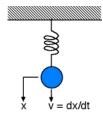
Excitability refers to the phenomenon where a system has but a single stable attractor, but it has two modes of returning to the equilibrium state. For small perturbations away from the equilibrium, the return is monotonic; however, for perturbations beyond a threshold value, the return is not monotonic, but undergoes a large excursion before settling down. The toilet is an example of an excitable system.

The harmonic oscillator

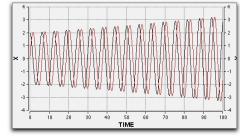
The simplest oscillator is a mass on a spring. The equation of motion is given by Newton's Law: $F = m \cdot a \Rightarrow F = m \cdot d^2x/dt^2$. Let m = 1, and v = dx/dt. The spring force is $F = -k \cdot x$. Then the equations of motion become

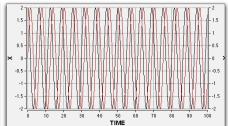


$$\frac{dx}{dt} = v, m\frac{dv}{dt} = -kx, \quad or \quad \tau \frac{dv}{dt} = -x, \text{ where } \tau = \text{m/k}$$

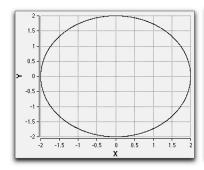
The simplest program to solve this system is:

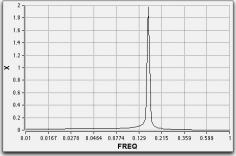
METHOD EULER STARTTIME = 0 STOPTIME = 100 DT = 0.01 d/dt(X) = vd/dt(v) = -K*XINIT v = 0INIT X = 2K = 1



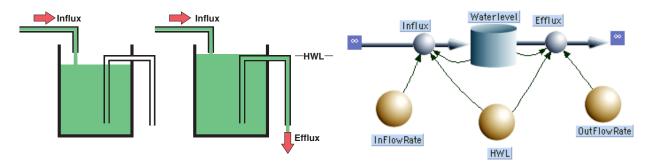


But the oscillation grows! The culprit is the Euler method, so switch to, say, RK4. A better way to plot the oscillation is to use the *phase plane* (x, v), on which the trajectory is a circle. The frequency of the oscillation can be gotten by pressing the Fourier Transform button and changing the x-axis to a *log* scale.





Exercise 1. Here is a simple oscillating system based on the siphon principle: When the fluid level reaches the High Water Level (HWL) the siphon empties the reservoir (i.e. Efflux >> Influx). Program the siphon oscillator using IF...THEN statements.



Exercise 2. The 'Brusselator'. The following system is a model for an oscillating chemical reaction. Find the approximate value for b for which the system becomes a limit cycle. Use the Rosenbrock stiff solver with DT = 0.002, DTMAX = 1, TOLORANCE = 0.01. Find the period of the oscillation using the Fourier transform button on the Graph window.

$$\frac{dx}{dt} = \left[1 - (b+1)x\right] + ax^2y, \quad x(0) = 1$$

$$\frac{dy}{dt} = bx - ax^2y, \quad y(0) = 1$$
(1)

A bistable switch

The first ingredient of an excitable system is a bistable switch. Consider the first order dx

system: $\frac{dx}{dt} = f(x)$, where f(x) has the shape shown in Figure 1. If the system is perturbed in

either direction from its stable points past the unstable point, then it quickly switches to the other equilibrium. A light switch works like this. f(x) defines a 'vector field' on the line showing which way the system will evolve when perturbed away from its stationary points. We will see that, by coupling this system to another 'slow' variable, one can convert the bistable system into an excitable or oscillatory system.

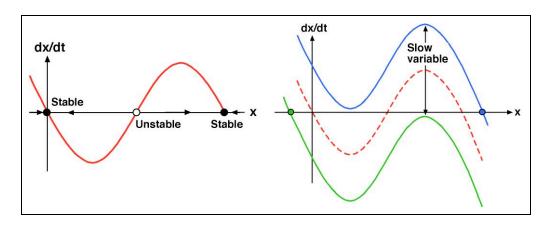


Figure 1. A bistable switch. A commonly used analytic form for f(x) is a cubic polynomial: f(x) = x(x-1)(x-2). This has an unstable root at x = 1, and stable points at x = 0, 2. (Alternative: $f(x) = -x^3/3 + x$, which has an unstable root at x = 0.)

Covalent modification can produce a bistable switch, as shown in Figure 2.

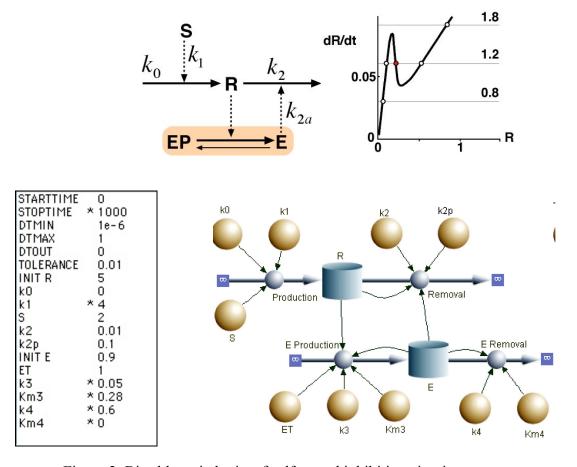


Figure 2. Bistable switch via a feedforward inhibition circuit.

Exercise 3: Figure 2 is taken from Figure 1f in reference [1]. Using the parameters given here, construct the model and plot the oscillations and the stimulusresponse curve.

The Fitzhugh-Nagumo Equations

The best example of an excitable phenomenon is the firing of a nerve: according to the Hodgkin and Huxley equations a subthreshold depolarization dies away monotonically, but a superthreshold depolarization initiates a spike potential. Fitzhugh and Nagumo devised a simplified version of the H-H equations that describes the essential features of the nerve impulse by only two differential equations.

I = g(V)C

The ionic current that flows through a nerve membrane is controlled by channels whose openings and closings are controlled by the local electrical field (voltage gated ion channels). For such a conductor, Ohms Law has the form I =

g(v), where v is the transmembrane voltage and g(v) is the voltage-dependent conductance. Since $Q = C \cdot v$, applying d/dt to each side the differential equation for the voltage change is:

$$C\frac{\mathrm{d}V}{\mathrm{dt}} = \frac{dQ}{dt} = I = -\hat{\mathbf{g}}(v)v \equiv -g(v)$$
 (2)

where C is the membrane capacitance and I = dQ/dt is the current. The voltage gate can be either open or shut; that is the conductance is bistable, so it has the S-shape shown in Figure 1a.

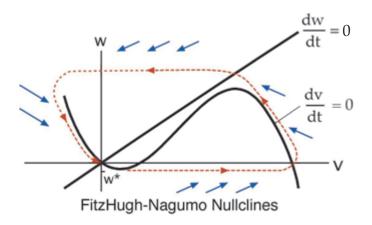
To turn the bistable conductance equation into an excitable system, Fitzhugh defined a *slow* depolarization variable, w(t), that can move the bistable curve up an down as shown in Figure 1b. This results in the following system:

$$\frac{dv}{dt} = -g(v) - w + I \tag{1.3}$$

$$\frac{dw}{dt} = \frac{1}{\tau} (v - kw - b) \tag{1.4}$$

$$\frac{dw}{dt} = \frac{1}{\tau} \left(v - kw - b \right) \tag{1.4}$$

where | > 1, and k > 0.



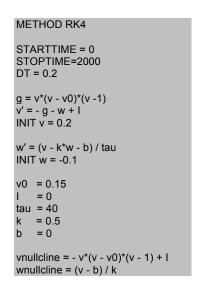


Figure 3. Phase plane for equations (2) and (3) showing the Nullclines that lead to excitable behavior.

The *phase portrait* for this system shows how an excitable system works: the single equilibrium at the origin is *locally* stable, but a small perturbation causes the system to make a large excursion before returning to rest. This sort of phase portrait is typical of excitable systems.



Note that by varying a parameter (e.g. *I*) the excitable system can be transformed into a bistable system in two variables. We will also see that, by adjusting the parameters, the system can oscillate in a *limit cycle*.

Exercise 4. Use the model equations at the right to make *time* and *phase plane* (w vs. v) plots and then

- 1. Make sliders for the parameters and find a parameter combination that makes the system oscillate.
- 2. Make a parameter plot of a critical parameter *I* vs. the amplitude of the oscillation to find the 'bifurcation point' where the oscillations suddenly appear.
- 3. Use the initial condition button, Ic, on the graph window to explore the pattern of trajectories.
- 4. Use the *Fourier Transform* button to estimate the period and frequency of the oscillation.
- 5. Try the RK2 and Stiff solver methods and compare how many iterations Madonna™ had to execute.

The simplest limit cycles

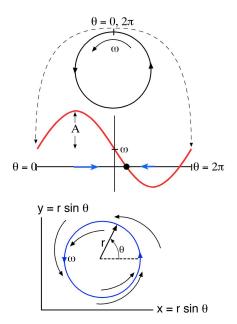
It is sometimes easier to think of periodic phenomena as taking place on a circle: $0 \le \theta \le 2\pi$: $d\theta/dt = \omega(\theta)$. Let $\omega(\theta) = \omega - A \cdot \sin(\theta)$. Sketch the vector field on the circle showing the stability of the equilibrium points and their stability as ω is varied. To do this, 'snip' the circle at $\theta = 0$ and unwrap it so it looks like this \longrightarrow

(Make the length of the vectors proportional to the speed of the 'phase point'.)

A slightly more elaborate version of the circular limit cycle is

$$\frac{dr}{dt} = r(1-r), \quad \frac{d\theta}{dt} = \omega$$

where the radius of the limit cycle, r, is governed by the simple logistic equation with amplitude = 1, and the speed around the cycle is ω = constant.



Calcium Oscillations and Cellular Signaling

Here we will learn how to model the oscillatory dynamics of the calcium second messenger system. The reference paper for this problem set is [2]. A reprint is on the course web site.

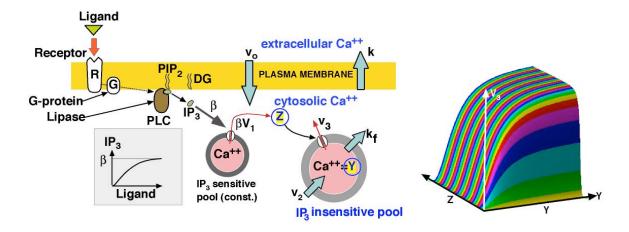


Figure 4. (a) The calcium oscillator. (b) The shape of the reaction velocity functions.

Many types of cells, when stimulated by hormones or neurotransmitters, burst into repetitive spikes of intracellular calcium release. The period of these oscillations ranges from less than 1 second to more than 30 minutes. These oscillations are thought to be an important attribute of

intra and intercellular signaling mechanisms. From our viewpoint they are a good example of "limit cycle" kinetics, and will give us an opportunity to learn how to model periodic chemical dynamics.

Consider the calcium transport system, shown in Figure 4. We write conservation equations for the concentration of intracellular calcium, \mathbf{Z} , and the concentration in the IP_{ζ}-insensitive pool (pool 2), \mathbf{Y} :

$$\frac{dZ}{dt} = \underbrace{v_0}_{\text{into}} + \underbrace{v_1 \beta}_{\text{into}} - \underbrace{v_2}_{\text{out of pool 2}} + \underbrace{v_3}_{\text{out of pool 2}} + \underbrace{k_f Y}_{\text{leak from pool 2}} - \underbrace{kZ}_{\text{transport out of cell}}$$
(5)

$$\frac{dY}{dt} = \underbrace{v_2}_{\text{transport}} - \underbrace{v_3}_{\text{out of pool 2}} - \underbrace{k_f Y}_{\text{leak from pool 2}}$$
(6)

The fluxes into and out of the IP3 insensitive pool (2) are the key nonlinearities controlling the behavior of the system. They are Michaelis-Menten type rate laws:

$$v_2 = V_{M2} \frac{Z^n}{K_2^n + Z^n} = 65 \frac{Z^2}{1 + Z^2}$$
 (7)

$$v_{3} = V_{M3} \frac{Y^{m}}{K_{R}^{m} + Y^{m}} \cdot \frac{Z^{p}}{K_{A}^{p} + Z^{p}}$$

$$= 500 \frac{Y^{2}}{2^{2} + Y^{2}} \frac{Z^{4}}{0.9 + Z^{4}}$$
(8)

Table 1 lists the parameters of the model, their units, and the values that produce oscillatory behavior.

PARAMETER	VALUE	Units
v_0	1	μM/s
k	10	1/s
kf	1	1/s
v1	7.3	μM/s
V	65	μM/s
V_{M3}	500	μM/s
K_2	1	μМ
K_R	2	μМ
K_{A}	0.9	μМ
m	2	1

Exercise 5. Make a Madonna Flowchart to simulate the system.

1. Show that the period of the oscillations decreases as β increases.

n	2	1
p	4	1
Yo	0.1	μM
Z_{o}	10	μΜ
β	0.3	

- 2. Start with a small value of the composite parameter ($v_0 + \beta v_1$) and show that as this quantity increases oscillations begin to appear only after a critical value is reached (this is called a "bifurcation point").
- 3. Note that the *nullclines* (dZ/dt = 0 = dY/dt) of the calcium regulation system look very such like those of the Fitzhugh-Nagumo equations. Indeed, an examination of the nullclines shows that, with appropriate tuning of parameters, the calcium model can exhibit excitable behavior.

Table 1. Parameter values

References

- 1. Tyson, J., Chen, K., and Novak, B. (2003). Sniffers, buzzers, toggles, and blinkers: dynamics of regulatory and signaling pathways in the cell. Curr Opin Cell Biol. *15*, 221-231.
- 2. Goldbeter, A., Dupont, G., and Berridge, M.J. (1990). Minimal model for signal-induced Ca⁺⁺ oscillations and for their frequency encoding through protein phosphorylation. Proc Natl Acad Sci U S A 87, 1461-1465.

A good elementary textbook on modeling of dynamical systems in biology is:

• Edelstein-Keshet, L. (1988). *Mathematical Models in Biology*. Ed.) New York: Random House.

Nullcline Analysis

http://www.sosmath.com/diffeq/system/qualitative/qualitative.html

http://www.sosmath.com/diffeq/system/nonlinear/linearization/linearization.html