

Simple Models in a Complex World

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What in the world do I do?

That's a tough question.

What in the world do I do?

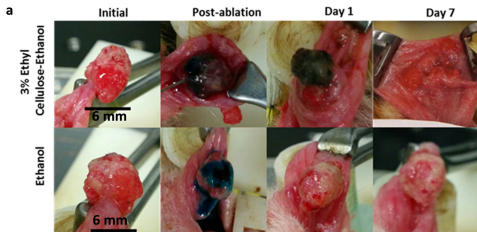
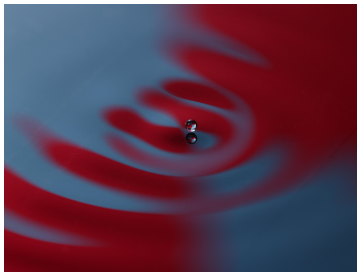
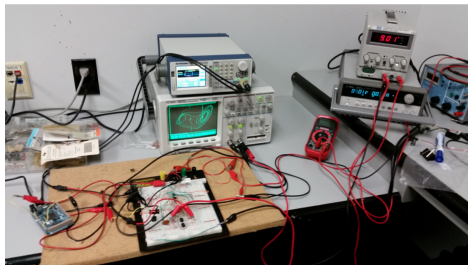
Mechanistically modeling problems arising from real world phenomena, often using the tools of dynamical systems and bifurcation theory, with an inclination for reduced (“simple”) models and a slight affinity for theory.

Why reduced modeling?

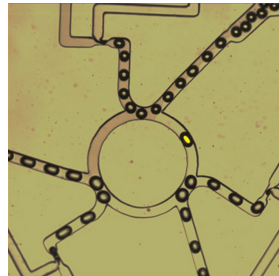
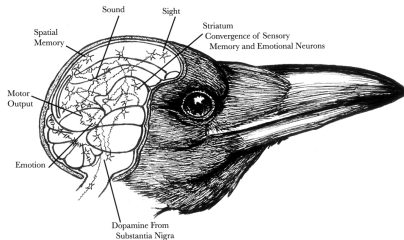
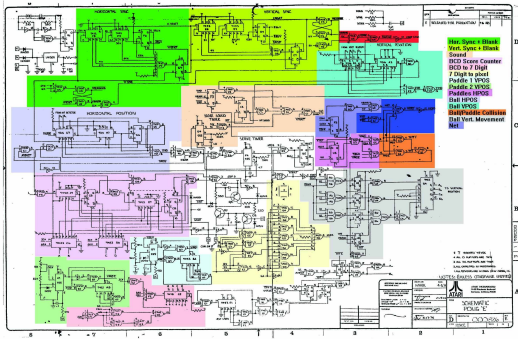
Everything should be made as simple as possible, but not simpler.

Albert Einstein

Stuff I'll be talking about today



Logical Circuits



History: Logic Gates

AND



x	y	$F(x,y) := x \wedge y$
0	0	0
0	1	0
1	0	0
1	1	1

NOT



x	$F(x,y) := \neg x$
0	1
1	0

OR



x	y	$F(x,y) := x \vee y$
0	0	0
0	1	1
1	0	1
1	1	1

NAND



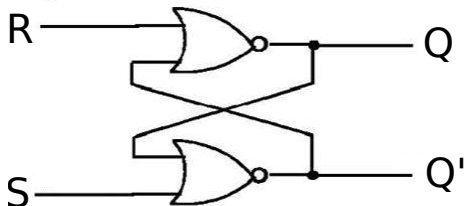
x	y	$F(x,y) := \neg(x \wedge y)$
0	0	1
0	1	1
1	0	1
1	1	0

NOR



x	y	$F(x,y) := \neg(x \vee y)$
0	0	1
0	1	0
1	0	0
1	1	0

History: RS flip-flop



Inputs		Outputs	
R	S	Q	Q'
0	1	1	0
1	0	0	1
0	0	keep	keep
1	1	undefined	undefined

History: Chaotic Circuits

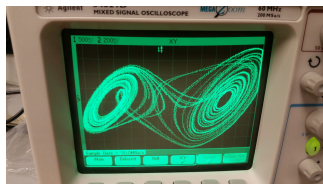
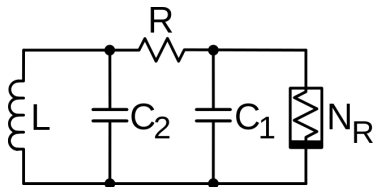


Figure: Source: [Chua 1983(Patent)]

$$\begin{aligned}C_1 \frac{dV_{C_1}}{dt} &= G(V_{C_2} - V_{C_1}) - g(V_{C_1}) \\C_2 \frac{dV_{C_2}}{dt} &= G(V_{C_1} - V_{C_2}) + \chi_L \\L \frac{d\chi_L}{dt} &= -V_{C_2}\end{aligned}\tag{1}$$

Chaotic NOR gate/RS flip-flop

- Some SPICE sims, but no experiments and no models (till now).
- System is too complex for traditional ODE models.

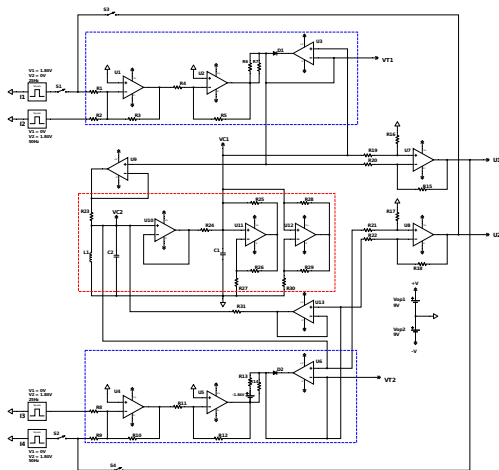
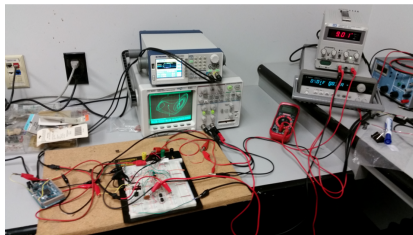
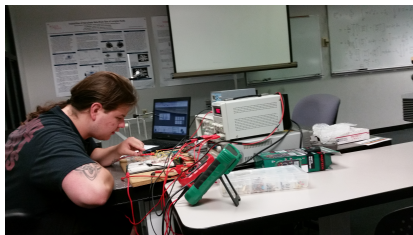


Figure: Source: [Rahman, Jordan, Blackmore. *Proc. Roy. Soc. A* (2018)]

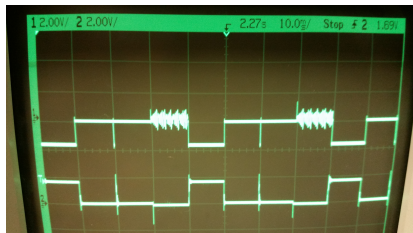
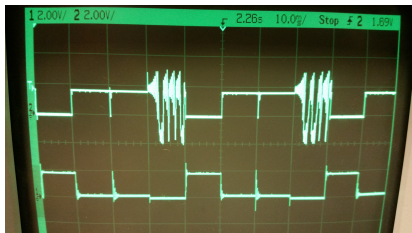
Experimentation



- Designed new version of chaotic RS flip-flop
- Built physical realization
- Simulations agree with the experiments
- Components caught on fire (oops)
- Testing the possibility of chaotic encryption

Modeling Thresholds

Experimental results:



Let ξ and η represent the threshold voltages

$$\begin{aligned}\xi_{n+1} &= f(l_1, l_2, \xi_n), \\ \eta_{n+1} &= g(l_3, l_4, \eta_n);\end{aligned}\tag{2}$$

What we know from direct observation:

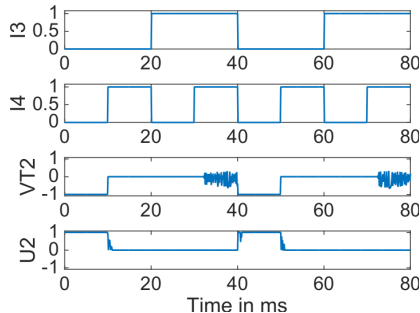
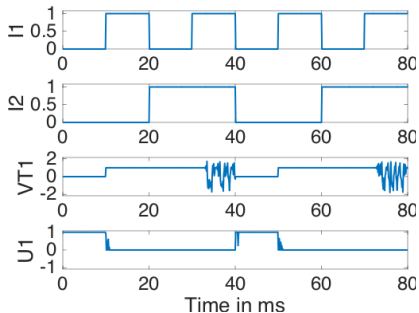
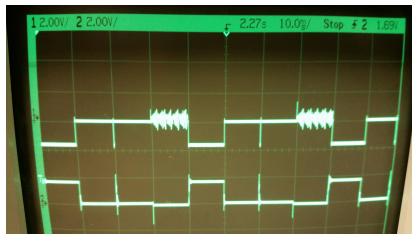
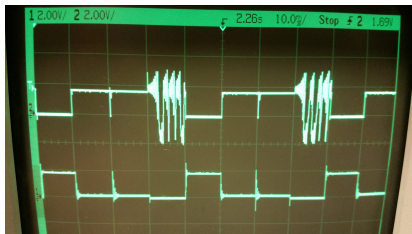
$$\begin{array}{lll}f(0, 0, 0) = 0 & f(1, 0, 1) = f(0, 1, 1) = 1 & f(1, 1, \star) = \star, \\ g(0, 0, -1) = -1 & g(1, 0, 0) = g(0, 1, 0) = 0 & g(1, 1, \star) = \star.\end{array}\tag{3}$$

To satisfy the above observations we let

$$\begin{aligned} f(l_1, l_2, x_1) &:= |l_1 - l_2| + l_1 l_2 y_f(x_1), \\ g(l_3, l_4, x_2) &:= |l_3 - l_4| - 1 + l_3 l_4 y_g(x_2); \end{aligned} \tag{4}$$

- y are combinations or modifications of tent maps.
- $y_g - 1$ has a f.p. at $x = 0$.
- y_f has two nonzero f.p.'s near ± 1 .

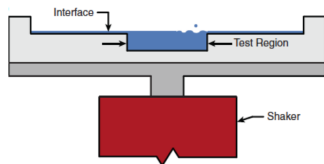
Simulations vs. Experiments



Current and future work

- Develop a capacitor-based (rather than our TCU-based) model and compare.
- Build neuronal analogs.
- Model microfluidic circuits.

Walking in confined geometry



(“Free” Space)

Figure: [Harris *et al.*, PRE 2013]

Figure:

http://math.mit.edu/~bush/?page_id=252

- (Couder *et al.* *Nature* (2005)) observed walking dynamics.
- Groups at Paris, MIT, and others have made significant experimental and theoretical contributions since.
- Analogous to wave-particle duality.
- Can produce quantum-like behavior.

Walking in rotating frame

(Corral)

Figure: http://math.mit.edu/~bush/?page_id=252

Discrete Dynamical Model

Originally developed by Gilet (PRE 2014).

$$\begin{bmatrix} w_{n+1} \\ x_{n+1} \end{bmatrix} = \begin{bmatrix} \mu[w_n + \Psi(x_n)] \\ x_n - Cw_n\Psi'(x_n) \end{bmatrix} \quad (5)$$

Simplifying assumptions:

- Droplet shifted proportionally to the gradient of the wave field.
- Droplet moving in a straight line.
- Only one eigenmode is excited in a confined geometry.

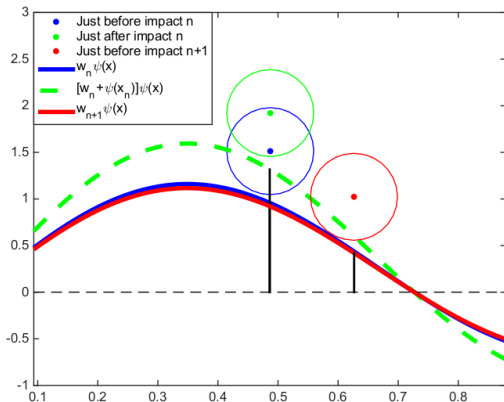
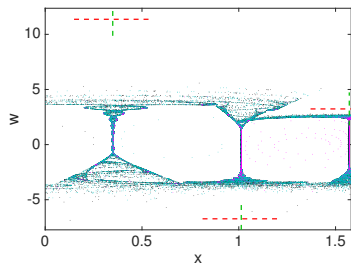


Figure: Source: [Rahman and Blackmore, CSF 2016]

Simulations



(Chaos)

Figure: $\mu = 0.94$

Confining the walker

$$w(n+1) = \mu[w_n + \Psi(x_n)] \quad x(n+1) = [x_n - Cw_n\Psi'(x_n)] \mod 2\pi$$

$$\Psi(x, \beta) = \frac{\cos \beta}{\sqrt{\pi}} \sin 3(x - \pi/2) + \frac{\sin \beta}{\sqrt{\pi}} \sin 5(x - \pi/2)$$

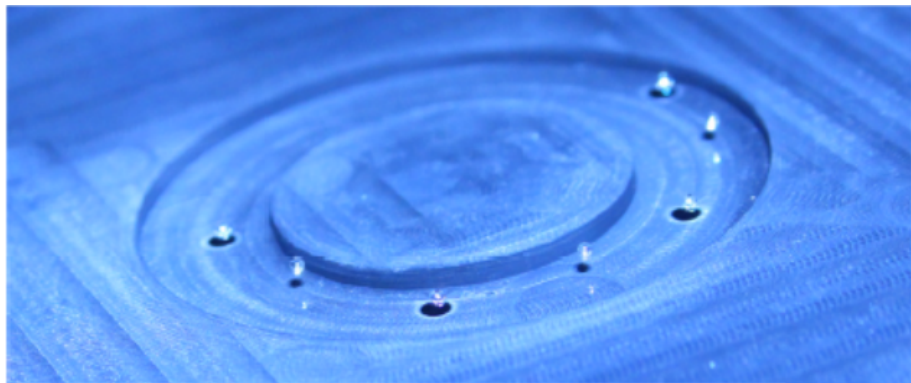


Figure: Source: [Filoux, Hubert, Vanderwalle. *Phys. Rev. E*(2015)]

Current and future work

- Working on kicked rotator - like model on the annulus.
- Working on model for elliptical corral.
- Connections between dynamics and statistics.

Drug diffusion: Story time

Two Bengalis, a grant, and Reddit.

- Souparno's NIH grant on individualized treatment.
- Lots of procrastination.
- Redditors write bad science headlines.

THE NEW REDDIT



JOURNAL OF SCIENCE

 **comments** other discussions (10) filter by field ▾

 **Trending:** **Scientists in Japan and Australia have developed a blood test that can det...**



Cancer

Duke University scientists have created a "lethal injection" for tumors. When injected into them, their ethanol-based gel cured 100% of the oral tumors in a small sample of hamsters. This treatment might work for some kinds of breast, liver, and other cancers, and it only costs about \$5.

 [acsh.org](https://www.acsh.org)

5 months ago by [cellular2013](#) [R&D](#) [Medicine](#)


 56.0k  

Drug diffusion: Story time

Two Bengalis, a grant, and Reddit.

- This title is much better.
- Using alcohol to treat solid tumors.

Development of enhanced ethanol ablation as an alternative to surgery in treatment of superficial solid tumors

Robert Morhard , Corrine Nief, Carlos Barrero Castedo, Fangyao Hu, Megan Madonna, Jenna L. Mueller, Mark W. Dewhirst, David F. Katz & Nirmala Ramanujam

Scientific Reports **7**, Article number: 8750

(2017)

doi:10.1038/s41598-017-09371-2

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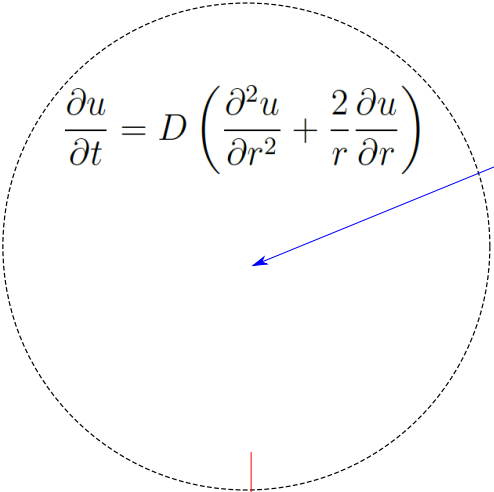
Published online: 18 August 2017

Simplest possible mechanistic model

Assumptions:

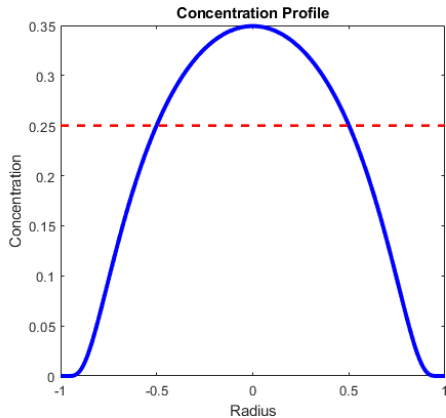
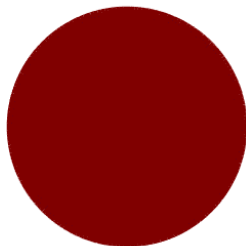
- Spherical solid tumor.
- Diffusion with constant diffusivity.
- Leaky boundary.
- Injection into the center.
- Diffusion begins after injection ends.
- There exists a minimum drug concentration required to kill a cell.

Simplest possible mechanistic model


$$\frac{\partial u}{\partial t} = D \left(\frac{\partial^2 u}{\partial r^2} + \frac{2}{r} \frac{\partial u}{\partial r} \right)$$
$$u(r, t = 0)$$
$$D \frac{\partial u}{\partial r} \bigg|_{r=R} = -\gamma u(r = R, t)$$

Cell death

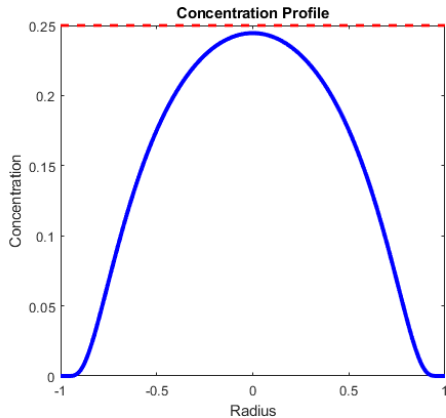
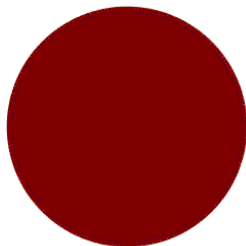
Radial Diffusion



Diffusion

No ablation

Radial Diffusion



Diffusion

Partial ablation

(Partial Ablation)

Diffusion

Full ablation

(Full Ablation)

Current and future work

- Model the relationship between time and threshold.
- Using statistics to connect the model with population data.
- Making individualized predictions.
- Analyze our other models.

List of Problems

November 14, 2017

1 Artificial Reservoir

Diffusion of drugs in uniform solid tumor. Leakage after tumor boundary activation. Initial injection into center. Radially symmetric.

$$\frac{\partial c}{\partial t} = D \left(\frac{\partial^2 c}{\partial r^2} + \frac{2}{r} \frac{\partial c}{\partial r} \right) \quad \frac{\partial c}{\partial r} = -\gamma_0 \begin{cases} 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ \frac{1}{r} & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \end{cases} \quad (1)$$

where

$$h_0 = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume.

1. Constant threshold for amount of drug it takes to kill a cell. Assumes zero oxygen gradient.
2. Constant threshold for amount of drug it takes to kill a cell. Assumes variation in oxygen gradient.
3. Can produce dose-response curve.
4. Use data to go from a dose-response curve to dose-response threshold.
5. Assume the tumor acts as a constant buffer binding.

2 Natural Leak

Diffusion of drugs in uniform solid tumor. Constant leakage. Initial injection into center. Radially symmetric.

$$\frac{\partial c}{\partial t} = D \left(\frac{\partial^2 c}{\partial r^2} + \frac{2}{r} \frac{\partial c}{\partial r} \right) \quad \frac{\partial c}{\partial r} = -\gamma_0 \begin{cases} 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ \frac{1}{r} & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \end{cases} \quad (2)$$

where

$$h_0 = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume.

1. Constant threshold for amount of drug it takes to kill a cell. Assumes zero oxygen gradient.
2. Constant threshold for amount of drug it takes to kill a cell. Assumes variation in oxygen gradient.
3. Can produce dose-response curve.
4. Use data to go from a dose-response curve to dose-response threshold.

3 Buffer Zone

Diffusion of drugs in uniform solid tumor. Leakage on outer buffered boundary. Initial injection into center. Radially symmetric.

$$\frac{\partial c}{\partial t} = D \left(\frac{\partial^2 c}{\partial r^2} + \frac{2}{r} \frac{\partial c}{\partial r} \right) \quad \frac{\partial c}{\partial r} = -\gamma_0 \begin{cases} 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ \frac{1}{r} & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \end{cases} \quad (3)$$

where

$$h_0 = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume.

1. Assumes no leak threshold, but neither diffusion up to some buffer and leakage after.
2. Constant threshold for amount of drug it takes to kill a cell. Assumes zero oxygen gradient.
3. Constant threshold for amount of drug it takes to kill a cell. Assumes variation in oxygen gradient.
4. Can produce dose-response curve.
5. Use data to go from a dose-response curve to dose-response threshold.

4 Absorption - Diffusion

Diffusion of drugs in uniform solid tumor. Constant leakage. Initial injection into center. Radially symmetric.

$$\frac{\partial c}{\partial t} = D \left(\frac{\partial^2 c}{\partial r^2} + \frac{2}{r} \frac{\partial c}{\partial r} \right) \quad \frac{\partial c}{\partial r} = -\gamma_0 \begin{cases} 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ \frac{1}{r} & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \end{cases} \quad (4)$$

where

$$h_0 = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume.

1. Constant threshold for amount of drug it takes to kill a cell. Assumes zero oxygen gradient.
2. Constant threshold for amount of drug it takes to kill a cell. Assumes variation in oxygen gradient.
3. Can produce dose-response curve.
4. Use data to go from a dose-response curve to dose-response threshold.

5 Off-center Injection

Diffusion of drugs and oxygen in uniform solid tumor. Leakage after tumor boundary activation. Initial injection of center tumor.

$$\frac{\partial c}{\partial t} = D \left(\frac{\partial^2 c}{\partial r^2} + \frac{2}{r} \frac{\partial c}{\partial r} \right) \quad \frac{\partial c}{\partial r} = -\gamma_0 \begin{cases} 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ \frac{1}{r} & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \end{cases} \quad (5)$$

where

$$h_0 = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume.

1. Constant threshold for amount of drug it takes to kill a cell. Assumes zero oxygen gradient.
2. Constant threshold for amount of drug it takes to kill a cell. Assumes variation in oxygen gradient.
3. Can produce dose-response curve.
4. Use data to go from a dose-response curve to dose-response threshold.

6 Drugs and Oxygen Diffusion

Diffusion of drugs and oxygen in uniform solid tumor. Leakage after tumor boundary activation. Initial injection into center. Radially symmetric.

$$\frac{\partial c}{\partial t} = D \left(\frac{\partial^2 c}{\partial r^2} + \frac{2}{r} \frac{\partial c}{\partial r} \right) \quad \frac{\partial c}{\partial r} = -\gamma_0 \begin{cases} 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ \frac{1}{r} & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \\ 0 & \text{for } x(t) \in R(t) \leq r_{\text{dead}} \end{cases} \quad (6)$$

where

$$h_0 = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume.

Initial condition:

$$c_{\text{dead}}(x, t) = c_0 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) \quad \text{for } x \in V_0 \quad c_{\text{dead}}(x, t) = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume of the drug itself and

$$c_{\text{dead}}(x, t) = c_0 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) \quad \text{for } x \in V_0 \quad c_{\text{dead}}(x, t) = \frac{V_0}{4\pi r_0^2 \left(1 - \frac{r_0}{r_{\text{dead}}} \right) D}$$

where V_0 is the initial volume of the oxygen.

1. Assume a superposition between oxygen and drug after injection.

2. Oxygen profile will determine kill threshold of the drug.

3. Can produce dose-response curve.

4. Use data to go from a dose-response curve to dose-response threshold.

5. Use optimization techniques to maximize efficacy with least toxicity.

7 Coupled Population and Concentration Model

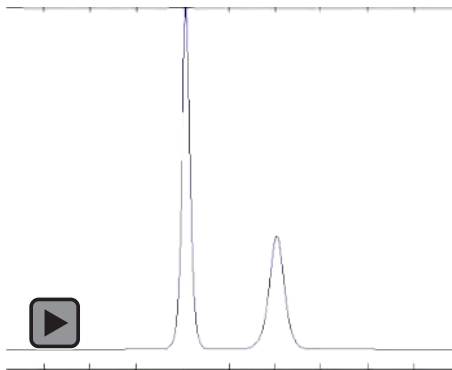
8 Polydispersed Tumor

What came first, the experiment or the model?

Solitary Waves



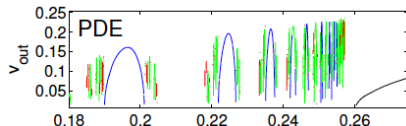
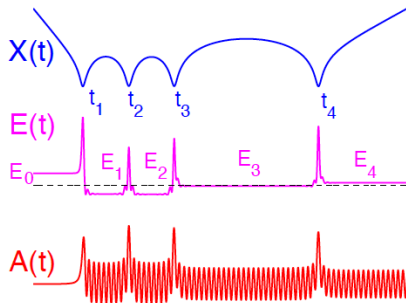
- First discovered by John Scott Russell.
- Not just water waves;
 - Generally solitary waves are solutions to non-linear evolutionary PDEs that translate at a constant speed while maintaining its profile.



- Kruskal-Zabusky(1960): Solitary waves of KdV
($u_t + 6uu_x + u_{xxx} = 0$) preserve shape and velocity after collisions.
- Soon after similar studies on NLS ($iu_t + u_{xx} + |u|^2u = 0$) and Sine-Gordon ($u_{tt} - u_{xx} + \sin u = 0$).
 - “Easy” because they are completely integrable.

History

- In the 70s and 80s many studies on φ^4 ($\varphi_{tt} - \varphi_{xx} + \varphi - \varphi^3 = 0$) were conducted.
 - Much more difficult because it's not integrable.
 - Many numerical experiments on kink-antikink (colliding in a special way) solutions (some observations dubbed “chaotic scattering”).



Recent History

- Reduction of PDE (φ^4) to a system of ODEs that approximates the behavior.

$$\begin{aligned}m\ddot{X}(t) + U'(X) + cAF'(X) &= 0, \\ \ddot{A}(t) + \omega^2 A + cF(X) &= 0; \\ U(X) = e^{-2X} - e^{-X}, \quad F(X) &= e^{-X}\end{aligned}\tag{6}$$

where X is the position of the center of a wave and A is the amplitude.

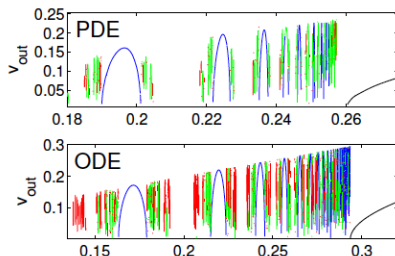


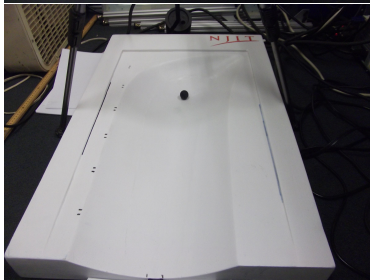
Figure: Source: [Goodman and Haberman. *Phys. Rev. Lett.*(2007)]

Experiment

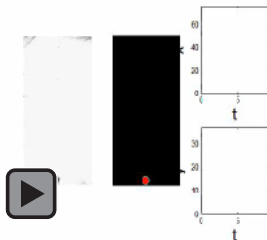
Source: [Goodman, Rahman, Bellanich, Morrison. *Chaos* (2015)]

Experimental set up:

Movies:



(1-bounce movie)



If anyone is interested in knowing more about my work please visit my website: myweb.ttu.edu/aminrahm

Aminur Rahman

Welcome to my website.

A summary of my academic and extracurricular activities. If you don't find what you are looking for feel free to contact me. Also, you may call me **Amin** for short.

[Research Statement \[PDF\]](#)

[Curriculum Vitae \[PDF\]](#)

[Dissertation \[PDF\]](#)

Erdős number: ≤ 5 (source: [MathSciNet](#))

Contact Information

Email: [amin.rahman \[at\] ttu \[dot\] edu](mailto:amin.rahman@ttu.edu)
Office: Math Building 222
Office Phone: 806-834-2545
Vehicle: Bicycle

Research

Publications and Presentations

Teaching

Extracurricular

Videos/Media

I am interested in modeling and analyzing various real world phenomena. Some of the topics I have worked on and/or have an interest in working on are: Hydrodynamic Quantum Analogs, Logical circuit dynamics, Chaotic scattering, Cancer modeling, and Particle Accelerator Physics. To find out more about my current research please refer to my research statement.

Cancer Drug Response:

Model Diffusion

Concentration Profile

And, thank you for your attention!