Problem 1: Noise-induced oscillations (Ingalls 7.8.27, 45 points)

Stochastic systems can exhibit a range of oscillatory behaviors, ranging from near-perfect periodicity to erratic cycles. To explore this behavior, consider a stochastic relaxation oscillator studied by José Vilar and colleagues (Vilar et al., 2002). The system involves an activator and a repressor. The activator enhances expression of both proteins. The repressor acts by binding the activator, forming an inert complex. A simple model of the system is:

Name and description	Reaction	Reaction propensity
R_1 (activator synthesis)	$\varnothing \longrightarrow b_A A$	$rac{\gamma_A}{b_A} \cdot rac{lpha_0 + N_A/K_A}{1 + N_A/K_A}$
R_2 (repressor synthesis)	$\varnothing \longrightarrow b_R R$	$rac{\gamma_R}{b_R} \cdot rac{N_A/K_R}{1+N_A/K_R}$
R_3 (activator decay)	$A \longrightarrow \varnothing$	$\delta_A N_A$
R_4 (repressor decay)	$R \longrightarrow \varnothing$	$\delta_R N_R$
R_5 (association)	$A + R \longrightarrow C$	$k_C N_A N_R$
R_6 (dissociation w/ activator decay)	$C \longrightarrow R$	$\delta_A N_C$

Here, N_A , N_R , and N_C are the molecular counts for the activator, repressor, and activator-repressor complex. The parameters b_A and b_R characterize the expression burst size. The Hill-type propensities of the synthesis reactions are not well-justified at the molecular level, but these expressions nevertheless provide a simple formulation of a stochastic relaxation oscillator.

- a) Take parameter values $\gamma_A = 250$, $b_A = 5$, $K_A = 0.5$, $\alpha_0 = 0.1$, $\delta_A = 1$, $\gamma_R = 50$, $b_R = 10$, $K_R = 1$, $k_C = 200$, and $\delta_R = 0.1$. Run a Gillespie SSA simulation of this model and verify its quasi-periodic behavior.
- b) The deterministic version of this model is:

$$\frac{da}{dt} = \gamma_A \frac{\alpha_0 + a/K_A}{1 + a/K_A} - k_C ar - \delta_A a$$

$$\frac{dr}{dt} = \gamma_R \frac{a/K_R}{1 + a/K_R} - k_C ar + \delta_A c - \delta_R r$$

$$\frac{dc}{dt} = k_C ar - \delta_A c$$

where a, r, and c are the concentrations of activator, repressor, and complex. Run a simulation with the same parameter values as in part (a). Does the system exhibit oscillations? How is the behavior different if you set $\delta_R = 0.2$?

c) The contrast between the behavior of the models in parts (a) and (b), for $\delta_R = 0.1$, can be explained by the excitability of this relaxation oscillato. Run two simulations of the deterministic model ($\delta_R = 0.1$), one from initial conditions (a, r, c) = (0,10,35) and another from initial conditions (a, r, c) = (5,10,35). Verify that in the first case, the activator is quenched by the

repressor, and the system remains at a low-activator steady state, whereas in the second case, this small quantity of activator is able to break free from the repressor and invoke a (single) spike in expression. Explain how noise in the activator abundance could cause repeated excitations by allowing the activator abundance to regularly cross the threshold. This is referred to as *noise-induced oscillation*.

Problem 2: Effect of autorepression (55 points)

Consider the open system

$$\emptyset \xrightarrow{k_1} A \xrightarrow{k_2} \emptyset$$

- a) Find the master equation for this system.
- b) Show that, at steady state,

$$P(N_A = n) = \frac{k_1}{k_2 n} P(N_A = n - 1, t)$$

c) Use the Taylor series for e^x to derive the steady-state probability distribution:

$$P(N_A = n) = \frac{\left(\frac{k_1}{k_2}\right)^n e^{-k_1/k_2}}{n!}$$

d) Perform a Gillespie SSA simulation of the open system above with $k_1 = 10$, $k_2 = 1$. Estimate the probability distribution P(n) from the timecourse A(t). (Exercise your judgment in ignoring data early in the simulation and choosing an appropriate timescale for the simulation.) Plot the Poisson probability density function with $\lambda = k_1/k_2$ on the same axes for comparison.

Now consider a related molecule that exhibits hyperbolic autorepression, i.e.

$$\varnothing \xrightarrow{k_1/(1+k_3B)} B \xrightarrow{k_2} \varnothing$$

e) Perform a Gillespie SSA simulation of this system with $k_1 = 100$, $k_2 = k_3 = 1$ and estimate the probability distribution P(n) from the timecourse B(t). Fit a Poisson distribution to P(n) and plot both on the same axes (e.g. using poissfit () and poisspdf () in MATLAB). Does the negative autoregulation system have more or less variance than expected for the simple regulation system?