

Modelling RNA Circuits

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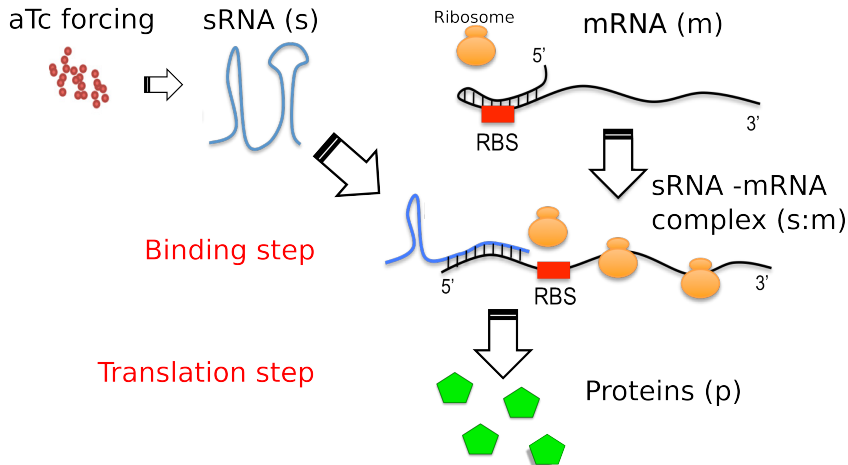
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Objectives

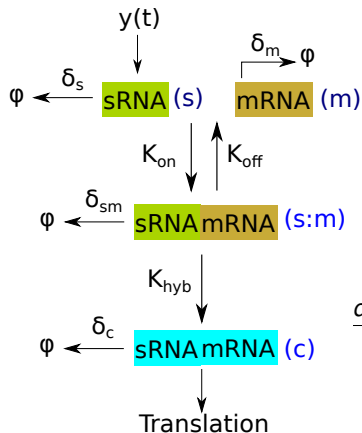
- ▶ Rational design of genetic circuits now possible
 - ▶ ODE models used for prediction.
 - ▶ Model parameters unknown.
- ▶ Project studied a recently designed genetic circuit, attempted parameter estimation.
- ▶ Parameter estimates would inform future design/experiments.

The RNA System



- ▶ mRNA produced 'self repressed' - tail folded over the RBS
- ▶ sRNA binds to it - new complex has RBS uncovered.
- ▶ Complex translated into proteins.

ODE model for system (1) - Binding step



- Binding of sRNA (s) and mRNA (m), into unstable complex ($s : m$) then stable one (c).

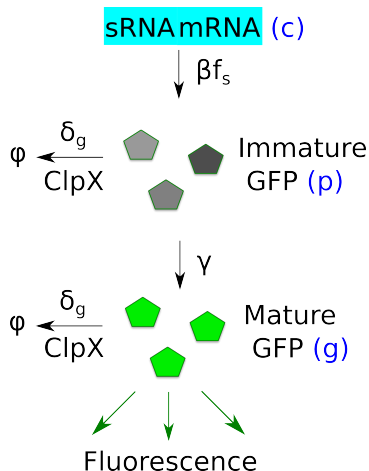
$$\frac{ds}{dt} = \frac{N_{\alpha_T}}{f_T} y(t) - (\mu + \delta_s)s - k_{on}sm + k_{off}s : m$$

$$\frac{dm}{dt} = \frac{N_{\alpha_L}}{f_L} x(t) - (\mu + \delta_m)m - k_{on}sm + k_{off}s : m$$

$$\frac{ds : m}{dt} = k_{on}sm - (k_{off} + k_{hyb})s : m - (\mu + \delta_{sm})s : m$$

$$\frac{dc}{dt} = k_{hyb}s : m - (\mu + \delta_c)c$$

ODE model for system (2) - Translation step



- ▶ Translation of stable complex (c) into immature GFP (p).
- ▶ maturation of GFP (g), machine calibration giving measured fluorescence (z).

$$\frac{dp}{dt} = \beta m + f_s \beta c - (\gamma + \mu + \delta_g)p - \frac{v_z p}{K_z + p + g}$$

$$\frac{dg}{dt} = \gamma p - (\mu + \delta_g)g - \frac{v_z g}{K_z + p + g}$$

$$z = z_0 + \frac{g}{\Theta}$$

Full model, with parameters to be estimated

$$\frac{ds}{dt} = \frac{N_{\alpha_T}}{f_T} y(t) - (\mu + \delta_s)s - k_{\text{on}}sm + k_{\text{off}}s : m \quad (1)$$

$$\frac{dm}{dt} = \frac{N_{\alpha_L}}{f_L} x(t) - (\mu + \delta_m)m - k_{\text{on}}sm + k_{\text{off}}s : m \quad (2)$$

$$\frac{ds : m}{dt} = k_{\text{on}}sm - (k_{\text{off}} + k_{\text{hyb}})s : m - (\mu + \delta_m)s : m \quad (3)$$

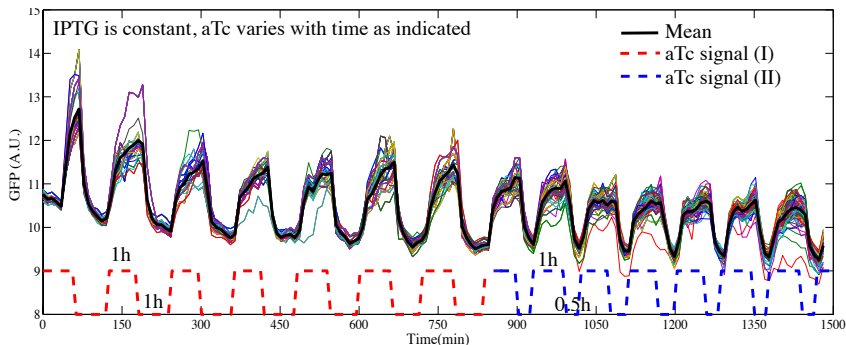
$$\frac{dc}{dt} = k_{\text{hyb}}s : m - (\mu + \delta_m)c \quad (4)$$

$$\frac{dp}{dt} = \beta m + f_s \beta c - (\gamma + \mu + \delta_g)p - \frac{v_z p}{K_z + p + g} \quad (5)$$

$$\frac{dg}{dt} = \gamma p - (\mu + \delta_g)g - \frac{v_z g}{K_z + p + g} \quad (6)$$

$$z = z_0 + \frac{g}{\ominus} \quad (7)$$

Recent experimental data



- ▶ Single cell fluorescence time series data (Jaramillo Lab, unpublished).
- ▶ Data records $z = z_0 + \frac{g}{\Theta}$

Method for parameter estimation in ODE model

Goal

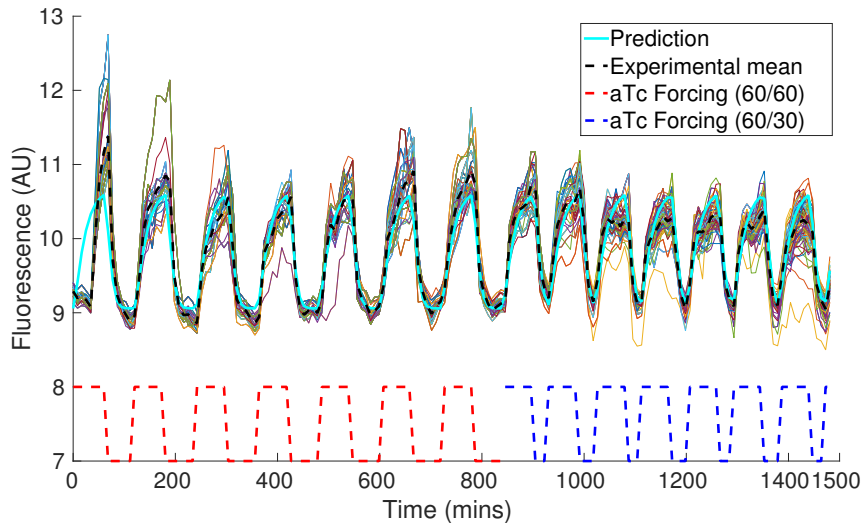
Estimate unknown parameters in model using time series data.

- ▶ Least squares error minimization approach

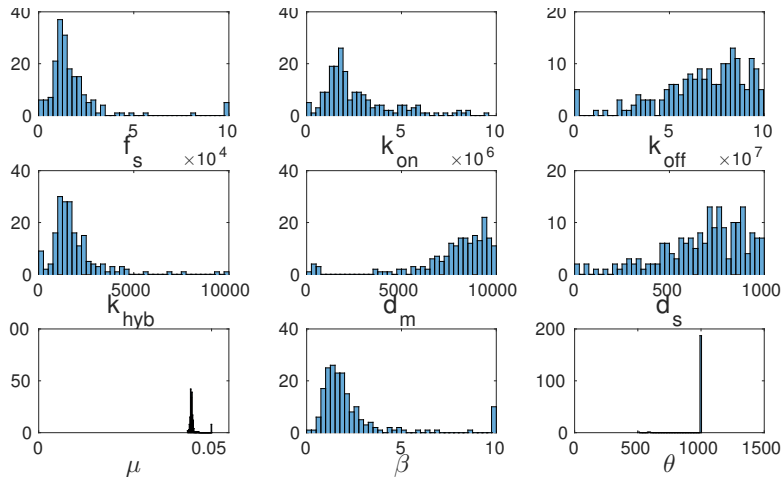
$$\arg \min_{\theta} \sum_{i=1}^n (z_{\text{exp,mean}}(t_i) - z(t_i, \theta))^2.$$

- ▶ Used evolutionary algorithm to perform minimisation.

Initial estimation results



Initial estimation results

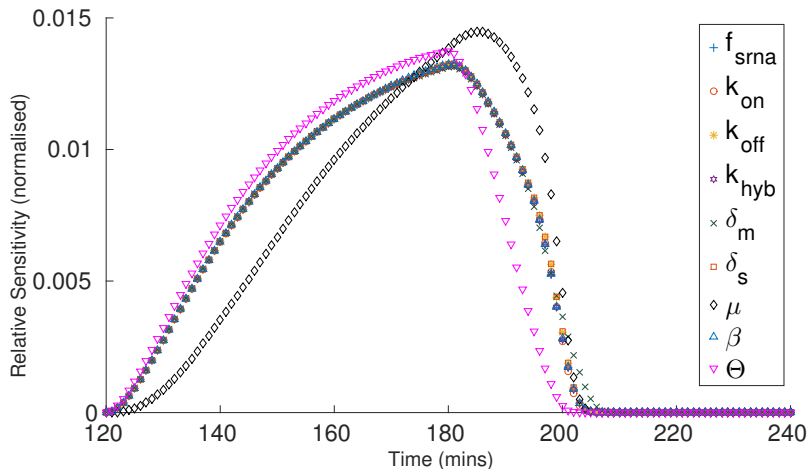


- ▶ Similar error values for all results.
- ▶ Many parameters poorly estimated.

Sensitivity analysis

$$S_{ij} = \left. \frac{\partial z}{\partial \theta_j} \right|_{t_i}$$

Perturbations of parameters all give similar effects on model output - **hard to distinguish**

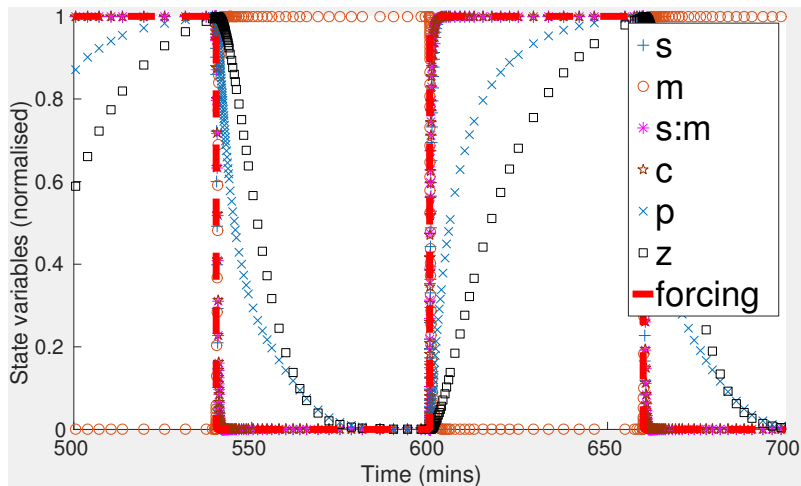


Full model, with parameters to be estimated

$$\begin{aligned}\frac{ds}{dt} &= \frac{N_{\alpha T}}{f_T} y(t) - (\mu + \delta_s)s - k_{\text{on}}sm + k_{\text{off}}s : m \\ \frac{dm}{dt} &= \frac{N_{\alpha L}}{f_L} x(t) - (\mu + \delta_m)m - k_{\text{on}}sm + k_{\text{off}}s : m \\ \frac{ds : m}{dt} &= k_{\text{on}}sm - (k_{\text{off}} + k_{\text{hyb}})s : m - (\mu + \delta_m)s : m \\ \frac{dc}{dt} &= k_{\text{hyb}}s : m - (\mu + \delta_m)c\end{aligned}$$

$$\begin{aligned}\frac{dp}{dt} &= \beta m + f_s \beta c - (\gamma + \mu + \delta_g)p - \frac{v_z p}{K_z + p + g} \\ \frac{dg}{dt} &= \gamma p - (\mu + \delta_g)g - \frac{v_z g}{K_z + p + g} \\ z &= z_0 + \frac{g}{\ominus}\end{aligned}$$

Model output for all state variables



- ▶ $s, m, s : m, c$ (binding step) respond instantly to forcing.
- ▶ p, z (translation step) on timescale of forcing.

Translation rate limiting

$$\begin{aligned}\frac{ds}{dt} &= \frac{N\alpha_T}{f_T} y(t) - (\mu + \delta_s)s - k_{\text{on}}sm + k_{\text{off}}s : m \\ \frac{dm}{dt} &= \frac{N\alpha_L}{f_L} x(t) - (\mu + \delta_m)m - k_{\text{on}}sm + k_{\text{off}}s : m \\ \frac{ds : m}{dt} &= k_{\text{on}}sm - (k_{\text{off}} + k_{\text{hyb}})s : m - (\mu + \delta_m)s : m \\ \frac{dc}{dt} &= k_{\text{hyb}}s : m - (\mu + \delta_m)c\end{aligned}$$

$$\begin{aligned}\frac{dp}{dt} &= \beta m + f_s \beta c - (\gamma + \mu + \delta_g)p - \frac{v_z p}{K_z + p + g} \\ \frac{dg}{dt} &= \gamma p - (\mu + \delta_g)g - \frac{v_z g}{K_z + p + g} \\ z &= z_0 + \frac{g}{\Theta}\end{aligned}$$

- ▶ Binding equations respond instantly.
- ▶ $\beta m + f_s \beta c$ term links binding and translation. It flips between fixed point values.
- ▶ All parameters 'upstream' have identical effects.

Simplified model

$$\frac{ds}{dt} = \frac{N\alpha_T}{f_T} y(t) - (\mu + \delta_s)s - k_{\text{on}}sm + k_{\text{off}}s : m$$

$$\frac{dm}{dt} = \frac{N\alpha_L}{f_L} x(t) - (\mu + \delta_m)m - k_{\text{on}}sm + k_{\text{off}}s : m$$

$$\frac{ds : m}{dt} = k_{\text{on}}sm - (k_{\text{off}} + k_{\text{hyb}})s : m - (\mu + \delta_{sm})s : m$$

$$\frac{dc}{dt} = k_{\text{hyb}}s : m - (\mu + \delta_c)c$$

$$\frac{dp}{dt} = Fy(t) - (\gamma + \mu + \delta_g)p - \frac{v_z p}{K_z + p + g}$$

$$\frac{dg}{dt} = \gamma p - (\mu + \delta_g)g - \frac{v_z g}{K_z + p + g}$$

$$z = z_0 + \frac{g}{\Theta}$$

Simplified model

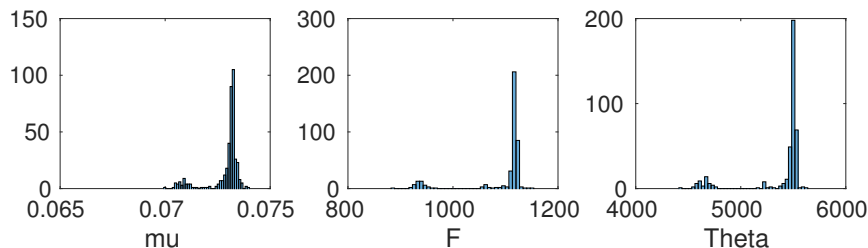
$$\frac{dp}{dt} = \textcolor{red}{F}y(t) - (\gamma + \textcolor{red}{\mu} + \delta_g)p - \frac{v_z p}{K_z + p + g} \quad (8)$$

$$\frac{dg}{dt} = \gamma p - (\mu + \delta_g)g - \frac{v_z g}{K_z + p + g} \quad (9)$$

$$z = z_0 + \frac{g}{\textcolor{red}{\Theta}} \quad (10)$$

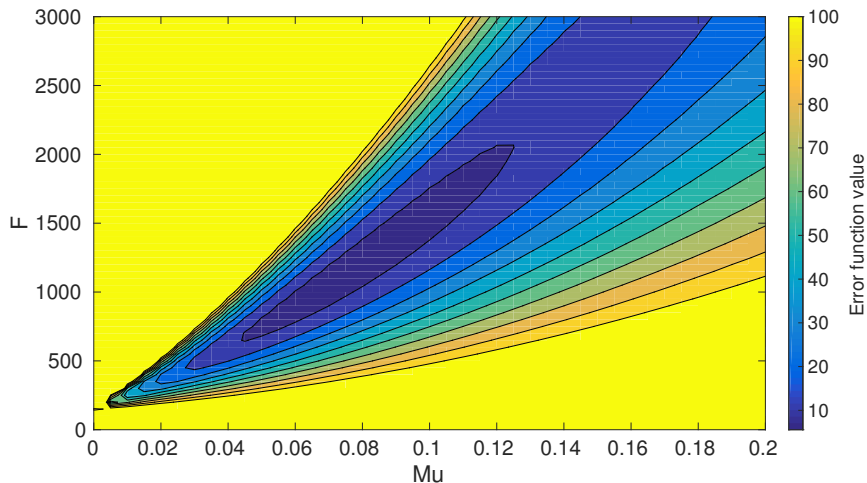
- ▶ Model only rate limiting steps - force translation term directly.
- ▶ 3 unknown parameters

Simplified model results (1)



- ▶ Error values as low as full model
- ▶ Clearer parameter estimation results
- ▶ Two peak structure due to local minima

Simplified model results (2)



- Error landscape, $\Theta = 5400$.

Future work

- ▶ Fluorescence data **not enough** to estimate all unknown parameters
- ▶ Further experiments needed - observe fast timescale.
- ▶ Investigate methods of McAuley et al.
- ▶ Change methodology - Bayesian Methods?

Acknowledgements

- ▶ Manish Kushawaha, Jaramillo Lab - School of Life Sciences
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