

Stochastic self-regulated gene expression model

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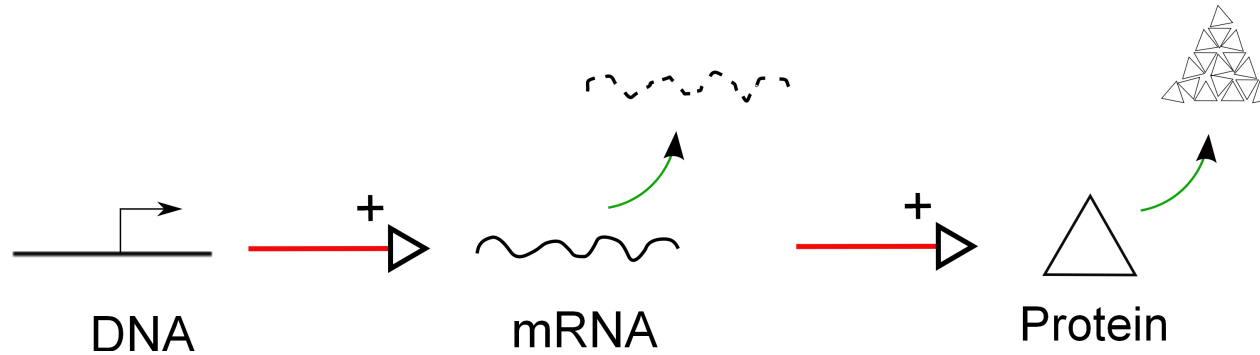
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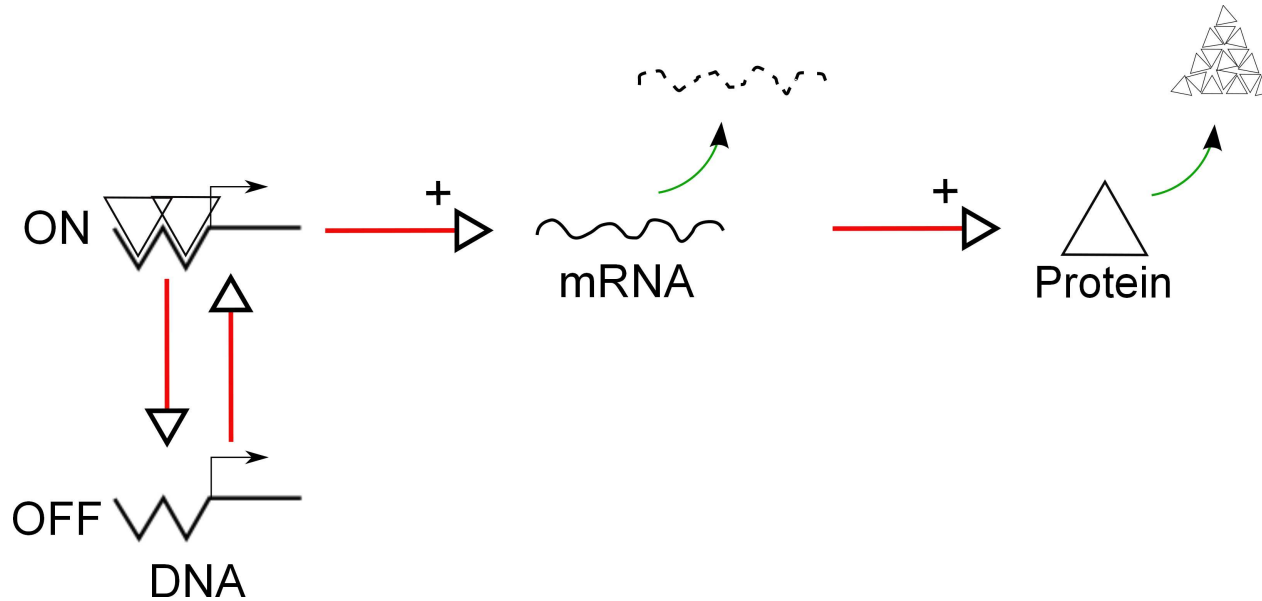
Outline

- Quick review of the standard stochastic models of gene expression
- Can we explain the **Bursting phenomena** through an adiabatic reduction of the standard model?

Central dogma

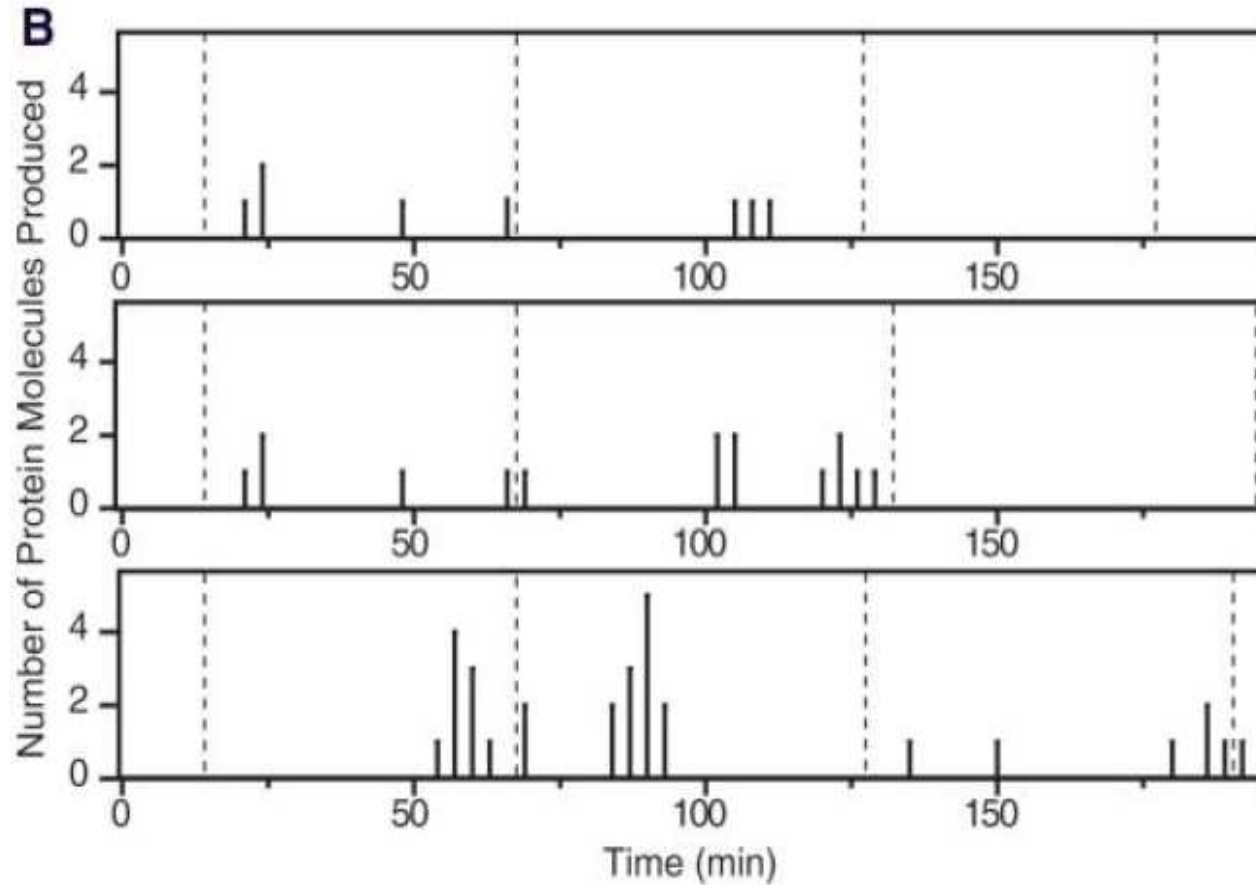


Central dogma



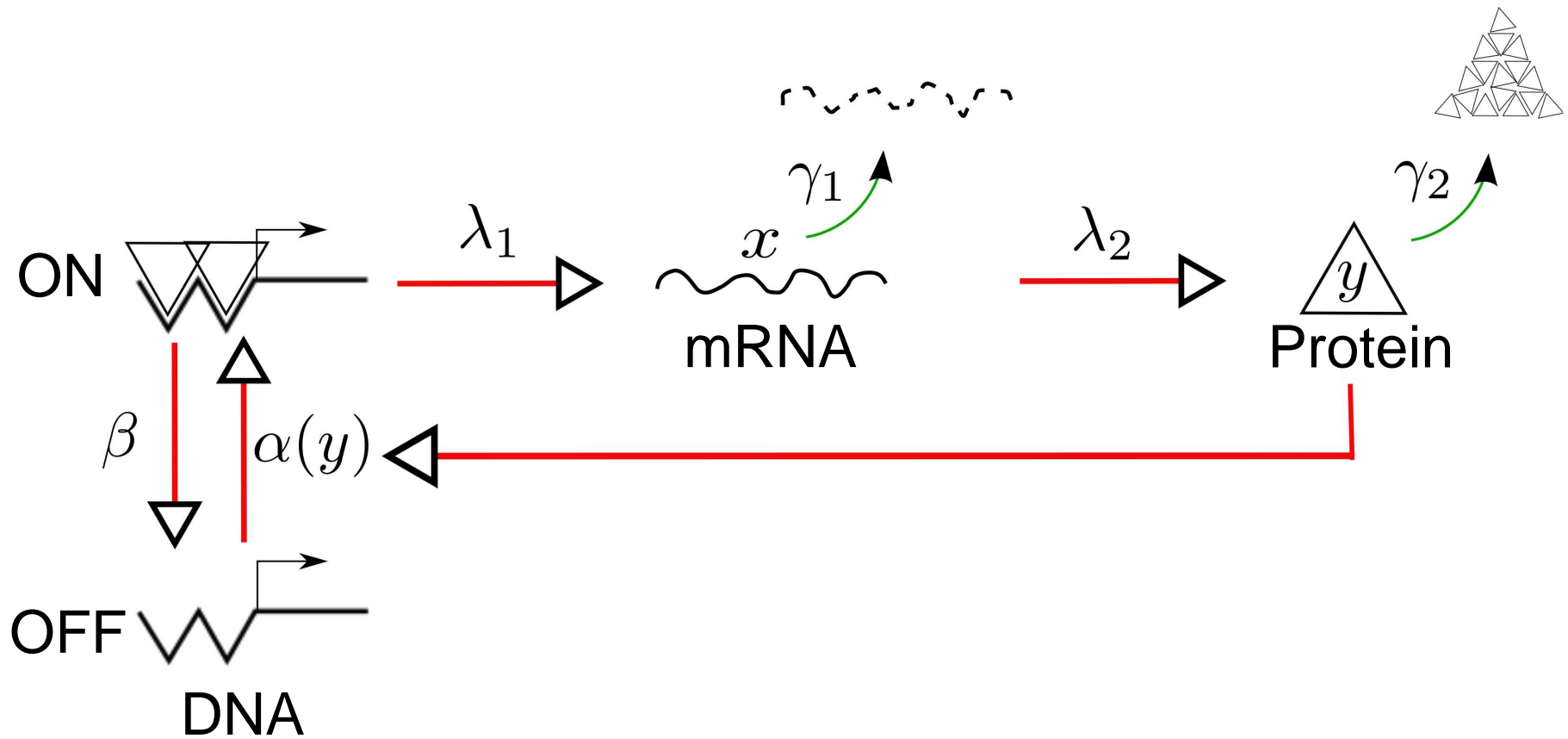
Studied either in a discrete state-space (Swain and Shahrezaei 08, Innocentini and Hornos 06) or in a continuous state-space (Lipniacki and Paszek 06, Mackey and Tyran and Y 2010)

The bursting phenomena

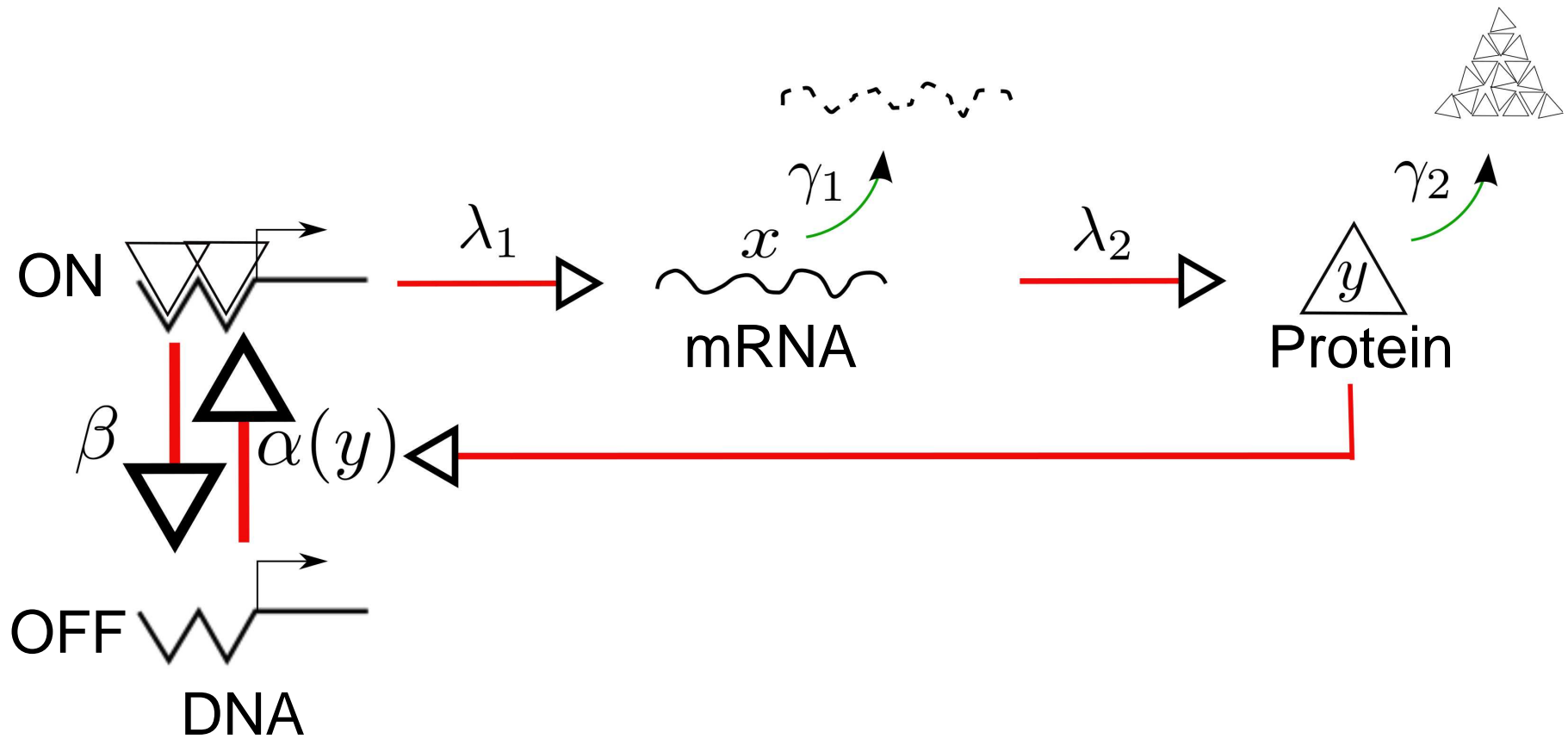


Yu et al. 06

Self-regulated gene

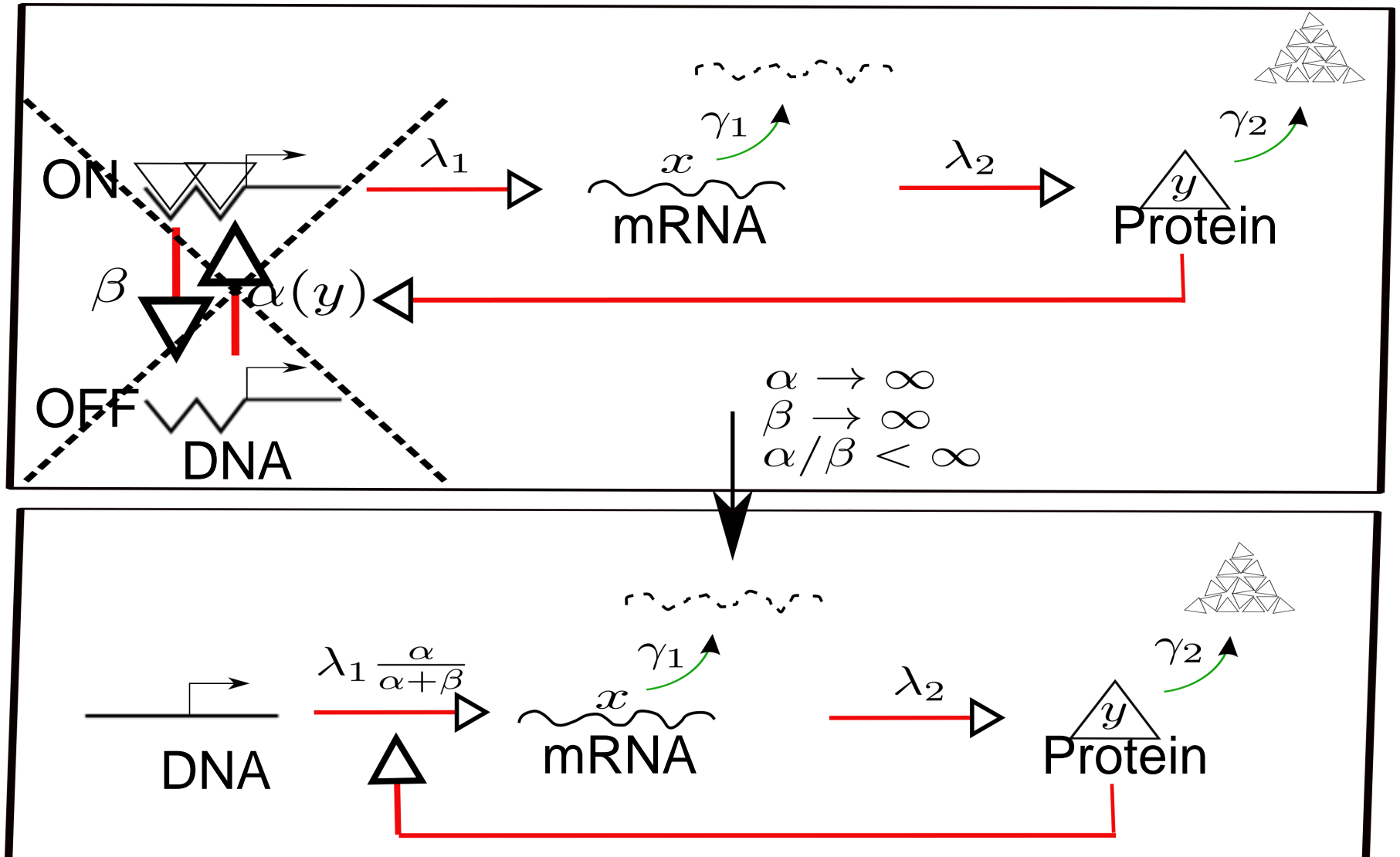


Reduction 1

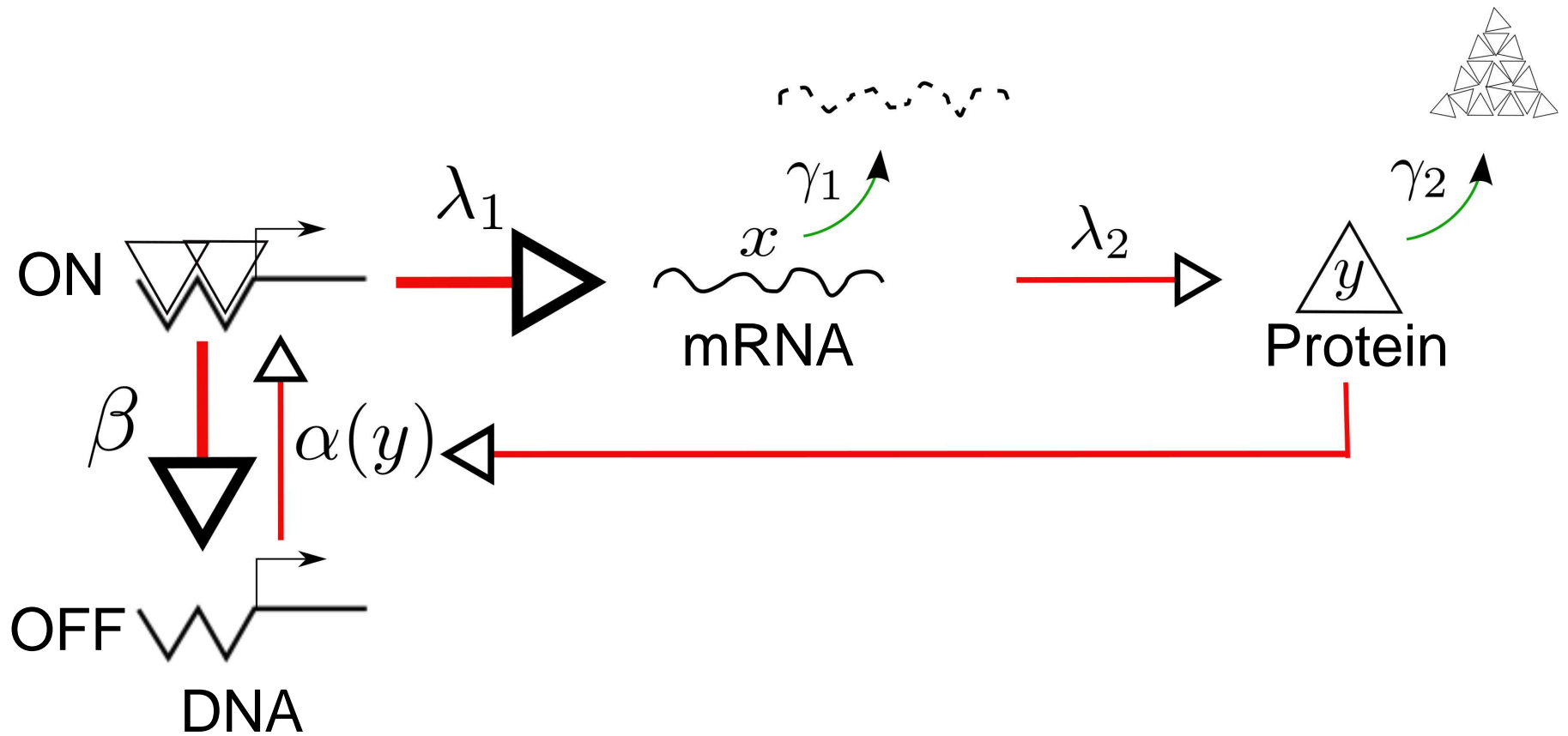


Bobrowski 06: Degenerate convergence of Semigroups.

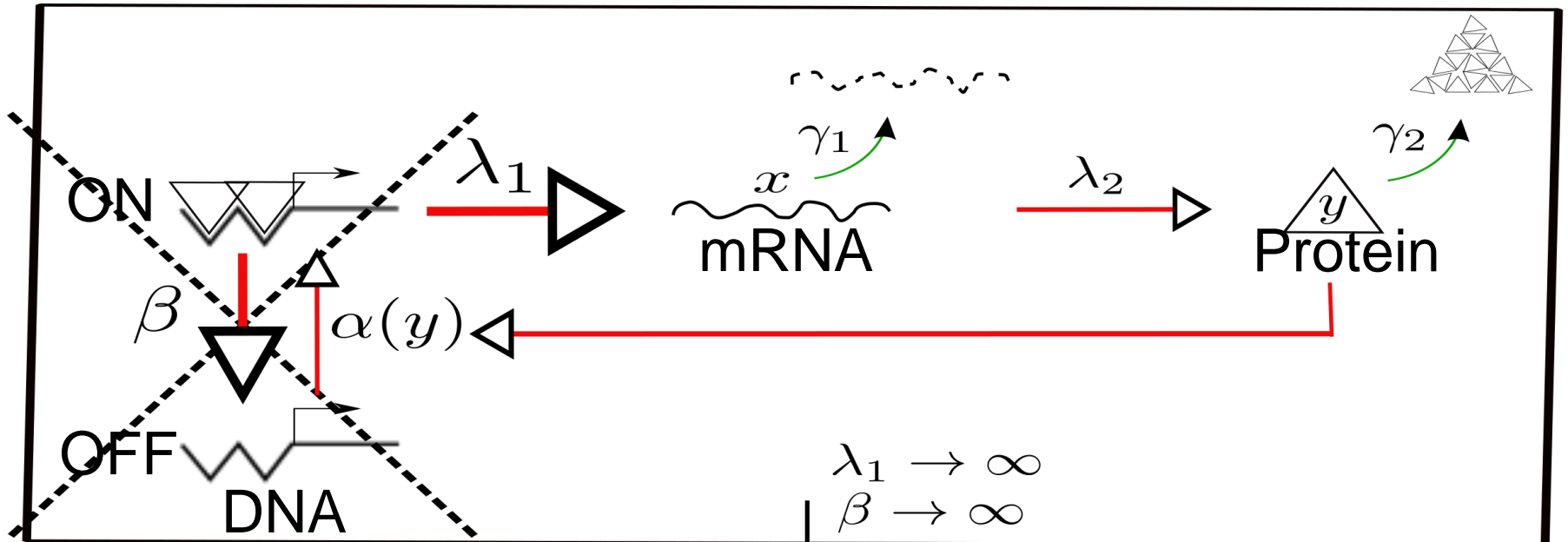
Reduction 1



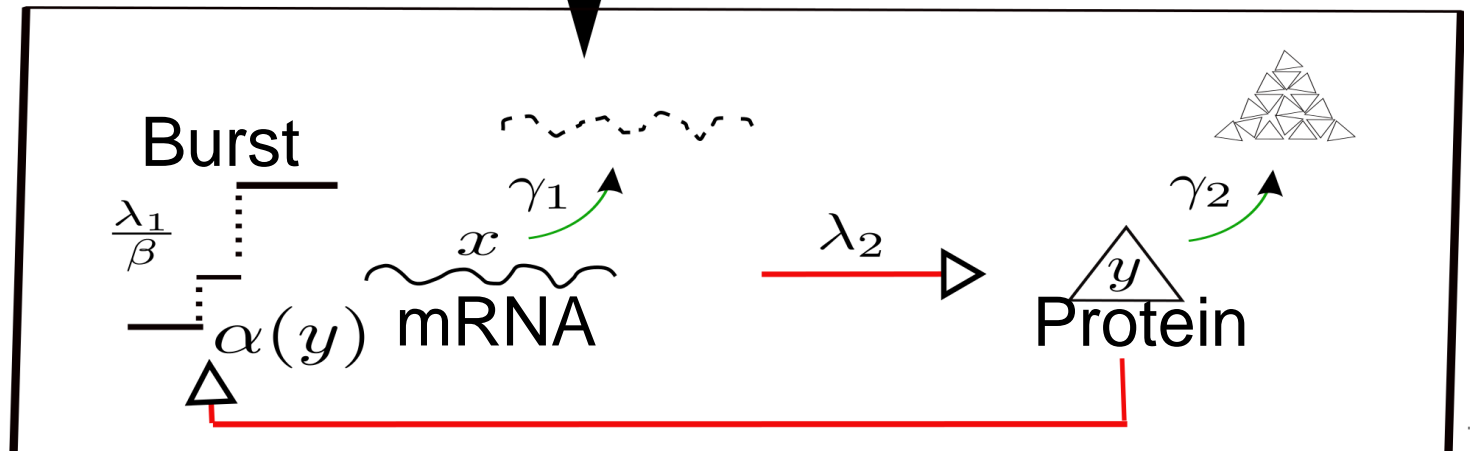
Reduction 2



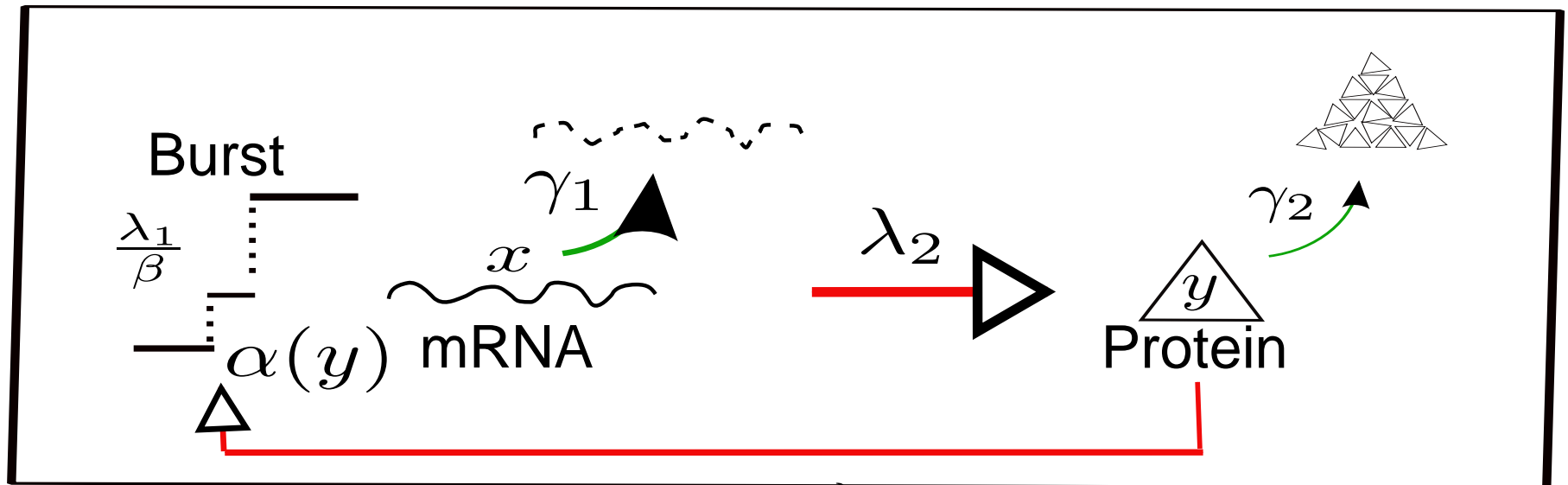
Reduction 2



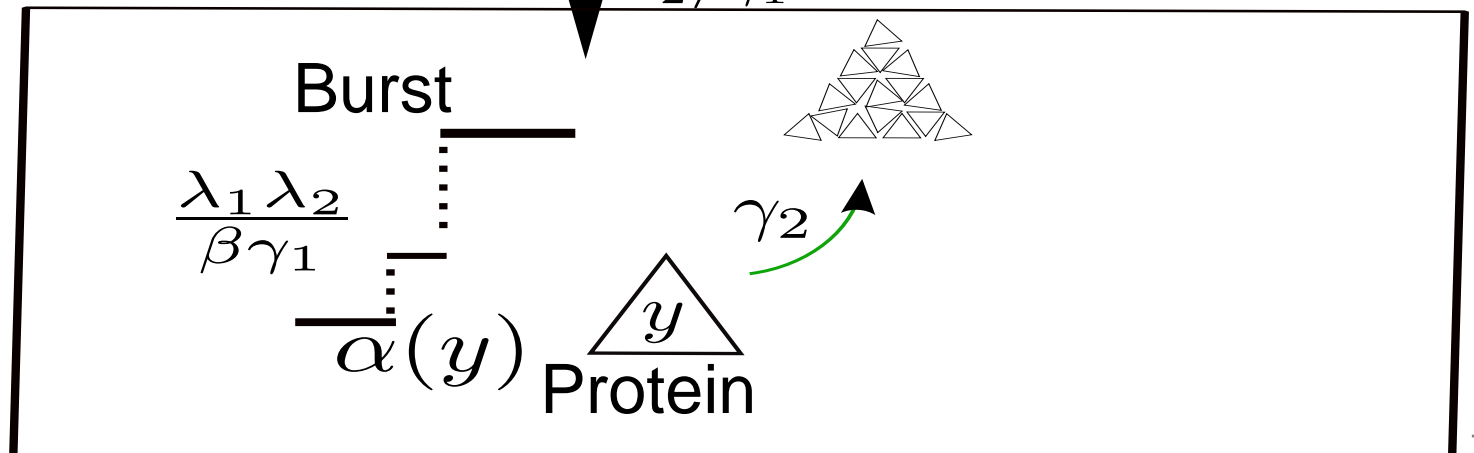
$$\begin{aligned} \lambda_1 &\rightarrow \infty \\ \beta &\rightarrow \infty \\ \lambda_1 / \beta &< \infty \end{aligned}$$



Reduction 3



$$\begin{aligned} \lambda_2 &\rightarrow \infty \\ \gamma_1 &\rightarrow \infty \\ \lambda_2 / \gamma_1 &< \infty \end{aligned}$$



Sketch of the proof

Initial process

$$(1) \quad \frac{dX}{dt} = \xi - \gamma X$$

where ξ is a dichotomous random process, which takes the values 0 or $\lambda > 0$ and switches at rate $\alpha(X)$ and β .

Final process, when $\beta \rightarrow \infty$, $\lambda \rightarrow \infty$, $\beta/\lambda \rightarrow b < \infty$

$$(2) \quad dX = dN(\alpha(X), h) - \gamma X$$

where N is a compound poisson process, of intensity $\alpha(X)$ and jump h exponentially distributed of mean b .

Semi-group proof

Let's define the semigroup $T_t f(x) = \mathbb{E}_x f(X_t)$ then the Dynkin's formula reads $T_t f(x) - f(x) = \int_0^t T_s \mathcal{A}f(X_s) ds$. For any test function f in the domain of A

$$\begin{aligned} \frac{d}{dt} \int_0^\infty p_t^0(x) f^0(x) + p_t^1(x) f^1(x) dx &= -\gamma \int_0^\infty x p_t^0(x) \frac{df^0}{dx} dx \\ - \int_0^\infty (\gamma x - \lambda) p_t^1(x) \frac{df^1}{dx} dx &+ \int_0^\infty (\alpha(x) p_t^0(x) - \beta(x) p_t^1(x)) (f^1(x) - f^0(x)) dx \end{aligned}$$

For any test function such that $f^0(x) = f^1(x) = f(x)$,

$$\begin{aligned} \frac{d}{dt} \int_0^\infty (p_t^0(x) + p_t^1(x)) f(x) dx &= - \gamma \int_0^\infty x (p_t^0(x) + p_t^1(x)) \frac{df}{dx} dx \\ &+ \lambda \int_0^\infty p_t^1(x) \frac{df}{dx} dx \end{aligned}$$

A particular choice for a test function

For any test function such that $f^0(x) = 0$, $f^1(x) = g(x)$,

$$\begin{aligned} \frac{d}{dt} \int_0^\infty p_t^1(x) g(x) dx &= -\gamma \int_0^\infty x p_t^1(x) \frac{dg}{dx} dx + \lambda \int_0^\infty p_t^1(x) \frac{dg}{dx} dx \\ &+ \int_0^\infty \alpha(x) p_t^0(x) g(x) dx - \beta \int_0^\infty p_t^1(x) g(x) dx \end{aligned}$$

One can perform a ***quasi steady-state approximation***, which gives

$$\begin{aligned} \lambda \int_0^\infty p_t^1(x) g(x) dx &= -\frac{\gamma\lambda}{\beta} \int_0^\infty x p_t^1(x) \frac{dg}{dx} dx + \frac{\lambda}{\beta} \int_0^\infty \alpha(x) p_t^0(x) g(x) dx \\ &+ \frac{\lambda^2}{\beta} \int_0^\infty p_t^1(x) \frac{dg}{dx} dx \end{aligned}$$

Iterating

Iterating the process, one can find,

$$\begin{aligned} \int_0^\infty \lambda p_t^1(x) \frac{df}{dx} dx &= \sum_{i \geq 1} \left(\frac{\lambda}{\beta} \right)^i \int_0^\infty \alpha(x) (p_t^0(x) + p_t^1(x)) \frac{d^i f}{dx^i} \\ &- \sum_{i \geq 1} \left(\frac{\lambda}{\beta} \right)^i \int_0^\infty \gamma x p_t^1(x) \frac{d^i f}{dx^i} - \sum_{i \geq 1} \left(\frac{\lambda}{\beta} \right)^i \int_0^\infty \alpha(x) p_t^1(x) \frac{d^i f}{dx^i} \end{aligned}$$

The first sum give the jump kernel

$$\int_0^\infty \alpha(x) (p_t^0(x) + p_t^1(x)) \left(\int_x^\infty h(y-x) f(y) dy - f(x) \right) dx,$$

where h is a distribution whose moments are given by

$$\mathcal{E}^i[h] = i! \left(\frac{\lambda}{\beta} \right)^i$$

The two others sum are shown to be arbitrary small. \square

summary

- The same is working in a discrete formalism
- Powerful tools to perform adiabatic reduction in Hybrid systems
- The "full-model" can now be analysed according its different limit behaviour.
- Other scaling are to be investigated

Thank you for your attention!