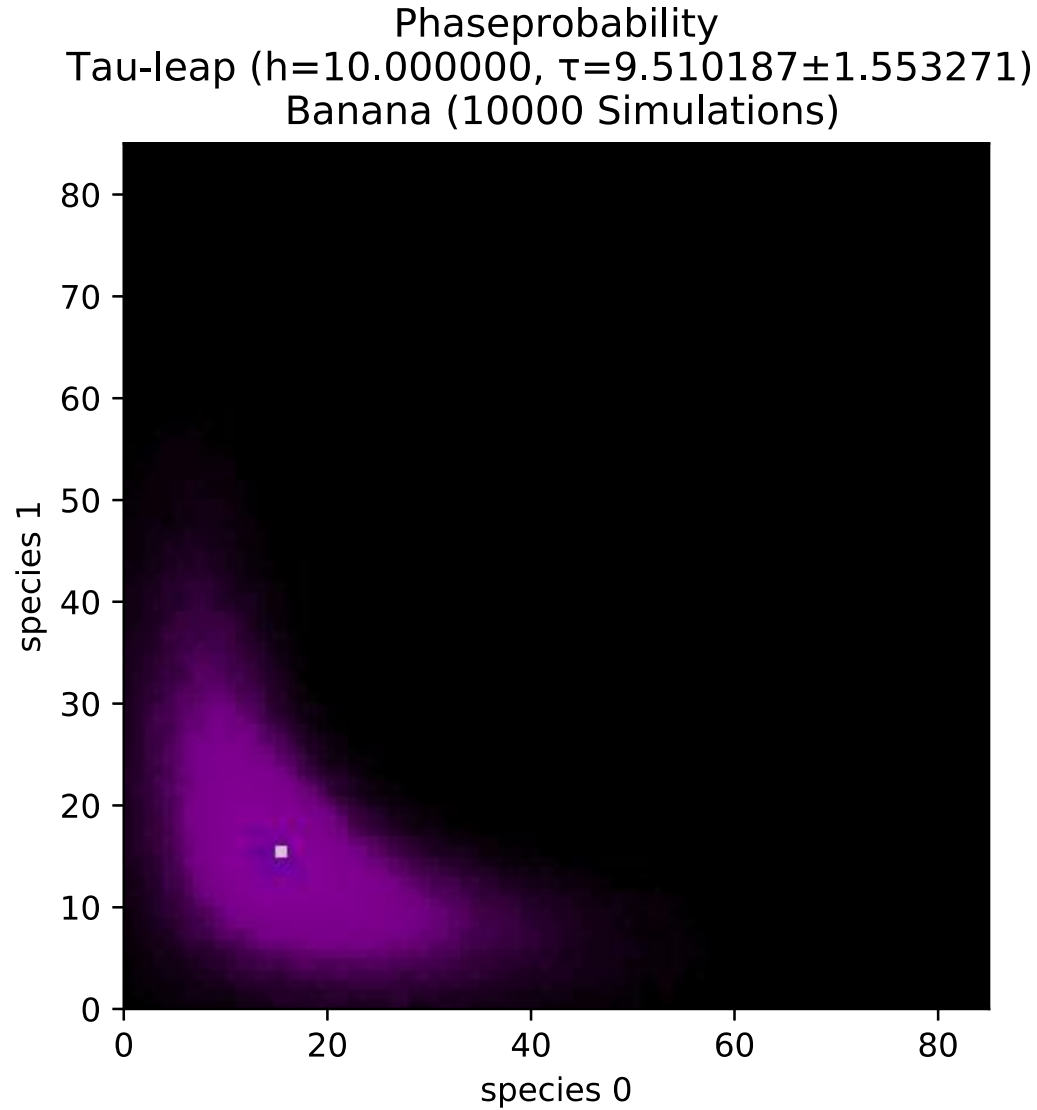


Bending the Banana

Phaseprobability
Tau-leap ($h=50.000000$, $\tau=25.671700 \pm 21.203371$)
Banana (10000 Simulations)

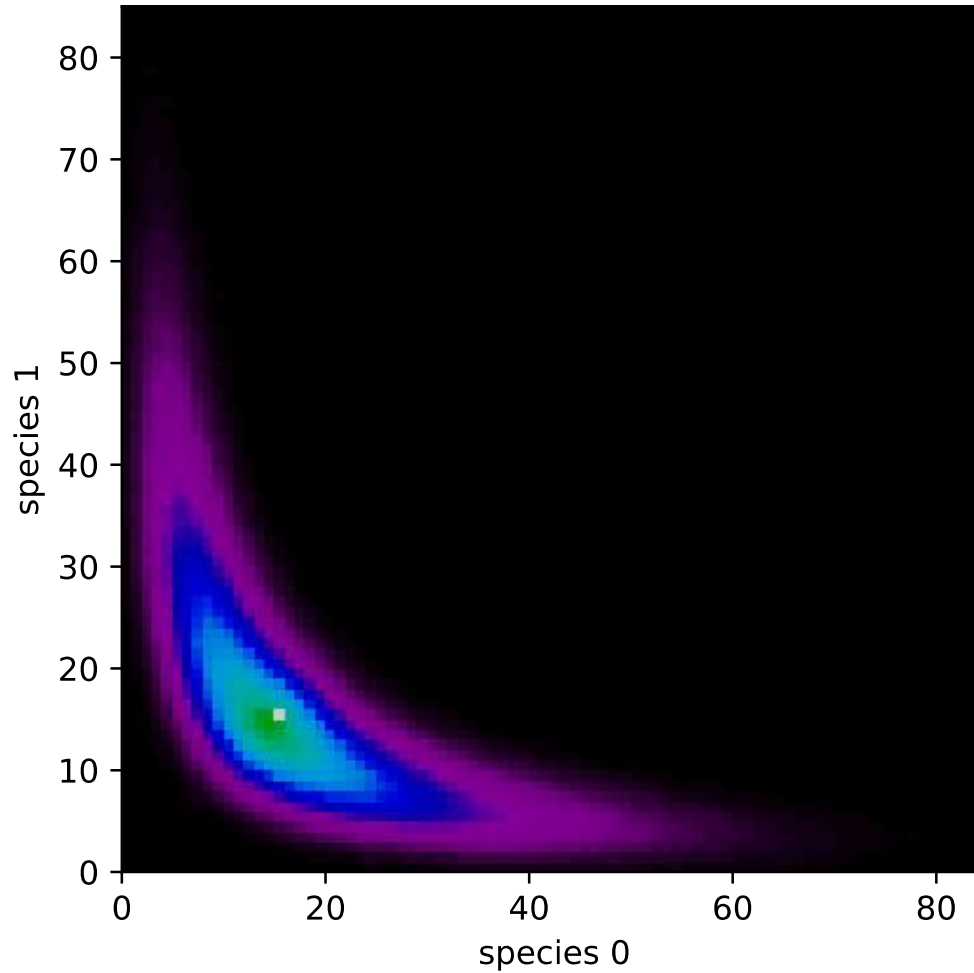


Bending the Banana



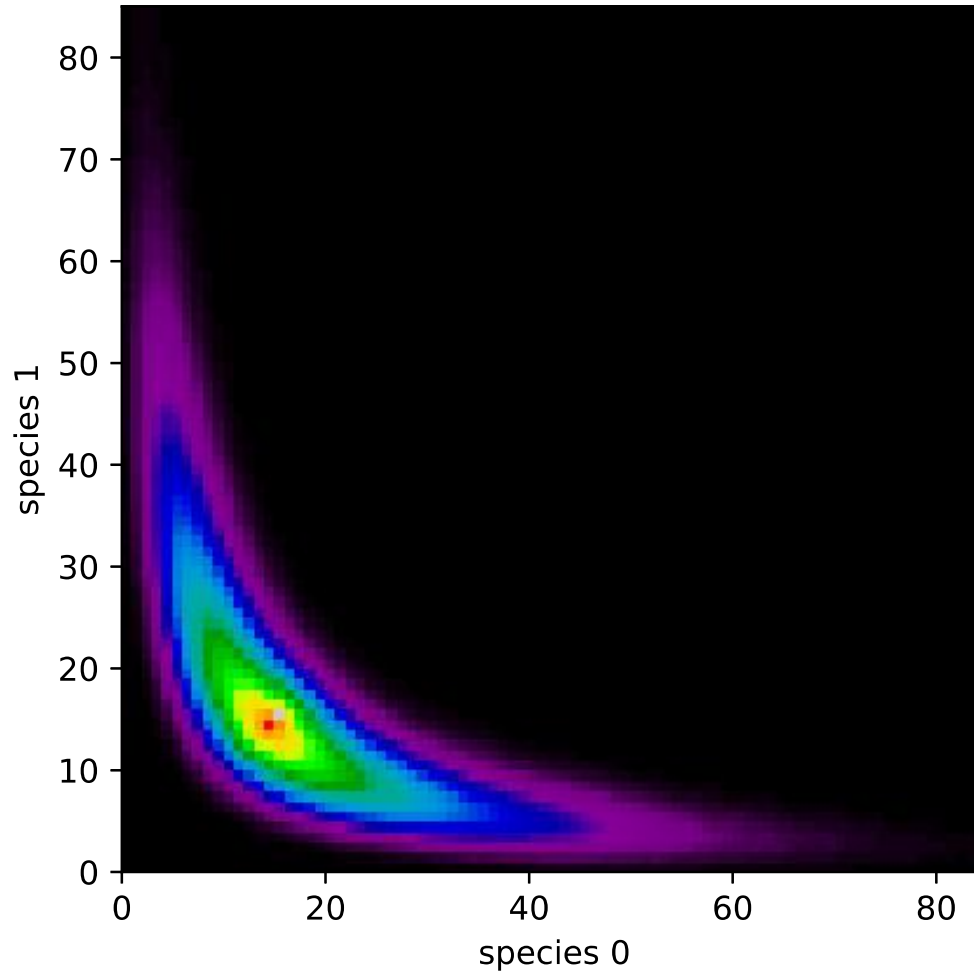
Bending the Banana

Phaseprobability
Tau-leap ($h=2.500000$, $\tau=2.494563 \pm 0.082507$)
Banana (10000 Simulations)

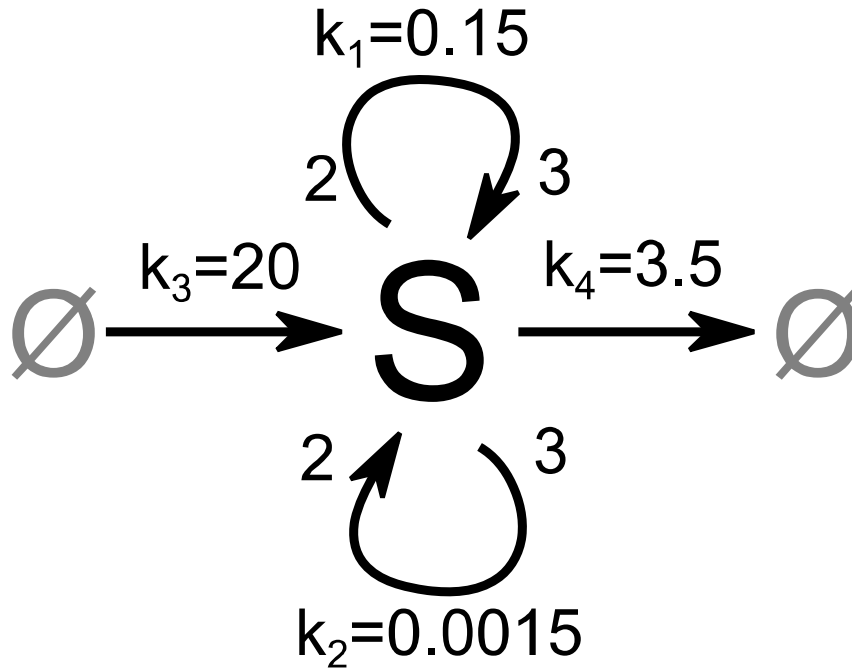


Bending the Banana

Phaseprobability
Tau-leap ($h=0.500000$, $\tau=0.499953\pm 0.003413$)
Banana (10000 Simulations)



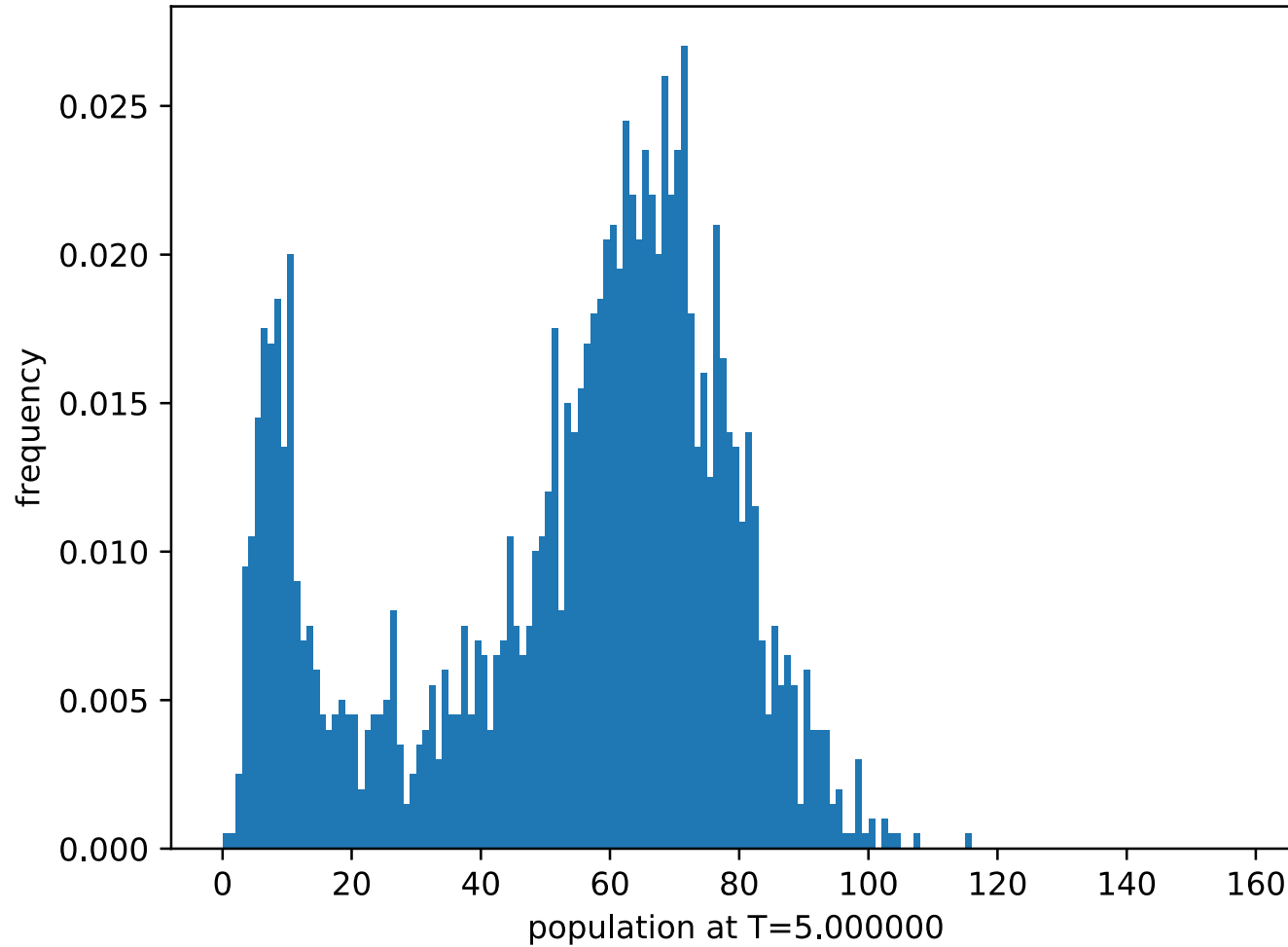
Schlögl model



- $S(0) = 40$
- bistability in time

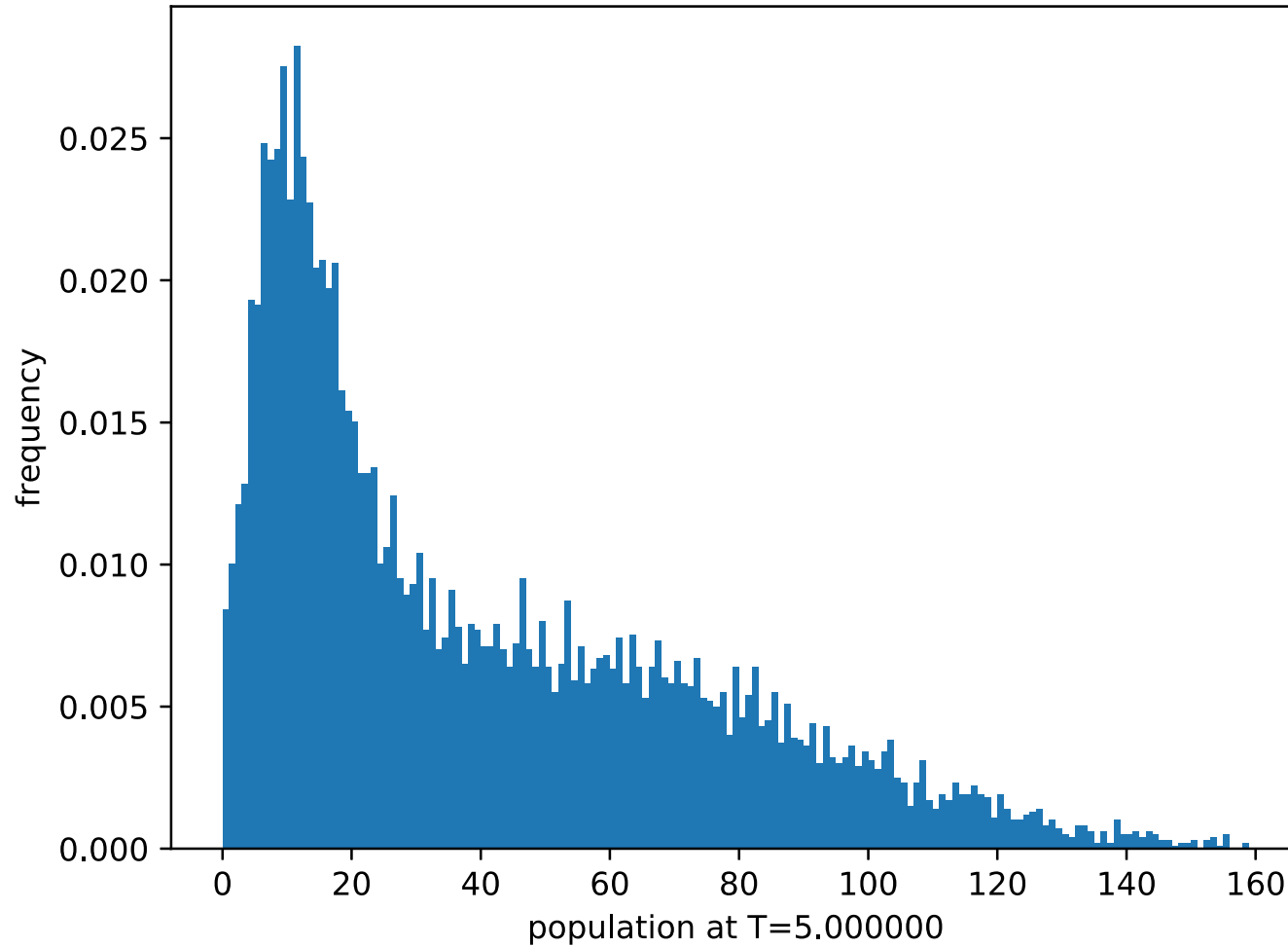
Schlögl Bistability

Population Histogram
SSA $\tau=0.000978\pm 0.002264$
Schlögl (2000 Simulations)



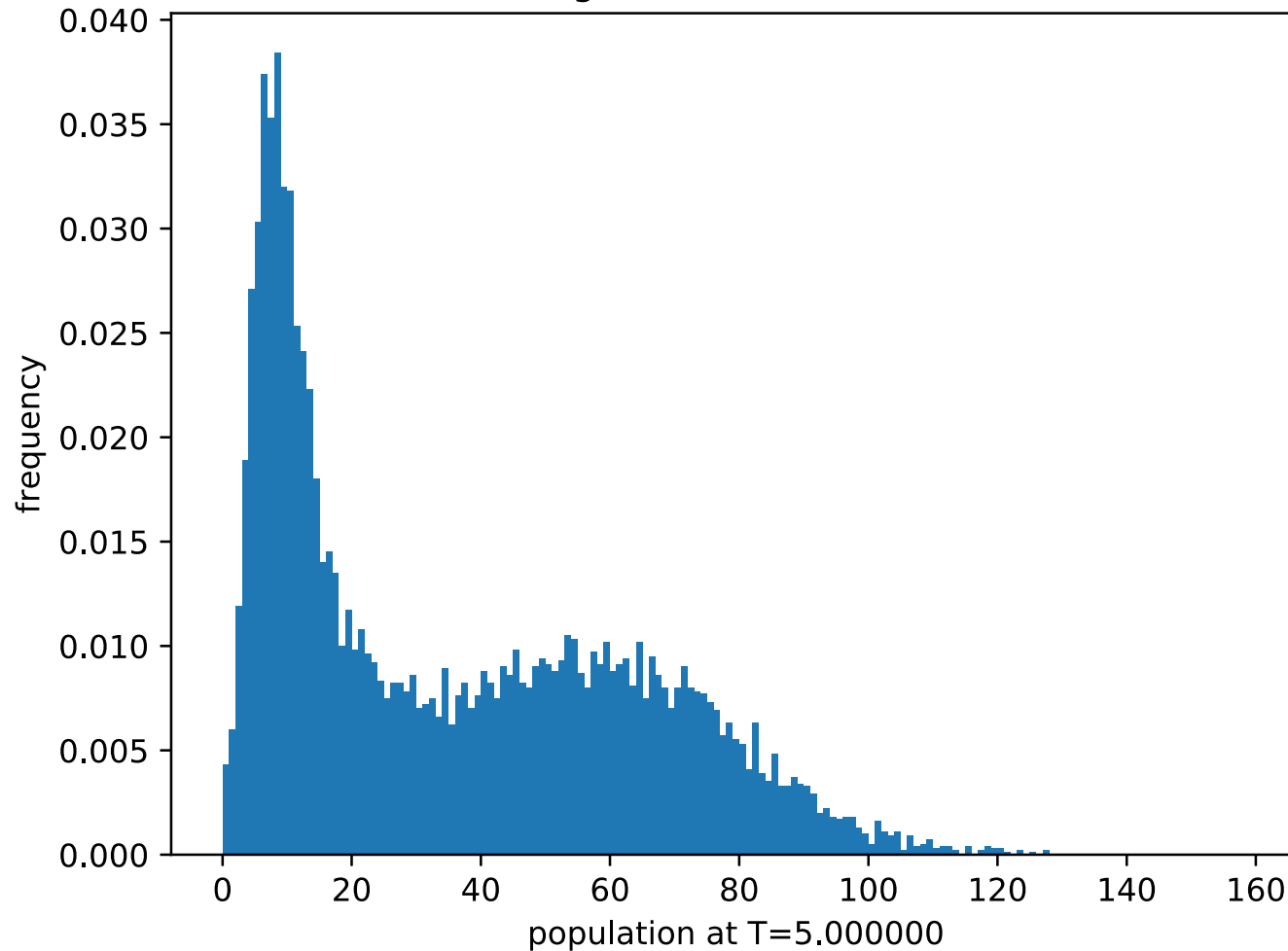
Schlögl Bistability

Population Histogram
Tau-leap ($h=1.000000$, $\tau=0.817688\pm 0.304819$)
Schlögl (10000 Simulations)



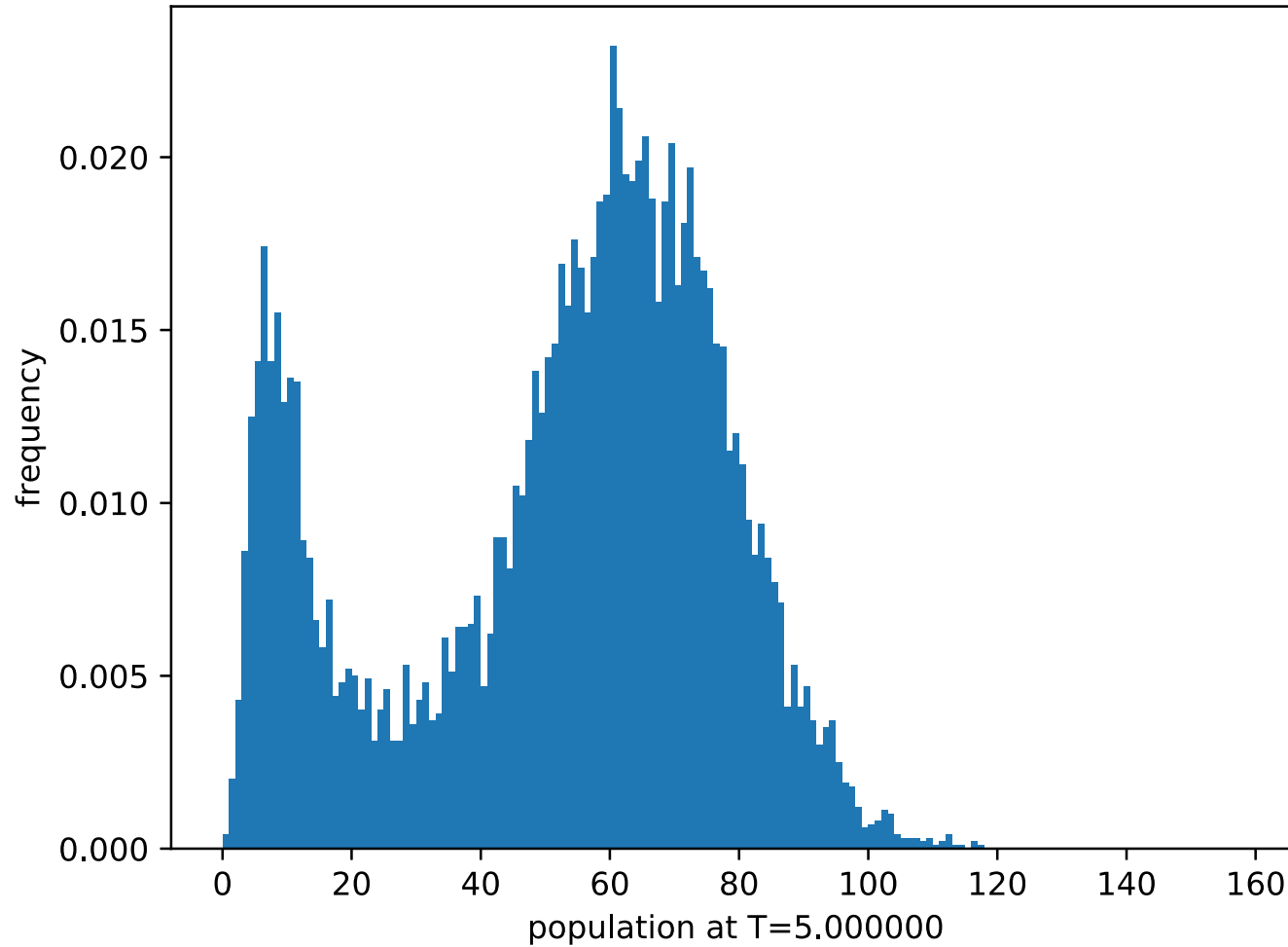
Schlögl Bistability

Population Histogram
Tau-leap ($h=0.250000$, $\tau=0.247143 \pm 0.018938$)
Schlögl (10000 Simulations)



Schlögl Bistability

Population Histogram
Tau-leap ($h=0.050000$, $\tau=0.049505 \pm 0.004949$)
Schlögl (10000 Simulations)



Conclusions

Birth-death model

- Tau-Leaping is a good approximation
- converges, but retains higher variance than SSA when stepsize is high

Banana model

- difficult, only similar to SSA solutions when stepsize approaches SSA-level
- → no gain from Tau-Leaping

Schlögl model

- although especially stepsize dependent, yields good results
- can benefit from Tau-Leaping

→ **model dependent performance of Tau-Leaping**

Discussion - so far

Have looked into:

- experimental behavior of Tau-Leaping for some elementary models
- → some models can be simulated a lot faster with Tau-Leaping, conserving dynamics, for other models dynamics break or are hard (i.e. small stepsize necessary)
- similar to the numerical integrators for ODEs, there is a trade-off between performance and accuracy

Discussion - go further

Further steps:

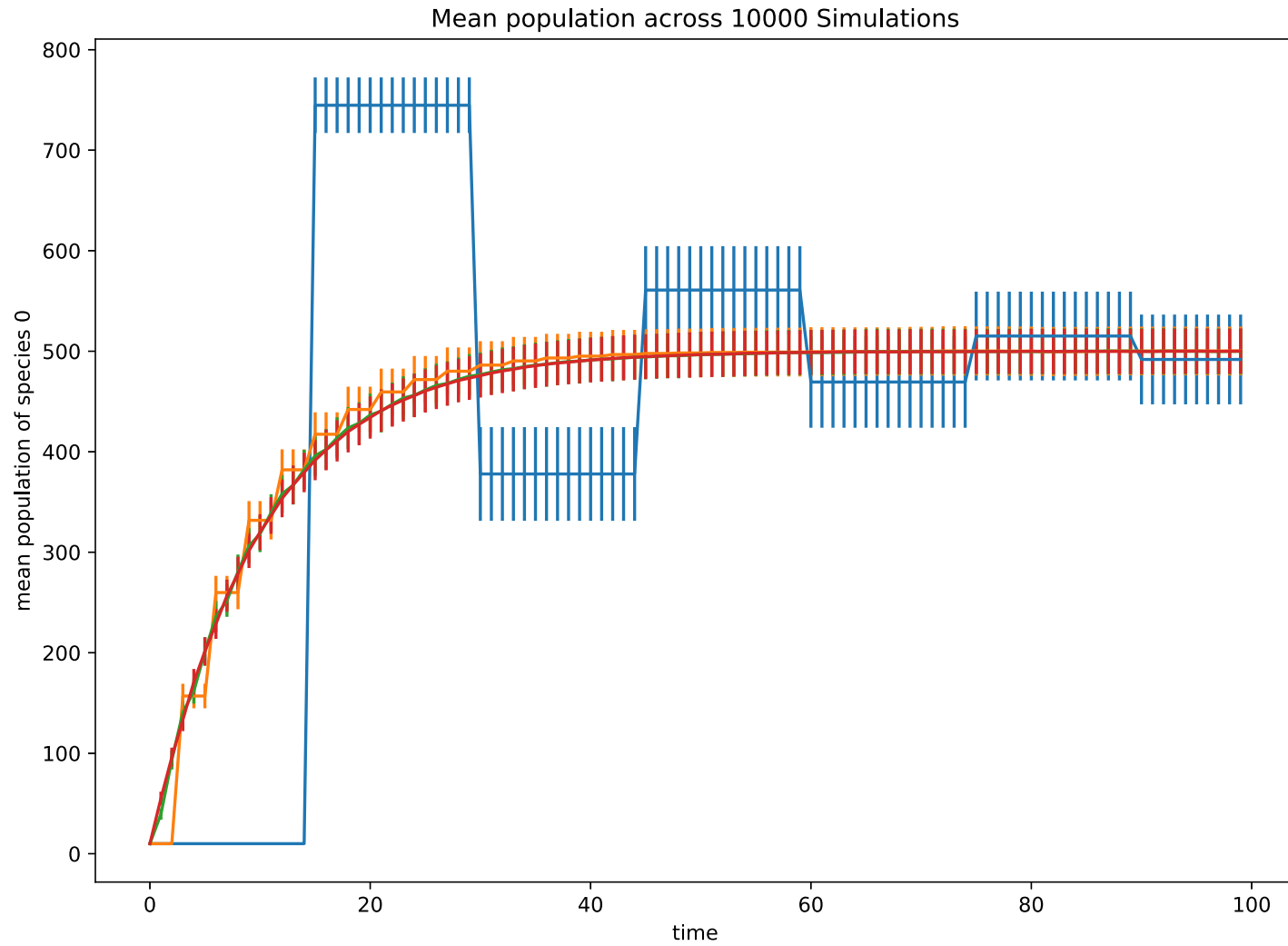
- evaluate Tau-Leaping for more complex models (more reactions, more species)
- look into numerical perturbation and maybe find generalizing principles from comparison to ODE integrators
- more rigorously test congruence of different simulations

References

1. Approximate accelerated stochastic simulation of chemically reacting systems (Gillespie 2001)
2. Efficient step size selection for the tau-leaping simulation method (Cao et al. 2005)
3. Numerics for Bioinformaticians course material (12.07.17)
http://systems-pharmacology.de/?page_id=724
4. Applied Numerics in Systemsbiology course material (12.07.17)
http://systems-pharmacology.de/?page_id=971
5. Download more RAM
<https://downloadmoreram.com/>

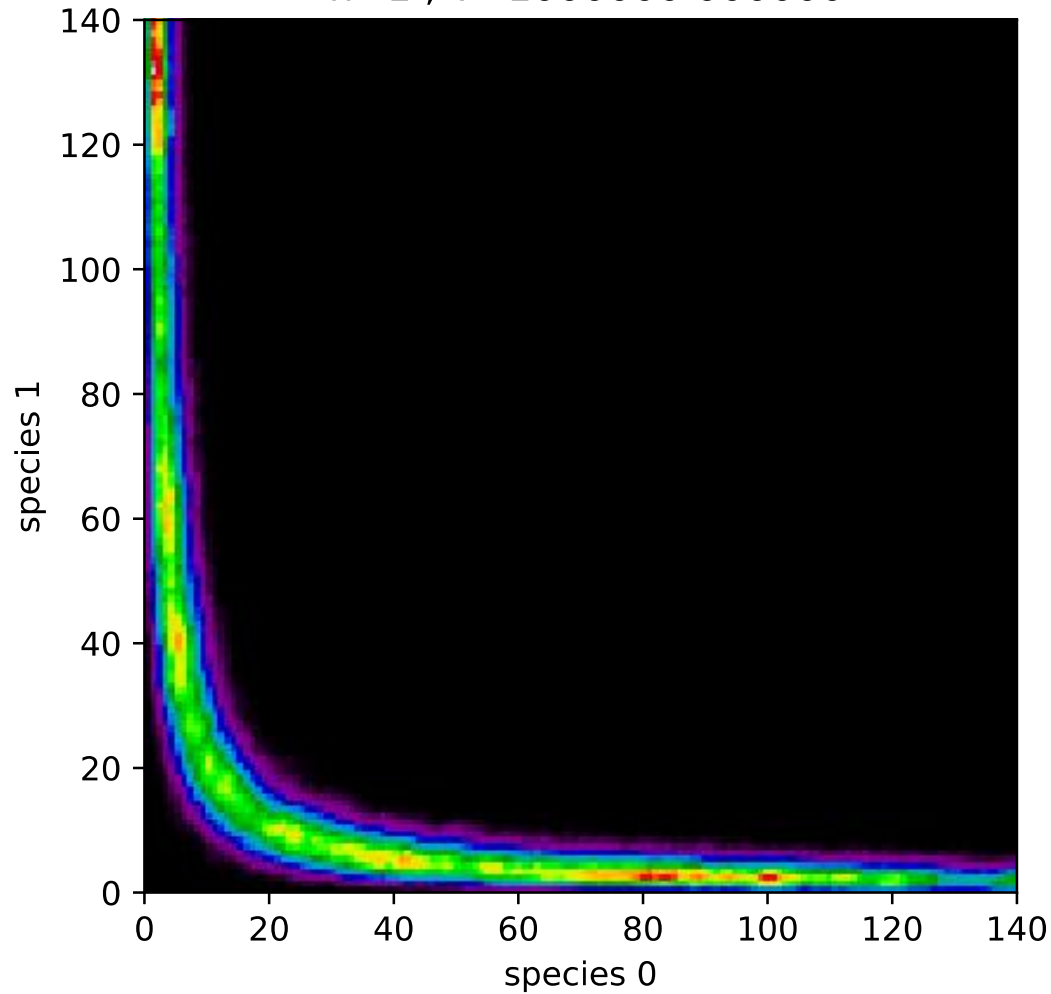
Additional Slides

Birth-death continued



Banana continued

Phase probability plot SSA
 $n=1$; $T=1000000.000000$



Banana continued

