

Modeling Type 1 Diabetes in NOD Mice

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Sabbatical leave with host
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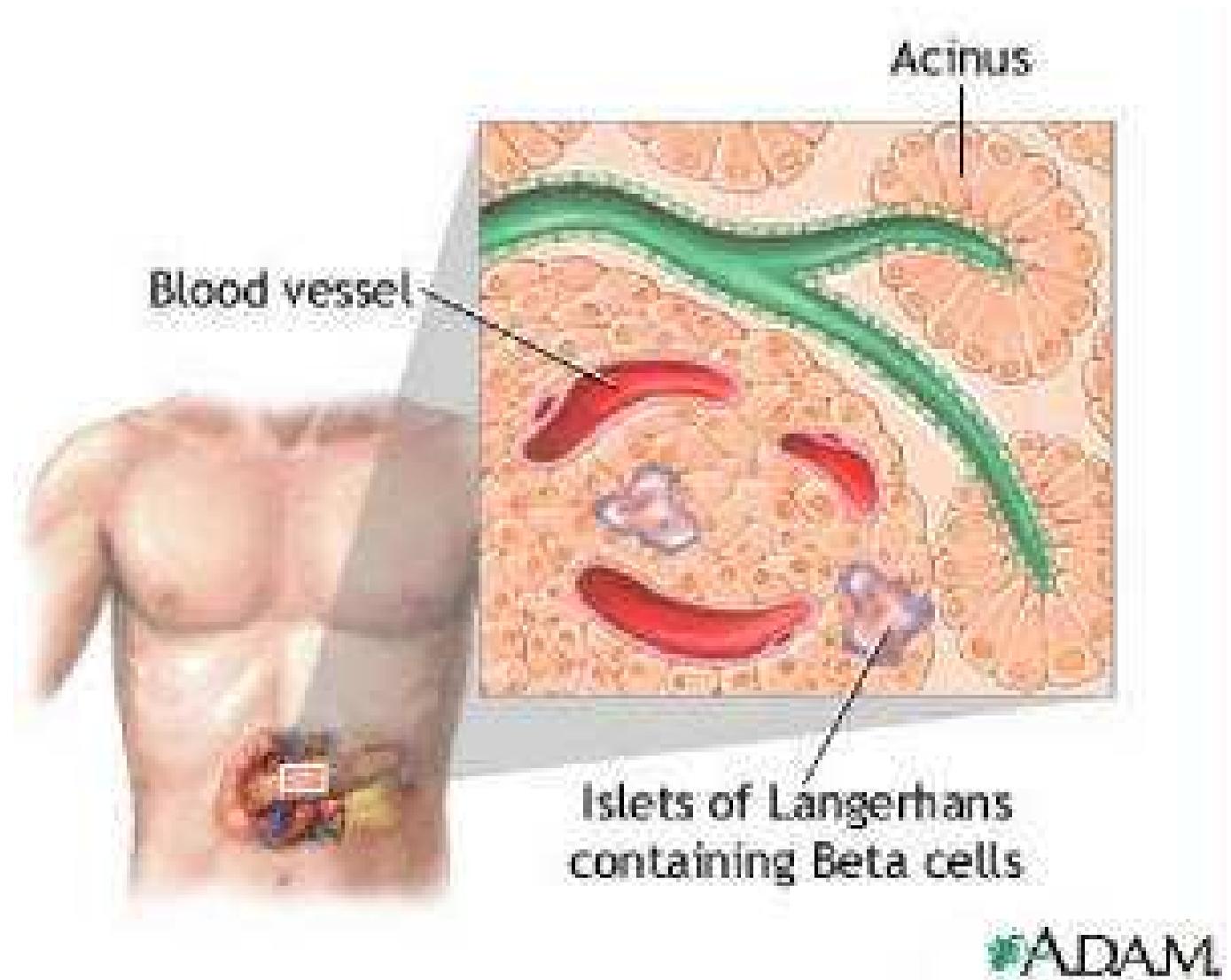
Outline

- Biology - Diabetes and Immune Response
- Mathematical Model
- Bifurcation Analysis
- Simulations
- Discussion and Conclusions

Glucose Metabolism

- Ingest food
 - Breaks down to simple sugars
- Blood absorbs sugar
 - Raises blood glucose concentration
- β cells in pancreas respond
 - Insulin released
- Cells increase glucose uptake
 - Insulin facilitates glucose transport across cell membranes, especially in skeletal muscles
- Glucose converted to glycogen
 - Preferred energy storage of cells
- Blood sugar level decreases
 - Body tightly regulates glucose levels

β Cells - Insulin Release



Type 1 or Juvenile Diabetes - Overview

- Diabetes mellitus results from the loss of β cells
 - An auto-immune disease
- Insulin production is severely reduced
- Hereditary disease - about 4-20 per 100,000 people
- Peak diagnosis occurs around age 14
- 10% of diabetes cases are Type 1, while 90% are Type 2 (where cells become insulin resistant, mostly in obese individuals)
- Treatment is regular injections of insulin - transplants are usually attacked by immune system

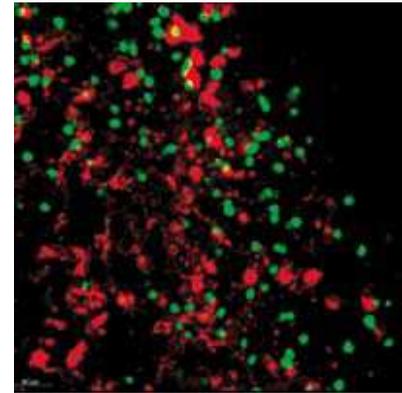
Type 1 or Juvenile Diabetes - Symptoms

- Classic Symptoms
 - Polyphagia (hungry)
 - Polydipsia (thirsty)
 - Polyuria (frequent urination)
- Other Symptoms
 - Blurred vision
 - Fatigue
 - Weight loss
 - Poor wound healing

Type 1 or Juvenile Diabetes - Disease

- Increased heart disease
 - Atherosclerosis from low insulin
- Blindness (retinopathy)
 - Increased pressure in eye
- Nerve damage (neuropathy)
- Kidney damage (nephropathy)

T Cell Activation

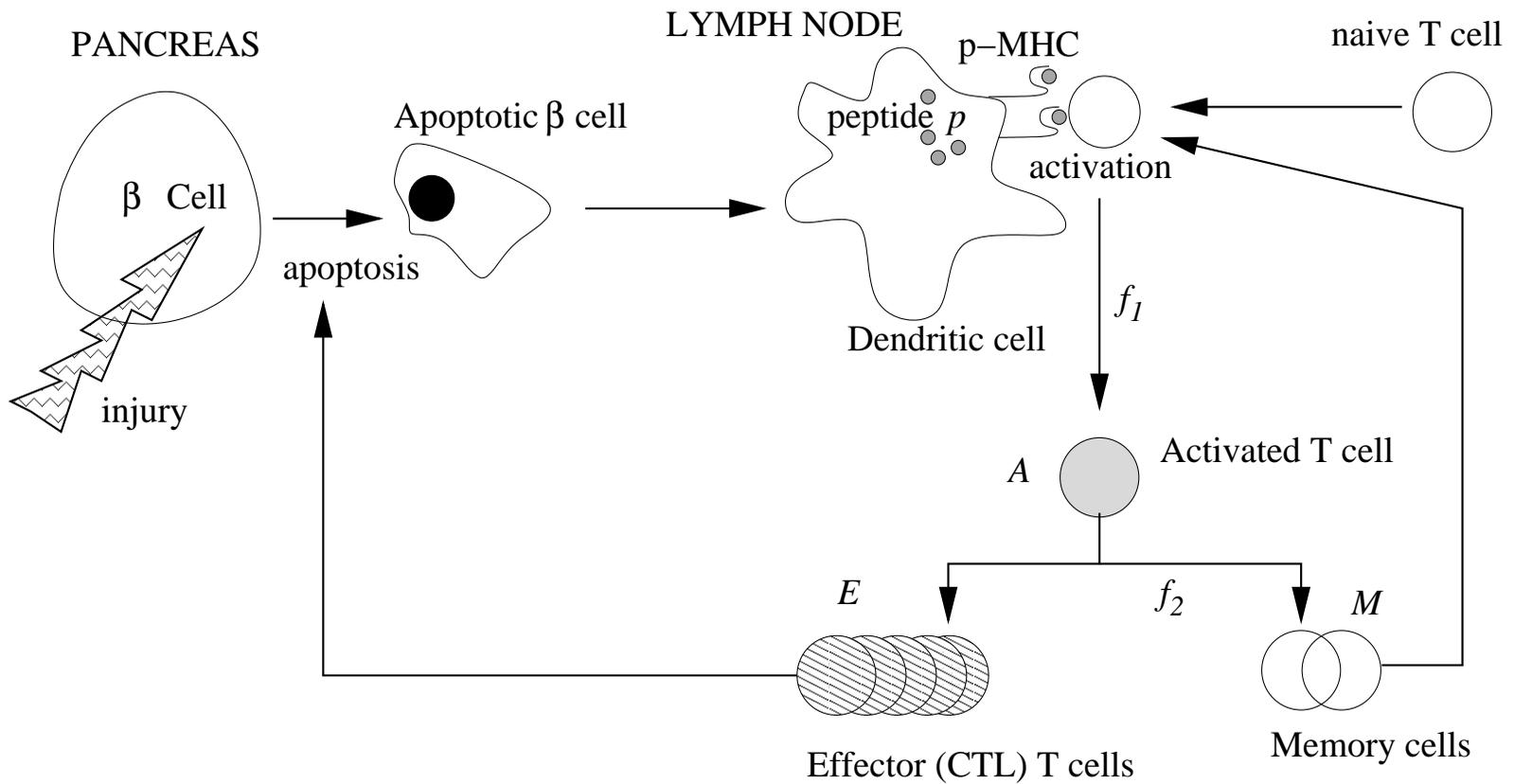


- T cells mature in the thymus
 - Cross-react with self-protein to prevent autoimmunity
- T cells migrate to Lymph nodes
 - Interact with antigen presenting cells (APCs)
 - APCs present antigen protein fragment (about 9 AAs) inside MHC (major histocompatibility complex)
 - The peptide-MHC complex interacts with T cells surface receptors
 - T cells with appropriate specificity become activated
- Most antigens are foreign proteins from viruses and bacteria

T Cell Immune Response

- Activated T cells proliferate about 6 cell divisions
- Most become Effector cells (cytotoxic T-lymphocytes or CTLs)
 - CTLs are efficient specific killers, destroying target cells
 - Relatively short-lived
- Some become Memory cells
 - No immediate effect
 - Long-lived cells
 - New exposure to same antigen, rapidly activated
 - Strategy for vaccines
- Type 1 diabetes when CTLs attack β cells in pancreas
- Other autoimmune diseases are similar

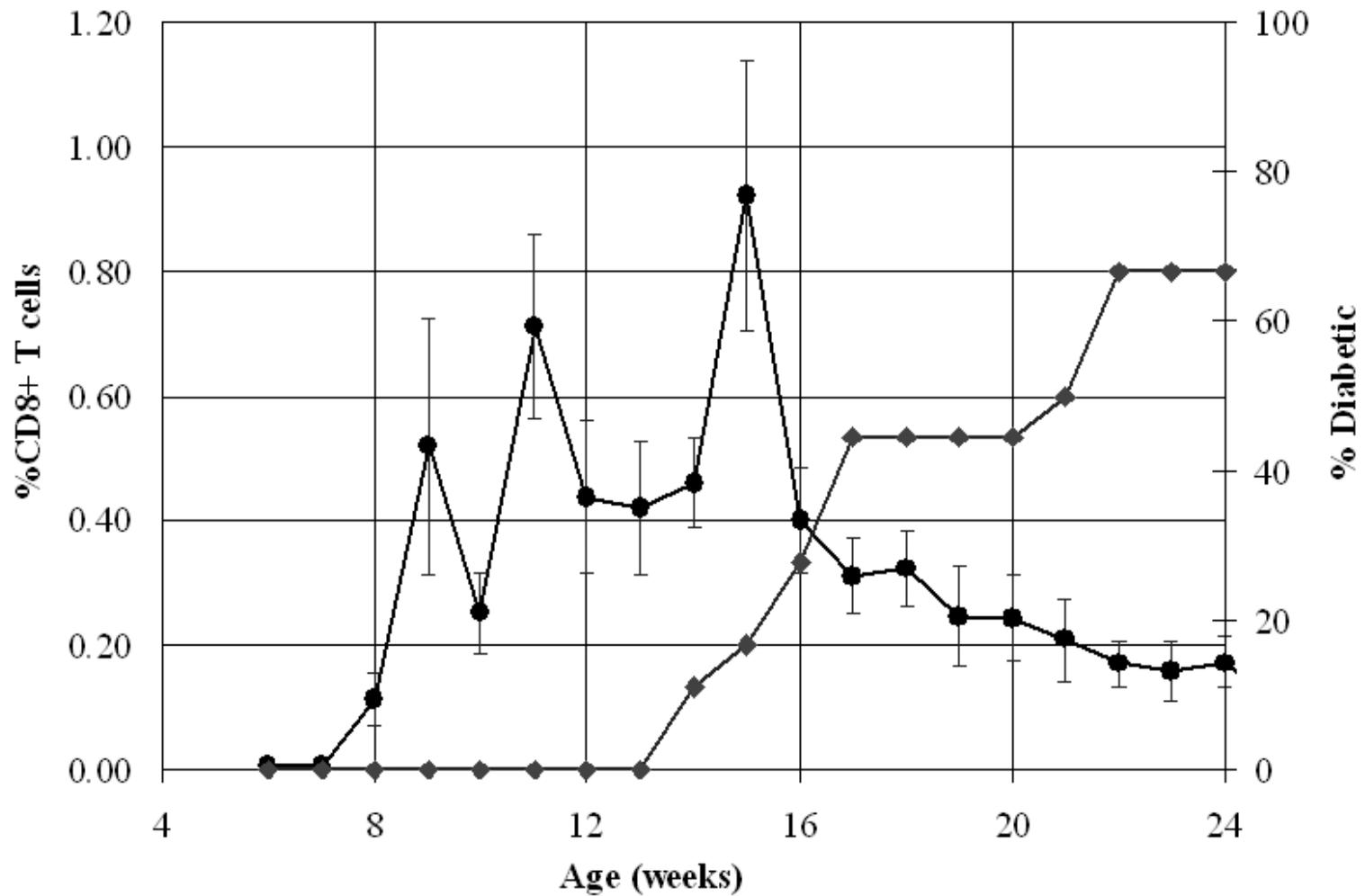
Model Scheme for Diabetes



Animal Model for Diabetes

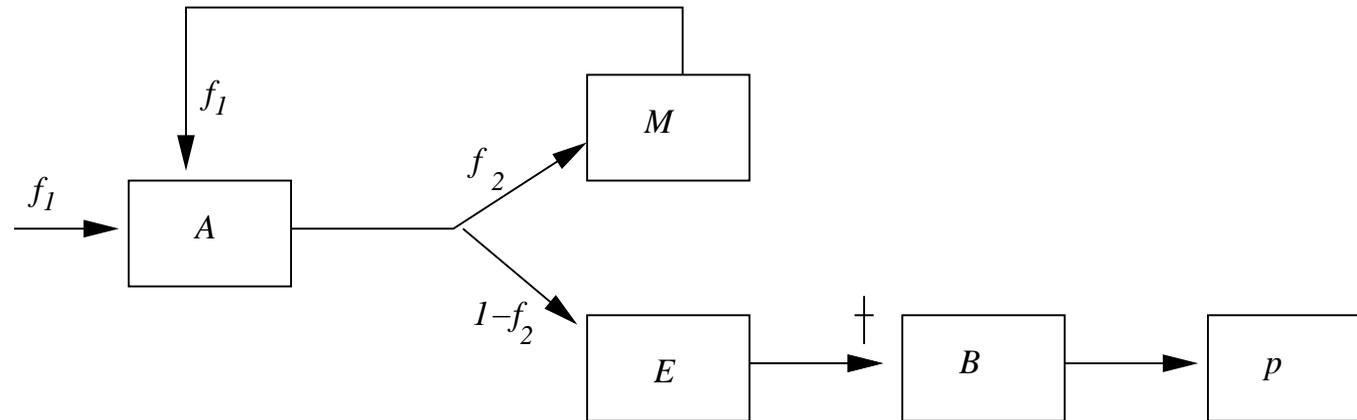
- Non-Obese Diabetic or NOD mice undergo apoptosis or programmed cell death of β cells in the pancreas shortly after birth
- Clearance of apoptotic cells by macrophages is reduced
 - Possibly forms self-antigen
 - Experiments suggest a fragment from IGRP (glucose-6-phosphate catalytic subunit-related protein) produces a dominant antigen
- Experiments designed to find autoreactive CD8⁺ T cells in pancreas of NOD mice
- Observed three waves of CD8⁺ T cells before mice became diabetic around 16 weeks

NOD Mice Data



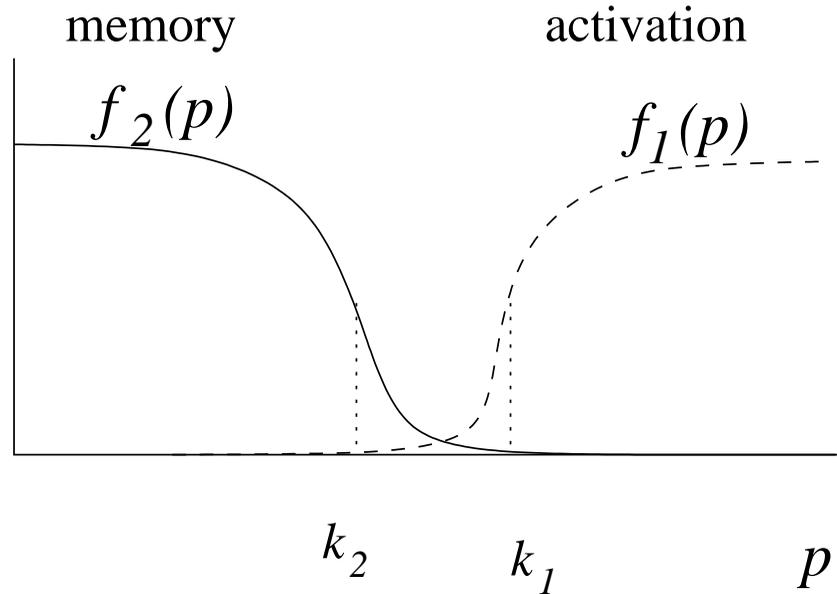
[Link to Model Simulation](#)

Simple Model Schematic



- A = Activated T cells
- M = Memory cells
- E = Effector or killer T cells
- p = peptide
- B = Fraction of remaining β cells

Feedback Functions



Activation function

$$f_1(p) = \frac{p^n}{k_1^n + p^n}$$

Inhibition function

$$f_2(p) = \frac{ak_2^m}{k_2^m + p^m}$$

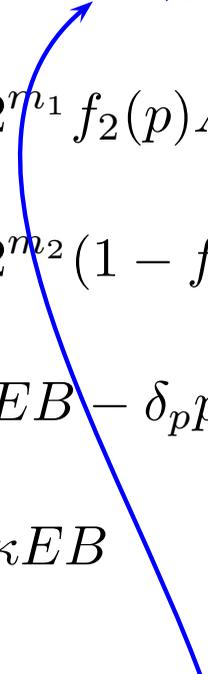
Complete Model

$$\begin{aligned}\frac{dA}{dt} &= (\sigma + \alpha M) f_1(p) - (\beta + \delta_A) A - \epsilon A^2 \\ \frac{dM}{dt} &= \beta 2^{m_1} f_2(p) A - f_1(p) \alpha M - \delta_M M \\ \frac{dE}{dt} &= \beta 2^{m_2} (1 - f_2(p)) A - \delta_E E \\ \frac{dp}{dt} &= REB - \delta_p p \\ \frac{dB}{dt} &= -\kappa EB\end{aligned}$$

with nonlinear feedback functions

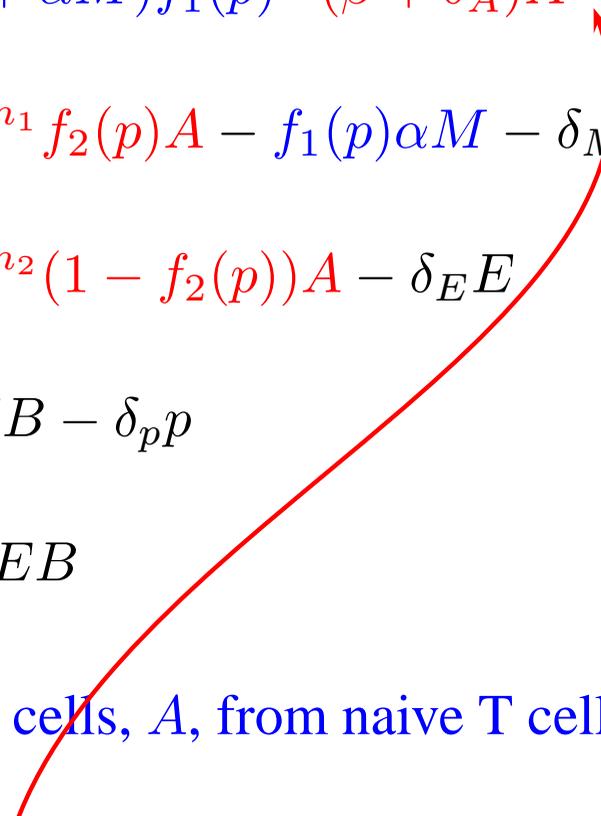
$$\begin{aligned}f_1(p) &= \frac{p^n}{k_1^n + p^n} \\ f_2(p) &= \frac{ak_2^m}{k_2^m + p^m}\end{aligned}$$

Activated T cells

$$\begin{aligned}\frac{dA}{dt} &= (\sigma + \alpha M)f_1(p) - (\beta + \delta_A)A - \epsilon A^2 \\ \frac{dM}{dt} &= \beta 2^{m_1} f_2(p)A - f_1(p)\alpha M - \delta_M M \\ \frac{dE}{dt} &= \beta 2^{m_2} (1 - f_2(p))A - \delta_E E \\ \frac{dp}{dt} &= REB - \delta_p p \\ \frac{dB}{dt} &= -\kappa EB\end{aligned}$$


The production of activated T cells, A , from naive T cells and memory cells.

Activated T cells

$$\begin{aligned}\frac{dA}{dt} &= (\sigma + \alpha M)f_1(p) - (\beta + \delta_A)A - \epsilon A^2 \\ \frac{dM}{dt} &= \beta 2^{m_1} f_2(p)A - f_1(p)\alpha M - \delta_M M \\ \frac{dE}{dt} &= \beta 2^{m_2} (1 - f_2(p))A - \delta_E E \\ \frac{dp}{dt} &= REB - \delta_p p \\ \frac{dB}{dt} &= -\kappa EB\end{aligned}$$


The production of activated T cells, A , from naive T cells and memory cells.

The loss of activated T cells, A , becoming effector and memory T cells, decaying, and competing with others.

Effector T Cells and β Cells

$$\begin{aligned}\frac{dA}{dt} &= (\sigma + \alpha M) f_1(p) - (\beta + \delta_A)A - \epsilon A^2 \\ \frac{dM}{dt} &= \beta 2^{m_1} f_2(p)A - f_1(p)\alpha M - \delta_M M \\ \frac{dE}{dt} &= \beta 2^{m_2} (1 - f_2(p))A - \delta_E E \\ \frac{dp}{dt} &= REB - \delta_p p \\ \frac{dB}{dt} &= -\kappa EB\end{aligned}$$

The effector T cells, E , destroy β cells producing the protein that activates T cells.

Complete Model - Discussion

- 5-Dimensional Model
 - Highly nonlinear
 - Difficult to analyze
- 17 Physiological parameters
 - Many are known or have good estimates
 - Constrains possible solutions
- Time Scale
 - The peptide, p , has fast reaction kinetics
 - This allows Quasi-Steady State Approximations
 - The β cells, B , have slow dynamics
 - This allows consideration of slow changing parameter

Quasi-Steady State Model

The model for analysis consists of three equations:

$$\begin{aligned}\frac{dA}{dt} &= (\sigma + \alpha M) f_1(p) - (\beta + \delta_A) A - \epsilon A^2 \\ \frac{dM}{dt} &= \beta 2^{m_1} f_2(p) A - f_1(p) \alpha M - \delta_M M \\ \frac{dE}{dt} &= \beta 2^{m_2} (1 - f_2(p)) A - \delta_E E\end{aligned}$$

together with the QSS peptide expression

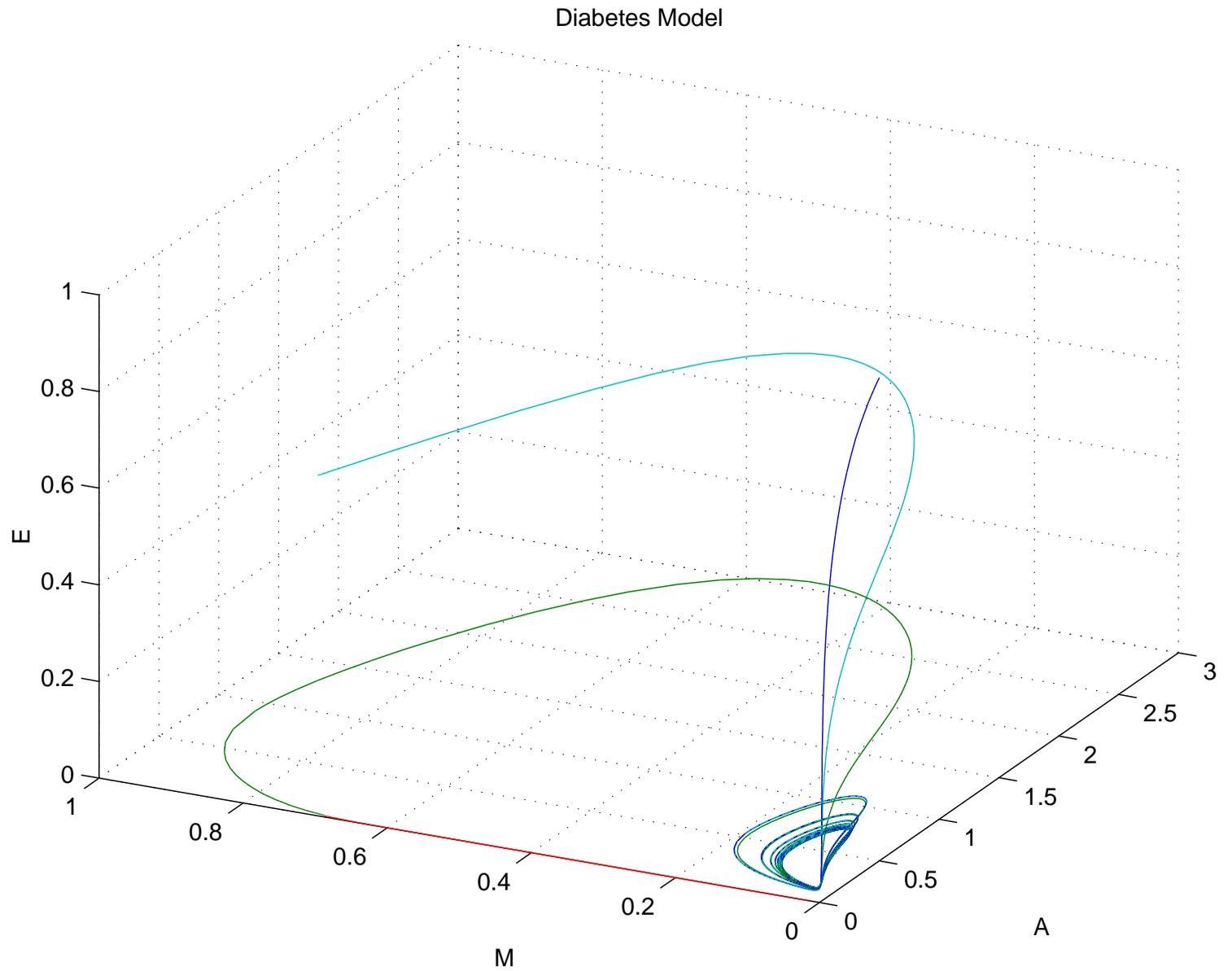
$$p \approx (RB/\delta_p)E$$

3-D system of differential equations permits a more complete analysis.

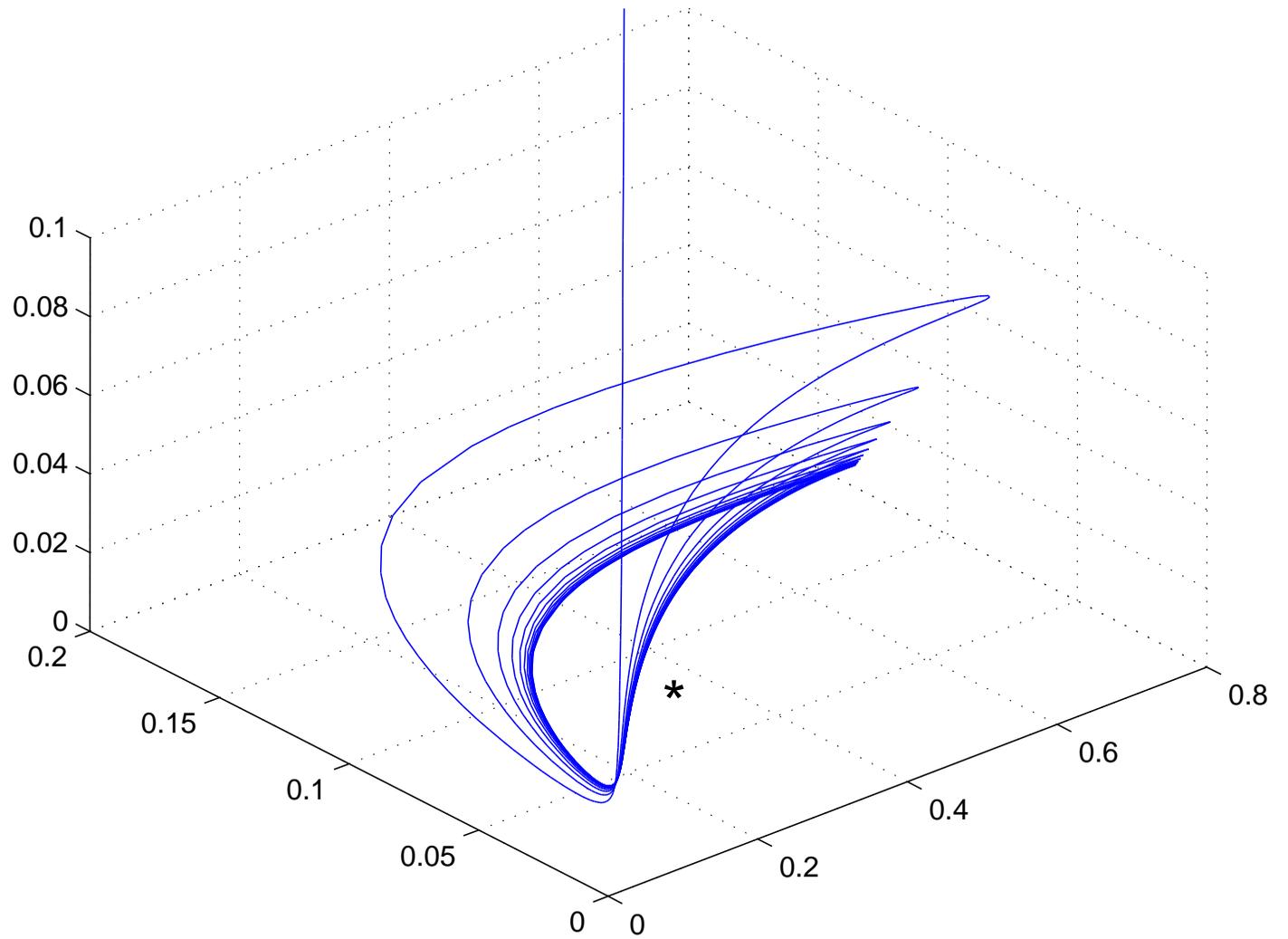
QSS Model - Discussion

- **Equilibria**
 - There are 1-5 equilibria
 - With physiological parameters, 3 equilibria exist
- **Linear Analysis of 3 Equilibria**
 - **Origin** is a **Stable Node**
 - **Diseased State** is a **Node** that is **Stable** or **Unstable** depending on parameters
 - Third equilibrium is a **Saddle Node**

3D Phase Portrait



3D Phase Portrait



QSS Model - Parameter Study

- **Parameters**
 - Experimental data compiled by Marée, Santamaria, and Edelstein-Keshet
 - Physiological range of parameters limited by their study for most parameters in the model
 - Several parameters remain unknown, so varied to obtain desired behaviour
 - Sensitivity of the parameters was studied

QSS Model - Bifurcation Study

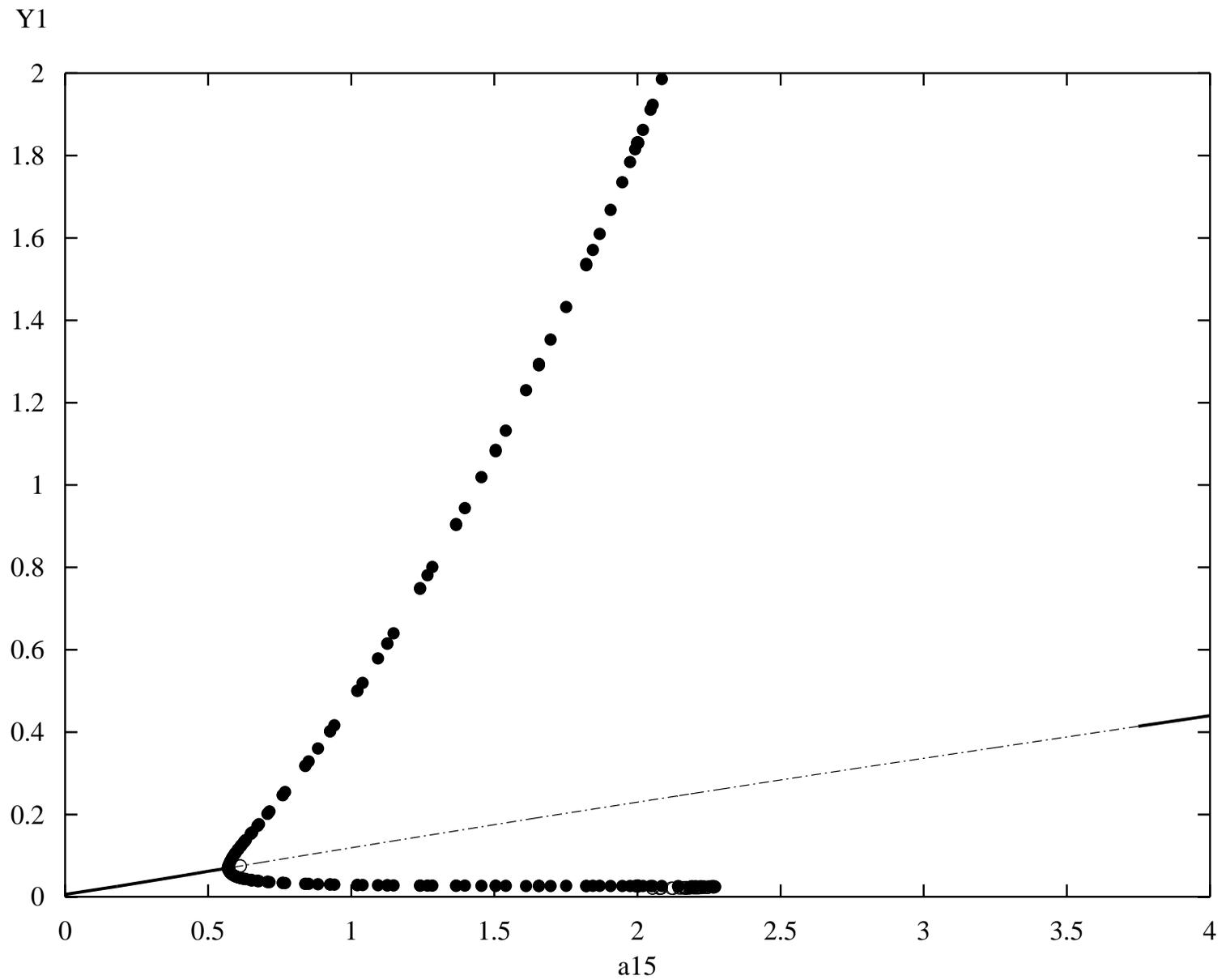
- **Bifurcation Analysis**

- Many parameters investigated using **AUTO** with **XPP**
- Chose peptide clearance rate δ_p as it is believed that poor clearance could induce diabetes
- In the normal range of clearance, the most solutions approached the **Origin**
- When halved, the many solutions oscillated about the **Diseased State**
- The **QSS** approximation is

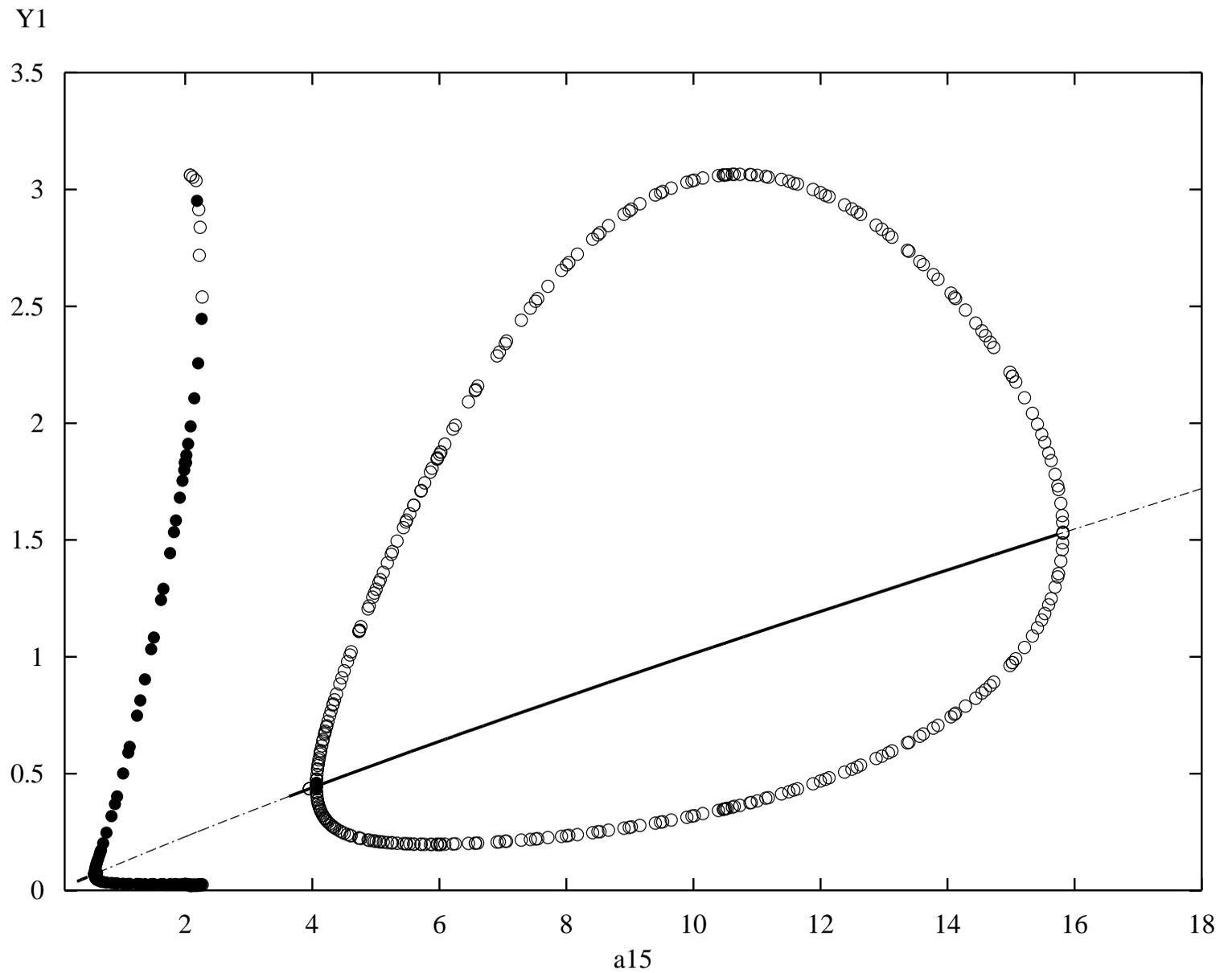
$$p \approx \frac{RB}{\delta_p} E,$$

so δ_p increasing is similar to B decreasing

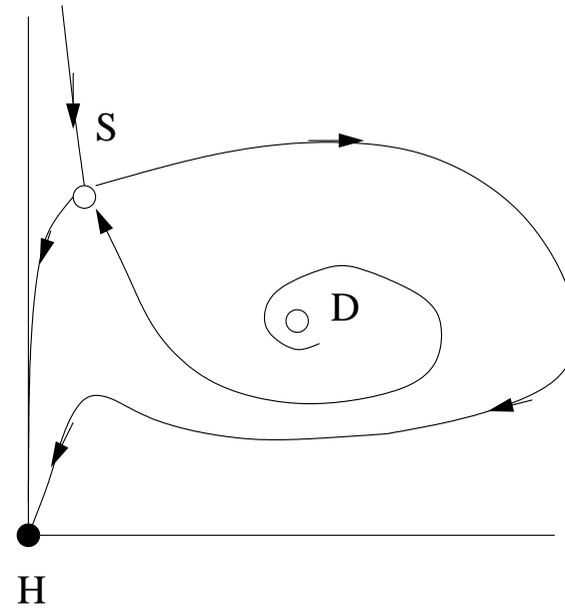
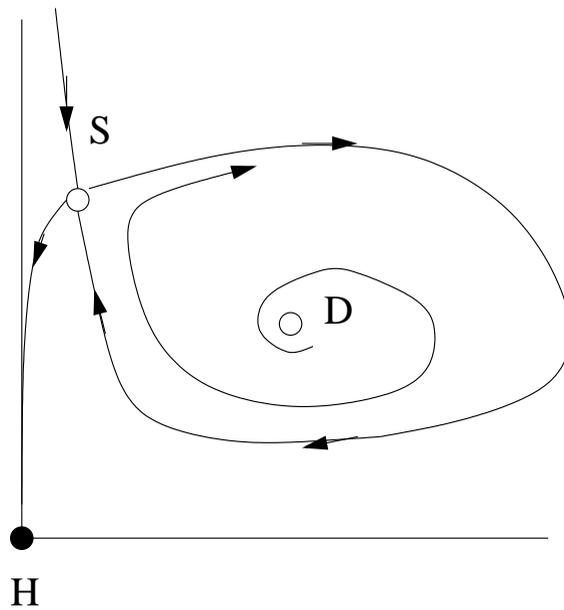
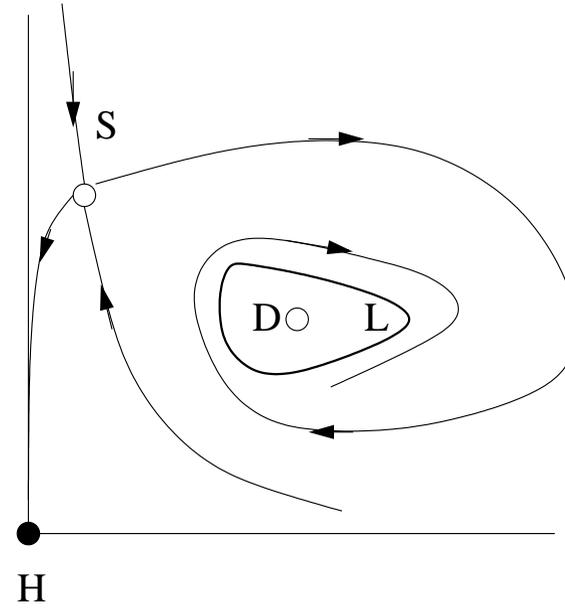
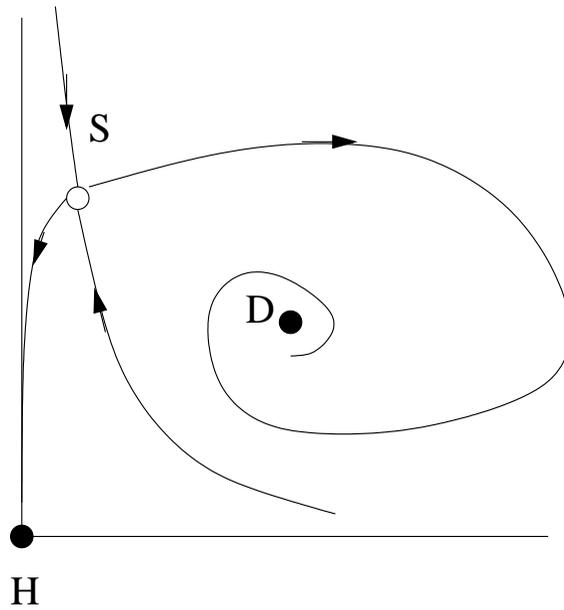
Bifurcation Diagram



Bifurcation Diagram



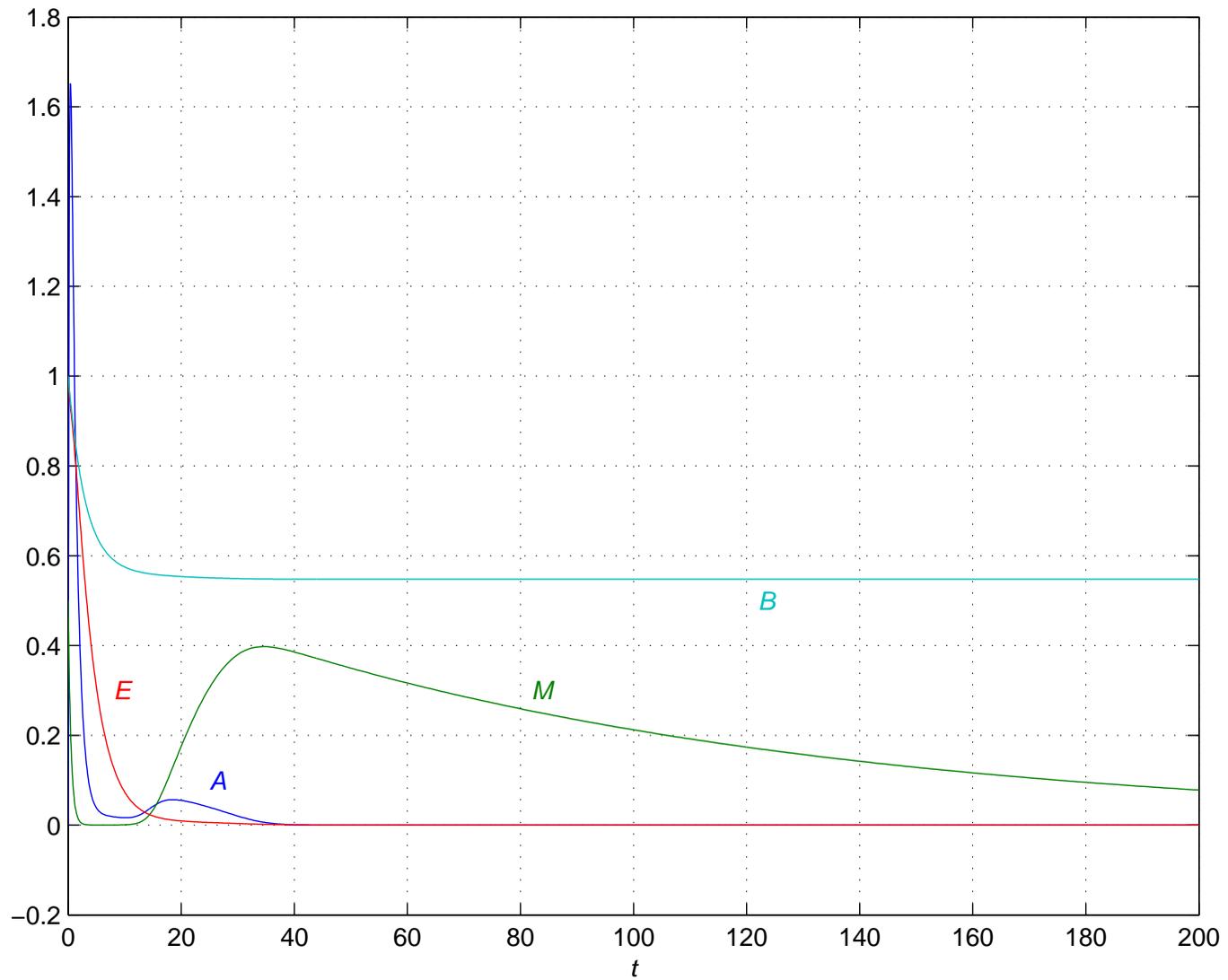
Homoclinic Bifurcation



Complete Model - Normal Mouse

- Simulated complete model for a normal mouse
 - Assumed an initial response of **Effector T cells**
 - Normal parameter values
 - Some β cells die, but levels at high equilibrium

Simulation - Normal Mouse

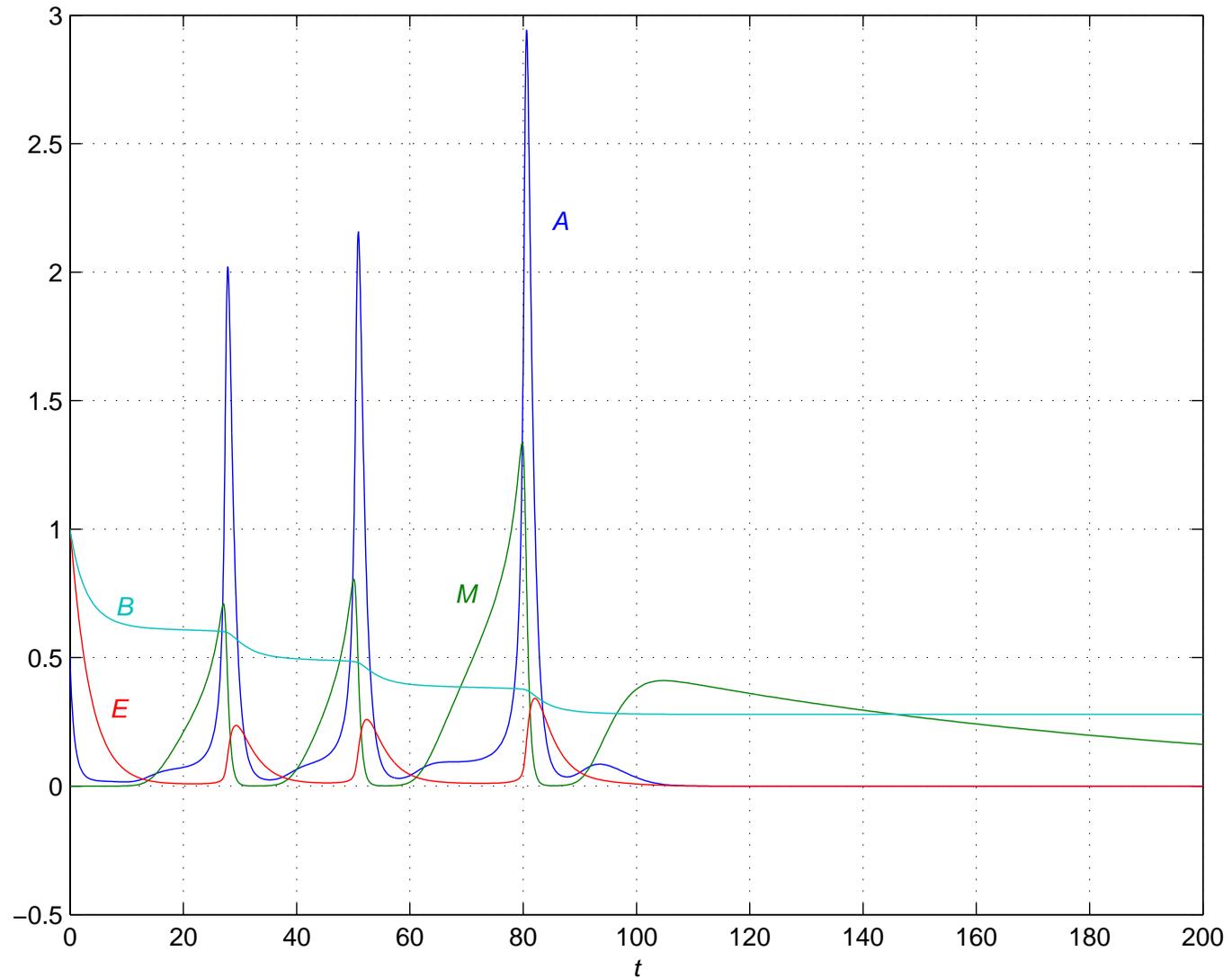


[Link to Homoclinic Diagram](#)

Complete Model - Diabetic Mouse

- Simulated complete model for a diabetic mouse
 - Assumed an initial response of **Effector T cells**
 - Lower peptide clearance
 - Increasing spikes of **Activated T cells**
 - Waves of short-lived **Effector T cells**
 - High **Memory cell** populations allow new response
 - Slow decline of β cells until diabetic

Simulation - Diabetic Mouse



Discussion and Conclusions

- Designed a reasonable model for NOD mice
- Parameters fit physiological data
- Simulations indicate parameters and initial conditions may be too sensitive
- Excellent qualitative behaviour of the model
- Good example of a homoclinic bifurcation
- Model supports biological theory of defective clearance after apoptosis