

Influence of Catabolite Repression and Inducer Exclusion on the Bistable Behavior of the *lac* Operon

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ABSTRACT A mathematical model of the *lac* operon which includes all of the known regulatory mechanisms, including external-glucose-dependent catabolite repression and inducer exclusion, as well as the time delays inherent to transcription and translation, is presented. With this model we investigate the influence of external glucose, by means of catabolite repression and the regulation of lactose uptake, on the bistable behavior of this system.

INTRODUCTION

Given their intrinsic nonlinearity, simple biochemical systems regulated at the level of gene expression are capable of complex dynamic behavior. Among the various patterns of behavior emerging from the regulation associated with nonlinear kinetics, bistability is extremely interesting. Bistability allows a true discontinuous switching between alternate steady states that can convert graded inputs into switch-like responses. Another important feature associated with bistability is hysteresis: if, in order for the system state to switch from one steady value to another, the input signal must surpass a given threshold. To switch back to the original state value, the input signal must be decreased below another (smaller) threshold. This permits a discontinuous evolution of the system along different possible pathways, which can be either reversible or irreversible, and may provide the system with an epigenetic (nongenetic) memory. The evolutionary significance of bistability, as well as its possible role in explaining some basic processes of life, like cell differentiation or the maintenance of phenotypic differences in the absence of genetic and environmental differences, has recently been discussed elsewhere (Laurent and Kellershohn, 1999; Casadesús and D'Ari, 2002; Ferrell, 2002).

Although it was not realized at the time, the lactose operon in *Escherichia coli* was one of the first molecular systems in which the bistability was experimentally demonstrated (Novick and Wiener, 1957). See Laurent and Kellershohn (1999) for a detailed discussion on this issue. Laurent and Kellershohn (1999) proposed a simple model of the lactose operon that, with a proper choice of the parameter values, was able show a bistable behavior. More recently, Yildirim and Mackey (2003) developed a more detailed mathematical

model, in which the parameters were all estimated from reported experimental data, and showed that, indeed, there is bistability in the lactose operon dynamics for realistic extracellular lactose concentration values.

The model of Yildirim and Mackey does not consider the regulatory mechanisms at the transcriptional level in detail. Instead, it assumes that all of them can be lumped into a single Hill-type equation despite the fact that the available experimental data allow a more detailed modeling approach. Furthermore, the Yildirim and Mackey model fails to include two important regulatory mechanisms that depend on the extracellular glucose concentration: catabolite repression and inducer exclusion. These mechanisms are essential to understand the lactose operon performance when the bacterial culture grows in a glucose-rich medium. Thus, it is important to investigate their influence on the dynamic system behavior.

In this study, we develop a more detailed mathematical model of the lactose operon that takes into account all of its known regulatory mechanisms, including catabolite repression and inducer exclusion, as well as the time delays inherent to transcription and translation. All of the model parameters are estimated from the existing experimental literature. The model equations are numerically solved to investigate the influence of the glucose-dependent regulatory mechanisms (catabolite repression and lactose uptake) on the system bistable behavior. Our results show that bistability is maintained for a large range of realistic extracellular glucose concentrations. It is known that the presence of glucose in the bacterial medium affects the induction of the lactose operon by external lactose. According to our results, this effect is attained in two different ways: first, by increasing the external lactose concentration threshold value at which the system shifts from the uninduced to the induced state, and second by decreasing the activation level of the lactose operon after induction has taken place.

In the next section (The *lac* Operon) we describe the lactose operon regulatory mechanisms and the way they interact to control operon performance. The mathematical model is developed in Model Development. The Numerical Results section outlines the numerical procedure used to solve the

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delay-differential equations. The numerical experiments performed to investigate the influence of catabolite repression and inducer exclusion on the lactose operon bistable behavior are also described in Model Development, together with the results they give. Summarizing comments are provided in Concluding Remarks. All of the model parameters are estimated from reported experimental data in the Appendix.

THE *lac* OPERON

The *lac* operon comprises the genes whose encoded proteins allow lactose metabolism in bacteria like *E. coli*. For the *lac* operon to be activated, two conditions must be fulfilled: 1), an activator (lactose) and 2), cAMP must be present in the intracellular medium. Until the early 1980s the *lac* operon (see Fig. 1) was thought to consist of one operator (O1), one promoter (P1), and one CAP complex binding site (C1), which control the activation of the *lac* operon as explained below (Beckwith, 1987).

In the absence of extracellular glucose, cAMP molecules are synthesized and they bind to free cAMP receptor proteins (CRP), forming the complex CAP. This complex binds to site C1, enhancing the binding affinity of the promoter P1 for mRNA polymerase (mRNAP) molecules. After binding to P1, some mRNAP start transcription of genes *lacZ*, *lacY*, and *lacA*. The product of gene *lacZ* is a monomer of the enzyme β -galactosidase, whereas the product of gene *lacY* is the protein *lac* permease. The protein of gene *lacA* does not play a role in the regulation of the *lac* operon, and it will not be considered further. In the presence of extracellular glucose, the production of cAMP is inhibited, and, therefore, the complex CAP cannot be formed.

If there is lactose in the extracellular medium and it is transported into the bacterium by permease proteins present in the cell membrane, some of the lactose is transformed into allolactose by β -galactosidase. Allolactose molecules in turn bind to *lac* repressor molecules, inactivating them and preventing their binding to the operator O1 and their further repression of the production of efficient mRNAs by mRNA polymerase molecules bound to P1. In short, there must be lactose inside the cell and no glucose in the extracellular medium to activate the lactose operon. The failure to fulfill either of these two requirements prevents the full activation of the operon.

There is a positive feedback loop in this regulatory pathway. The more permease molecules there are in the cell membrane, the higher the lactose uptake rate and the higher the intracellular lactose concentration. The higher the internal lactose concentration, the higher the intracellular allolactose concentration, the higher the operon activation level, and the faster the rate of β -galactosidase and *lac* permease production. This positive feedback loop may ultimately lead to system bistability.

Until the early 1980s, this was believed to be a complete picture of the *lac* operon functioning. However, thanks to a series of ingenious experiments, almost all of them performed by the group of Reznikoff (see Reznikoff, 1992, and references therein), our understanding of the situation has changed dramatically. We now know that the *lac* operon regulatory machinery is much more complex than described above.

In addition to O1, there are two other operators (denoted O2 and O3) in the *lac* operon (see Fig. 1). Active repressor molecules can also bind to O2 and O3, although with a smaller affinity than that of O1. The DNA can also fold in such a way that a single repressor molecule can bind to two different operators, as shown in Fig. 2. This has the effect of stabilizing the repressor-DNA complex.

There is also duplicity in the CAP binding sites. Two of them, denoted C1 and C2, are found in the *lac* operon; their position is shown in Fig. 1. The CAP complex can bind both of them, but its affinity for C2 is smaller than its affinity for C1.

Given their position, a repressor bound to O1 blocks the ability of C2 to be bound by one CAP complex, and vice versa. Although they do not intersect, it is known that a repressor bound to O3 bends the DNA chain in the same direction as a CAP complex bound to C1 does (Vossen et al., 1996). From this, we assume that once a repressor is bound to O3, C1 cannot be bound by a CAP, and once C1 is bound by a CAP, O3 cannot be bound by a repressor.

The *lac* operon also has more than one promoter. In Fig. 1 we show the main promoter (P1), together with a secondary promoter (P2). There are at least three more promoters in the *lac* operon. One of them is upstream from the *lacZ* gene starting point, and two more are downstream. Of all of them, only P2 has been characterized. In vitro studies (Peterson and Reznikoff, 1985) revealed that the affinity of promoter P2 for

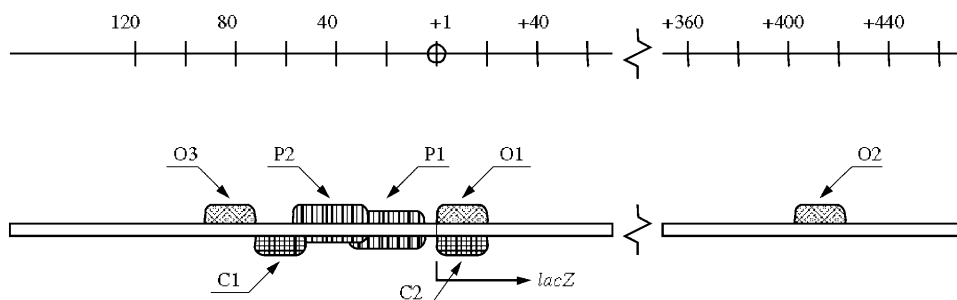


FIGURE 1 Regulatory elements of the *lac* operon. The upper scale identifies the position, in basepair units, of every element in the *lac* repressor genome. Position +1 signals the first basepair of gene *lacZ*.

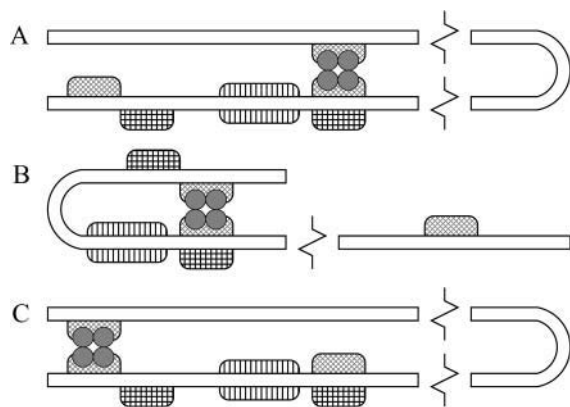


FIGURE 2 DNA folds in such a way that a single repressor molecule (which consists of four monomeric units) can simultaneously bind to two different operators as shown in the figure. (A) The repressor is bound to O1 and O2; (B) The repressor is bound to O1 and O3; and (C) the repressor is bound to O2 and O3.

mRNAP is higher than that of promoter P1. In the absence of cAMP, mRNAP prefers to bind promoter P2. This situation changes when cAMP is added. In such case, most C1 sites are bound by CAP complexes preventing the binding of mRNAPs to P2 (as seen in Fig. 1, *C1* and *P2* overlap), and, thus, the fraction of P1s bound by an mRNA polymerase is much higher than the corresponding P2 fraction. This suggests that P2 plays an indirect role in the activation of P1 by cAMP. More recent experimental *in vivo* studies (Donnelly and Reznikoff, 1987) show that mutations that abolish the activity of promoter P2, without affecting the C1 binding site, fail to activate P1. Moreover, CAP mutants that repress P2 and P3 but do not activate P1 have also been isolated (Eschenlauer and Reznikoff, 1991). The conclusion from these experiments is that P2 (and thus P3) is unlikely to make a major contribution to P1 activation (Reznikoff, 1992). Reznikoff (1992) speculates that one possible explanation for this promoter clustering is the tendency of mRNA polymerase to concentrate near the active promoter P1.

MODEL DEVELOPMENT

For the purposes of this model, we consider the lactose operon to consist of four different binding sites. Let us label them as sites 1, 2, 3, and 4. Site 1 comprises O3 and C1, and so its possible binding states are empty (e), bound by a repressor (r), or bound by a CAP complex (c). Site 2 consists only of P1, and so its possible binding states are empty (e) or bound by a mRNAP (p). Site 3 is made up of O1 and C2; its possible binding states are the same as those of site 1. Finally, site 4 comprises O2 only, and its binding states are e or c. If there is a repressor molecule simultaneously bound to sites 1 (O3) and 3 (O1), their binding states shall be denoted as r and 1, respectively. Similarly, if a repressor is simultaneously bound to sites 1 and 4, their states will, respectively, be denoted as r and 1. If the repressor is bound to sites 3 and 4 (O2), the states of these binding sites will, respectively, be denoted as r and 3. Notice that we ignore promoter P2 (and all of the other additional promoters). We have done so because, as discussed in the previous section (The *lac* Operon), these promoters seem to not play a role in the regulation of the *lac* operon *in vivo*.

With the introduction of this notation, all of the possible binding states of the *lac* operon can be represented by a four-character string. A simple counting reveals that there are 50 of these binding states. In this enumeration, it must be taken into account that if a repressor is bound to O1 (site 3) and O3 (site 1) there is no room for a mRNAP to bind P1 (site 2); see Oehler et al. (1994). The list of all of these 50 *lac* operon binding states is tabulated in Table 1.

Under the quasi-steady-state assumption that the *lac* repressor, mRNAP, and cAMP binding reactions are sufficiently rapid, relative to the transcription and translation rates, the probability of every one of the 50 *lac* operon binding states can be calculated as (Ackers et al., 1982)

$$P_i = \frac{e^{-E_i/RT} [\text{mRNAP}]^{\alpha_i} [\text{CAP}]^{\beta_i} [\text{R}]^{\gamma_i}}{Z}, \quad (1)$$

where the partition function is given by

$$Z = \sum_{i=1}^{50} e^{-E_i/RT} [\text{mRNAP}]^{\alpha_i} [\text{CAP}]^{\beta_i} [\text{R}]^{\gamma_i}. \quad (2)$$

In Eqs. 1 and 2, P_i and E_i represent the probability and energy of the i th binding state; α_i , β_i , and γ_i are, respectively, the number of mRNAP, CAP, and *lac* repressor molecules bound to the *lac* operon in that particular state; $[\text{mRNAP}]$ is the concentration of mRNAP; $[\text{CAP}]$ is the concentration of CAP; and $[\text{R}]$ is the concentration of *lac* repressor molecules.

Of all the 50 binding states of the *lac* operon, only 14 of them, those in which a mRNAP is bound to site 2 while site 3 is either empty or bound by a CAP complex, are able to start transcription and produce efficient mRNA chains. These 14 states are: epee, eper, rpee, rper, cpee, cper, rpe1, epce, epcr, rpce, rprc, cpce, cpcr, and rpe1. From these considerations, the transcription initiation rate of the *lac* operon genes can be modeled as

$$\psi = [\text{P1}] k_m (P_{\text{cpee}} + P_{\text{cper}} + P_{\text{epce}} + P_{\text{eper}} + P_{\text{rpee}} + P_{\text{rper}} + P_{\text{rpe1}} + P_{\text{cpce}} + P_{\text{cpcr}} + P_{\text{epce}} + P_{\text{eper}} + P_{\text{rpce}} + P_{\text{rprc}} + P_{\text{rpe1}}), \quad (3)$$

where $[\text{P1}]$ is the concentration of promoter P1, and k_m is the rate of transcription initiation at such promoter.

The *lac* operon encodes three genes: *lacZ*, *lacY*, and *lacA*. The product of gene *lacZ* is a monomer of enzyme β -galactosidase (B); the product of gene *lacY* is protein lactose permease (P); and the product of *lacA* is thiogalactoside transacetylase, which is thought not to play a role in the regulation pathway of the *lac* operon (Beckwith, 1987) and will not be considered further.

The transcription of the *lac* operon produces a mRNA with three ribosome binding sites, one for each of the encoded genes. We are concerned with the concentration of ribosome binding sites corresponding to *lacZ* ($[\text{M}_B]$) and *lacY* ($[\text{M}_P]$). From the considerations in the previous paragraphs, the dynamics of these quantities can be modeled as

TABLE 1 The 50 possible binding states of the *lac* operon

eeee	reee	ceee	ree1
eeer	reer	ceer	rer1
eere	rere	cere	rec1
eerr	rerr	cerr	rpe1
eece	rece	cece	rpr1
eeer	recr	cecr	rpe1
epee	rpee	cpce	eer3
eper	rper	cper	epr3
epre	rpre	cpce	rer3
eprr	rprc	cpce	rpr3
epce	rpce	cpce	cer3
epcr	rpcr	cpce	cpr3
			re1e
			re1r

$$\frac{d[M_B]}{dt} = \psi_{\tau_B} - (\mu + \xi_M)[M_B], \quad (4)$$

and

$$\frac{d[M_P]}{dt} = \psi_{\tau_P} - (\mu + \xi_M)[M_P], \quad (5)$$

where the symbol ψ_τ denotes the variable ψ delayed in time by an amount τ . τ_B and τ_P are, respectively, the delays between the initiation of transcription and the appearance of the ribosome binding sites corresponding to *lacZ* and *lacY*; ξ_M is the mRNA degradation rate; and μ is the bacterial growth rate.

The dynamic equations for the β -galactosidase and *lac* permease concentrations are, respectively, given by

$$\frac{d[B]}{dt} = \frac{1}{4}\kappa_B[M_B]_{T_B} - (\mu + \xi_B)[B], \quad (6)$$

and

$$\frac{d[P]}{dt} = \kappa_P[M_P]_{T_P} - (\mu + \xi_P)[P], \quad (7)$$

with κ_B and κ_P the respective translation initiation rates at the *lacZ* and *lacY* ribosome binding sites, T_B the time it takes to translate the *lacZ* mRNA, T_P the time it takes to translate the *lacY* mRNA, and ξ_B and ξ_P the β -galactosidase and *lac* permease degradation rates, respectively. The factor $1/4$ in Eq. 6 accounts for the fact that the active form of β -galactosidase is a tetramer.

Lactose is cotransported into the cell with a hydrogen ion by *lac* permease. Transport of lactose by the permease is inhibited by external glucose, a phenomenon known as inducer exclusion. It has been reported that glucose affects the transport rate constant ϕ_{L_1} , rather than the corresponding saturation constant Φ_{L_1} . Following Wong et al. (1997), we assume that the lactose transport rate can be modeled as

$$\nu_{L_1}([L_T]) = \phi_{L_1} \left(\frac{[L_E]}{[L_E] + \Phi_{L_1}} \frac{\Phi_{G_1}}{\Phi_{G_1} + [G_E]} - \frac{[L_T]}{[L_T] + \Phi_{L_1}} \right) [P], \quad (8)$$

where Φ_{G_1} is the lactose transport constant for inhibition by glucose, $[L_E]$ is the lactose concentration in the external medium, $[L_T]$ is the total intracellular lactose concentration (lactose plus allolactose, see below), and $[G_E]$ is the external glucose concentration. Since lactose transport is reversible, a term was included to account for lactose efflux dependent on the internal lactose concentration $[L_T]$. It is known that the lactose efflux does not depend on the external glucose concentration (Wong et al., 1997).

Once inside the cell, a fraction of the lactose is transformed by β -galactosidase to allolactose and the remainder is hydrolyzed to glucose and galactose. Allolactose is an excellent substrate of β -galactosidase and is also hydrolyzed to glucose and galactose by this enzyme. According to Martínez-Bilbao et al. (1991), the total lactose hydrolysis rate can be modeled as

$$\nu_{L_2}([L_T]) = \phi_{L_2} \left(\frac{[L_T]}{[L_T] + \Phi_{L_2}} \right) [B], \quad (9)$$

where ϕ_{L_2} is the rate constant for lactose hydrolysis, and Φ_{L_2} is the corresponding saturation constant.

Following Dean (1989), we assume that the conversion of lactose into allolactose and the further hydrolysis of allolactose take place in a quasi-steady state. Based on the quasi-steady-state assumption and the fact that 50% of the incoming lactose is converted into allolactose (Dean, 1989), the total allolactose $[A_T]$ concentration is related to $[L_T]$ by

$$[A_T] = [L_T]/2. \quad (10)$$

From Eqs. 8–10, the equation governing the total allolactose concentration dynamics is

$$\frac{d[A_T]}{dt} = \frac{(\nu_{L_1}(2[A_T]) - \nu_{L_2}(2[A_T]))}{2} - (\xi_A + \mu)[A_T], \quad (11)$$

where ξ_A is the allolactose degradation rate.

The synthesis of carbohydrates other than glucose is inhibited when glucose is plentiful. This phenomenon is known as catabolite repression. The primary signal molecule for catabolite repression is cAMP. In the absence of extracellular glucose, the production rate and consequently the intracellular concentration of cAMP increase. The exact mechanism controlling cAMP synthesis has not been elucidated. However, according to Wong et al. (1997), the cAMP production rate can be modeled as

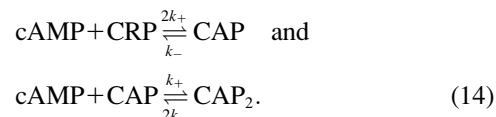
$$\nu_{cAMP} = \phi_{cAMP} \left(\frac{\Phi_{cAMP}}{[G_E] + \Phi_{cAMP}} \right), \quad (12)$$

where ϕ_{cAMP} is the cAMP synthesis rate constant, Φ_{cAMP} is the inhibition constant for the effect of glucose on cAMP, and $[G_E]$ is the extracellular glucose concentration. From Eq. 12 and assuming that cAMP removal through degradation or transport out of the cell follows first-order kinetics, the equation governing the dynamics of the concentration of cAMP is

$$\frac{d[cAMP_T]}{dt} = \nu_{cAMP} - \xi_{cAMP}[cAMP] - \mu[cAMP_T], \quad (13)$$

where ξ_{cAMP} is the cAMP removal rate, $[cAMP]$ is the concentration of free cAMP, and $[cAMP_T]$ is the total (free plus bound) cAMP concentration.

One or two cAMP molecules can bind to a cAMP receptor protein (CRP) to, respectively, form the complexes CAP and CAP₂. Of these, only CAP has high affinity for specific DNA binding sites (Pyles and Lee, 1996). The reactions leading to complexes CAP and CAP₂ can be written as



The equilibrium equations for the reactions in Eq. 14 are

$$\begin{aligned} 2[cAMP][CRP] &= K_{CAP}[CAP], \quad \text{and} \\ [cAMP][CAP] &= 2K_{CAP}[CAP_2], \end{aligned} \quad (15)$$

where $[cAMP]$ denotes the concentration of free cAMP and $K_{CAP} = k_-/k_+$ is the reaction dissociation constant.

Given that the production of CRP is regulated by an operon in the bacterium different from the *lac* operon, we assume a constant concentration $[CRP]$. cAMP participates in the regulation of many genes. For the purpose of this study, we assume that the concentration of cAMP bound to molecules involved in the regulation on genes other than the *lac* operon ones is constant. Thus, it can be stated that the total cAMP concentration is given by

$$[cAMP_T] = [cAMP] + [CAP] + 2[CAP_2]. \quad (16)$$

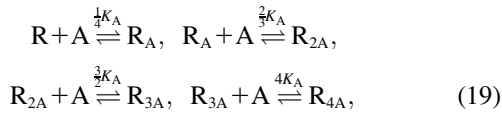
Eqs. 15 and 16 constitute a complete set of algebraic equations for variables $[cAMP]$, $[CAP]$, and $[CAP_2]$. By solving it, considering that $[CAP]$ and $[cAMP]$ must be zero when $[cAMP_T] = 0$, we obtain the following formulas to calculate $[CAP]$ and $[cAMP]$ in terms of K_{CAP} , $[CRP]$, and $[cAMP_T]$:

$$[CAP] = \frac{2K_{CAP}[cAMP_T]}{(K_{CAP} + [cAMP_T])^2} [CRP], \quad (17)$$

and

$$\begin{aligned} [cAMP] &= \frac{[cAMP_T] - K_{CAP} - 2[CRP]}{2} \\ &\quad + \frac{1}{2} \sqrt{([cAMP_T] - K_{CAP} - 2[CRP])^2 + 4[cAMP_T]K_{CAP}}. \end{aligned} \quad (18)$$

The *lac* repressor is a tetramer made up of the product of gene *lacI*. It has a high affinity for its specific DNA binding sites (operators). If there is allolactose present in the cell, it binds the repressor tetramer, decreasing its affinity for the operator sites. Up to four allolactose molecules can bind one repressor molecule according to the following sequential reactions:



where K_A is the equilibrium dissociation constant of the allolactose-repressor binding reaction.

The equilibrium conditions for the chemical reactions of Eq. 19 are

$$\begin{aligned} [R][A] &= \frac{1}{4}K_A[R_A], & [R_A][A] &= \frac{2}{3}K_A[R_{2A}], \\ [R_{2A}][A] &= \frac{3}{2}K_A[R_{3A}], & [R_{3A}][A] &= 4K_A[R_{4A}]. \end{aligned} \quad (20)$$

The total concentrations of allolactose and *lac* repressor are, respectively, given by

$$\begin{aligned} [A_T] &= [A] + [R_A] + 2[R_{2A}] + 3[R_{3A}] + 4[R_{4A}], & \text{and} \\ [R_T] &= [R] + [R_A] + [R_{2A}] + [R_{3A}] + [R_{4A}]. \end{aligned} \quad (21)$$

By solving Eqs. 20 and 21 for $[A]$ and $[R]$, and taking into consideration that when $[A_T] = 0$, $[A]$ must be zero and $[R]$ must be equal to $[R_T]$, we obtain the following expressions, which permit us calculate $[A]$ and $[R]$ in terms of $[R_T]$, $[A_T]$, and K_A :

$$[R] = \left(\frac{K_A}{K_A + [A]} \right)^4 [R_T], \quad (22)$$

and

$$\begin{aligned} [A] &= \frac{[A_T] - K_A - 4[R_T]}{2} \\ &+ \frac{1}{2} \sqrt{([A_T] - K_A - 4[R_T])^2 + 4[A_T]K_A}. \end{aligned} \quad (23)$$

NUMERICAL RESULTS

Equations 4–7 and 13 constitute a complete set of delay-differential equations, which govern the dynamics of variables $[M_B]$, $[M_P]$, $[B]$, $[P]$, $[A_T]$, and $[cAMP_T]$. All of the parameters in these equations are estimated from published experimental data in the Appendix. To numerically solve these equations, a fourth-order Runge-Kutta algorithm, adapted to deal with time delays, was implemented in Fortran.

To test the feasibility of the model presented here, an experiment by Knorre (1968) is simulated. In this experiment, Knorre let a bacterial culture grow, for a long time, in a glucose-rich and lactose-free medium, so the lactose operon is uninduced. Then, the bacteria were washed and transferred to a lactose-rich and glucose-free medium so induction of the lactose operon would take place. The temporal evolution of the operon induction, after the medium change, was determined by periodically measuring the β -galactosidase activity.

The Knorre (1968) experiment was simulated as follows. First, we set the external lactose and glucose concentration values as $[L_E] = 1.0$ mM and $[G_E] = 0.0$ mM. This situation corresponds to the lactose-rich and glucose-free medium and is enough to fully induce the lactose operon. With these conditions, the model equations were integrated over a long enough time interval such that the system reached the induced steady state. Let $[M_B]$, $[M_P]$, $[B]$ and $[P]$ denote the fully induced steady-state values of the corresponding variables. Since, in the uninduced state, the lactose operon activity is one thousandth that of the induced state (Savageau, 1999), we select the following initial conditions to account for the bacterial culture growing in the glucose-rich and lactose-free medium: $[M_B]_0 = [M_B]/1000$, $[M_P]_0 = [M_P]/1000$, $[B]_0 = [B]/1000$, $[P]_0 = [P]/1000$. To complete the initial condition set we considered $[A_T]_0 = 0$ and $[cAMP_T]_0 = 0$, given that the absence of external lactose avoids the accumulation of internal allolactose, whereas the external glucose inhibits the synthesis of cAMP. Finally, with these initial conditions and external lactose and glucose concentration values of $[L_E] = 1.0$ mM and $[G_E] = 0.0$ mM (corresponding to the lactose-rich and glucose-free environment), the model time-delay-differential equations were numerically solved for 200 min. The result of our simulation is compared against the experimental data of Knorre (1968) in Fig. 3, where it can be seen that there is good agreement between the model predictions and the experimental data. Note in particular that we have not adjusted parameters to fit the data, but rather have been able to use the model in conjunction with our parameter estimation to predict the time course shown in Fig. 3.

Now, we turn to a numerical investigation of the system bistable behavior. We start with a simulation of an experiment in which a bacterial culture grows in a medium with a constant concentration of glucose, and the lactose concentration is slowly increased (starting from zero), letting the bacterial culture lactose operon relax to equilibrium after

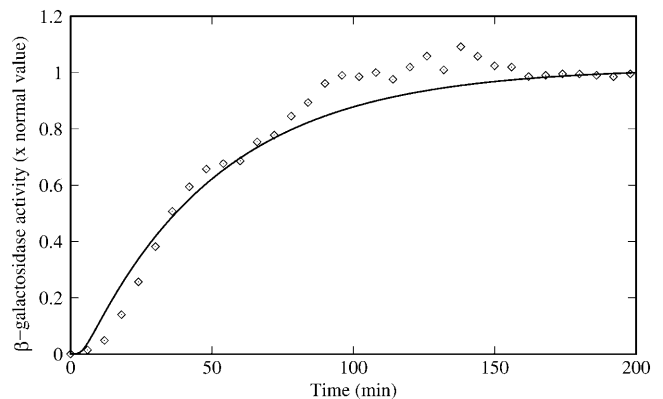


FIGURE 3 Time evolution of the β -galactosidase activity after an *E. coli* bacterial culture is changed from a glucose to a lactose growth situation. Comparison of the Knorre (1968) experimental data (diamonds) with the numeric simulation described in the text (solid line).

every incremental step, until the lactose operon is fully induced. Afterward, the medium lactose concentration is decreased back to zero in the same quasi-static way. To simulate this experiment, we started by setting the external lactose concentration ($[L_E]$) and all the model variable initial conditions to $0.0 \mu\text{M}$. Then, we numerically solved the model equations until a steady state was reached. After this, $[L_E]$ was incremented in steps of $0.25 \mu\text{M}$ and the model equations were solved again, with the previous steady-state values as initial conditions, until the system reaches a new steady state. This procedure was repeated until $[L_E] = 100.0 \mu\text{M}$. Afterward, $[L_E]$ was decreased in steps of $0.25 \mu\text{M}$, and the whole process was repeated until $[L_E] = 0.0 \mu\text{M}$. In every step, we recorded the values of $[L_E]$ and the corresponding β -galactosidase steady-state concentration. The experiment was repeated for different values of the external glucose concentration $[G_E]$.

The results of our simulations are presented in Fig. 4. It is clear that, in agreement with the results of Yildirim and Mackey (2003), the expanded model of the lactose operon of *E. coli* presented here also shows bistable behavior for realistic values of external glucose and lactose concentrations. The two main characteristics of bistability are shown in these plots. First, the steady-state β -galactosidase concentration has a discontinuous transition, from the uninduced to the induced steady state, when the external lactose concentration surpasses a threshold value; and second, in order for the system to switch back to the uninduced state, the external lactose concentration must be decreased below a second (smaller) threshold level.

Yildirim and Mackey (2003), with a model that considered neither catabolite repression nor inducer exclusion, found that the transition from the uninduced to the induced state took place around $L_E \approx 60 \mu\text{M}$. As seen in Fig. 4, there is a qual-

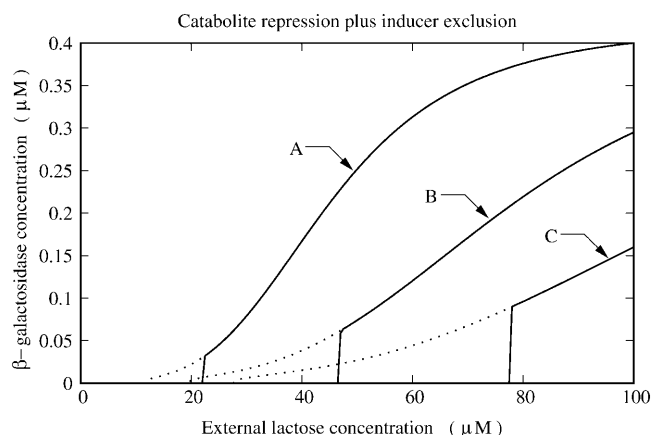


FIGURE 4 Plots of steady-state β -galactosidase concentration versus external lactose concentration for various values of external glucose concentration ($[G_E]$): (A) $[G_E] = 0.0 \mu\text{M}$; (B) $[G_E] = 140.0 \mu\text{M}$; and (C) $[G_E] = 280.0 \mu\text{M}$. The black and gray lines correspond to the increasing and decreasing external lactose concentration pathways, respectively.

itative agreement between our results, which account for both glucose-dependent mechanisms, and those of Yildirim and Mackey. It is known that the presence of glucose in the external medium affects the induction of the lactose operon, making it more difficult. There are two different ways in which the external glucose could influence the system induction: by increasing the induction threshold and by decreasing the activation level of the already induced operon. From Fig. 4 we can observe that both phenomena occur.

Extracellular glucose affects the lactose operon activation by inhibiting the production of cAMP (catabolite repression) and by reducing the efficiency of lactose permease to transport lactose molecules into the cell (inducer exclusion). Both mechanisms are taken into account in this model, and, thus, the results in Fig. 4 reflect both effects. To figure out what the separate effects of catabolite repression and inducer exclusion are, we repeated the same numerical experiments with two hypothetical mutant strains of *E. coli*. In one of them (let us call it *nlac*), catabolite repression works as in the wild strain, but inducer exclusion is absent, and thus the efficiency of glucose transport by lactose permease is independent of the extracellular glucose concentration. In the second hypothetical mutant strain (*ncat*), inducer exclusion is normal, but catabolite repression has been shut down so the cAMP synthesis rate is independent of the extracellular glucose concentration.

In Fig. 5, the results of the numerical experiments with strain *nlac* are shown. There, we see that catabolite repression is capable, on its own, of increasing the external lactose concentration induction threshold, although not as efficiently as when it is combined with inducer exclusion. Two other interesting features of the plots in Fig. 5 are that the external

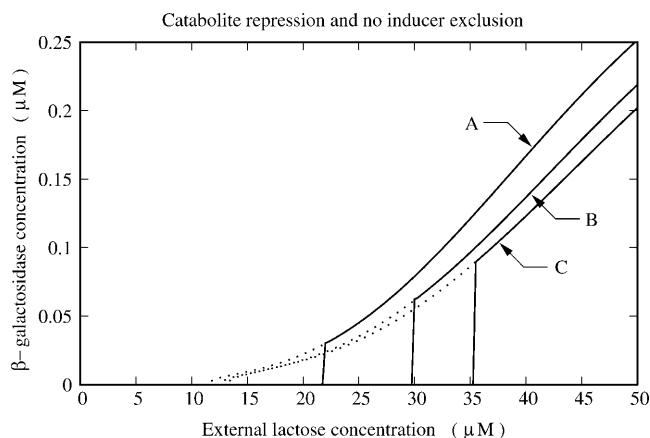


FIGURE 5 Plots of steady-state β -galactosidase concentration versus external lactose concentration for various values of external glucose concentration ($[G_E]$): (A) $[G_E] = 0.0 \mu\text{M}$; (B) $[G_E] = 140.0 \mu\text{M}$; and (C) $[G_E] = 280.0 \mu\text{M}$. The black and gray lines correspond to the increasing and decreasing external lactose concentration pathways, respectively. The experiments were carried out with the hypothetical mutant strain *nlac*, in which inducer exclusion is absent so the efficiency of glucose transport by lactose permease is independent of extracellular glucose.

lactose concentration uninduction threshold is almost independent of the extracellular glucose concentration, and that the effect of glucose on the activation level (the steady-state β -galactosidase concentration) of the induced operon is not as pronounced as we observed in Fig. 4.

The results of the numerical simulations with strain *ncat* are plotted in Fig. 6. In this case, we see that inducer exclusion affects the *lac* operon inductions by increasing the extracellular lactose concentration induction threshold and decreasing the activation level of the induced operons. The uninduction threshold is also affected, although not as strongly as are the two other features.

After comparing the results in Figs. 4–6, we observe that neither the effects of catabolite repression nor inducer exclusion are dominant in the range of external glucose concentration we explored. Neither is capable of accounting for the intensity of the global effect. On the other hand, the saturation constant of catabolite repression is $40.0\ \mu\text{M}$, whereas that of inducer exclusion is $271.0\ \mu\text{M}$. This means that catabolite repression is more sensitive at external glucose concentrations $\sim 50\ \mu\text{M}$, whereas the sensitivity of inducer exclusion is optimized at $\sim 300\ \mu\text{M}$ of external glucose. This can be seen by comparing Figs. 5 and 6. The effects of catabolite repression are not very different between external glucose concentrations of $140\ \mu\text{M}$ and $280\ \mu\text{M}$, whereas the effects of inducer exclusion are significantly enhanced in this range. Thus, catabolite repression and inducer exclusion seem to be complementary mechanisms. The above assertion follows from the facts that the bacterial response to the external glucose is, somehow, the sum of the catabolite repression and the inducer exclusion individual responses, and that the glucose sensitivities of both mech-

anisms are such that they enhance the bacterial glucose sensitivity range.

CONCLUDING REMARKS

The mathematical and computational modeling of biological systems is a subject of increasingly intense interest. The accelerating growth of biological knowledge, in concert with a growing appreciation of the spatial and temporal complexity of such systems, threatens to overwhelm our capacity to integrate, understand, and reason about biology and biological function. The construction, analysis, and simulation of formal mathematical models is a useful way to manage such problems. Genetic regulation is an area in which this approach is particularly promising. The galactose, tryptophan, and lactose operons in *E. coli*, as well as the lysis/lysogeny switch of phage lambda, are examples of molecular systems in which the extant amount of experimental data concerning their functioning permit one to construct detailed mathematical models, capable of making precise dynamic predictions.

Experimental evidence of bistability in the *lac* operon was first found by Novick and Wiener (1957). Previous mathematical models of the lactose operon have demonstrated the possibility of bistable behavior with a proper choice of the model parameters (Laurent and Kellershohn, 1999), and that this behavior is indeed predicted by the models with realistic parameter values (Yildirim and Mackey, 2003). However, none of these models has taken into account two glucose-dependent mechanisms that play an important role in the regulation of the lactose operon regulatory pathway: catabolite repression and inducer exclusion.

In this paper, we developed a mathematical model of the lactose operon which considers both catabolite repression and inducer exclusion, as well as all other known regulatory mechanisms and the time delays inherent to transcription and translation. We have paid special attention to the estimation of all of the model parameters from published experimental data. The accuracy of the model was tested by simulating an experiment of Knorre (1968). In this experiment, a bacterial culture feeding on glucose, and thus having the lactose operon uninduced, was suddenly switched to a lactose-rich medium. Then, the β -galactosidase activity was periodically measured for $>3\ \text{h}$, until the lactose operon was fully activated. The results of our simulation (the details are given in Numerical Results) are compared with the experimental data of Knorre in Fig. 3. From that comparison, we see that there is good agreement between the simulation predictions and the experimental data. Thus, we conclude that the model is reliable enough to numerically analyze the system bistable behavior and explore the individual effects of catabolite repression and inducer exclusion.

It is known that the presence of glucose in the extracellular medium makes induction of the lactose operon more difficult in a given *E. coli* culture. This could happen in two different

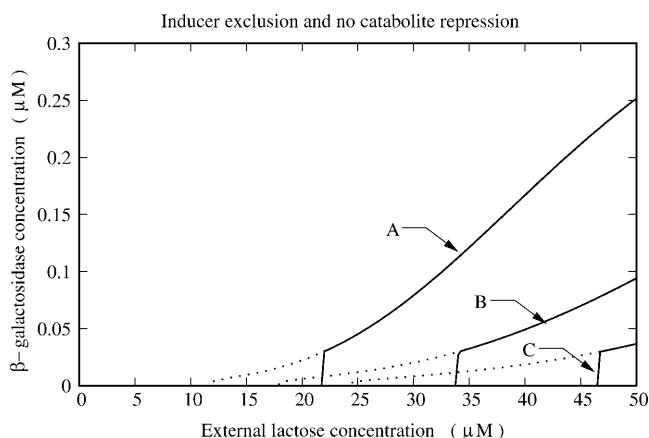


FIGURE 6 Plots of steady-state β -galactosidase concentration versus external lactose concentration for various values of external glucose concentration ($[G_E]$): (A) $[G_E] = 0.0\ \mu\text{M}$; (B) $[G_E] = 140.0\ \mu\text{M}$; and (C) $[G_E] = 280.0\ \mu\text{M}$. The black and gray lines correspond to the increasing and decreasing external lactose concentration pathways, respectively. The experiments were carried out with the hypothetical mutant strain *ncat*, in which the cAMP synthesis rate is independent of the extracellular glucose concentration.

ways: by increasing the threshold value the external lactose concentration must surpass to make the system switch from the uninduced to the induced state, and by decreasing the activation level of the already induced operon. Our results indicate that both phenomena take place. When the individual responses of catabolite repression and inducer exclusion are separately studied, both mechanisms are complementary in the sense that their individual responses add up to account for the system global response to external glucose, and their combined effect enlarges the system external glucose sensitivity range. In conclusion, our modeling approach helps us to understand the dynamic response, via the mechanisms of catabolite repression and induced exclusion, of the lactose operon to extracellular glucose, and provides quantitative predictions as well. These predictions may, in principle, be tested experimentally.

APPENDIX: PARAMETER ESTIMATION

Energy of the *lac* operon 50 possible binding states

The binding energy of any one of the 50 binding states of the *lac* operon can be calculated from

$$E_i = \sum_{\lambda=1}^4 \varepsilon_{i,\lambda} + \varepsilon_{i,12} + \varepsilon_{i,13} + \varepsilon_{i,14} + \varepsilon_{i,34}, \quad (24)$$

where $\varepsilon_{i,\lambda}$ is the binding energy of site λ in the i th state, and $\varepsilon_{i,12}$, $\varepsilon_{i,13}$, $\varepsilon_{i,14}$, and $\varepsilon_{i,34}$ are cooperativity energies, due to the interaction between a CAP complex and a mRNAP simultaneously bound to C1 and P1 and to a repressor simultaneously binding two different operators. If sites ν and κ ($\nu < \kappa$) are bound by the same repressor molecule, only the individual binding energy $\varepsilon_{i,\nu}$ is considered in the sum of Eq. 24. The additional energy due to the repressor also binding site κ is accounted for as the cooperativity energy $\varepsilon_{i,\lambda\kappa}$.

For a given i , the value of $\varepsilon_{i,\lambda}$ is defined as

$$\varepsilon_{i,\lambda} = \begin{cases} 0, & \text{if site } \lambda \text{ is empty} \\ \Delta G_{\lambda c}, & \text{if site } \lambda \text{ is bound by a cAMP} \\ \Delta G_{\lambda r}, & \text{if site } \lambda \text{ is bound by a repressor} \\ \Delta G_{\lambda p}, & \text{if site } \lambda \text{ is bound by a mRNAP.} \end{cases} \quad (25)$$

The energies $\Delta G_{\lambda c}$, $\Delta G_{\lambda r}$, and $\Delta G_{\lambda p}$ are estimated below. The cooperativity energies can be calculated as

$$\varepsilon_{i,12} = \begin{cases} \Delta G_{cp}, & \text{if a cAMP and a mRNAP are} \\ & \text{bound to sites 1 and 2,} \\ & \text{respectively;} \\ 0, & \text{otherwise;} \end{cases} \quad (26)$$

$$\varepsilon_{i,13} = \begin{cases} \Delta G_{13}, & \text{if a repressor is bound to sites 1 and 2,} \\ & \text{simultaneously;} \\ 0, & \text{otherwise;} \end{cases} \quad (27)$$

$$\varepsilon_{i,14} = \begin{cases} \Delta G_{14}, & \text{if a repressor is bound to sites} \\ & \text{1 and 4, simultaneously;} \\ 0, & \text{otherwise;} \end{cases} \quad (28)$$

and

$$\varepsilon_{i,34} = \begin{cases} \Delta G_{34}, & \text{if a repressor is bound to sites} \\ & \text{3 and 4, simultaneously} \\ 0, & \text{otherwise.} \end{cases} \quad (29)$$

Energies ΔG_{13} , ΔG_{14} , and ΔG_{34} are estimated below, as well.

Cooperativity and individual binding energies

In this section, the following relation between the binding energy (ΔG) and the association constant (K_B) of a given chemical reaction is used:

$$\Delta G = -RT \ln K_B.$$

Here, we take $T = 37^\circ\text{C}$, which corresponds to

$$RT = 0.617 \text{ kcal/mol.}$$

ΔG_{2p} and ΔG_{cp}

Malan et al. (1984) measured the association constant of mRNA polymerase binding to P1 in the absence and in the presence of cAMP. They report the following values, respectively: $K_B^{-\text{cAMP}} \approx 1.5 \times 10^7 \text{ M}^{-1}$ and $K_B^{+\text{cAMP}} \approx 2.0 \times 10^8 \text{ M}^{-1}$. From this, ΔG_{2p} and ΔG_{cp} can be calculated as

$$\Delta G_{2p} = -RT \ln K_B^{-\text{cAMP}} \approx -10.20 \text{ kcal/mol,}$$

and

$$\Delta G_{cp} = -RT \ln K_B^{+\text{cAMP}} - \Delta G_{2p} \approx -1.59 \text{ kcal/mol.}$$

ΔG_{1c}

Pyles and Lee (1996) report the following value for the CAP-C1 binding association constant: $4.1 \times 10^7 \text{ M}^{-1}$. Baker et al. (2001) found $2.5 \times 10^7 \text{ M}^{-1}$. Here, we take the mean value ($3.3 \times 10^7 \text{ M}^{-1}$), from which

$$\Delta G_{1c} \approx -10.68 \text{ kcal/mol.}$$

ΔG_{3c}

According to Hudson and Fried (1991), the affinity of CAP for site C2 is $\sim 1/30$ that for site C1, i. e., the corresponding association constant is of the order of $1.1 \times 10^6 \text{ M}^{-1}$. The binding energy calculated from this association constant is

$$\Delta G_{3c} \approx -8.58 \text{ kcal/mol.}$$

ΔG_{3r}

Falcon and Matthews (2000) report for the association constant of the repressor-O1 binding reaction a value of $1.0 \times 10^{11} \text{ M}^{-1}$. This association constant corresponds to the following binding energy:

$$\Delta G_{3r} \approx -16.97 \text{ kcal/mol.}$$

ΔG_{1r} and ΔG_{4r}

According to Oehler et al. (1994), the affinities of O2 and O3 for the repressor are $1/10$ and $1/300$ that of O1, respectively. This means that the corresponding association constants are $1.0 \times 10^{10} \text{ M}^{-1}$ and $3.33 \times 10^8 \text{ M}^{-1}$. Therefore, the binding energies are

$$\Delta G_{4r} \approx -14.21 \text{ kcal/mol,}$$

and

$$\Delta G_{1r} \approx -12.11 \text{ kcal/mol.}$$

ΔG_{34}

From Oehler et al. (1994), when a repressor molecule is simultaneously bound to O1 (site 3) and O2 (site 4), the affinity of the complex is 5 times that of the repressor-O1 complex. From this

$$\Delta G_{34} = -RT \ln 5 \approx -1.0 \text{ kcal/mol.}$$

ΔG_{13}

Oehler et al. (1994) also report that when a repressor is bound to O1 (site 3) and O3 (site 1), the affinity is 100 times that of the repressor-O1 complex. This further implies that

$$\Delta G_{13} = -RT \ln 100 + \Delta G_{3r} - \Delta G_{1r} \approx -7.70 \text{ kcal/mol.}$$

ΔG_{14}

Finally, Oehler et al. (1994) assert that when a repressor is simultaneously bound to O2 (site 4) and O3 (site 1), the affinity is 2000 times that of the repressor-O2 complex. This means that

$$\Delta G_{14} = -RT \ln 2000 + \Delta G_{4r} - \Delta G_{1r} \approx -6.78 \text{ kcal/mol.}$$

The energies of the 50 binding states calculated from the above estimated values and Eqs. 24–29 are tabulated in Table 2.

Transcription and translation parameters

E. coli volume

E. coli are rod-like bacteria 3–5 μm long and 0.5 μm in diameter, so they have a volume in the range from 6.0×10^{-16} L to 9.8×10^{-16} L. We take a mean volume of 8.0×10^{-16} L.

Growth rate, μ

The growth rate of a bacterial culture depends strongly on the growth medium conditions. Typically, the mass doubling time varies from 20 to >40 min (Bremer and Dennis, 1996). For the purpose of this study, we consider a doubling time of 30 min, which corresponds to the following growth rate:

$$\mu \approx 0.02 \text{ min}^{-1}.$$

TABLE 2 Energies of the 50 *lac* operon possible binding states

$E_{eeee} \approx 0.0$	$E_{reec} \approx -12.11$	$E_{ceee} \approx -10.68$	$E_{ree1} \approx -18.90$
$E_{eeer} \approx -14.21$	$E_{reer} \approx -26.32$	$E_{ceer} \approx -24.89$	$E_{rer1} \approx -35.87$
$E_{eere} \approx -16.97$	$E_{rere} \approx -29.08$	$E_{cere} \approx -27.65$	$E_{rec1} \approx -27.48$
$E_{eerr} \approx -31.18$	$E_{rerr} \approx -43.29$	$E_{cerr} \approx -41.86$	$E_{rpe1} \approx -29.10$
$E_{eece} \approx -8.58$	$E_{reec} \approx -20.69$	$E_{cece} \approx -19.26$	$E_{rpr1} \approx -46.07$
$E_{eerc} \approx -22.79$	$E_{reer} \approx -34.90$	$E_{ceer} \approx -33.47$	$E_{rpe1} \approx -37.68$
$E_{epee} \approx -10.20$	$E_{rpee} \approx -22.31$	$E_{cpce} \approx -22.47$	$E_{eer3} \approx -17.97$
$E_{eper} \approx -24.41$	$E_{rper} \approx -36.52$	$E_{cper} \approx -36.68$	$E_{ep3} \approx -28.17$
$E_{epre} \approx -27.17$	$E_{rpre} \approx -39.28$	$E_{cpre} \approx -39.44$	$E_{rer3} \approx -30.08$
$E_{eprr} \approx -41.38$	$E_{rpr} \approx -53.49$	$E_{cpr} \approx -53.65$	$E_{rpr3} \approx -40.28$
$E_{epce} \approx -18.78$	$E_{rpe} \approx -30.89$	$E_{cpce} \approx -31.05$	$E_{eer3} \approx -28.65$
$E_{eper} \approx -32.99$	$E_{rper} \approx -45.10$	$E_{cper} \approx -45.26$	$E_{cpr3} \approx -40.44$
			$E_{re1e} \approx -19.81$
			$E_{re1r} \approx -34.02$

All of these energies are expressed in units of kcal/mol.

mRNA polymerase concentration, $[mRNAP]$

According to Bremer and Dennis (1996), there are ~ 1500 active RNA polymerase molecules per cell in *E. coli* bacterial cultures growing at the rate μ estimated above. This leads to a concentration

$$[mRNAP] \approx 3.0 \mu\text{M.}$$

Promoter concentration, $[P1]$

According to Bremer and Dennis (1996), there are ~ 2.5 genome equivalents per average *E. coli* cell at the growth rate determined by μ . Assuming one promoter P1 per genome equivalent, the right promoter concentration can be estimated as

$$[P1] \approx 5.0 \times 10^{-3} \mu\text{M.}$$

CRP concentration, $[CRP]$

According to Anderson et al. (1971), there are ~ 1300 molecules of cAMP receptor protein concentration per *E. coli* cell. This corresponds to

$$[CRP] \approx 2.6 \mu\text{M.}$$

Total repressor concentration, $[R_T]$

From Yagil and Yagil (1971), the product of the total repressor concentration and the repressor-operator reaction association constant has an average value of 2.9×10^3 . On the other hand, Falcon and Matthews (2000) report for the association constant of the repressor-O1 binding reaction a value of $1.0 \times 10^{11} \text{ M}^{-1}$. Therefore, we can estimate the total repressor concentration as

$$[R_T] \approx 2.9 \times 10^{-2} \mu\text{M.}$$

This value agrees with the experimental results of Gilbert and Müller-Hill (1966), who estimated that ~ 10 – 20 copies of the *lac* repressor are present per cell. This corresponds to a concentration between $2.0 \times 10^{-2} \mu\text{M}$ and $4.0 \times 10^{-2} \mu\text{M}$.

Transcription initiation rate, k_m

Malan et al. (1984) measured the transcription initiation rate at P1 and report the following value:

$$k_m \approx 0.18 \text{ min}^{-1}.$$

mRNA degradation rate, ξ_M

Kennell and Riezman (1977), measured a *lacZ* mRNA half-life of 1.5 min. From this

$$\xi_M = \frac{\ln 2}{1.5 \text{ min}} = 0.46 \text{ min}^{-1}.$$

lacZ mRNA translation initiation rate, κ_B

From Kennell and Riezman (1977), translation starts every 3.2 s at the *lacZ* mRNA. This leads to the following translation initiation rate:

$$\kappa_B \approx 18.8 \text{ min}^{-1}.$$

***lacY* mRNA translation initiation rate, κ_P**

According to Beckwith (1987), the production rate of *lac* permease is smaller than that of β -galactosidase monomers even though, as Kennell and Riezman (1977) report, there are similar levels of both mRNA species. This suggests that *lacY* mRNAs are translated at a lower rate. Nevertheless, to our knowledge, there are no reported measurements of the *lacY* mRNA translation initiation rate. Thus, we assume it is equal to that of *lacZ*:

$$\kappa_P \approx 18.8 \text{ min}^{-1}.$$

 ***β* -galactosidase degradation rate, ξ_B**

The breakdown rate of β -galactosidase was measured by Mandelstam (1957), who found it to be 0.05 per hour. This corresponds to

$$\xi_B \approx 8.33 \times 10^{-4} \text{ min}^{-1}.$$

***lac* permease degradation rate, ξ_P**

According to Kennell and Riezman (1977), the degradation rate of this protein is

$$\xi_P \approx 0.01 \text{ min}^{-1}.$$

Equilibrium dissociation constant between CRP and cAMP, K_{CAP}

From the experimental results of Baker et al. (2001)

$$K_{CAP} \approx 3.0 \mu\text{M}.$$

Time delay between transcription initiation and appearance of a *lacZ* ribosome binding site, τ_B

Once a RNA polymerase has transcribed a mRNA chain long enough for a *lacZ* ribosome to bind to it, translation can start. According to Draper (1996), efficient mRNAs can initiate translation every 3 s. From this and the fact that the mRNA chain elongation rate is of the order of 50 nucleotide/s (Bremer and Dennis, 1996), <150 nucleotides are required for a ribosome to bind a mRNA and start translation. Furthermore, the DNA chain elongation rate is at least 490 nucleotide/s (Bremer and Dennis, 1996). Thus it takes <0.31 s after transcription initiation to have a *lacZ* ribosome binding site, i.e.,

$$\tau_B \approx 5.1 \times 10^{-3} \text{ min}.$$

Time delay between transcription initiation and appearance of a *lacY* ribosome binding site, τ_P

Since gene *lacZ* precedes gene *lacY*, the former one has to be completely transcribed before we have a *lacY* ribosome binding site. Since gene *lacZ* is 2994 basepairs long and the DNA chain elongation rate is at least 490 nucleotide/s (Bremer and Dennis, 1996), we can estimate

$$\tau_P \approx 0.1 \text{ min}.$$

Time delay due to translation of genes *lacZ* (T_B) and *lacY* (T_P)

The monomers of β -galactosidase (the product of gene *lacZ*) and *lac* permease (the product of gene *lacY*) are, respectively, 998 and 417 amino acids long. This means that gene *lacZ* is 2994 basepairs long, whereas *lacY* is 1251 basepairs long. From this and taking into account that, according to Bremer and Dennis (1996), the mRNA chain elongation rate is ~ 50 nucleotide/s, the times it takes for genes *lacZ* and *lacY* to be translated are

$$T_B \approx 1.0 \text{ min},$$

and

$$T_P \approx 0.42 \text{ min}.$$

Lactose and allolactose dynamics parameters***Lactose transport rate and saturation constants, ϕ_{L_1} and Φ_{L_1}***

From Lolkema et al. (1991), these constants can be estimated as

$$\phi_{L_1} \approx 1.08 \times 10^3 \text{ min}^{-1},$$

and

$$\Phi_{L_1} \approx 5.0 \times 10^{-2} \mu\text{M}.$$

Lactose hydrolysis rate and saturation constants, ϕ_{L_2} and Φ_{L_2}

We estimate these parameters from the data reported in Martínez-Bilbao et al. (1991) as

$$\phi_{L_2} \approx 3.60 \times 10^3 \text{ min}^{-1},$$

and

$$\Phi_{L_2} \approx 1.4 \times 10^3 \mu\text{M}.$$

Lactose transport constant for inhibition by glucose, Φ_{G_2}

This parameter can be estimated from the data reported by Winkler and Wilson (1967):

$$\Phi_{G_2} \approx 2.71 \times 10^2 \mu\text{M}.$$

Allolactose degradation rate constant, ξ_A

Following Wong et al. (1997) we consider this parameter to be negligible:

$$\xi_A \approx 0.0 \text{ min}^{-1}.$$

Catabolite repression and operon induction parameters***cAMP synthesis rate constant, ϕ_{cAMP}***

This parameter was estimated from the data reported by Epstein et al. (1975):

$$\phi_{cAMP} \approx 5.5 \mu\text{M min}^{-1}.$$

cAMP synthesis saturation constant, Φ_{cAMP}

From Notley and Ferenci (1995), the saturation constant for cAMP synthesis can be estimated as

$$\Phi_{cAMP} \approx 40.0 \mu\text{M}.$$

cAMP excretion and degradation rate, ξ_{cAMP}

The compound excretion and degradation rate for cAMP was measured by Epstein et al. (1975):

$$\xi_{cAMP} \approx 2.1 \text{ min}^{-1}.$$

Repressor-allolactose dissociation constant, K_A

From the data in Jobe and Bourgeois (1972) and von Hippel et al. (1974), K_A was estimated as

$$K_A \approx 1.0 \mu\text{M}.$$

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