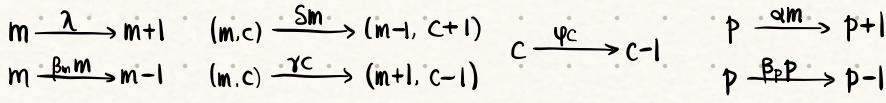


Appendix A : Protein variability for constant transcription



The standard linear noise approach is described in detail in [2].

Note that in this article the authors defined $M_{ij} = \frac{\langle x_i \rangle}{\langle x_j \rangle} K_{ij}$ instead of $M_{ij} = -\frac{\langle x_i \rangle}{\langle x_j \rangle} K_{ij}$ in [2].

Now we have

$$\hat{K} = \begin{pmatrix} -\beta_m - S & Y & 0 \\ S & -(r + \psi) & 0 \\ \alpha & 0 & -\beta_p \end{pmatrix} \xrightarrow[\text{with flux balance}]{\text{normalize}} \hat{M} = \begin{pmatrix} -\frac{1}{\tau_m} & \frac{R}{\tau_m} & 0 \\ \frac{1}{\tau_c} & -\frac{1}{\tau_c} & 0 \\ \frac{1}{\tau_p} & 0 & -\frac{1}{\tau_p} \end{pmatrix} \quad (25)$$

$$D_{ij} = \frac{\sum_k S_i^k S_j^k r_k(\bar{x})}{\langle x_i \rangle \langle x_j \rangle}$$

$$\hat{D} = \begin{pmatrix} \frac{2}{\tau_m \langle m \rangle} & -\frac{R}{\tau_m \langle c \rangle} - \frac{1}{\tau_c \langle m \rangle} & 0 \\ -\frac{R}{\tau_m \langle c \rangle} - \frac{1}{\tau_c \langle m \rangle} & \frac{2}{\tau_c \langle c \rangle} & 0 \\ 0 & 0 & \frac{2}{\tau_p \langle p \rangle} \end{pmatrix} \quad (25)$$

$$\eta_{ij} = \frac{\text{Cov}(x_i, x_j)}{\langle x_i \rangle \langle x_j \rangle}$$

$$\text{Lyapunov equation } \hat{M} \hat{\eta} + \hat{\eta} \hat{M}^T + \hat{D} = \hat{0} \quad (24)$$

Plug (25) into (24), we can get (26).

A.2 mRNA autocorrelations

$$A_{ij}(t) = \frac{\langle x_i(t) x_j(0) \rangle - \langle x_i(t) \rangle \langle x_j(0) \rangle}{\langle x_i \rangle \langle x_j \rangle}$$

Note that $A_{ij}(t)$ is normalized, where $x_i = x_j = m \Rightarrow A_{mm}(t)$.

$$\frac{\partial}{\partial t} \hat{A}(t) = \hat{M} \hat{A}(t), \quad \hat{A}(t) = \hat{\eta} \exp(\hat{M}t)$$

In Eq.(30), $A_m(t) = \frac{\langle m(0)m(t) \rangle - \langle m \rangle^2}{\text{Var}(m)}$ is not my $A_{mm}(t)$!

$$\frac{\partial}{\partial t} \begin{pmatrix} A_{mm}(t) & A_{mc}(t) & A_{cp}(t) \\ A_{cm}(t) & A_{cc}(t) & A_{cp}(t) \\ A_{pm}(t) & A_{pc}(t) & A_{pp}(t) \end{pmatrix} = \begin{pmatrix} -\frac{1}{\tau_m} & \frac{R}{\tau_m} & 0 \\ \frac{1}{\tau_c} & -\frac{1}{\tau_c} & 0 \\ \frac{1}{\tau_p} & 0 & -\frac{1}{\tau_p} \end{pmatrix} \begin{pmatrix} A_{mm}(t) & A_{mc}(t) & A_{cp}(t) \\ A_{cm}(t) & A_{cc}(t) & A_{cp}(t) \\ A_{pm}(t) & A_{pc}(t) & A_{pp}(t) \end{pmatrix}$$

$$\frac{d}{dt} A_{pm}(t) = \frac{1}{\tau_p} A_{mm}(t) - \frac{1}{\tau_p} A_{pm}(t)$$

Denote the linear relationship $A_{mp}(t) = \mathbb{L}_1[A_{mm}(t)]$

Use linear response, if $A_{mp}(t) = \int_{-\infty}^t A_{mm}(t') I_{mp(A)}(t-t') dt'$, then $I_{mp(A)}(t) = \mathbb{L}_1[\delta(t)]$.

Take $t=0$, we get

Note that $I_{mp(A)}(t)=0$ when $t<0$.

$$\eta_{mp} = \int_{-\infty}^0 A_{mm}(t') I_{mp(A)}(-t') dt'$$

$$= \int_0^\infty A_{mm}(-t) I_{mp(A)}(t) dt'$$

$$= \int_0^\infty A_{mm}(t) I_{mp(A)}(t) dt' \quad (\text{stationary}) \quad (30)$$

Solve $I_{mp(A)}(t)$.

$$\frac{d}{dt} I_{mp(A)}(t) = \frac{1}{\tau_p} S(t) - \frac{1}{\tau_p} I_{mp(A)}(t).$$

For $t > 0$, we have $I_{mp(A)}(t) = C_0 e^{-\frac{t}{\tau_p}}$, where C_0 should be determined below.

Take $\varepsilon > 0$ and $\varepsilon \rightarrow 0$, integrate from $-\varepsilon$ to ε .

$$I_{mp(A)}(\varepsilon) - I_{mp(A)}(-\varepsilon) = \frac{1}{\tau_p} - \frac{1}{\tau_p} \int_{-\varepsilon}^{\varepsilon} I_{mp(A)}(t) dt$$

From causality, $I_{mp(A)}(t) = 0$ for $t < 0$. $I_{mp(A)}$ doesn't diverge.

$$I_{mp(A)}(\varepsilon) - 0 = \frac{1}{\tau_p} - 0 \Rightarrow I_{mp(A)}(\varepsilon) = \frac{1}{\tau_p} \Rightarrow C_0 = \frac{1}{\tau_p}, \text{ that is, } I_{mp(A)}(t) = \frac{1}{\tau_p} e^{-\frac{t}{\tau_p}} \text{ for } t > 0.$$

$$\frac{\text{Cov}(m, p)}{\langle m \rangle \langle p \rangle} = \int_0^\infty \frac{\langle m(t) m(t') \rangle - \langle m \rangle^2}{\langle m \rangle^2} I_{mp(A)}(t) dt'$$

$$\text{In (30), } I(t) = \frac{\langle p \rangle}{\langle m \rangle} I_{mp(A)}(t) = \frac{\langle p \rangle}{\langle m \rangle} \frac{e^{-\frac{t}{\tau_p}}}{\tau_p} \text{ As (33).}$$

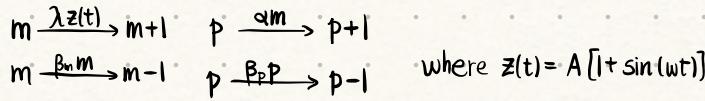
To explain (34), Since we have $\hat{A}(t) = \hat{\eta} \text{Exp}(\hat{M}t)$

Diagonalize $\hat{M} = \hat{P}^{-1} \hat{\Lambda} \hat{P}$, $\hat{\Lambda} = \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix}$, where $\lambda_1, \lambda_2, \lambda_3$ are eigenvalues of \hat{M} .

$\hat{A}(t) = \hat{\eta} \hat{P}^{-1} \text{Exp}(\hat{\Lambda}t) \hat{P} = \hat{\eta} \hat{P}^{-1} \begin{pmatrix} e^{\lambda_1 t} & 0 & 0 \\ 0 & e^{\lambda_2 t} & 0 \\ 0 & 0 & e^{\lambda_3 t} \end{pmatrix} \hat{P}$. The weird μ_1 & μ_2 in (36) are just two of those eigenvalues of \hat{M} .

Appendix C: Protein variability with periodic upstream transcription rates

I only show the case without mRNA interactions.



Denote the time-dependent ensemble averages conditioned on the history of $z(t)$.

$$\tilde{m}(t) := \langle m | z[-\infty, t] \rangle, \quad \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \frac{d\tilde{m}(t)}{dt} dt = \lambda \langle z \rangle - \beta_m \langle m \rangle$$

$$\tilde{p}(t) := \langle p | z[-\infty, t] \rangle, \quad \langle m \rangle = \frac{\lambda}{\beta_m} \langle z \rangle = \lambda \tilde{m}(t)$$

$$\frac{d\tilde{m}(t)}{dt} = \lambda z(t) - \beta_m \tilde{m}(t) \Rightarrow \tilde{m}(t) = \mathbb{L}_2[z(t)]. \text{ Linear Respond: If } \tilde{m}(t) = \int_{-\infty}^t z(t') I_{zm}(t-t') dt',$$

$$\text{then } I_{zm}(t) = \mathbb{L}_2[\delta(t)].$$

Denote the unsynchronized ensemble averages and (co)variances as

↳ time-averaged time dependent ensemble averages and (co)variances

$$\langle m \rangle = \frac{1}{T} \lim_{T \rightarrow \infty} \int_0^T \tilde{m}(t) dt, \quad \langle p \rangle, \text{Var}(m), \text{Var}(p), \text{Cov}(m, p)$$

Time-averaged signal.

$$\langle z \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T z(t) dt = A$$

$$\text{Var}(z) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \underbrace{(z(t) - \langle z \rangle)^2}_{\hookrightarrow A^2 \sin^2(\omega t)} dt = \frac{A^2}{2}$$

Solve $I_{zm}(t)$ for $t > 0$.

$$\frac{d}{dt} I_{zm}(t) = \lambda S(t) - \beta_m I_{zm}(t)$$

$$I_{zm}(t) = C_0 e^{-\beta_m t}, \text{ where } C_0 \text{ to be determined.}$$

Take $\varepsilon > 0$ and $\varepsilon \rightarrow 0$, integrate from $-\varepsilon$ to ε .

$$I_{zm}(\varepsilon) - I_{zm}(-\varepsilon) = \lambda - \beta_m \int_{-\varepsilon}^{\varepsilon} I_{zm}(t) dt$$

$$I_{zm}(\varepsilon) - 0 = \lambda - 0 \Rightarrow C_0 = \lambda \Rightarrow I_{zm}(t) = \lambda e^{-\beta_m t} \text{ for } t > 0.$$

Then solve for $Cov(m, z)$.

$$\tilde{m}(t) = \int_{-\infty}^t z(t') I_{zm}(t-t') dt'$$

$$\tilde{m}(t) z(t) = \int_{-\infty}^t z(t') z(t) I_{zm}(t-t') dt'$$

$$\begin{aligned} \tilde{m}(t) z(t) - \langle z \rangle^2 &= \int_{-\infty}^t (z(t') z(t) - \langle z \rangle^2) I_{zm}(t-t') dt' \\ &= \int_{-\infty}^t (A^2 \sin(\omega t') \sin(\omega t) - A^2) I_{zm}(t-t') dt' \\ &= \int_{-\infty}^t A^2 (\sin(\omega t') \sin(\omega t) + \sin(\omega t') + \sin(\omega t)) I_{zm}(t-t') dt' \\ &\stackrel{t=t-t'}{=} + \int_0^{+\infty} A^2 (\sin(\omega(t-\tau)) \sin(\omega t) + \sin(\omega(t-\tau)) + \sin(\omega t)) I_{zm}(\tau) d\tau \\ &= \lambda A^2 \int_0^{+\infty} (\underbrace{\sin(\omega(t-\tau)) \sin(\omega t)} + \underbrace{\sin(\omega(t-\tau))} + \underbrace{\sin(\omega t)}) e^{-\beta_m \tau} d\tau \\ &= \lambda A^2 \left[\frac{\beta_m}{\beta_m^2 + \omega^2} \sin^2(\omega t) - \frac{\beta_m}{\beta_m^2 + \omega^2} \sin \omega(t-\tau) \right] \Big|_{\tau=0}^{+\infty} \end{aligned}$$

$$\begin{aligned} Cov(m, z) &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T (\tilde{m}(t) z(t) - \langle z \rangle^2) dt \\ &= \lambda A^2 \frac{\beta_m}{\beta_m^2 + \omega^2} \frac{1}{T} \lim_{T \rightarrow \infty} \left(\int_0^T \sin^2(\omega t) dt - \text{finite} \right) = \lambda A^2 \frac{\beta_m}{\beta_m^2 + \omega^2} \cdot \frac{1}{2} \end{aligned}$$

$$\begin{aligned} \eta_{mz} &= \frac{Cov(m, z)}{\langle m \rangle \langle z \rangle} = \frac{\beta_m}{\lambda A^2} Cov(m, z) = \frac{1}{2} \frac{\beta_m^2}{\beta_m^2 + \omega^2} \\ &= \frac{1}{2} \frac{1}{1 + \omega^2 \tau_m^2} \quad (58) \end{aligned}$$