

Raphael Sofaer

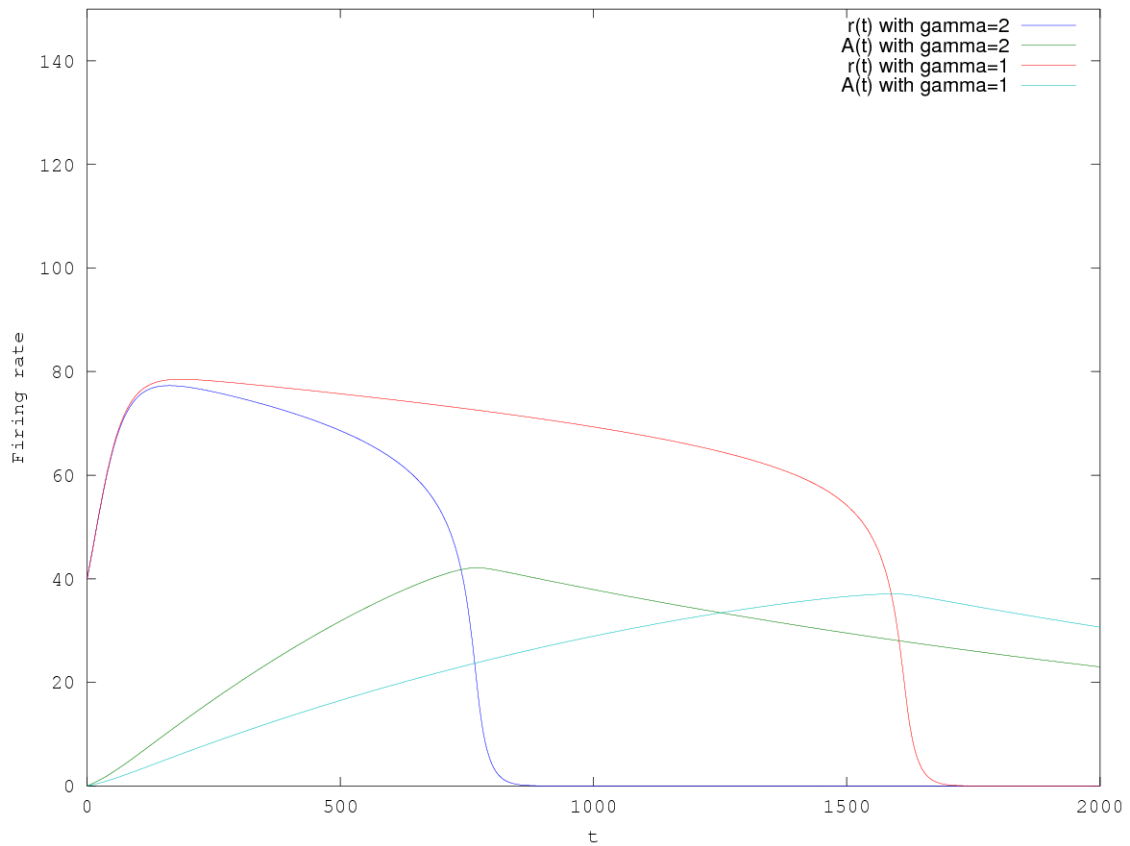
Computational Neuroscience: Homework 4

Due on April 25, 2012

John Rinzel

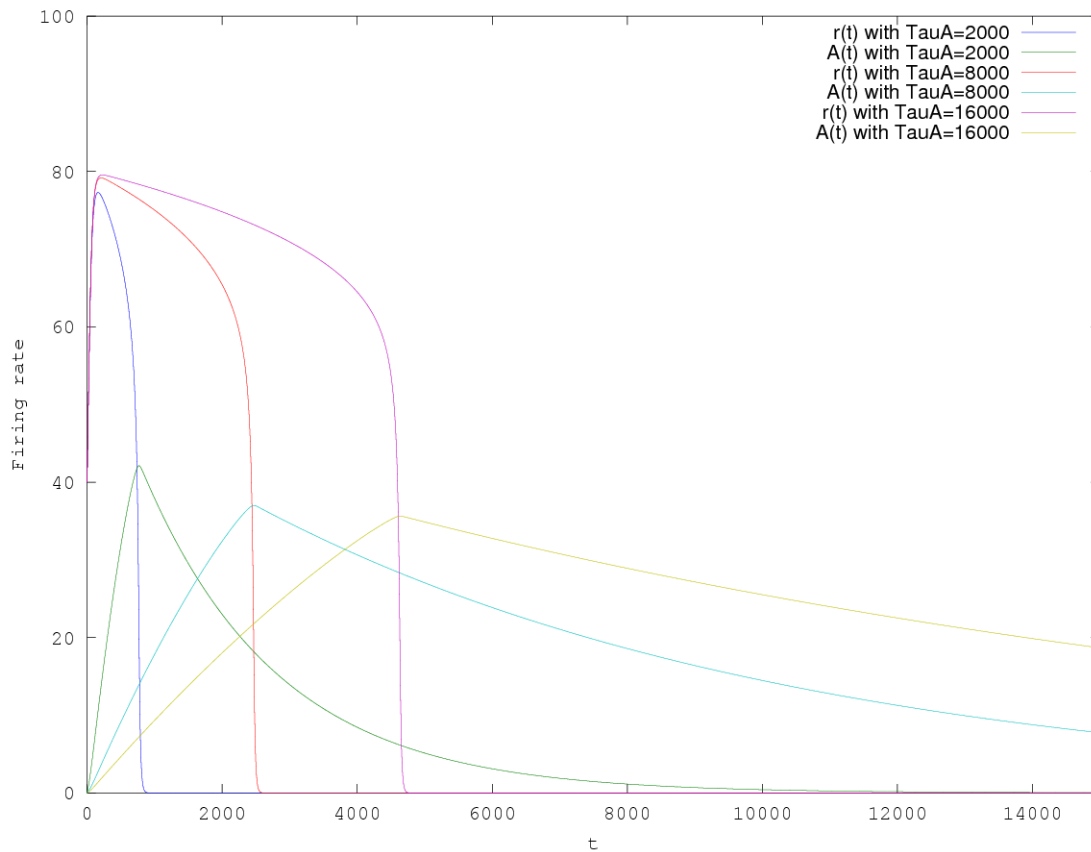
Short-Term Memory model with Adaptation: 1

I based my Matlab script (run in Octave) on STM_MF.m. For the first run, I used $\gamma = 2$ to match the graph in the homework description. With $\gamma = 1$, $A(t)$ increased more slowly, and therefore took longer to grow large enough to drive down $r(t)$.



2

For the second set of trials, I left γ at 2. As I increased τ_A , the system slowed down. $A(t)$ changed more slowly, so it took longer for $A(t)$ to increase to drive $R(t)$ down, and it took longer for $A(t)$ to decrease again after $R(t)$ plummeted. The shape of the curves was qualitatively similar, however.

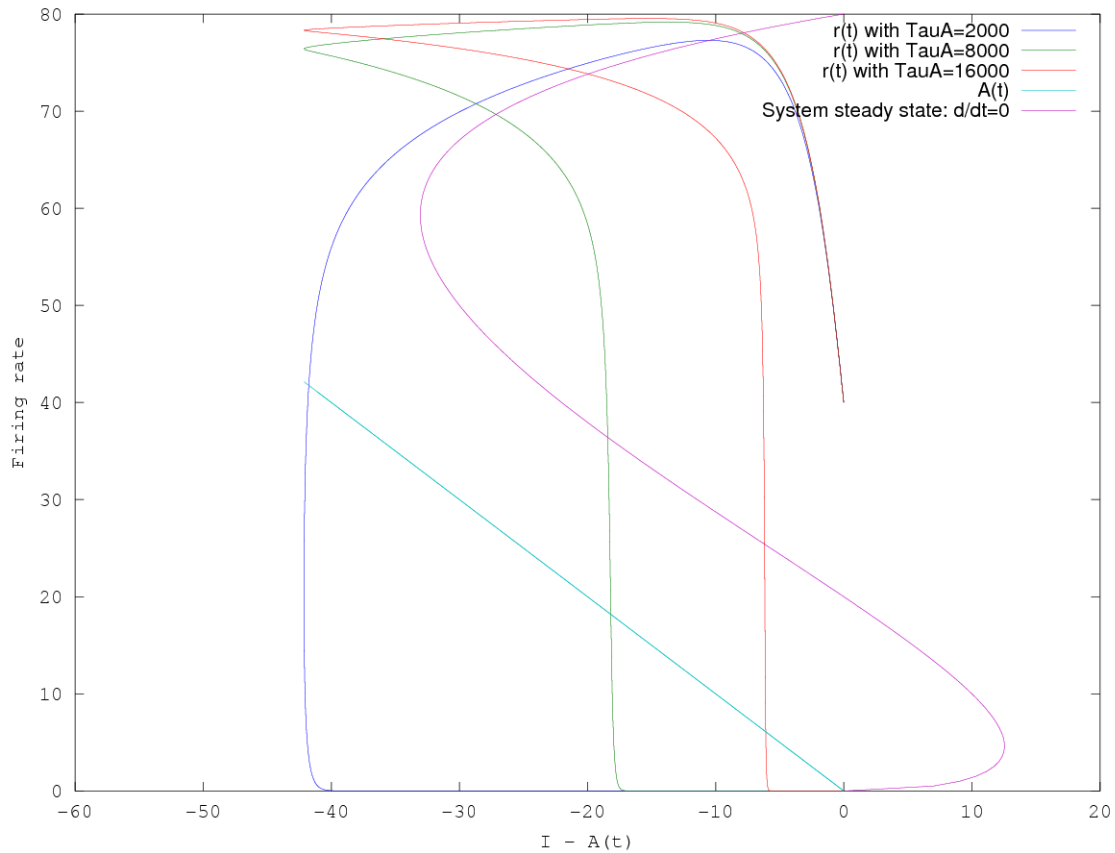


3

I approached finding the steady state curve analytically for individual values of r . Given: $\tau(dr/dt) = -r + S_{NR}(ar - A)$ ($I = 0$), we can set dr/dt to 0 and calculate from there.

$$\begin{aligned}
 r &= S_{NR}(ar - A) \\
 r &= \frac{M * (ar - A)^2}{\sigma^2 + (ar - A)^2} \\
 \delta &= (ar - A)^2 \\
 r &= \frac{M * \delta}{\sigma^2 + \delta} \\
 r\sigma^2 + r\delta &= M * \delta \\
 \frac{r\sigma^2}{\delta} &= M - r \\
 \frac{1}{\delta} &= \frac{M - r}{r\sigma^2} \\
 \delta &= \frac{r\sigma^2}{M - r} \\
 A^2 - 2arA + a^2r^2 - \frac{r\sigma^2}{M - r} &= 0
 \end{aligned}$$

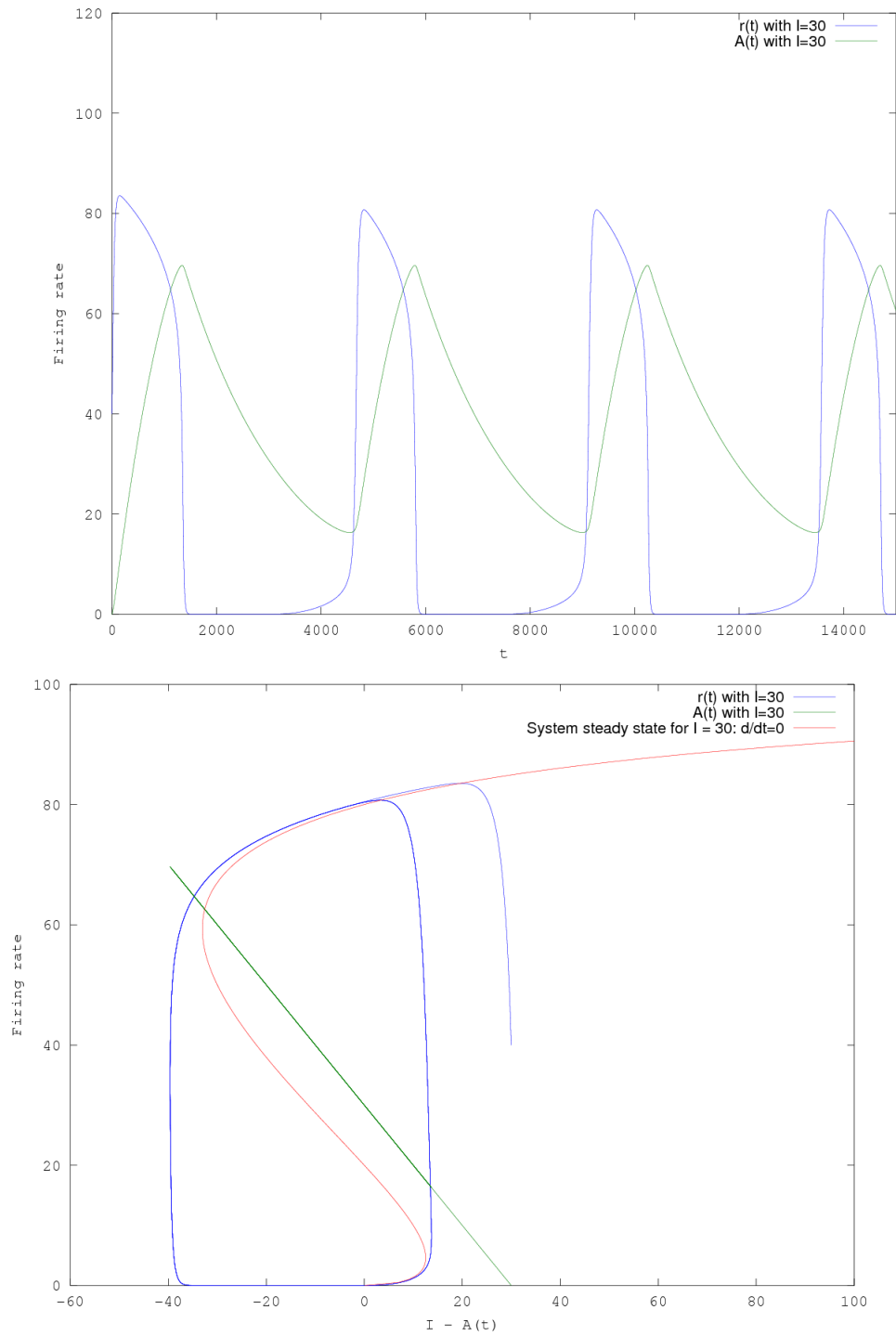
Since everything other than A is constant for a given r , this is a quadratic in A , easily solved. In the graphs below, we can see R jump to the upper part of the steady state, stay there for a little while as $A(t)$ grows (STM), then fall to the bottom as $A(t)$ dominates the input. When τ_A is lowest, we can see $A(t)$ immediately forcing down $r(t)$, so it doesn't spend much time at the top of the graph, but as we raise τ_A , $r(t)$ remains for much of $A(t)$'s rise.

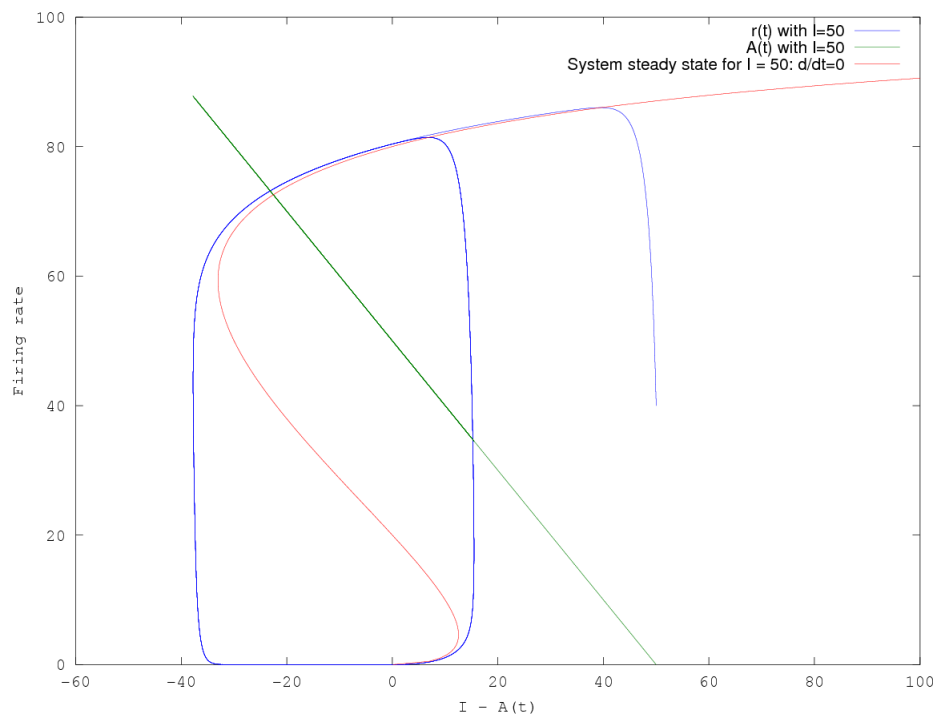
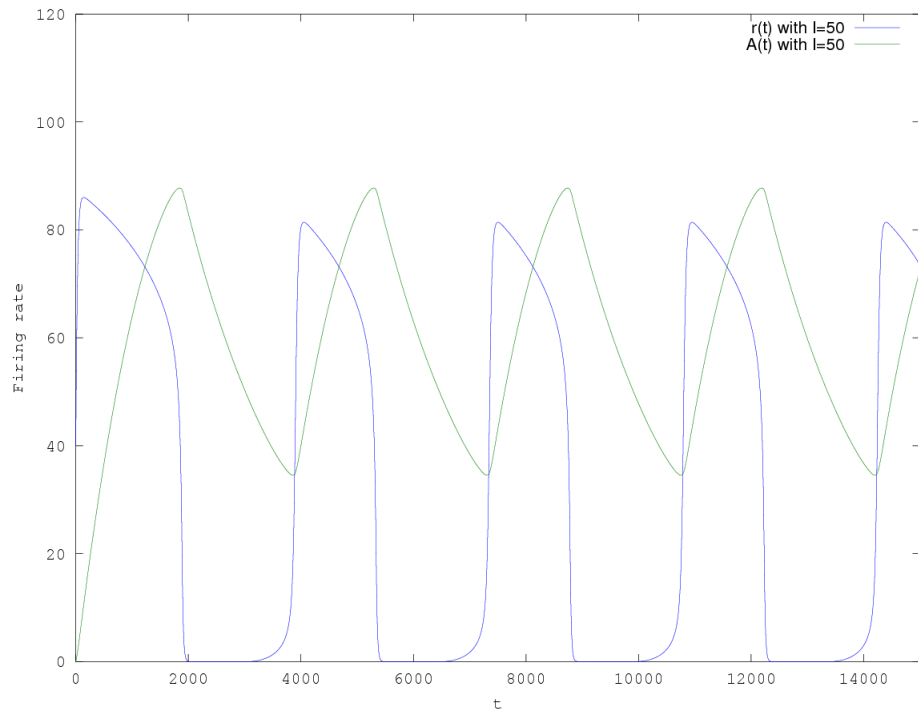


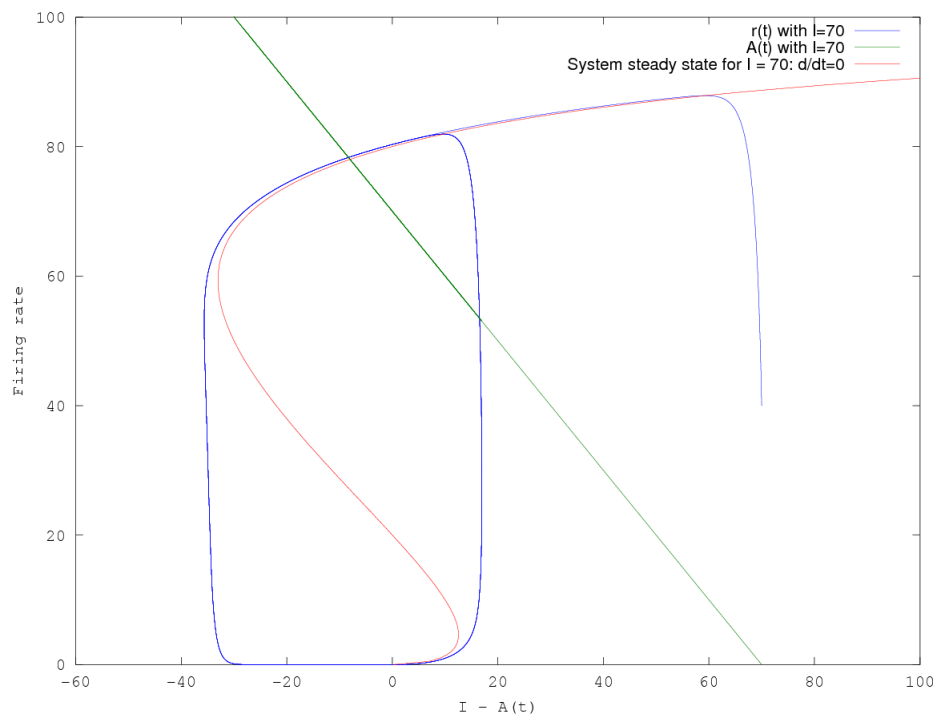
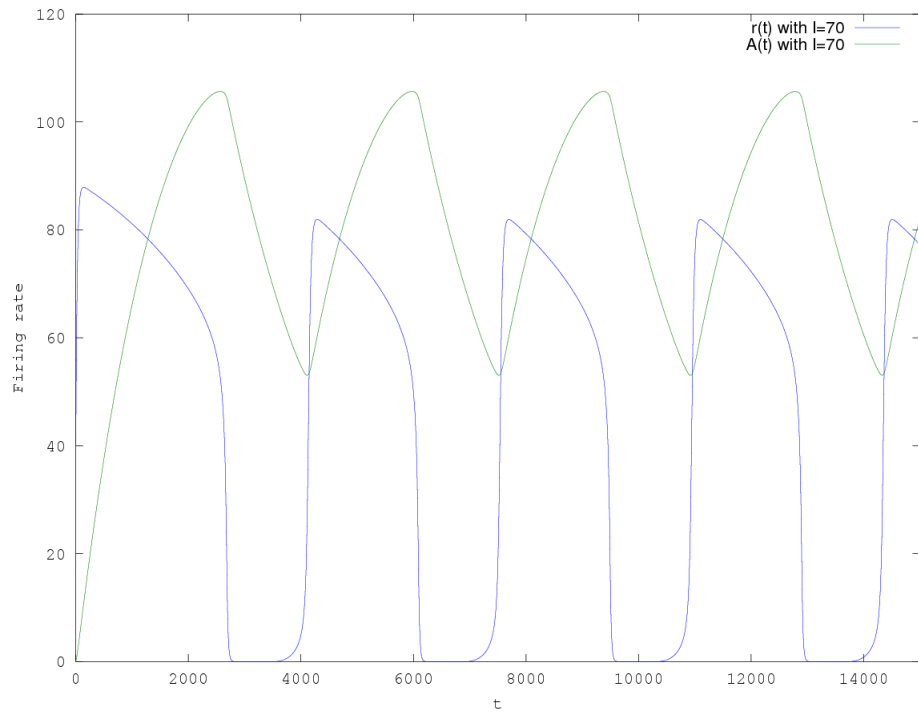
4

For varying the values of I , I set τ_A back to 2000. In these time courses, instead of the firing rate falling to nothing, the constant input current causes the neuron to reach a steady state solid firing rate. The steady state is about 62 for $I=30$, 72 for $I=50$, and 78 for $I=70$. For any given I , we could find what steady state it would go to by first checking the sign of dr/dt , to know whether it would go to a zero or positive steady state, then if $dr/dt > 0$, we can solve for r in $r = S_{NA}(ar - A(r))$.

In the six graphs below, we can see that the system oscillates for each of these values of I , fastest for $I=50$, slightly slower for $I=70$ since A is greater, and slowest for $I=30$. In the phase planes, we can see that for each input value, the firing rate first jumps upwards, then goes into a cycle of slowly decreasing as the input decreases, falling fast as $A(t)$ rises, then jumping again as $A(t)$ falls.







In order to find the threshold where the oscillating regime ends and a forced constant firing rate begins, I hypothesized that $r(t)$ would stabilize at the left side of the upper part of the steady state curve. Then, to stop oscillation, I tried to choose an I such that $dA/dt = 0$ when $r(t) \approx 60$. Taking $dA/dt = 0$ gives $A = \gamma r = 2r$, so A should stabilize at about $A = 120$. Then, since we know $dr/dt = 0$, we can look at that equation.

$$\begin{aligned}
 r &= S_{NA}(ar + I - A) \\
 60 &= S_{NA}(180 + I - 120) \\
 60 &= S_{NA}(60 + I) \\
 60 &= S_{NA}(60 + I) \\
 60 &= \frac{M * (60 + I)^2}{\sigma^2 + (60 + I)^2} \\
 60 &= \frac{100 * (60 + I)^2}{120^2 + (60 + I)^2} \\
 60 &= \frac{100 * (60 + I)^2}{120^2 + (60 + I)^2} \\
 0 &= 40I^2 + 4800I - 720000 \\
 I &\approx 87
 \end{aligned}$$

I looked in the region around $I = 87$, and found that the regime change actually happens between $I = 85$ and $I = 86$.

