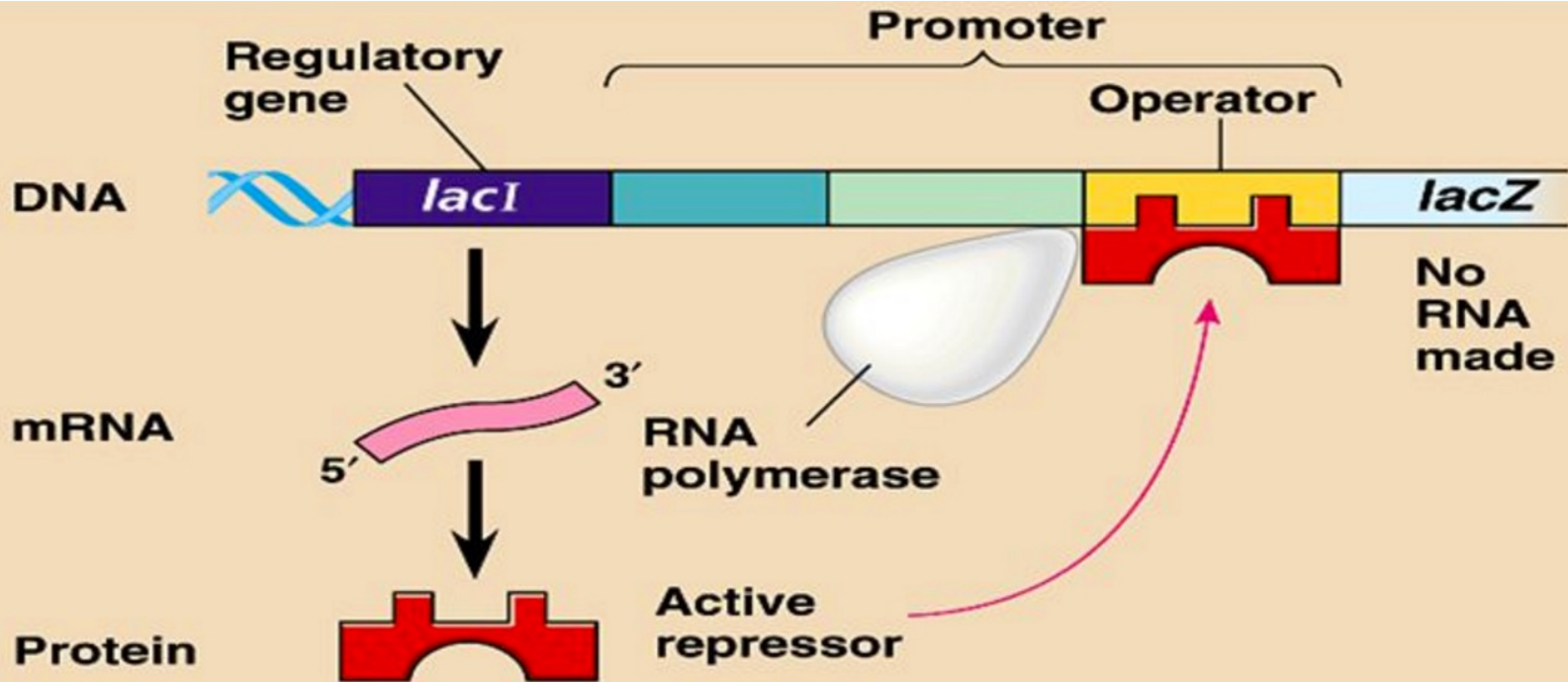


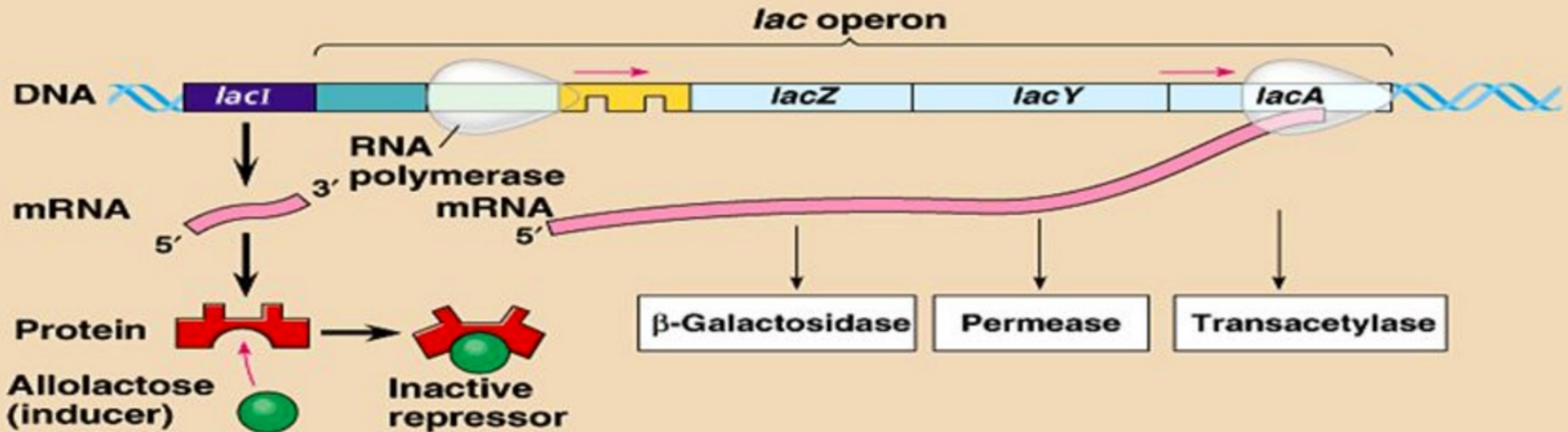
# Boolean models of the *lac* operon in *E. coli*

Matthew Macauley  
Clemson University

# The *lac* operon



(a) Lactose absent, repressor active, operon off



(b) Lactose present, repressor inactive, operon on

# Downsides of an ODE model

- Very mathematically technical.
- Too hard to solve explicitly. Numerical methods are needed.
- MANY experimentally determined “rate constants” (I count 22...)
- Often, these rate constants aren't known even up to orders of magnitude.

$$\frac{dM}{dt} = \alpha_M \frac{1 + K_1 (e^{-\mu\tau_M} A_{\tau_M})^n}{K + K_1 (e^{-\mu\tau_M} A_{\tau_M})^n} + \Gamma_0 - \tilde{\gamma}_M M$$

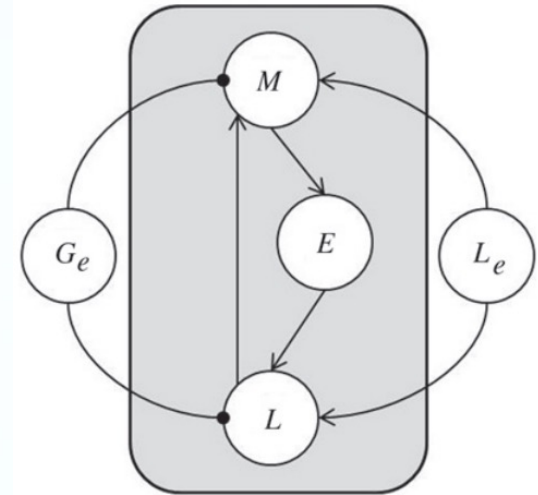
$$\frac{dB}{dt} = \alpha_B e^{-\mu\tau_B} M_{\tau_B} - \tilde{\gamma}_B B$$

$$\frac{dA}{dt} = \alpha_A B \frac{L}{K_L + L} - \beta_A B \frac{A}{K_A + A} - \tilde{\gamma}_A A$$

$$\frac{dP}{dt} = \alpha_P e^{-\mu(\tau_B + \tau_P)} M_{\tau_B + \tau_P} - \tilde{\gamma}_P P$$

$$\frac{dL}{dt} = \alpha_L P \frac{L_e}{K_{L_e} + L_e} - \beta_{L_1} P \frac{L}{K_{L_1} + L} - \alpha_A B \frac{L}{K_L + L} - \tilde{\gamma}_L L$$

# A Boolean approach



- Let's assume everything is “Boolean” (0 or 1):
  - Gene products are either present or absent
  - Enzyme concentrations are either high or low.
  - The operon is either ON or OFF.
- mRNA is transcribed ( $M=1$ ) if there is no external glucose ( $G=0$ ), and either internal lactose ( $L=1$ ) or external lactose ( $L_e=1$ ) are present.

$$x_M(t+1) = f_M(t+1) = \overline{G_e} \wedge (L(t) \vee L_e)$$

- The LacY and LacZ gene products ( $E=1$ ) will be produced if mRNA is available ( $M=1$ ).

$$x_E(t+1) = f_E(t+1) = M(t)$$

- Lactose will be present in the cell if there is no external glucose ( $G_e=0$ ), and either of the following holds:
  - ✓ External lactose is present ( $L_e=1$ ) and *lac* permease ( $E=1$ ) is available.
  - ✓ Internal lactose is present ( $L=1$ ), but  $\beta$ -galactosidase is absent ( $E=0$ ).

$$x_L(t+1) = f_L(t+1) = \overline{G_e} \wedge \left[ (L_e \wedge E(t)) \vee (L(t) \wedge \overline{E(t)}) \right]$$

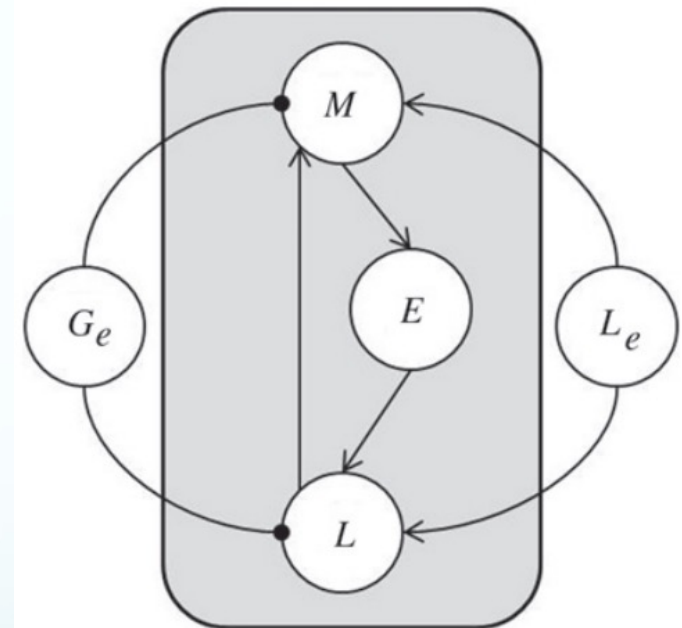
# Comments on the Boolean model

- We have two “types” of Boolean quantities:
  - mRNA (M), lac gene products (E), and internal lactose (L) are **variables**.
  - External glucose ( $G_e$ ) and lactose ( $L_e$ ) are **parameters** (constants).
- Variables and parameters are drawn as **nodes**.
- Interactions can be drawn as **signed edges**.
- A signed graph called the **wiring diagram** describes the dependencies of the variables.
- Time is discrete:  $t = 0, 1, 2, \dots$

$$x_M(t+1) = f_M(t+1) = \overline{G_e} \wedge (L(t) \vee L_e)$$

$$x_E(t+1) = f_E(t+1) = M(t)$$

$$x_L(t+1) = f_L(t+1) = \overline{G_e} \wedge \left[ (L_e \wedge E(t)) \vee (L(t) \wedge \overline{E(t)}) \right]$$



- Assume that the variables are updated **synchronously**.



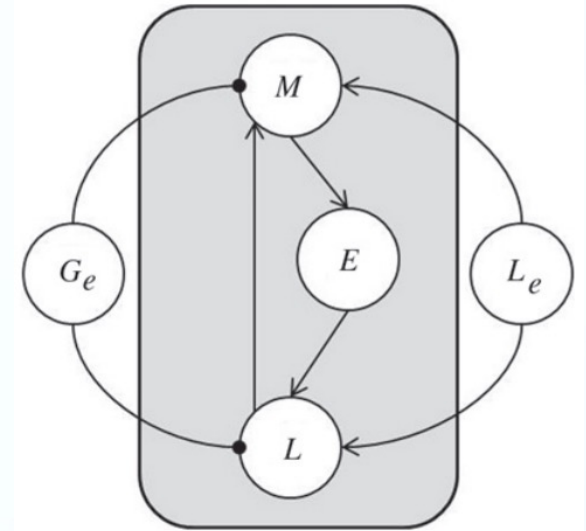
# How to analyze a Boolean model

- At the bare minimum, we should expect:
  - Lactose absent => operon OFF.
  - Lactose present, glucose absent => operon ON.
  - Lactose and glucose present => operon OFF.

$$x_M(t+1) = f_M(t+1) = \overline{G_e} \wedge (L(t) \vee L_e)$$

$$x_E(t+1) = f_E(t+1) = M(t)$$

$$x_L(t+1) = f_L(t+1) = \overline{G_e} \wedge \left[ (L_e \wedge E(t)) \vee (L(t) \wedge \overline{E(t)}) \right]$$



- The state space (or phase space) is the directed graph  $(V, T)$ , where

$$V = \{(x_M, x_E, x_L) : x_i \in \{0,1\}\} \quad T = \{(x, f(x)) : x \in V\}$$

- We'll draw the state space for all four choices of the parameters:
  - $(L_e, G_e) = (0, 0)$ . We hope to end up in a fixed point  $(0,0,0)$ .
  - $(L_e, G_e) = (0, 1)$ . We hope to end up in a fixed point  $(0,0,0)$ .
  - $(L_e, G_e) = (1, 0)$ . We hope to end up in a fixed point  $(1,1,1)$ .
  - $(L_e, G_e) = (1, 1)$ . We hope to end up in a fixed point  $(0,0,0)$ .

# How to analyze a Boolean model

- We can plot the state space using several software packages:
  - The “Cyclone” software package: at <http://cyclone.algorun.org/>.
  - The BoolNet library in R.
- For Cyclone, we need to convert our logical functions into polynomials.

$$x_M(t+1) = f_M(t+1) = \overline{G_e} \wedge (L(t) \vee L_e)$$

$$x_E(t+1) = f_E(t+1) = M(t)$$

$$x_L(t+1) = f_L(t+1) = \overline{G_e} \wedge \left[ (L_e \wedge E(t)) \vee (L(t) \wedge \overline{E(t)}) \right]$$

- Here is the relationship between Boolean logic and polynomial algebra:

<u>Boolean operations</u>	<u>logical form</u>	<u>polynomial form</u>
○ AND	$z = x \wedge y$	$z = xy$
○ OR	$z = x \vee y$	$z = x + y + xy$
○ NOT	$z = \overline{x}$	$z = 1 + x$

Everything is modulo 2, so  $1+1=0$ , and  $1=-1$ , and  $x^2=x$ , and thus  $x(x+1)=0$ .

# Cyclone

Calculate Dynamics of a discrete dynamical system using exhaustive search

More Information: <https://github.com/PlantSimLab/cyclone>

Packaged by: [Abdelrahman Hosny](#)

# Authors



Reinhard Laubenbacher (PI)

Center for Quantitative Medicine

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input	output	visualization
<pre> 1 MODEL NAME: 3-variable lac operon model 2 SIMULATION NAME: Lactose but not glucose 3 NUMBER OF VARIABLES: 3 4 VARIABLE NAMES: x1 x2 x3 5 NUMBER OF STATES: 2 2 2 6 SPEED OF VARIABLES: 1 1 1 7 8 f1 = 1 9 f2 = x1 10 f3 = x2 + x3 + x2*x3 </pre>	<pre> 1 digraph test { 2 node0 [label=" 0 0 0"]; 3 node1 [label=" 0 0 1"]; 4 node2 [label=" 0 1 0"]; 5 node3 [label=" 0 1 1"]; 6 node4 [label=" 1 0 0"]; 7 node5 [label=" 1 0 1"]; 8 node6 [label=" 1 1 0"]; 9 node7 [label=" 1 1 1"]; 10 node0 -&gt; node0 11 node1 -&gt; node5 12 node2 -&gt; node1 13 node3 -&gt; node5 14 node4 -&gt; node2 15 node5 -&gt; node7 16 node6 -&gt; node3 17 node7 -&gt; node7 18 } 19 </pre>	

Load sample data

Change parameters

Reset computation

RUN COMPUTATION



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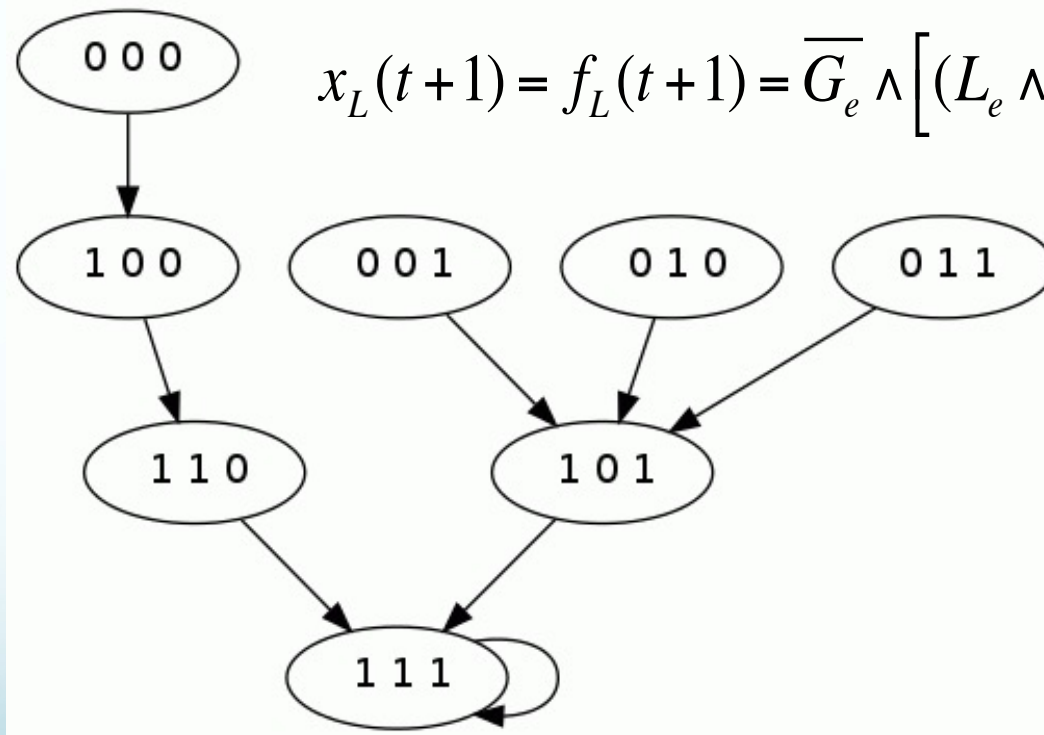


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State space when  $(G_e, L_e) = (0, 1)$ . The operon is ON.

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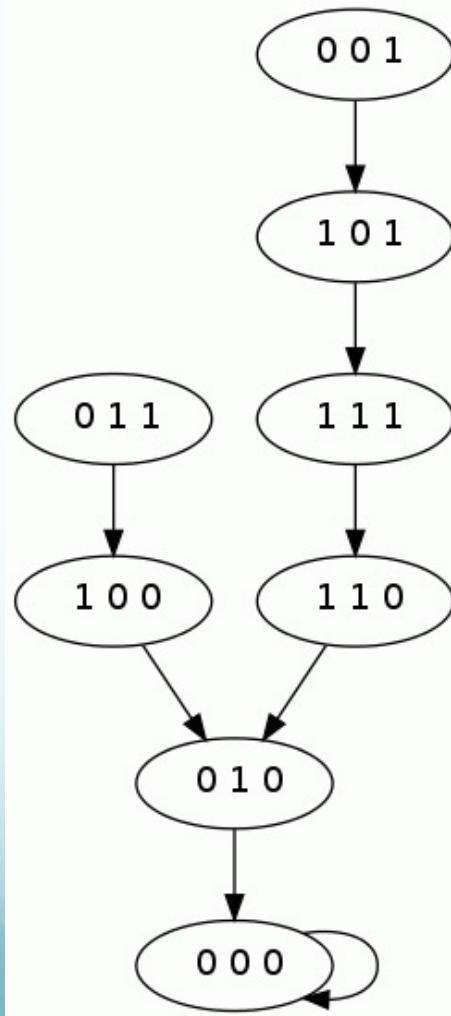
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State space when  $(G_e, L_e) = (0, 0)$ .

The operon is OFF.

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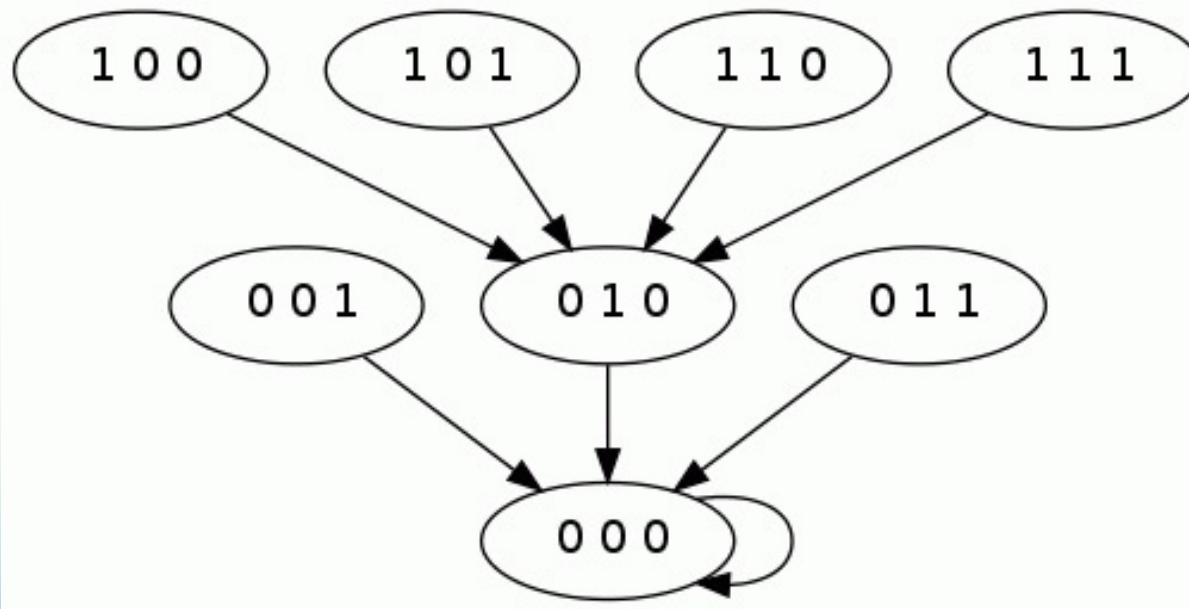


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State space when  $(G_e, L_e) = (1, 0)$ . The operon is OFF.

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## Authors

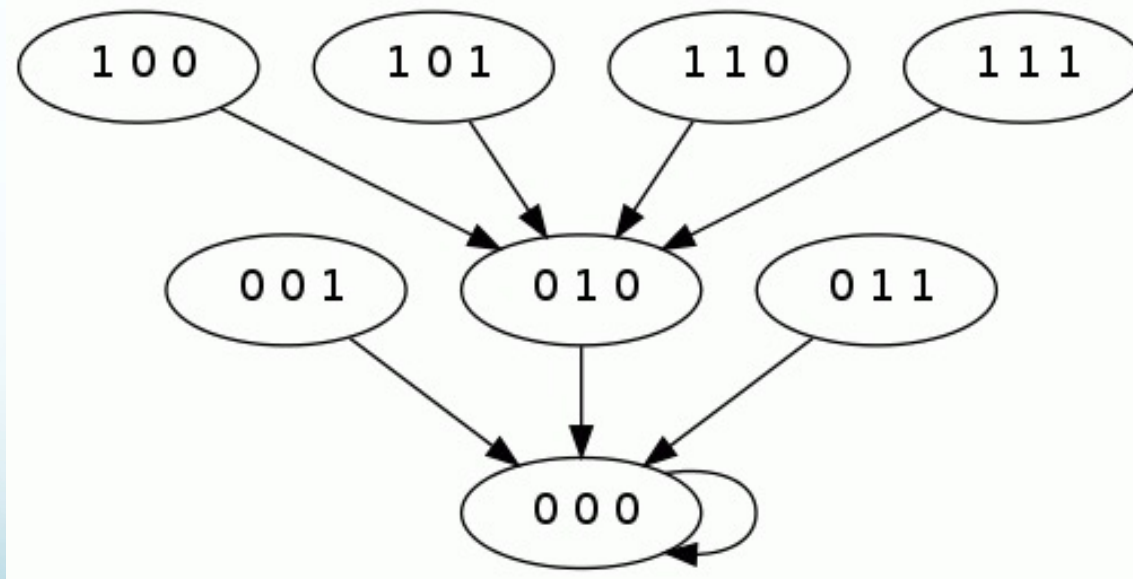


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$$x_L(t+1) = f_L(t+1) = \overline{G_e} \wedge [(L_e \wedge E(t)) \vee (L(t) \wedge \overline{E(t)})]$$



State space when  $(G_e, L_e) = (1, 1)$ . The operon is OFF.



# Summary so far

- **Gene regulatory networks** consist of a collection of gene products that interact with each other to control a specific cell function.
- Classically, these have been modeled quantitatively with differential equations (**continuous models**).
- **Boolean models** take a different approach. They are **discrete models** that are inherently qualitative.
- The **state space** graph encodes all the dynamics. The most important features are the **fixed points**, and a necessary step in model validation is to check that they are biologically meaningful.
- The model of the *lac* operon shown here is a “toy model”. Next, we will see more complicated models of the *lac* operon that capture intricate biological features of these systems.
- Modeling with Boolean logic is a relatively new concept, first done in the 1970s. It is a popular research topic in the field of **systems biology**.



# A more refined model

- Our first model only used 3 variables: mRNA (M), enzymes (E), and lactose (L).

- Let's propose a new model with 5 variables:

- M: mRNA

$$f_M = A$$

- B:  $\beta$ -galactosidase

$$f_B = M$$

- A: allolactose

$$f_A = A \vee (L \wedge B)$$

- L: intracellular lactose

$$f_L = P \vee (L \wedge \bar{B})$$

- P: *lac* permease (transporter protein)

$$f_P = M$$

- Assumptions

- Extracellular lactose is always available.
- Extracellular glucose is always unavailable.
- Translation and transcription require one unit of time.
- Protein and mRNA degradation require one unit of time
- Lactose metabolism require one unit of time

# Using Macaulay2 to generate polynomials

- Let's see how to analyze this model two ways:
  - Using “Cyclone” in AlgoRun (polynomial algebra)
  - Using BoolNet in R
- For Cyclone, we need to convert these into polynomials.
- It will be convenient to do this using a software package.
- We'll use Macaulay2, an open-source computational algebra package that can be run online: <http://www2.macaulay2.com/Macaulay2/>
- Here's how to use Macaulay2 to turn these functions into polynomials:

$$f_M = A$$

$$f_B = M$$

$$f_A = A \vee (L \wedge B)$$

$$f_L = P \vee (L \wedge \bar{B})$$

$$f_P = M$$

```
Q = ZZ/2[M,B,A,L,P] / ideal(M^2-M,B^2-B,A^2-A,L^2-L,P^2-P)
```

```
RingElement | RingElement :=(x,y)->x+y+x*y;
```

```
RingElement & RingElement :=(x,y)->x*y;
```

```
fM = A;
```

```
fB = M;
```

```
fA = A | (L & B);
```

```
fL = P | (L & (1+B));
```

```
fP = M;
```

```
(fM, fB, fA, fL, fP)
```

# Using Cyclone to compute the state space

$$f_M = A$$

$$f_B = M$$

$$f_A = A \vee (L \wedge B)$$

$$f_L = P \vee (L \wedge \bar{B})$$

$$f_P = M$$



Cyclone

Run

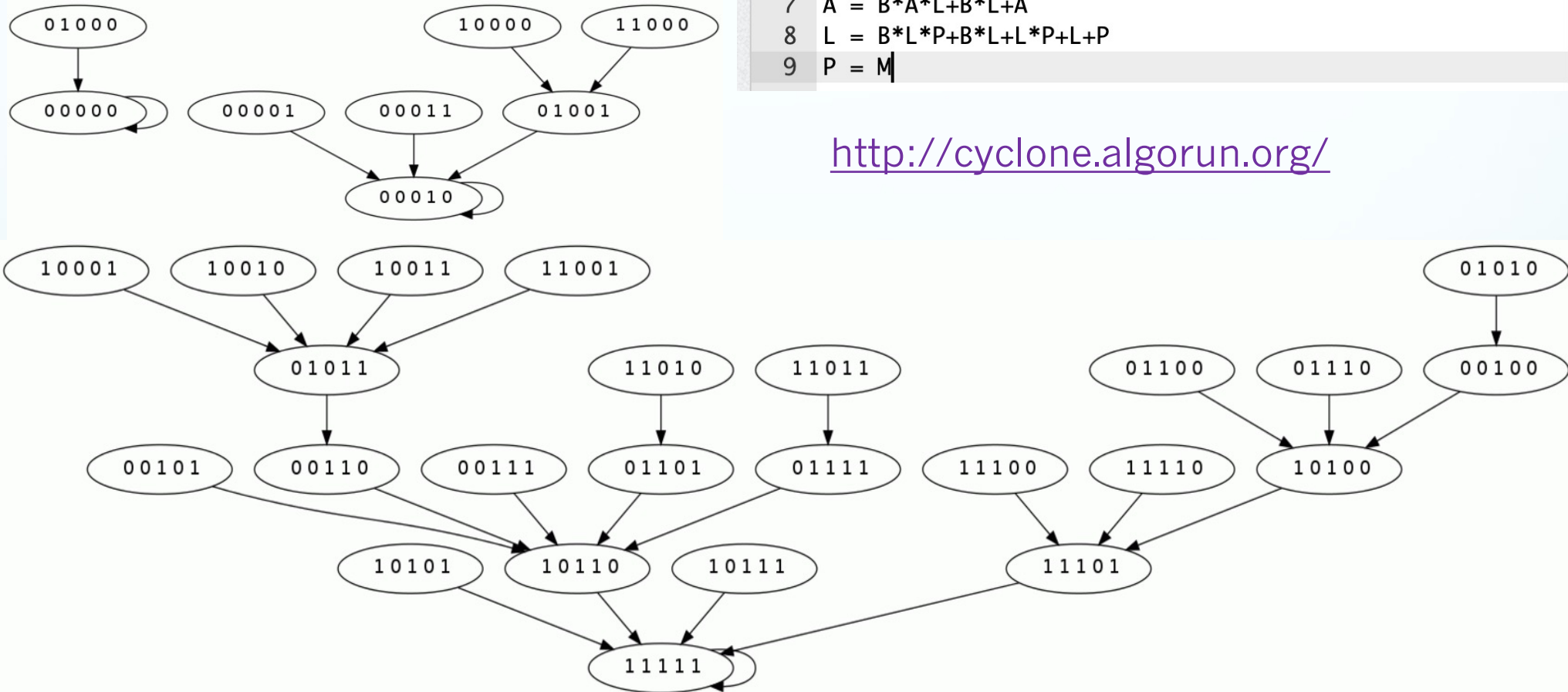
About

input

```

1 # Example Boolean Model
2 NUMBER OF VARIABLES: 5
3 NUMBER OF STATES : 2
4
5 M = A
6 B = M
7 A = B*A*L+B*L+A
8 L = B*L*P+B*L+L*P+L+P
9 P = M
    
```

<http://cyclone.algorun.org/>



# Using BoolNet in R to compute the state space

```
lac-operon x
1 targets, factors
2 M, A
3 B, M
4 A, A | (L & B)
5 L, P | (L & !B)
6 P, M
7 |
7:1 Text File

Console Terminal x
~/code/BoolNet/
> lacModel <- loadNetwork("lac-operon")
> getAttractors(lacModel)
Attractor 1 is a simple attractor consisting of 1 state(s) and has a basin of 2 state(s):

|---<----|
V      |
00000  |
V      |
|--->----|

Genes are encoded in the following order: M B A L P

Attractor 2 is a simple attractor consisting of 1 state(s) and has a basin of 6 state(s):

|---<----|
V      |
00010  |
V      |
|--->----|

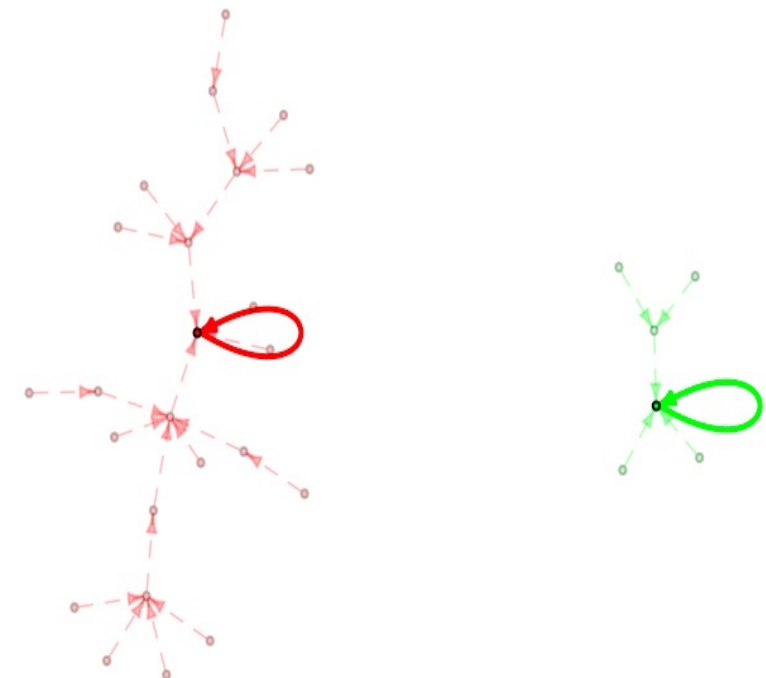
Genes are encoded in the following order: M B A L P

Attractor 3 is a simple attractor consisting of 1 state(s) and has a basin of 24 state(s):

|---<----|
V      |
11111  |
V      |
|--->----|

Genes are encoded in the following order: M B A L P
```

```
> lacModel <- loadNetwork("lac-operon")
> lacAttractors <- getAttractors(lacModel)
> plotStateGraph(lacAttractors)
> |
```



- Attractor 1
- Attractor 2
- Attractor 3



# Problems with our refined model

- Model variables:

- M: mRNA
- B:  $\beta$ -galactosidase
- A: allolactose
- L: intracellular lactose
- P: *lac* permease (transporter protein)

$$f_M = A$$

$$f_B = M$$

$$f_A = A \vee (L \wedge B)$$

$$f_L = P \vee (L \wedge \bar{B})$$

$$f_P = M$$

- Problems:

- The fixed point (M,B,A,L,P) = (0,0,0,0,0) should not happen with lactose present but not glucose. [though let's try to justify this...]
- The fixed point (M,B,A,L,P) = (0,0,0,1,0) is not biologically feasible: it would describe a scenario where the bacterium does not metabolize intracellular lactose.

- Conclusion: *The model fails the initial testing and validation, and is in need of modification.* (Homework!)