Brilliant: Differential Equations II

Dave Fetterman

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Note: Latex reference: http://tug.ctan.org/info/undergradmath/undergradmath.pdf

1 Chapter 1: Basics

1.1 Chapter 1: Nonlinear Equations

The two types of problems in this course are:

- Nonlinear equations (several equations on one independent variable)
- Partial differential equations (single equation with several independent variables)

Linear equations have solutions like y_1, y_2 that can be combined using any $c \in \mathbb{R}$ like $y_1 + cy_2$.

Example: Bacteria in a dish with a lot of food, no deaths

- $b'(t) = r_b b(t), r_b > 0.r_b$ would be the rate of growth.
- This is linear. Reason 1: $\frac{d}{dt}(y_1+cy_2)=y_1'+cy_2'=r_b(y_1+c_y2)$ since $y'=r_by(t)$, and same for y2.
- Also, this works because the solution is $b(t) = b(0)e^{r_b t}$, so $b_1(t) + cb_2(t) = b_1(0)e^{r_b t} + cb_2(0)e^{r_b t} = (b_1(0) + cb_2(0))e^{r_b t}$

Example: Logistic equation: Bacteria in a dish with a lot of food, limited by carrying capacity M.

- $b'(t) = r_b b(t) [M b(t)].$
- This is nonlinear. Reason: $\frac{d}{dt}(y_1'+cy_2')=y_1'+cy_2'=r_b[y_1+cy_2][M-y_1-cy_2]=My_1+Mcy_2-y_1^2-2cy_1y_2-cy_1^2y_2^2$
- $\neq My_1 y_1^2 + Mcy_2 c^2y_2^2$ because of the extra $-2cy_1y_2$ term.

Sidebar: Note that this equation $b' = r_b b[M - b]$ is separable, so it can be solved.

- $\frac{db}{dt} = rb[M-b]$
- $\bullet \ \frac{db}{b(M-b)} = rdt$
- $\frac{1}{M}(\frac{1}{b} + \frac{1}{M-b})db = rdt$ after partial fractions work
- $(\ln(b) \ln(M b)) = Mrt + C \Rightarrow \ln(\frac{b}{M b}) = Mrt + C$
- $\frac{b}{M-b} = e^{Mrt}e^C$
- Initial conditions $b=b(0), t=0 \Rightarrow \frac{b}{M-b} = \frac{b(0)}{M-b(0)} e^{Mrt}$
- $b(1 + \frac{b(0)}{M b(0)}e^{Mrt}) = M\frac{b(0)}{M b(0)}e^{Mrt}$
- $b(M b(0) + b(0)e^{Mrt}) = Mb(0)e^{Mrt}$
- $b = \frac{Mb(0)e^{Mrt}}{M+b(0)[e^{Mrt}-1]}$

This logistic solution will taper off to M at some point. Note that $\lim_{t\to\infty} b(t) = M$ since the non-exponential terms stop mattering. Also b(t) = M sticks as a constant solution or equilibrium immediately. These equilibria tell us what matters - the long-term behavior of solutions!

Another **Example**: Lotka-Volterra equation pairs: Bacteria (b) and bacteria-killing phages (p), with kill rate k.

- The "product" kb(t)p(t) measures the interactions and kills resulting from this.
- $b'(t) = r_b b(t) kp(t)b(t)$, or the normal growth rate minus kill rate
- p'(t) = kp(t)b(t) since its population grows as it kills bacteria.
- Equilibria include b = 0, p = 0 and b = 0, p > 0, since these are *constant* solutions, or places where b'(t) = 0, p'(t) = 0.

Direction fields, with vector pointing towards $\langle b'(t), p'(t) \rangle$ (TODO - I think) let us follow the arrows to determine the curve over time. In this case, the bacteria will always go extinct.

However, if we add a new death rate term $-d_p p(t)$ so $p'(t) = -d_p p(t) + k p(t) b(t)$:

- We get an equilibrium at $b = \frac{d_p}{k}$, $p = \frac{r_b}{k}$. (Since 0 = b'(t) = rb kpb, $(\Rightarrow pk = r)$, 0 = p'(t) dp + kpb, $(\Rightarrow bk = d)$)
- But otherwise the solutions swirl around this point. This is called a **cycle**. TODO What is a **limit cycle**?

Note that there are systems where the "solution particle" neither reaches an equilibrium or cycles around one point. The **Lorenz system** famously has this owl-eye shaped double attractor (an example of **strange sets**) where initially close particles diverge unpredictably if the constants ρ , σ , b are chosen right:

- $x'(t) = \sigma(y x)$
- $y'(t) = x(\rho z) y$
- z'(t) = xy bz
- TODO

1.2 Chapter 1.2: PDEs

Many methods of attack for PDEs

- Separation of variables
- Power series (Note: did we actually touch on this?)
- Fourier Transform

Example: Standing wave, where one end of a rope is fixed.

- Vertical displacement from a line of rope: u(x,t) depends on where (x) and when (tt).
- Rope's wave equation is $u_{tt} = v^2 u_{xx}$, where v is the "constant wave speed", and the others are the space, time partials.
- Note that $u = \cos(vt)\sin(x)$ and $u = \sin(vt)\cos(x)$ both work.
- If you guess the solution has split variables like u = X(x)Y(y)T(t), then, upon substitution and division by X(x)Y(y)T(t), $\frac{\delta^2 u}{\delta t^2} = v^2 \left[\frac{\delta^2 u}{\delta x^2} + \frac{\delta^2 u}{\delta y^2}\right]$ yields $\frac{T''(t)}{T(t)} = v^2 \left[\frac{X''(x)}{X(x)} + \frac{Y''(y)}{Y(y)}\right]$
- This method may or may not work. But if it does, it means that since x, y, and t are independent variables, each individual piece must be constant.
- So, for example, if we know $\frac{X''(x)}{X(x)} = -4\pi^2$, we can get to $X(x) = \sin(2\pi x)$
- The wave equation is similar in 3D: $u_{tt} = v^2[u_{xx} + u_{yy} + u_{zz}]$, or using the Laplacian, $u_{tt} = v^2 \nabla^2 u$. Here, u measures not displacement but expansion/compression of air at (x, y, z), time t.

Using Fourier transforms helps turn difficult PDEs into an easier problem like an ODE. Fourier transforms work best when

- The domain is all of \mathbb{R}^n
- The function *u* vanishes at infinity.

The Fourier transform changes the domain of x to that of ω . It comes with the (highly simplified) rule (see Vector Calculus course): $F\left[\frac{\delta f}{\delta x}\right] = i\omega F[f]$. **Example**: Drunkard's walk.

- One dimensional: moves left or right in a random way. Starts at x = 0, t = 0.
- u(x,t) is probability of being at point x at time t. Naturally, $\int_{x=-\infty}^{x=\infty} u(x,t) dx = 1$.
- Also, it obeys the 1-dD diffusion equation $\frac{\delta u}{\delta t} = \frac{\delta^2 u}{\delta x^2}$
- The Fourier transform doesn't affect t at all.
- So by taking Fourier transform of both sides of diffusion equation we get

$$-F(u_t) = \frac{\delta}{\delta t}F(u)$$
 since F doesn't care about t.

$$-\frac{\delta^2 u}{\delta x^2} = i\omega F(\frac{\delta u}{\delta x}) = -\omega^2 F(u)$$

- So
$$\frac{\delta}{\delta t}F(u) = -\omega^2 F(u)$$

– This is solvable as
$$F(u) = ce^{-\omega^2 t}$$
. Take it on faith that $c = \frac{1}{2\pi}$ for now. TODO

– Known fact:
$$F[Ae^{-\frac{ax^2}{2}}] = \sqrt{\frac{1}{2\pi a}}Ae^{\frac{-\omega^2}{2a}}$$

– This means
$$t = \frac{1}{2a}$$
 and $a = \frac{1}{2t}$

$$-F(u) = \frac{1}{2\pi}e^{-\omega^2t}, F[Ae^{-\frac{ax^2}{2}}] = \sqrt{\frac{1}{2\pi a}}Ae^{\frac{-\omega^2}{2a}} \text{ so } u = Ae^{\frac{-ax^2}{2}}$$

– Solving, you get
$$A=\sqrt{\frac{1}{4\pi t}}, a=\frac{1}{2t},$$
 so $u(x,t)=\sqrt{\frac{1}{4\pi t}}e^{-\frac{x^2}{4t}}$

2 Chapter 2: Nonlinear Equations

2.1 2.1: Lotka-Volterra I

Major ideas:

• phase plane: TODO

• nullcline: TODO

• direction field: TODO

• equilibria: TODO

Example: Bacteria vs. phages (again)

- Bacteria unrestrained grow in proportion to their population, so $\frac{db}{dt} = r_b b(t)$ (solved: $b(t) = b(0)e^{r_b t}$)
- Phages unfed decrease in proportion to current size, so $\frac{dp}{dt} = -d_p p(t)$ (solved: $p(t) = p(0)e^{-d_p t}$)
- Bacteria die with likelihood of meeting a phage, and phages increase with likelihood of meeting a bacterium. So the set of equations, for constant k, becomes:
 - $-b'(t) = r_b b(t) k b(t) p(t)$
 - $-p'(t) = -d_n p(t) + kb(t)p(t)$
 - The product of p and b makes our equations nonlinear (WHY?)
 - I guess, very generally, $b_1p_1 = k$, $b_2p_2 = k$, but $(b_1 + b_2)(p_1 + p_2) = b_1p_1 + b_2p_2 + b_1p_2 + b_2p_1 = 2k + b_1p_2 + b_2p_1 \neq 2k$, so the last two "mixed" terms mean you can't just add solutions (b_1, p_1) and (b_2, p_2) .

General thoughts on this solution:

- So a solution (b(t), p(t)), traces out a curve on the bp-phase plane (b is x-axis, p is y-axis) as time (unrepresented in the plane) continues.
- If we add a unit tangent vector at every point (B, P) aligned with $(b'(t), p'(t)) = (r_b B k B P, -d_p P + k B P)$, we can follow the arrows to see the solution over time.
- The above is called a direction field
- This is sometimes hard to sketch analytically, so we can look to the **nullclines**: places where one of the components of the direction field is zero.
- In this case, $r_bB kBP = (r_b kP)B = 0$ when P = 0 or $P = \frac{r_b}{k}$, and $-d_pP + kBP = (kB d_p)P = 0$ when P = 0 or $B = \frac{d_p}{k}$.
- The **upshot of nullclines** (since we don't care about $P, B \leq 0$): The lines $B = \frac{d_p}{k}, P = \frac{r_b}{k}$ divide the plane into pieces where the components of this (continuous) function pair can't change sign.
- For instance, $B > \frac{d_p}{k}$, $P < \frac{r_b}{k}$ means $r_b b k b p > 0$, $-d_p p + k d p > 0$, so both populations are growing here. This helps to sketch the curve.
- The curve looks like a counterclockwise whirlpool around the $(B, P) = (\frac{d_p}{k}, \frac{r_b}{k})$. (bacteria grow with low but growing phages; bacteria decrease as phages overwhelm; both decrease as phages starve; bacteria start coming back)
- The center point is a (constant **equilibrium**) solution, and other solutions swirl around it but don't get attracted or repelled.

There are a few types of equilibria:

- This one is a **center** around which solutions circle.
- A stable equilibrum would see small upsets come back to an unchanging state.
- An unstable equilibrum would see small upsets create wildly divergent paths.

2.2 2.2: Lotka-Volterra II

In the Bacteria-Phage system, we can't yet prove everything rotates around the **center**. Let's do that.

Developing a **conserved quantity** will help to do that. **Example**: Block on a horizontal spring with mass m, spring constant k_s :

- x(t): Displacement from rest position.
- $v(t) = \frac{dx}{dt}$: Horizontal velocity
- $\frac{dv}{dt} = -\frac{k_s}{m}x(t)$ by Hooke's law, I think.
- Suppose there's some Energy function E(x,v). By chain rule $\frac{d}{dt}E(x(t),v(t)) = \frac{dE}{dx}\frac{dx}{dt} + \frac{dE}{dv}\frac{dv}{dt}$
- = $\frac{dE}{dx}v \frac{k_s}{m}\frac{dE}{dv}x$. If we set E as conserved, as in E'(t) = 0, then $\frac{dE}{dx}v = \frac{k_s}{m}\frac{dE}{dv}x$
- We can eyeball and see that $E = \frac{1}{2}k_sx^2 + \frac{1}{2}mv^2$ solves this equation, or we can assume $E(x,v) = F(x) + G(v) \Rightarrow 0 = E'(t) = F'(x)v \frac{k_s}{m}G'(v)x = 0$ from the above equations and guess from there.
- This means in the xv phase space, that there's a fixed E such that the particle follows the ellipse $E = \frac{1}{2}k_sx^2 + \frac{1}{2}mv^2$ in phase space around the solution point (0,0).

Extended Example: Continuing on finding a conserved quantity for Bacteria / Phage:

- We need to find U(b(t), p(t)) such that U'(t) = 0, or by chain rule $\frac{\delta U}{\delta b} \frac{\delta b}{\delta t} + \frac{\delta U}{\delta p} \frac{\delta p}{\delta t} = 0$
- Subbing in, $\frac{\delta U}{\delta b}[r_b b kbp] + \frac{\delta U}{\delta p}[-d_p p + kbp] = 0$
- A hint suggests finding U such that $\frac{\delta U}{\delta b} = -\frac{d_p}{b} + k$, $\frac{\delta U}{\delta p} = -\frac{r_b}{p} + k$ to make terms cancel.
- Integrating these gives us U as both $-d_p \ln(b) + kb + Q(p)$ and $-r_b \ln(p) + kp + R(b)$ so $U = -d_p \ln(b) r_b \ln(p) + kb + kp$. This weird curve consistutes a level set in pb-space upon which a solution sits.
- The spring example has an elliptic paraboloid solution. There's an absolute minimum (E = 0 at (0,0)) but level sets become closed loops away from it.

- For the Lotka example, there is a critical point $(\nabla U = \vec{0})$ when $\nabla U(b, p) = (\frac{\delta U}{\delta b}, \frac{\delta U}{\delta p}) = (k \frac{d_p}{b}, k \frac{r_b}{p})$, which is (0, 0) at our known center $(\frac{d_p}{k}, \frac{r_b}{k})$
- Showing that we always increase gong away from the point $(\frac{d_p}{k}, \frac{r_b}{k})$ should guarantee us closed level sets.
- One method: Assume we're picking a unit vector $\vec{v} = \langle \hat{v}_b, \hat{v}_p \rangle$ so that our line from our center is $\vec{v} = \langle \frac{d_p}{k} + tv_b, \frac{r_b}{k} + tv_b \rangle$. U = F(b) + G(p) in this case, so sub the b part into F to get $F(\frac{d_p}{k} + t\hat{v}_b) = d_p[1 \ln(\frac{d_p}{k} + t\hat{v}_b)] + kt\vec{v}_b$. Taking derivative of that w.r.t t shows it is always positive. Same goes for the G(p) portion of U.
- Another (DF) method: Note that $\nabla U = (k \frac{d_p}{b}, k \frac{r_b}{p})$'s grad (second derivative) is always positive. So derivative always has positive curvature (maybe using that term wrong), and we'll always increase around this point.
- Also, we know that the particle travels around the level set (loop) and doesn't reverse course, because then, b'(t) = p'(t) = 0, and we only have that at the center point (nullcline intersection).

2.3 2.3: Linearization

Extended Example: Suppose there's a limit to bacterial growth, so we cap our population at M_b .

- If $b(t) << M_b$, things should be similar. If b(t) is nearly M_b , then growth should approach 0. So, this implies $\frac{db}{dt} = r_b b(t) \to \frac{db}{dt} = r_b b(t) (1 \frac{b(t)}{M_b})$. Note: This isn't the only possibility but we'll use it.
- This updates our Lotka-Volterra model to something more complicated:

$$-b'(t) = r_b b(t) (1 - \frac{b(t)}{M_b}) - kb(t)p(t)$$

$$-p'(t) = -d_p p(t) + kb(t)p(t)$$

- Other than b = 0, p = 0, the meaningful nullclines are solved by setting b'(t) = 0 (yielding $r_b(1 \frac{b}{M_b}) kp = 0$) and p'(t) = 0 (yielding $b = \frac{d_p}{k}$)
- Note: We'll clean up through some MAGIC non-dimensionalization (how to derive?) to simplify:

$$-x(t) = \frac{1}{M_b}b(\frac{t}{r_b}), y(t) = \frac{k}{r_b}(\frac{t}{r_b}), \alpha = \frac{d_p}{r_b}, \beta = \frac{kM_b}{r_b}$$

- Gives us new equations: $\frac{dx}{dt} = x(t)[1-x(t)] x(t)y(t), \frac{dy}{dt} = -\alpha y(t) + \beta x(t)y(t)$
- And new nullclines: $x + y = 1, x = \frac{a}{b}$

- So there's an equilibrium point in the positive xy quadrant if: $y = 1 x = 1 \frac{\alpha}{\beta}$ and y > 0 implies $1 \frac{\alpha}{\beta} > 0 \Rightarrow \frac{\alpha}{\beta} < 1$
- Looking at the direction field, it appears solutions swirl around and are attracted into the center point $(\frac{\alpha}{\beta}, 1 \frac{\alpha}{\beta})$, making it a **stable equilibrum**

This is similar to the block-spring example, if a damping term $-\frac{\gamma}{m}v$ is added.

- $\frac{dx}{dt} = v, \frac{dv}{dt} = -\frac{k_s}{m}x \frac{\gamma}{m}v$
- This can be thought of in matrix terms: $\frac{d}{dt} \begin{pmatrix} x(t) \\ v(t) \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -\frac{k_s}{m} & -\frac{\gamma}{m} \end{pmatrix} \begin{pmatrix} x(t) \\ v(t) \end{pmatrix}$ Call the matrix A.
- From Diff Eq I, the solution is exp(tA) (matrix exponential), making x(t) a linear combination of $e^{\lambda t}$ or possibly $te^{\lambda t}$ terms, with the eigenvalues as λs .
- The eigenvalues in this case, using the quadratic formula, could be:
 - Two real, distinct, negative roots. So, these $e^{\lambda t}$ terms decay, and x(t) levels off.
 - Two distinct complex roots with real part $-\frac{\gamma}{2m} < 0$. This ends up being some sines and cosines multiplied by $e^{-\frac{\gamma t}{2m}}$, which decays too.
 - Finally, if we have a repeated negative real eigenvalue, we have solution $x(t) = Ae^{-\frac{\gamma t}{2m}} + Bte^{-\frac{\gamma t}{2m}}$, also decaying.
 - So any disturbance in the spring will oscillate and come to rest at x(t) = v(t) = 0 quickly.

So with linear systems $\vec{x}'(t) = A\vec{x}(t)$, the eigenvalues determine what happens around the equilibrium point. However, the **bacteria-phage model is non-linear**. Here is **how we linearize** for nearby solutions in a nonlinear system:

- Set small disturbance $\delta x(t) << 1, \delta y(t) <<$ so $x(t) = \frac{\alpha}{\beta} + \delta x(t), y(t) = 1 \frac{\alpha}{\beta} + \delta y(t)$
- Since they're small, all powers like $\delta x(t)^2$ and $\delta x(t)\delta y(t)$ are considered basically zero.
- So substitute $x(t) \to \frac{\alpha}{\beta} + \delta x(t), y(t) \to 1 \frac{\alpha}{\beta} + \delta y(t)$ into our $\frac{dx}{dt}$ and $\frac{dy}{dt}$ equations.
- This gives us the A solving $\frac{d}{dt} \begin{pmatrix} \delta x(t) \\ \delta y(t) \end{pmatrix} = A \begin{pmatrix} \delta x(t) \\ \delta y(t) \end{pmatrix}$, which is $A = \begin{pmatrix} -\frac{\alpha}{\beta} & -\frac{\alpha}{\beta} \\ \beta \alpha & 0 \end{pmatrix}$ after working through the substitution.

 Finding the eigenvalues here yields the same situation as the block-spring example: decays in all situations.

It turns out through the **Hartman-Grobman Theorem** that $\vec{x}'(t) = \vec{F}(\vec{x}(t))$, for some continuously differential vector field F, if we linearize near equilibrium x_0 , then what falls out of this A approach works if the eigenbalues aren't all purely imaginary.

It turns out the uncapped bacteria system from before looks like $\frac{d}{dt} \begin{pmatrix} \delta x(t) \\ \delta y(t) \end{pmatrix} =$

 $\begin{pmatrix} 0 & -1 \\ \alpha & 0 \end{pmatrix} \begin{pmatrix} \delta x(t) \\ \delta y(t) \end{pmatrix}$, with characteristic equation $\lambda^2 + \alpha = 0, \alpha > 0$. This means both values are imaginary, and we had to use the conserved quantity appraoach!

3 2.4: Hartman-Grobman Theorem

Extended Example: Consider a phage that dies off quicky:

- $\frac{db}{dt} = r_b b(t) k_b b(t) p(t)$, $\frac{dp}{dt} = -r_p p(t) = 0 \cdot b(t) p(t)$, where k_p is the zero (phages don't increase), and k_b is still the kill factor for the bacteria.
- In this base, b(t) = p(t) = 0 is the only equilibrium.
- Non-dimensionalize as $x(t) = b(\frac{t}{r_b}), y(t) = \frac{k_b}{r_b} p(\frac{t}{r_b}), \alpha = \frac{r_p}{r_b}$
- This makes the equations $x'(t) = x(t) x(t)y(t), y'(t) = -\alpha y(t)$, and the nullclines therefore x(t) = 0, y(t) = 1, y(t) = 0
- Looking at this six-section dierection field, we see that solutions exactly on the y-axis are attracted to equilibrium (0,0), and other are repelled.
- This makes sense since if the bacteria is 0, the phage die and approach (0,0), otherwise the bacteria multiply and win (so it's a *saddle point*)
- The way to tell: linearize the equations. $\frac{d}{dt} \begin{pmatrix} x(t) \\ y(t) \end{pmatrix} \approx \begin{pmatrix} 1 & 0 \\ 0 & -\alpha \end{pmatrix} \begin{pmatrix} x(t) \\ y(t) \end{pmatrix}$ since, if x(t), y(t) << 1, x(y)t(t) = 0.
- Then the eigenvalues are $\lambda = 1, -\alpha$ so the solution is $Ae^t, Be^{-\alpha t}$ for x(t), y(t) (TODO respectively?) Hartman-Grobman ensures this is the general solution.

However, let's solve directly and see if we come to the same result.

- $y'(t) = -\alpha y(t), y(0) = y_0 \Rightarrow y(t) = y_0 e^{-\alpha t}$
- With this in hand, $\frac{dx}{dt} = x(t) x(y)y(t) = x(t)[1 y_0e^{-\alpha t}], x(0) = x_0$ separates out to

$$-\frac{dx}{x} = [1 - y_0 e^{-\alpha t}] dt$$

$$-\ln(x) = [t + \frac{y_0}{\alpha} e^{-\alpha t}] + C$$

$$-x = e^C e^t \exp(\frac{y_0}{\alpha} e^{-\alpha t})$$

$$-x(0) = x_0 \Rightarrow e^C = x_0 e^{-\frac{y_0}{\alpha}}$$

$$-\Rightarrow x(t) = x_0 e^t \exp(\frac{y_0}{\alpha} (e^{-\alpha t} - 1))$$

But how do we deform the phase plane so this looks linear? We need some mapping $\vec{h}(x,y) = \langle u(x,y), v(x,y) \rangle$ that is continuous and invertible (so we don't "damage" the phase plane). This is called a **homeomorphism**.

- So near the equilibrium (0,0), the equations $y'(t) = -\alpha y(t), y(0) = y_0 \Rightarrow y(t) = y_0 e^{-\alpha t}$ linearized for $\delta x, \delta y$ must be similar to those for u(x(t), y(t), v(x(t), y(t)))
- This means we need $\frac{du}{dt} = u$, $\frac{dv}{dt} = -\alpha v$
- After doing the substitution, we see that $v = v_0 e^{-\alpha t}$ exactly mimics $y(t) = y_0 e^{-\alpha t}$ for the phage solution. So we take v = y.
- Therefore, we know that since $u = u_0 e^t$ and $x(ty) = x_0 \exp(t + \frac{y_0}{\alpha}(e^{-\alpha t} 1))$, that we need $u(x(t), y(t) = u(x_0, y_0)e^t$
- And this is satisfied if we guess $u(x,y) = xe^{-y}\alpha$ and work it out.
- This function $\vec{h}(x,y) = (u,v) = \langle xe^{-\frac{y}{\alpha}},y \rangle$ is invertible by $(x,y) = \langle ue^{\frac{v}{\alpha}},v \rangle$, which is continuous.

4 2.5: Application - Lasers

Lasers create excited atoms, which then emit photos while transitioning to an unexcited state. This system has a close analogue with the previous phages (like photons) and bacteria (like atoms) model.

- n(t): number of photons in the laser; r_g : rate of photons gained (created by excited atoms transitioning to unexcited state); r_l : rate of photons lost (emitted)
- $\Rightarrow \frac{dn}{dt} = r_g r_l$ by definition.
- We can assume we're losing a constant k (kill?) portion of photons per unit time, so $\frac{dn}{dt} = r_g kn(t)$
- e(t): number of excited atoms (that will maybe create photons). Atoms are excited by external energy pump.
- Excited atoms radiate when meeting a photon (which survives the meeting)

• So we can use the same setup from the bacteria: with I the constant of meeting (intersect?), $r_g = Ie(t)n(t) \Rightarrow n'(t) = Ie(t)n(t) - kn(t)$

Mini example: Assume no photons leave (cap the end of the laser)

- k = 0 in this scenario.
- So every meeting creates one more photon $(n \to n+1)$ while enervating one excited atom $(e \to e-1)$. This implies, equivalently:
 - -e+n is a conserved quantity,

$$-e(t) + n(t) = e(0) + n(0),$$

$$- [e(t) + n(t)]' = 0$$

- Then, if k = 0, n'(t) = Ie(t)n(t) - kn(t), and coupled with e'(t) + n'(t) = 0 above, we have e'(t) = -Ie(t)n(t)

Extended example: Atoms spontaneously lose energy. This is actually what happens

- \bullet From quantum physics, we have a rate s of atoms just (s)pontaneously losing energy.
- We also have an energy (p)ump that energizes atoms with quantity p.
- Then, our change in (e)xcited atoms is e'(t) = p s Ie(n)(t)
- If we want to find the smallest p guaranteeing $n \ge 1$ (there's at least one photo) at equilibrium (e'(t) = n'(t) = 0):

$$-n'(t) = 0 \Rightarrow Ien = kn \Rightarrow n(Ie - k) = 0.$$
 If $n \neq 0, \Rightarrow e = \frac{k}{I}$

$$-e'(t) = 0 \Rightarrow Ien = p - se$$

- Together,
$$p - se = Ien = kn \Rightarrow kn + se = p \Rightarrow kn + s\frac{k}{I} = p$$

$$-n \ge 1 \Rightarrow p \le k + \frac{ks}{I}$$

- Another tactic: We could also assume we start out at equilibrium, so n_0, e_0 are constant solutions.

- Solving
$$n' = 0 = Ie_0n_0 - kn_0, e' = 0 = Ie_0n_0 - se_0 + p$$
, we find equilibria $n_0 = \frac{p}{k} - \frac{s}{I}, e_0 = \frac{k}{I}$

- Then,
$$n_0 \ge 1 \Rightarrow \frac{p}{k} - \frac{s}{I} \ge 1 \Rightarrow p \ge k + \frac{ks}{I}$$

Non-dimensionalization time:

• Scale against $e_0(=\frac{k}{I}), n_0(=\frac{p}{k}-\frac{s}{I})$ like this: $x(t)=\frac{n(\alpha t)}{n_0}, y(t)=\frac{e(\alpha t)}{e_0}$

- NOTE: What does this do? This makes (1,1) the equilibrium, as $x(t) = \frac{n_0}{n_0} = 1, y(t) = \frac{e_0}{e_0} = 1$!
- What α lets us the n' = Ien kn, e' = -Ien se + p and write

$$-\frac{dx}{dt} = x(t)y(t) - x(t)$$

$$-\frac{dy}{dt} = \frac{1}{k}(\frac{pI}{k} - s)[1 - x(t)y(t)] + \frac{s}{k}[1 - y(t)]$$

$$-x' = \frac{\alpha n'(\alpha t)}{n_0} = xy - x = \frac{Ie(\alpha t)n(\alpha t)}{kn_0} - \frac{n(\alpha t)}{n_0}$$

$$-\frac{\alpha Ien - \alpha kn(\alpha t)}{n_0} = \frac{Ie(\alpha t)n(\alpha t)}{kn_0} - \frac{n(\alpha t)}{n_0}$$

$$-\alpha Ie - \alpha k = \frac{Ie(\alpha t)}{k} - 1 \Rightarrow \alpha (Ie - k) = \frac{Ie - k}{k} \Rightarrow \alpha = \frac{1}{k}$$

- This solves the x equation, and I suppose it can be validated in the y equation (tediously).
- If we chunk up our (somehow positive?) constants as $c = \frac{1}{k}(\frac{pI}{k} s), d = \frac{s}{k}$, we end up with y' = c[1 xy] + d[1 y]
- We only care about x, y > 0, so x' = 0 = xy x = x(y 1) implies y = 1 is a nullcline
- $-y'=0=c[1-xy]+d[1-y]=c-cxy+d-dy\Rightarrow c+d=y(d+cx)\Rightarrow y=\frac{c+d}{d+cx}$ a scaled and shifted hyperbola.

Look at the solutions:

- It appears we have a counterclockwise swirl around (1,1), and nearby solutions tend toward this equilibrium.
- Hartman-Grobman: rewrite our linearized solution in neighborhood of (1,1) as $x(t) = 1 + \delta x(t), y(t) = 1 + \delta y(t)$

- Using
$$x' = xy - x$$
, $y' = c[1 - xy] + d[1 - y]$ and $\frac{d}{dt} \begin{pmatrix} \delta x(t) \\ \delta y(t) \end{pmatrix} = A \begin{pmatrix} \delta x(t) \\ \delta y(t) \end{pmatrix}$, we can solve and write $A = \begin{pmatrix} 0 & 1 \\ -c & -c - d \end{pmatrix}$

- Eigenvalues: $\lambda = \frac{1}{2}(-c d \pm \sqrt{(c+d)^2 4c})$
 - * If square root term is zero, we have repeated eigenvalue, so $\delta x(t), \delta y(t)$ are combos of $e^{-\frac{c+d}{2}}, te^{-\frac{c+d}{2}}$, which decays
 - * If square root term is greater than zero, we have two distinct real, negative eigenvalues (since c, d are positive), so this decays.

- * If square root term is less than zero, we have distinct complex eigenvalues, but combos of $e^{-\frac{c+d}{2}}\cos(\frac{1}{2}\sqrt{-(c+d)^2+4c}), e^{-\frac{c+d}{2}}\sin(\frac{1}{2}\sqrt{-(c+d)^2+4c})$ decay too
- * Note: I suppose Hartman-Grobman can't work in purely imaginary scenario, because these kinds of functions don't converge or diverge without a term outside the sin or cos (?)
- \ast And in any case, since these lambdas aren't strictly imaginary, Hartman-Grobman works.

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