

## Solution of Exercise 1:

1.1 *A change in production rate.* A gene Y with simple regulation is produced at a constant rate  $\beta_1$ . The production rate suddenly shifts to a different rate  $\beta_2$ .

(a) Calculate and plot Y(t).

(b) What is the response time (time to reach halfway between the steady-states)?

Solution:

(a): Lets mark the time when the shift occurs as  $t = 0$ . Before the shift, Y reaches steady state at a level  $Y(t = 0) = Y_{st} = \beta_1 / \alpha$ . After the shift,

$$\frac{dY}{dt} = \beta_2 - \alpha Y.$$

The solution of such an equation is generally  $Y = \beta_2 / \alpha + C_1 \exp(-\alpha t)$ , where the constant  $C_1$  need to be determined so that  $Y(t = 0) = \beta_1 / \alpha$ . This yields the following solution

$$Y(t) = \frac{\beta_1}{\alpha} \exp(-\alpha t) + \frac{\beta_2}{\alpha} (1 - \exp(-\alpha t))$$

One can see that the initial condition  $\beta_1 / \alpha$  decays exponentially at the same rate as the new steady state  $\beta_2 / \alpha$  increases.

(b) To find the response time we have to solve the equation :

$$\frac{\beta_1}{\alpha} \exp(-\alpha \tau_{1/2}) + \frac{\beta_2}{\alpha} (1 - \exp(-\alpha \tau_{1/2})) = \frac{1}{2} \left( \frac{\beta_1}{\alpha} + \frac{\beta_2}{\alpha} \right)$$

After some algebra the response time is found to be  $\log(2) / \alpha$ .

1.2 *Cascades*. Consider a cascade of three activators,  $X \rightarrow Y \rightarrow Z$ . Protein X is initially present in the cell in its inactive form. The input signal of X,  $S_x$ , appears at time  $t = 0$ . As a result, X rapidly becomes active and binds the promoter of gene Y, so that protein Y starts to be produced at rate  $\beta$ . When Y levels exceed a threshold  $K_y$ , gene Z begins to be transcribed. All proteins have the same degradation/dilution rate  $\alpha$ . What is the concentration of protein Z as a function of time? What is its response time with respect to the time of addition of  $S_x$ ? What about a cascade of three repressors? Compare your solution to the experiments shown in Rosenfeld and Alon, 2003.

Solution:

We will assume all proteins have the same  $\alpha$ . After induction, Y is produced at rate  $\beta_y$  and degraded/diluted at rate  $\alpha$ :

$$\frac{dY}{dt} = \beta_y - \alpha Y$$

yielding the familiar exponential approach to steady-state:

$$Y(t) = \frac{\beta_y}{\alpha} (1 - \exp(-\alpha t))$$

Assuming a step function for the activation of gene Z by Y (logic input function), transcription of gene Z starts at time  $\tau_{yz}$  when  $Y(\tau_{yz}) = K_y$ :

$$Y(\tau_{yz}) = \frac{\beta_y}{\alpha} (1 - \exp(-\alpha \tau_{yz})) = K_y \Rightarrow \tau_{yz} = \frac{1}{\alpha} \log \left( \frac{Y_{st}}{Y_{st} - K_y} \right)$$

where  $Y_{st} = \beta_y / \alpha$ . Just for extra clarity, let's consider the limits of this equation to see if this makes sense. When  $K_y \ll Y_{st}$ ,  $Y_{st} - K_y \rightarrow Y_{st}$  and  $\tau_{yz} \rightarrow 0$ . In this case the threshold for Z activation is low, and Y levels cross it very fast. Conversely, if the activation threshold  $K_y$  is very high, approaching  $Y_{st}$ , Z is never activated because  $Y_{st} - K_y \rightarrow 0$  and  $\tau_{yz} \rightarrow \infty$ .

Production of Z starts after time  $t = \tau_{yz}$  at a constant rate of  $\beta_z$ :

$$\frac{dZ}{dt} = \begin{cases} 0 & t < \tau_{yz} \\ \beta_z - \alpha Z & t > \tau_{yz} \end{cases}$$

Solving we get

$$Z(t) = \begin{cases} 0 & t < \tau_{yz} \\ \frac{\beta_z}{\alpha} (1 - \exp(-\alpha(t - \tau_{yz}))) & t > \tau_{yz} \end{cases}$$

Solving for the response time, the time to reach half of the steady state of Z:

$$\frac{\beta_z}{\alpha} \left(1 - \exp\left(-\alpha(t_{1/2} - \tau_{yz})\right)\right) = \frac{1}{2} \frac{\beta_z}{\alpha} \Rightarrow t_{1/2} = \tau_{yz} + \log(2) / \alpha$$

Hence, there is an extra delay of  $\tau_{yz}$  in the response time of gene Z relative to simple regulation with no cascade. If Z activates a third gene W when it crosses a threshold  $K_z$ , this will occur at a time of  $\tau_{zW}$  found from:

$$\frac{\beta_z}{\alpha} (1 - e^{-\alpha(\tau_{zW} - \tau_{yz})}) = Z_{st} (1 - e^{-\alpha(\tau_{zW} - \tau_{yz})}) = K_z$$

solving for  $\tau_{zW}$  we obtain:

$$\tau_{zW} = \tau_{yz} + \frac{1}{\alpha} \log\left(\frac{Z_{st}}{Z_{st} - K_z}\right)$$

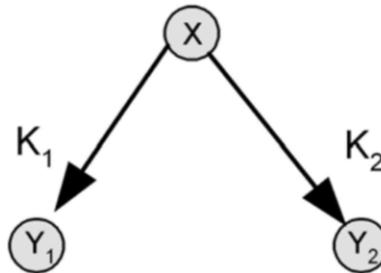
We can generalize this result: each step in a cascade, where a gene X activates a downstream gene after crossing a threshold  $K_x$  adds a delay of :

$$\tau_{delay} = \frac{1}{\alpha} \log\left(\frac{X_{st}}{X_{st} - K_x}\right)$$

In the special case in which the activation threshold is half the steady-state level (this can be shown to be in some cases an optimal value), the delay is  $\tau_{delay} = \log(2)/\alpha$ . In summary, since  $1/\alpha$  is often on the scale of a cell generation, a transcriptional cascade can be a slow process.

1.3 *Fan-out*. Transcription factor X regulates two genes  $Y_1$  and  $Y_2$ . Draw the resulting network, termed a fan-out with two target genes. The activation thresholds for these genes are  $K_1$  and  $K_2$ . The activator X begins to be produced at time  $t=0$  at rate  $\beta$ , and is degraded/diluted at rate  $\alpha$ , and its signal  $S_X$  is present throughout. What are the times at which  $Y_1$  and  $Y_2$  reach halfway to their maximal expression? Design a fan-out with three target genes in which the genes are activated with equal temporal spacing.

Solution:



Based on the previous problem:

$$\tau_1 = \frac{1}{\alpha} \log\left(\frac{X_{st}}{X_{st} - K_1}\right)$$

$$\tau_2 = \frac{1}{\alpha} \log\left(\frac{X_{st}}{X_{st} - K_2}\right)$$

After the corresponding delays in gene activation, denoted  $\tau_1$  and  $\tau_2$ , production of  $Y_1$  and  $Y_2$  starts at a constant rate reaching half the steady state after  $\log(2)/\alpha$ . The time to reach half maximum is therefore:  $\tau_{1/2} = t_i + \log(2)/\alpha$  ( $i=1,2$ ), where  $i=1,2$  for  $Y_1$  and  $Y_2$  respectively.

For three target genes, we require  $\tau_2 - \tau_1 = \tau_3 - \tau_2$ , or  $\tau_2 = \frac{1}{2}(\tau_1 + \tau_3)$ . This amounts to the following requirements on the thresholds,

$$\frac{1}{\alpha} \log\left(\frac{X_{st}}{X_{st} - K_2}\right) = \frac{1}{2} \left( \frac{1}{\alpha} \log\left(\frac{X_{st}}{X_{st} - K_1}\right) + \frac{1}{\alpha} \log\left(\frac{X_{st}}{X_{st} - K_3}\right) \right) \Rightarrow$$

$$X_{st} - K_2 = \sqrt{(X_{st} - K_1)(X_{st} - K_3)}$$

namely that the difference between  $X_{st}$  and the thresholds are arranged according to a geometric mean.

1.4. *Positive feedback.* What is the effect of positive auto-regulation on the response time? Use as a model the following linear equation:

$$dX/dt = \beta + \beta_1 X - \alpha X$$

Explain each term and solve for the response time. When might such a design be biologically useful?

Solution: The basal production rate is  $\beta$ , the positive effect of X on its own production (positive auto-regulation - PAR) is described in this model by the linear term  $\beta_1 X$ , and degradation/dilution is represented as usual by  $-\alpha X$ . Let's group the terms that multiply X in this linear model:

$$\frac{dX}{dt} = \beta - (\alpha - \beta_1) X$$

We see that the degradation/dilution rate is effectively reduced by positive auto-regulation, to an effective rate  $\alpha' = \alpha - \beta_1$ . Assuming that the auto-regulation is not too strong, that is that  $\beta_1 < \alpha$ , the term multiplying X is negative and we get an approach to a stable steady-state:

$$X(t) = X_{st} (1 - \exp(-\alpha' t)).$$

Where  $\alpha' = \alpha - \beta_1$ . The response time is defined as the time to reach half of the steady state:  $T_{1/2}(PAR) = \log(2)/\alpha'$ . The response time is longer than that for simple regulation due to the reduced effective degradation/dilution rate

$$T_{1/2}(PAR) = \log(2)/(\alpha - \beta_1) > \log(2)/\alpha = T_{1/2}(simple).$$

Thus positive auto-regulation has an effect that is opposite to that of negative auto-regulation. The former slows response time, whereas the latter speeds response times. Note that strong auto-regulation, in which  $\beta_1 > \alpha$ , can lead to instability and unchecked growth of X in the model. In real systems, this instability will be limited by other factors (such as saturation of the input function), locking Y in an ON state of high expression even after its activating input  $\beta$  vanishes. Hence, strong positive feedback creates a bi-stable system, in which X is either at a low or at a high fixed point. This is useful for commitment-type biological decisions, such as those made in development. Positive feedback characterizes developmental systems that make a switch that is either OFF or is locked ON (e.g., a cell commits to become a muscle cell rather than, say, a blood cell, by means of positive feedback loops on key transcription factors).

A different biological example is found in some regulatory systems that govern the transcription of protein parts of multi-protein structures that are assembled slowly. An example is the bacterial flagellum described in Chapter 6 that can take two cell generations to be completed. Such slow processes can benefit from weak positive auto-regulation to slow down responses and prolong delays (Kalir et al., 2005)