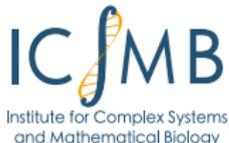


Role of the particle's stepping cycle in a TASEP: a model of mRNA translation

Luca Ciandrini

Institute for Complex Systems and Mathematical Biology,
University of Aberdeen

Edinburgh, 9th December 2011

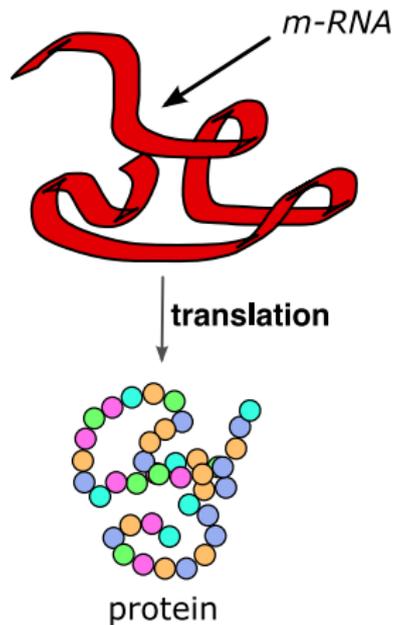
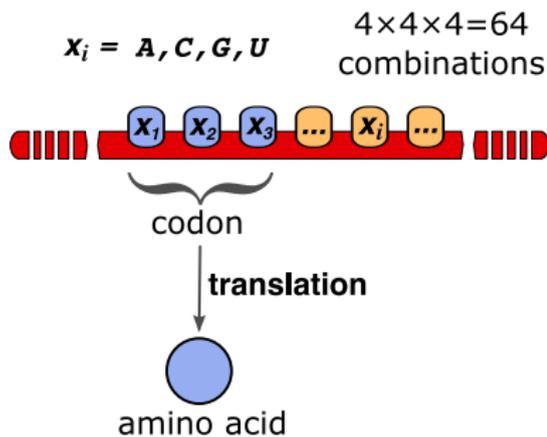


- 1 Introduction: mRNA translation for dummies
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- 5 Conclusions and further developments

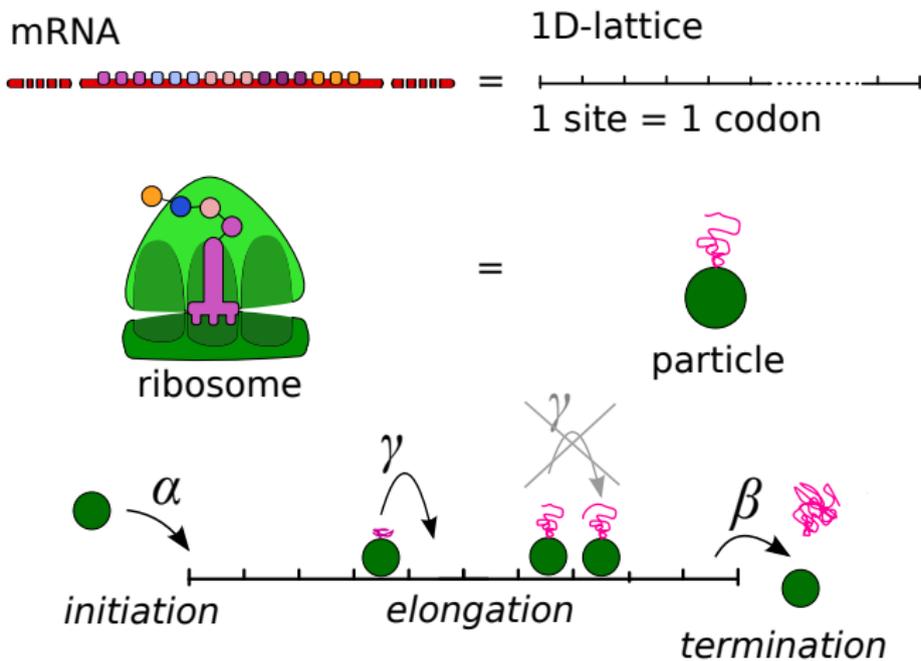
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The central dogma of molecular biology

DNA $\xrightarrow{\text{transcription}}$ mRNA $\xrightarrow{\text{mRNA translation}}$ proteins

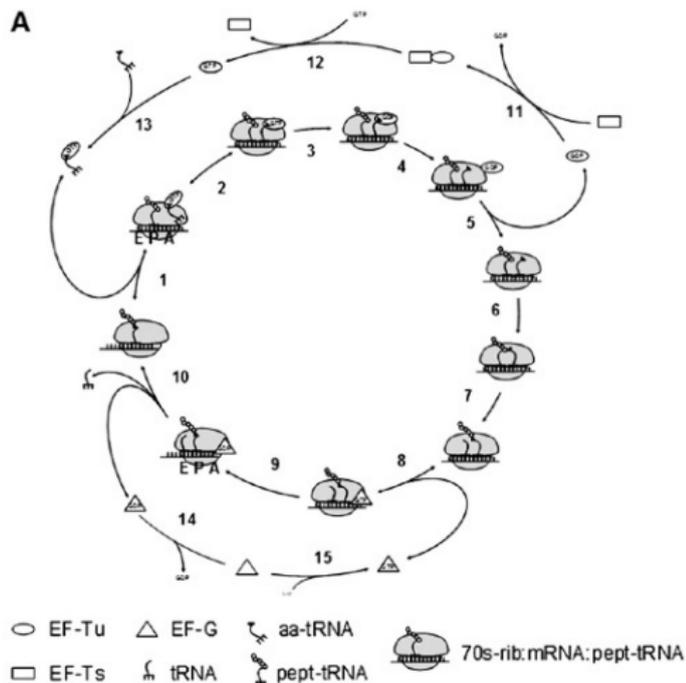


TASEP has been introduced to mimic mRNA translation



C. T. MacDonald, J. H. Gibbs, A. C. Pipkin. *Biopolymers*, 6(1):1-5, 1968

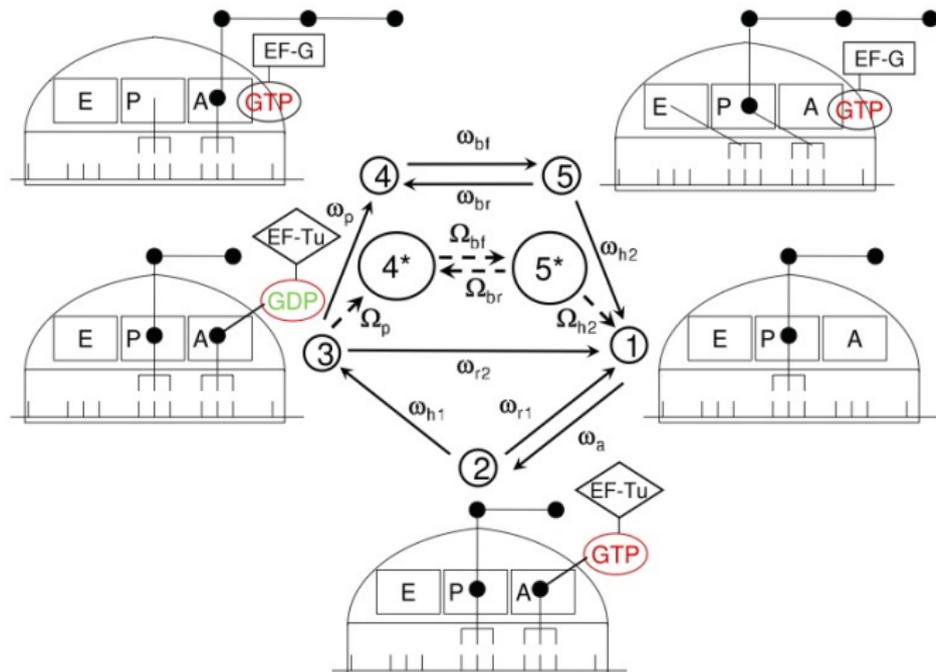
The ribosome's stepping cycle can be rather complicated...



Zouridis H, Hatzimanikatis V, Biophys J. 2007;92(3).

A model for protein translation: polysome self-organization leads to maximum protein synthesis rates.

The ribosome's stepping cycle can be rather complicated...

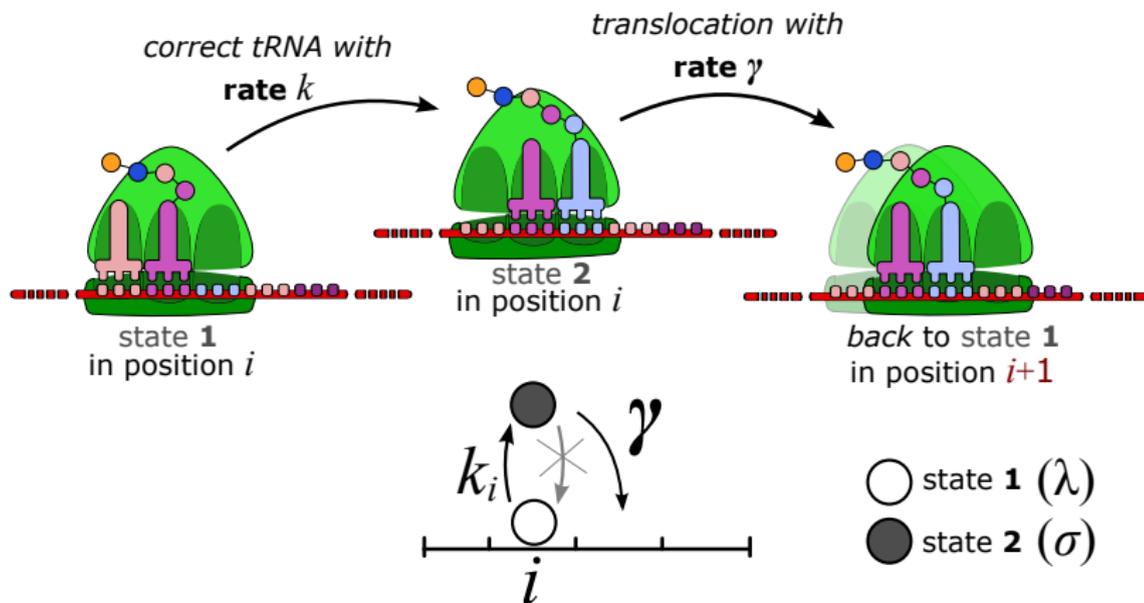


Sharma AK, Chowdhury D, Phys. Biol. 2011;8(2).

Distribution of dwell times of a ribosome: effects of infidelity, kinetic proofreading and ribosome crowding.

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Two-state model



S. Klumpp, Y. Chai, and R. Lipowsky, *Phys. Rev. E*, 78:041909, 2008

L. Ciandrini, I. Stansfield, and M. C. Romano, *Phys. Rev. E*, 81:051904, 2010

Occupation numbers and densities

The occupation number of site i is $n_i = 0, 1, 2$ (\square , \circ , \bullet).

The dynamical rules can be written as

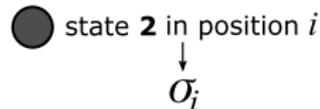
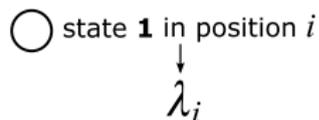


And we introduce the densities:

$$l_i = n_i(2 - n_i) \qquad s_i = \frac{n_i(n_i - 1)}{2}$$

$$\lambda_i = \langle l_i \rangle \qquad \sigma_i = \langle s_i \rangle$$

$$\rho_i = \lambda_i + \sigma_i .$$



Mean-field (MF) equations

○ state **1** in position i
↓
 λ_i

● state **2** in position i
↓
 σ_i

The flow equations (MF) for the densities are:

$$\begin{cases} \frac{d\lambda_i}{dt} = \sigma_{i-1}(1 - \lambda_i - \sigma_i)\gamma - k\lambda_i \\ \frac{d\sigma_i}{dt} = k\lambda_i - \sigma_i(1 - \lambda_{i+1} - \sigma_{i+1})\gamma \end{cases}$$

+ **steady state** ($\frac{d\lambda_i}{dt} = \frac{d\sigma_i}{dt} = 0$).

The incoming and outgoing currents at site i are:

$$J_+^i = \sigma_{i-1}(1 - \lambda_i - \sigma_i)\gamma, \quad J_-^i = \sigma_i(1 - \lambda_{i+1} - \sigma_{i+1})\gamma.$$

Our mean-field assumes $\langle s_i s_j \rangle \simeq \langle s_i \rangle \langle s_j \rangle$ and $\langle l_i s_j \rangle \simeq \langle l_i \rangle \langle s_j \rangle$, which is different than simply $\langle n_i n_j \rangle \simeq \langle n_i \rangle \langle n_j \rangle$

Periodic Boundary Conditions (PBC)

$$\begin{cases} \frac{d\lambda}{dt} = \sigma(1 - \lambda - \sigma)\gamma - k\lambda \\ \frac{d\sigma}{dt} = k\lambda - \sigma(1 - \lambda - \sigma)\gamma \end{cases}$$

↓

$$J = \sigma(1 - \lambda - \sigma)\gamma$$

$$\lambda = \frac{J}{k}$$

$$\sigma = \rho - \lambda = \rho - \frac{J}{k}$$

L. Ciandrini, I. Stansfield, and M. C. Romano, *Phys. Rev. E*, 81:051904, 2010

Periodic Boundary Conditions (PBC)

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$$\downarrow \\ J = \sigma(1 - \lambda - \sigma)\gamma$$

$$\lambda = \frac{J}{k}$$

$$\sigma = \rho - \lambda = \rho - \frac{J}{k}$$

$$w := k/\gamma$$

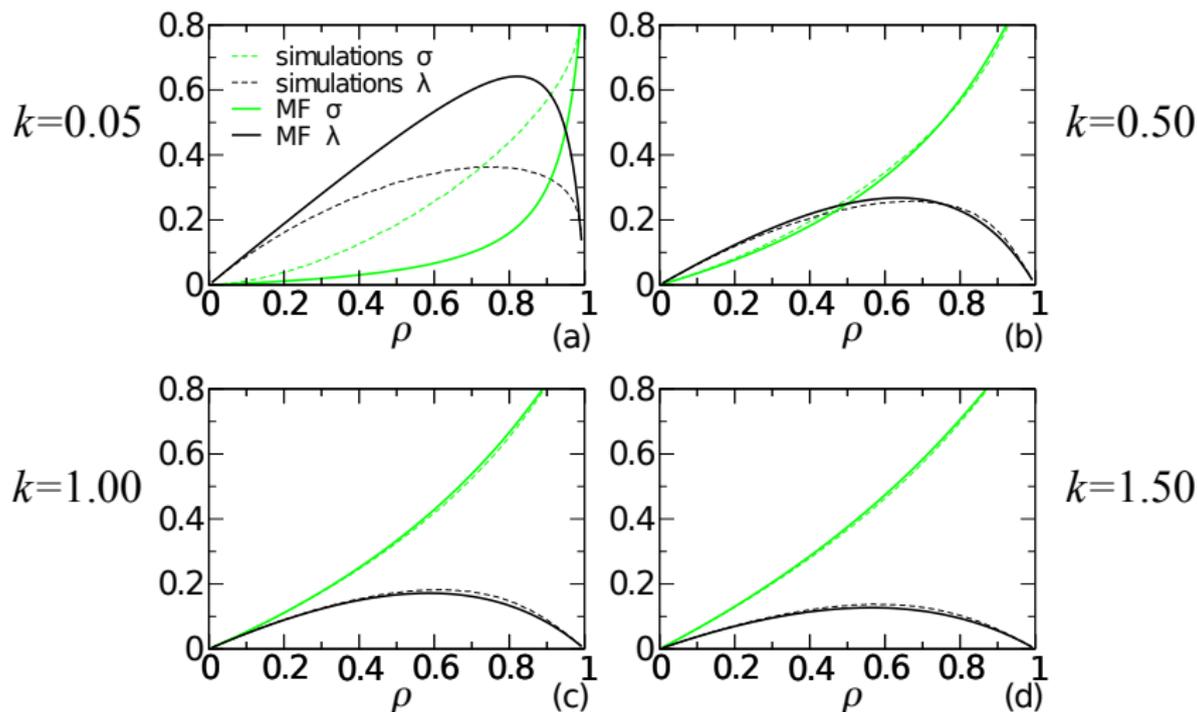
$$J = k \frac{\rho(1 - \rho)}{w + (1 - \rho)}$$

$$\lambda = \frac{\rho(1 - \rho)}{w + (1 - \rho)}$$

$$\sigma = w \left[\frac{\rho}{w + (1 - \rho)} \right]$$

Comparison between simulations (---) and MF (—)

For small k the MF underestimates transitions toward the state 2



The current is simply $J = k\lambda$.

Where the value of ρ for which the current is maximal is

$$\rho^* = 1 - \chi,$$

and the maximal value of the densities are

$$\lambda^* = 1 - 2\chi, \quad \sigma^* = \chi.$$

$$w := k/\gamma$$

$$\chi := w(\sqrt{1 + 1/w} - 1)$$

Open Boundary Conditions (OBC)

$$\begin{cases} \frac{d\lambda_i}{dt} = \sigma_{i-1}(1 - \lambda_i - \sigma_i)\gamma - k\lambda_i \\ \frac{d\sigma_i}{dt} = k\lambda_i - \sigma_i(1 - \lambda_{i+1} - \sigma_{i+1})\gamma \end{cases}$$

+ **boundary conditions:**

$$\begin{aligned} \frac{d\lambda_1}{dt} &= \alpha(1 - \lambda_1 - \sigma_1) - k\lambda_1 \\ \frac{d\sigma_L}{dt} &= k\lambda_L - \beta\sigma_L \end{aligned}$$

Open Boundary Conditions (OBC)

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we can write a recursive map for the density σ_i :

$$\rightarrow \sigma_{i+1} = 1 - J\left(\frac{1}{k} + \frac{1}{\gamma\sigma_i}\right)$$

The boundaries are substituted by reservoirs of particles and the dynamics between the reservoir and the lattice is assumed to be the same as in the bulk. J. Krug, *Phys. Rev. Lett.*, 67:1882, 1991

The *Maximal Current Principle* states that J in the MC regime is given by

$$J_{\text{MC}} = \max_{\rho \in [\rho_{L+1}, \rho_0]} J_{\text{PBC}}(\rho),$$

where ρ_0 and ρ_{L+1} are respectively the densities of the reservoirs of particles at the left and the right boundaries and are chosen to realize the injection and depletion parameters α and β .

$$\begin{aligned} \rho_0 &= \lambda_0 + \sigma_0, \\ \alpha &= \sigma_0 \gamma, & \beta &= \gamma(1 - \rho_{L+1}) \end{aligned}$$

The Maximal Current Principle locates the critical points

The critical points are obtained by equating σ_0 with σ^* and ρ_{L+1} with ρ^* . [$w := k/\gamma$, $\chi := w(\sqrt{1 + 1/w} - 1)$]

Low Density (LD) phase

$[\alpha < \beta, \alpha < \gamma\chi]$

$$\rho = \frac{\alpha(k + \gamma)}{\gamma(k + \alpha)}, \quad J = \alpha \frac{k(\gamma - \alpha)}{\gamma(k + \alpha)}$$

High Density (HD) phase

$[\beta < \alpha, \beta < \gamma\chi]$

$$\rho = 1 - \frac{\beta}{\gamma}, \quad J = \beta \frac{k(\gamma - \beta)}{\gamma(k + \beta)}$$

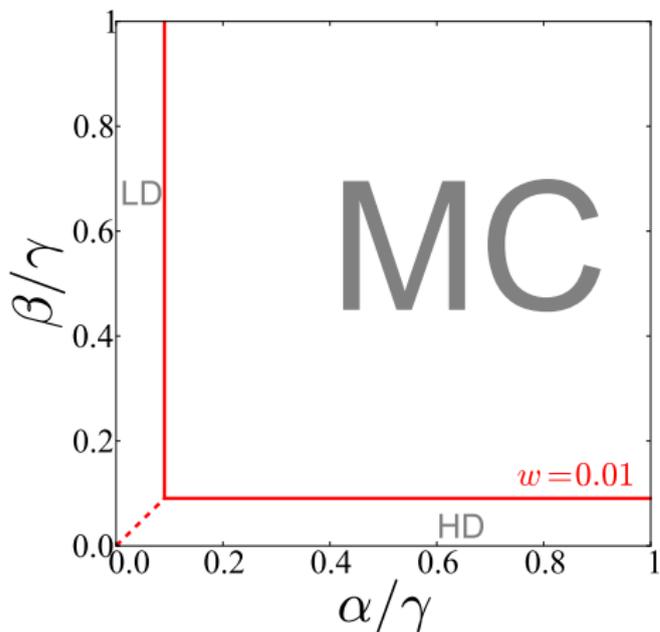
Maximal Current (MC) phase

$[\alpha, \beta \geq \gamma\chi]$

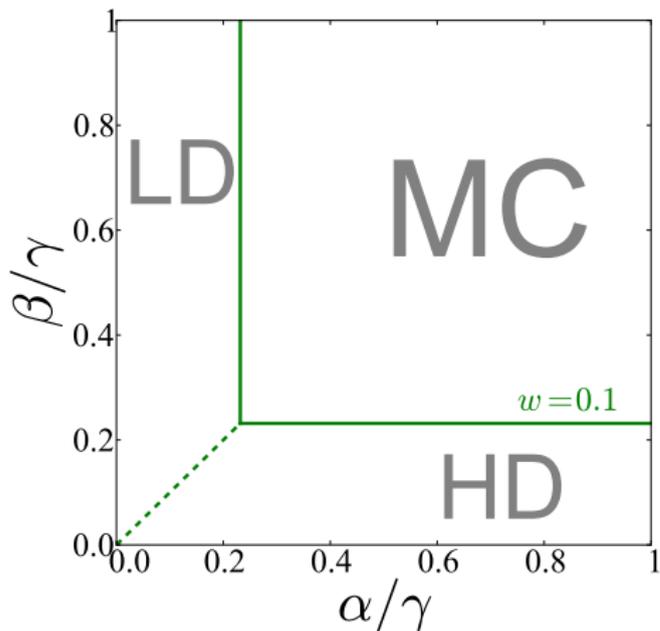
$$\rho = 1 - \chi, \quad J = k(1 - 2\chi)$$

$$[\sigma^* = \chi, \rho^* = 1 - \chi]$$

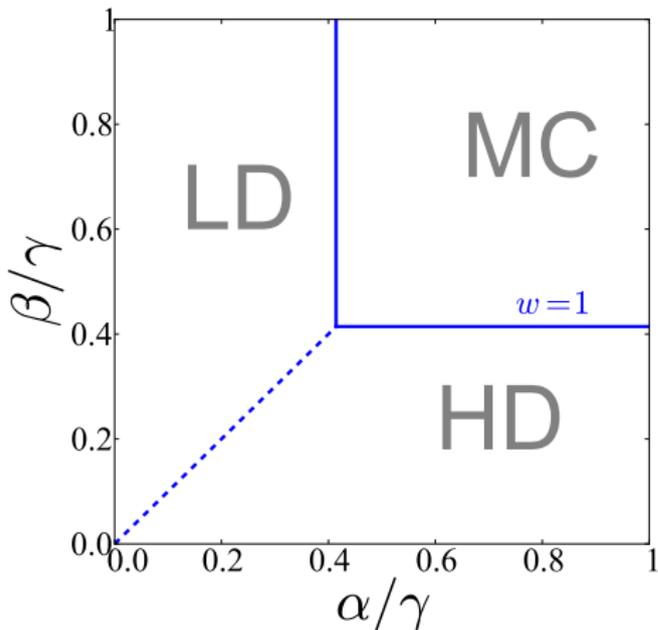
The critical points α_c and β_c depend on $w = k/\gamma$, but in the limit $w \rightarrow \infty$ (particles with only one state) we find the well known TASEP results.



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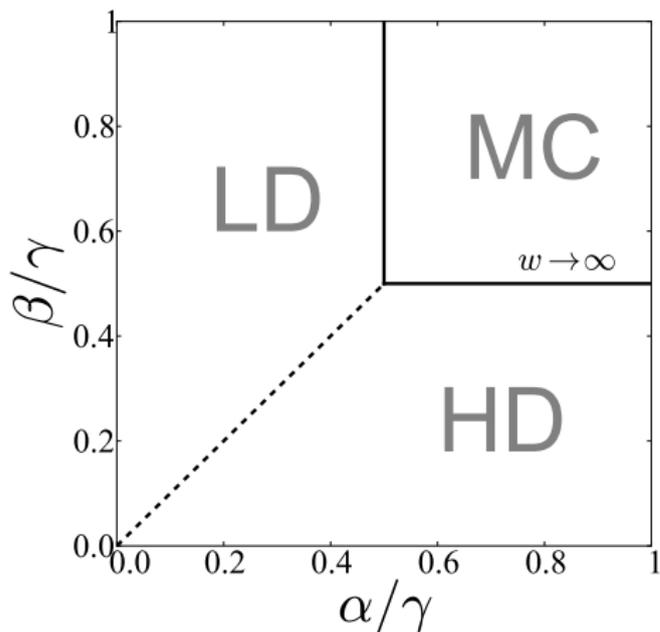


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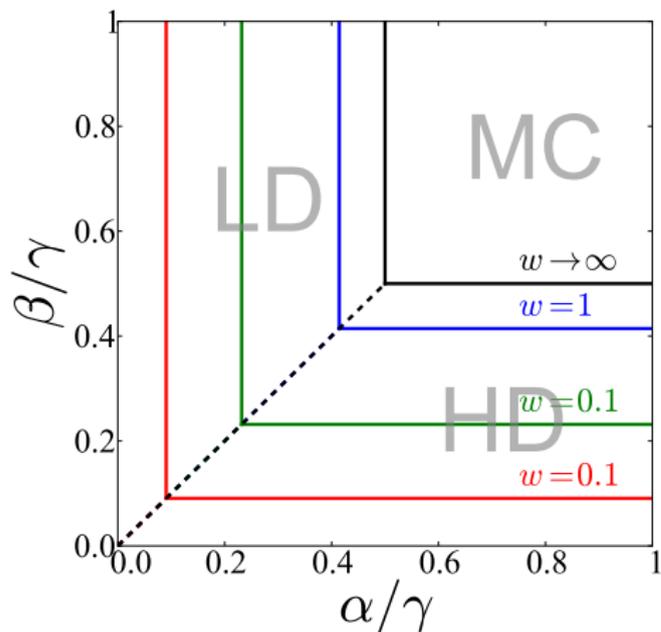
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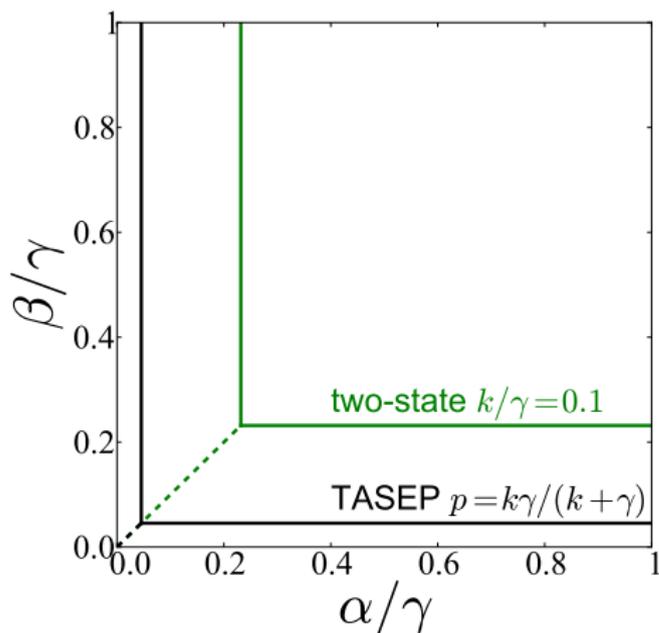


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The critical points α_c and β_c depend on $w = k/\gamma$, but in the limit $w \rightarrow \infty$ (particles with only one state) we find the well known TASEP results.



Two-state model is intrinsically different from the TASEP



TASEP with hopping rate
 $p = k\gamma/(k + \gamma)$, $k/\gamma = 0.1$

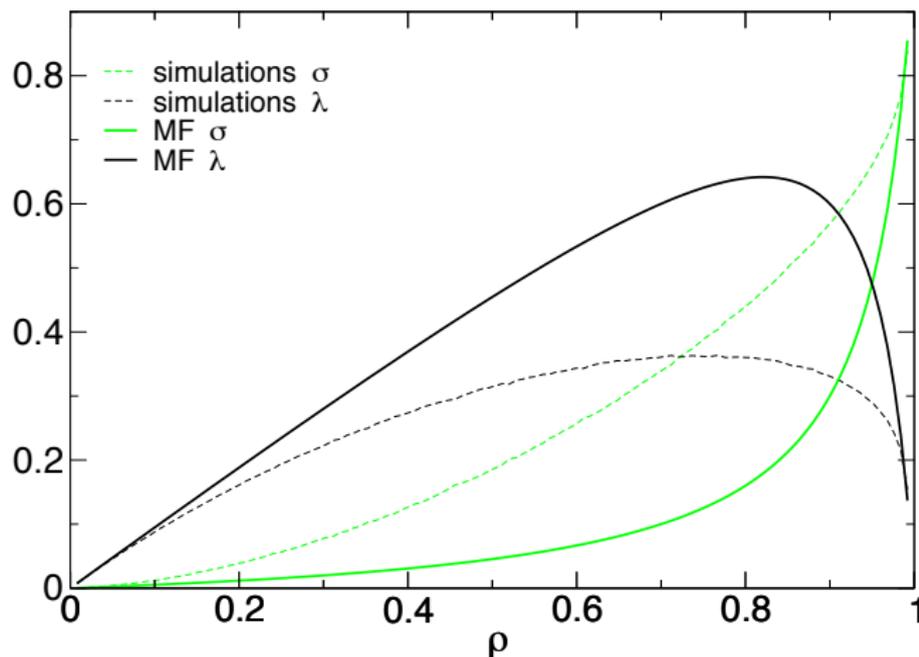
The two-state model cannot be mapped onto a TASEP with effective hopping rates.

Critical points and steady-state quantities depend on the internal dynamics, but there are dynamical effects too...

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Non-localised traffic jams

Starting point: deviation MF-theory for small k/γ



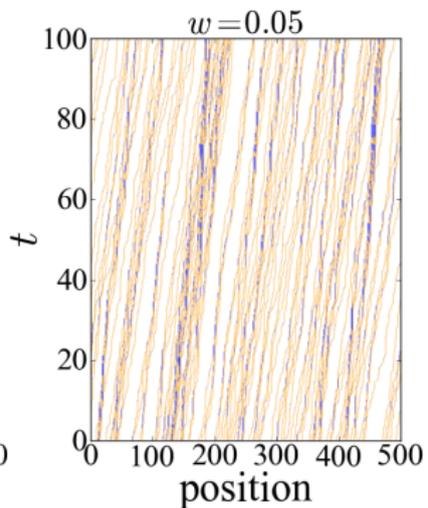
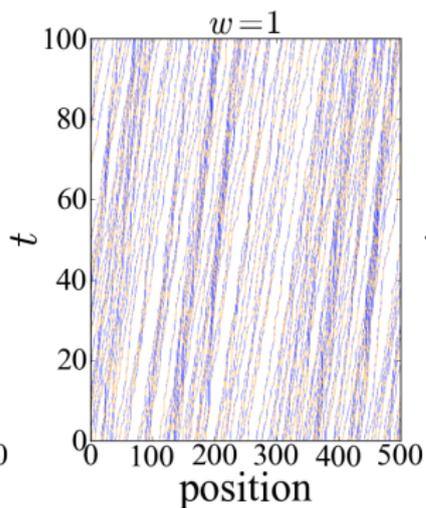
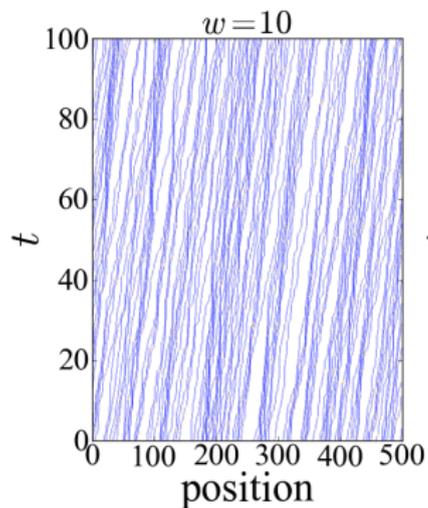
Work in progress...

(manuscript in preparation)

with M C Romano and A Parmeggiani (Université de Montpellier II)

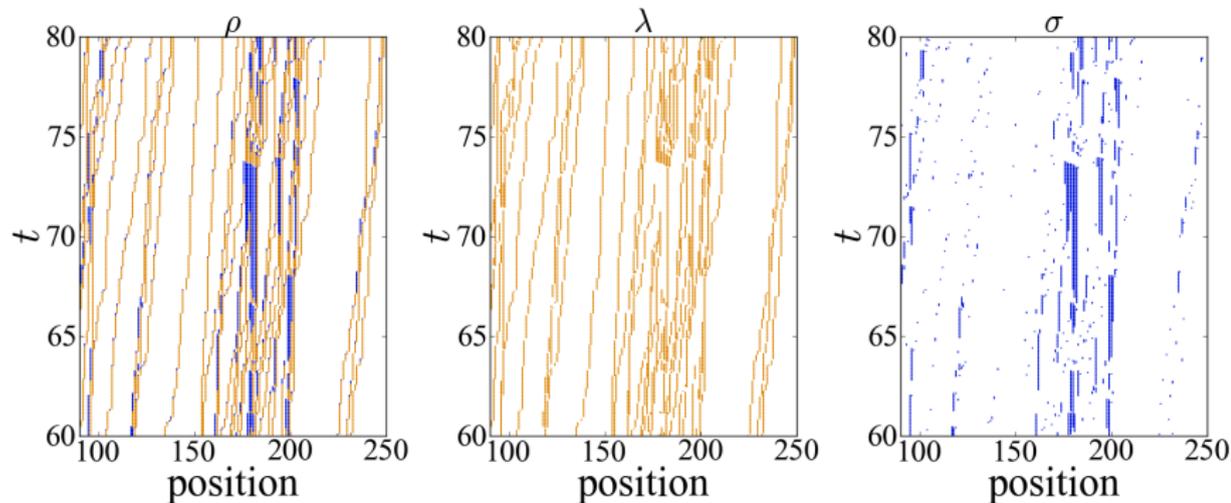
Kymographs of the two-state model

The kymographs show non-localised traffic jams...



Kymographs of the two-state model

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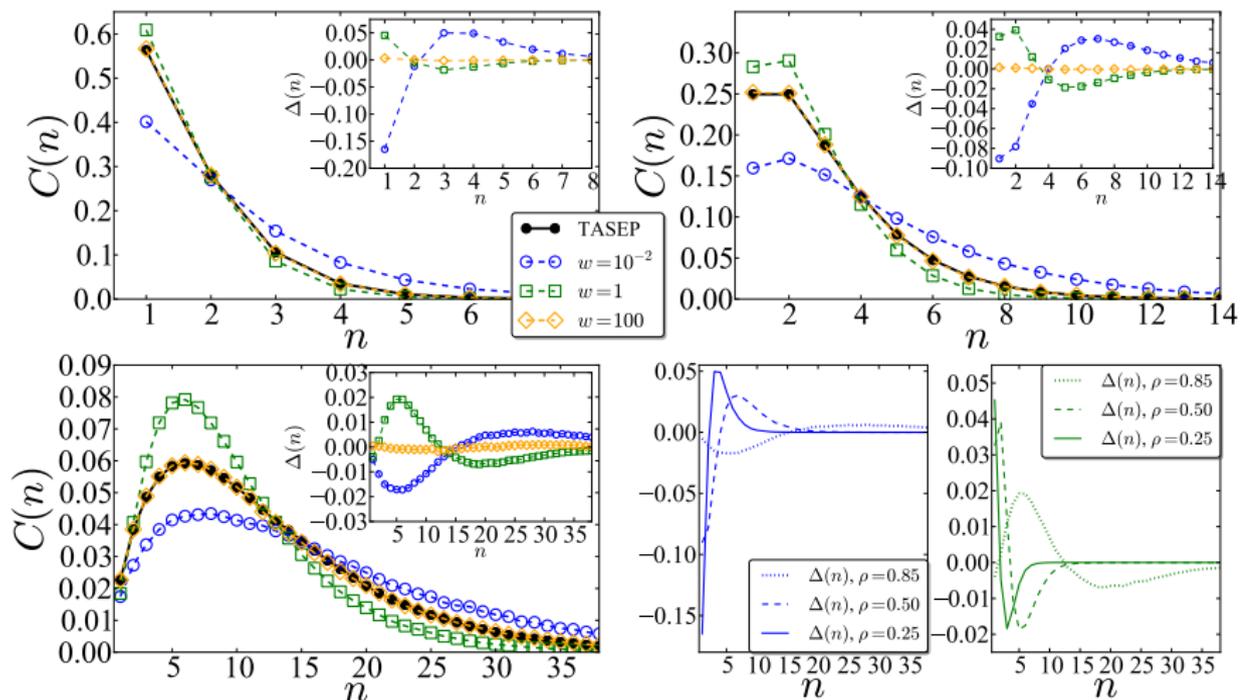


...and the presence of shocks in the density σ .

A more quantitative overview on the clustering

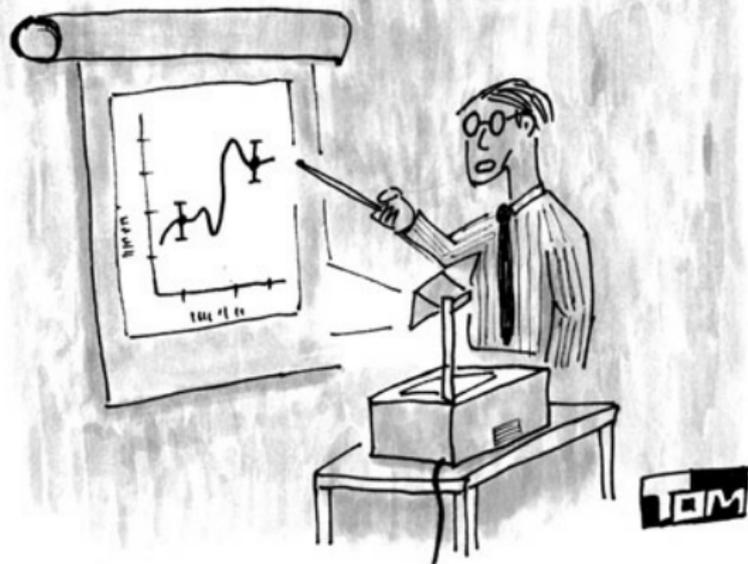
$C(n)$ = probability of finding a cluster of n particles

$\Delta(n)$ = difference between two-state model and TASEP



$(w = k/\gamma)$

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"WE CAN SEE HERE THAT THE
AGREEMENT WITH THEORY IS EXCELLENT"

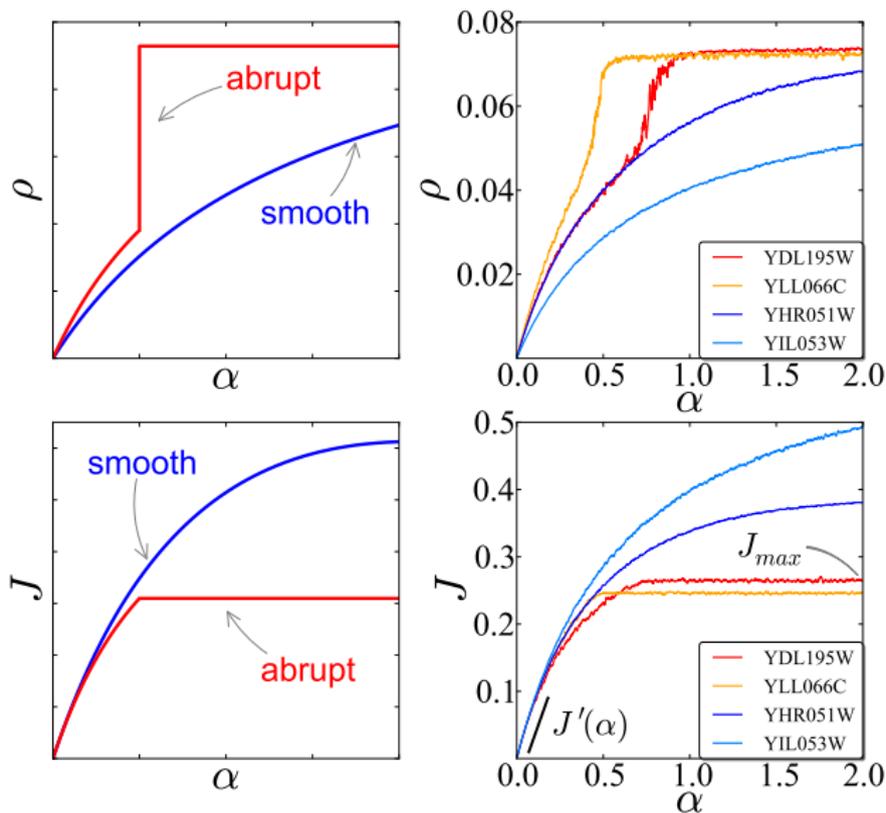
©2000 Tom Swanson

The final aim of this project is to predict the *transcriptome-proteome relation* and thereby to gain a deeper understanding of the physical and biological processes underlying translation.

We estimate all the necessary parameters ($\gamma, \{k_i\}, \beta$) and simulate the entire genome of *S.cerevisiae* (~ 6000 sequences). This allows us to:

- Classify the genes into two main types according to $\rho(\alpha)$ and $J(\alpha)$: significant correlation to biological functions.
- Estimate of the “operative” injection rates α for each mRNA (for the first time), which is experimentally not possible.

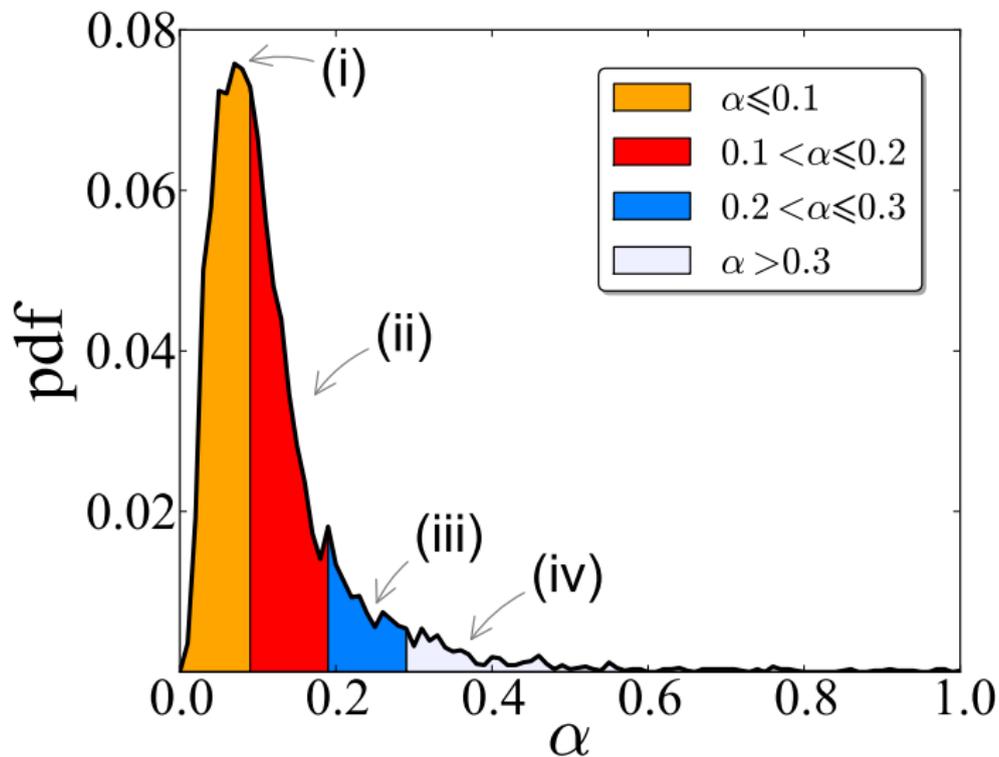
Classification of mRNA sequences in smooth/abrupt...



...and correlation with biological functions of encoded proteins.

L. Ciandrini, I. Stansfield, and M. C. Romano, (to be submitted)

Genome-wide estimates of α and biological functions



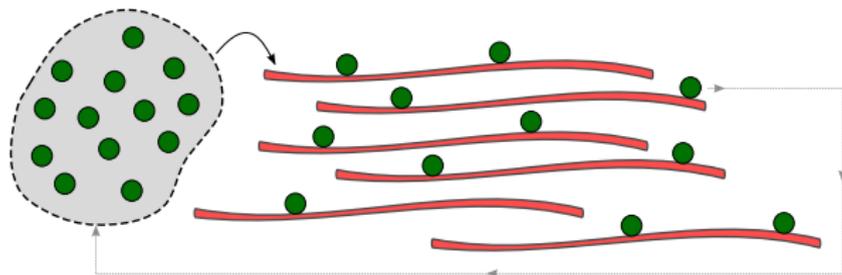
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To summarise, the two-state model condenses the particle's stepping cycle into two steps: internal transition (k) and translocation (γ).

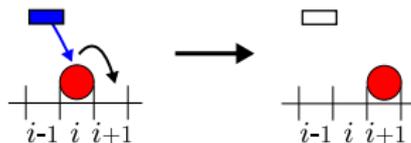
- The emerging dynamics substantially differs from the TASEP (e.g., no particle-hole symmetry, the phase diagram cannot be reproduced with effective rates, non-localised traffic jams);
- it recovers the TASEP in the limit $w \rightarrow \infty$ (but biologically $w \ll 1$);
- it allows direct application to genome-wide protein synthesis databases;
- the fundamental dynamics is also applicable to other biological processes (e.g., molecular motors).

■ Competition for resources:

- A mixed population of competing TASEPs with a shared reservoir of particles, P Greulich[†], L Ciandrini[†], R J Allen^{*}, M C Romano^{*}, *submitted* [arXiv:1111.1775].



- Multiple phase transitions in a system of exclusion processes with limited reservoirs of particles and fuel carriers, C A Brackley, L Ciandrini, M C Romano, *to be submitted*.



■ Folding TASEP

(*in preparation*, with F Turci, A Parmeggiani, E Pitard and M C Romano)

■ Modeling

- **M. C. Romano**

ICSMB and IMS, University of Aberdeen

- **R. J. Allen** and P. Greulich

Physics department, University of Edinburgh

- **A. Parmeggiani**

Biological Physics and Systems Biology Team, DIMNP, University of Montpellier II

- **F. Turci** and **E. Pitard**

Laboratoire Charles Coulomb, University of Montpellier II

- **C. A. Brackley**

ICSMB, University of Aberdeen

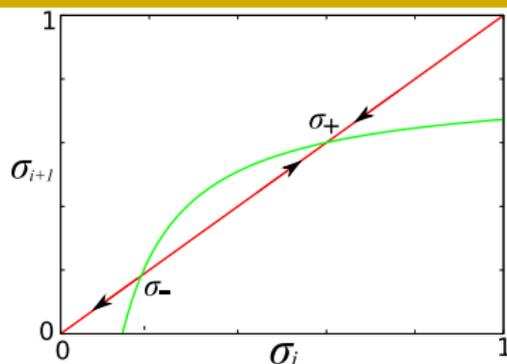
■ Experiments

- **I. Stansfield** and R. Betney

Institute of Medical Sciences, University of Aberdeen

Thank you!

We find the general term of the recursion: $\sigma_i = \sigma_i(\sigma_1, J)$



$$\sigma_{i+1} = 1 - J\left(\frac{1}{k} + \frac{1}{\gamma\sigma_i}\right)$$

which fixed points are

$$\sigma_{\pm} = \frac{1}{2} \left[\left(1 - \frac{J}{k}\right) \pm \sqrt{\left(1 - \frac{J}{k}\right)^2 - \frac{4J}{\gamma}} \right].$$

The general term is given by:

$$\sigma_i = \frac{-\sigma_- \sigma_+ (\sigma_+^{i-1} - \sigma_-^{i-1}) + \sigma_1 (\sigma_+^i - \sigma_-^i)}{-\sigma_- \sigma_+ (\sigma_+^{i-2} - \sigma_-^{i-2}) + \sigma_1 (\sigma_+^{i-1} - \sigma_-^{i-1})}.$$

Reasoning on the iterative map [B. Derrida, E. Domany, and D. Mukamel, *Journal of Statistical Physics*, 69(3-4):667–687, 1992] we are able to obtain the MF solutions

Low Density (LD) phase [$\alpha < \beta$, $\alpha \leq k(\sqrt{1 + \gamma/k} - 1)$]

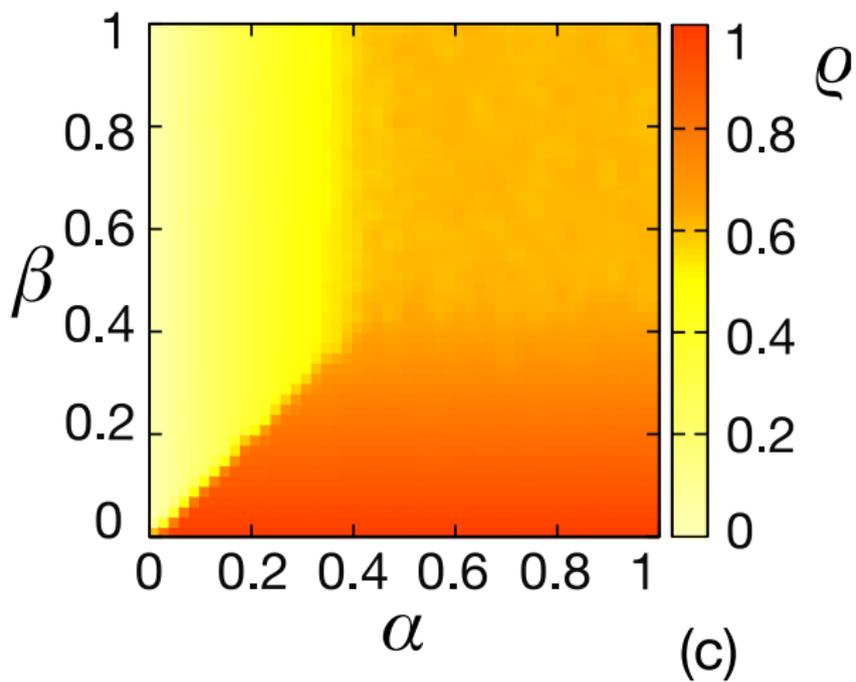
$$\rho = \frac{\alpha(k + \gamma)}{\gamma(k + \alpha)}, \quad J = \alpha \frac{k(\gamma - \alpha)}{\gamma(k + \alpha)}$$

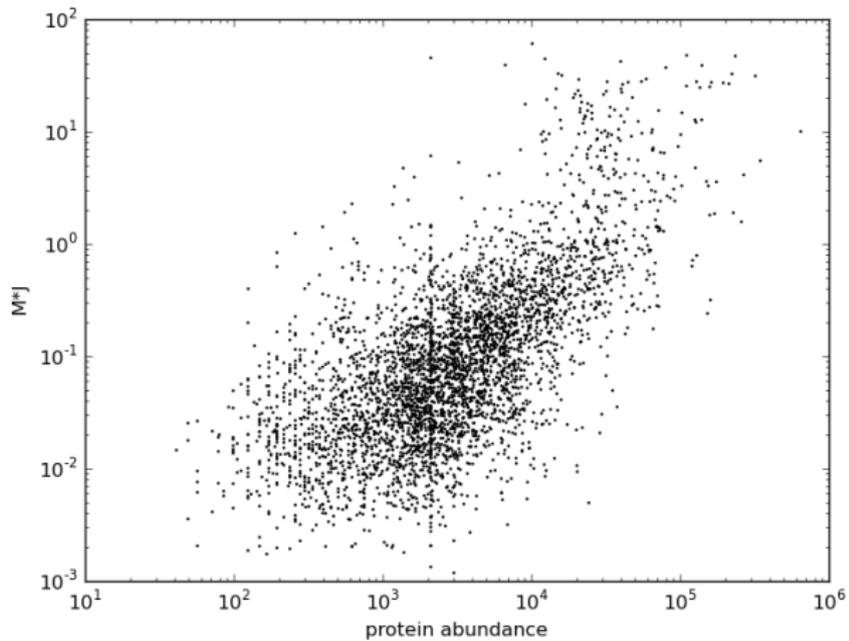
High Density (HD) phase [$\beta < \alpha$, $\beta \leq k(\sqrt{1 + \gamma/k} - 1)$]

$$\rho = 1 - \frac{\beta}{\gamma}, \quad J = \beta \frac{k(\gamma - \beta)}{\gamma(k + \beta)}$$

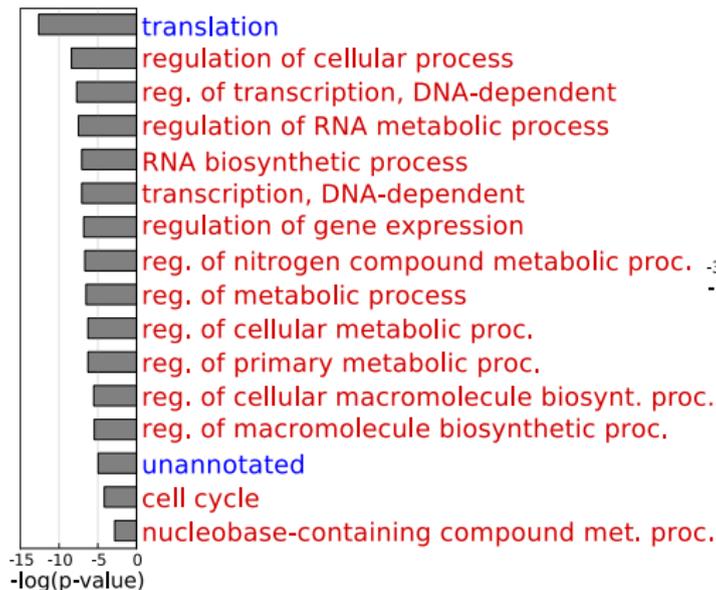
Maximal Current (MC) phase [$\alpha, \beta > k(\sqrt{1 + \gamma/k} - 1)$]

$$\rho = 1 - \frac{k}{\gamma} \left(\sqrt{1 + \gamma/k} - 1 \right), \quad J = k \left[1 - 2 \frac{k}{\gamma} \left(\sqrt{1 + \gamma/k} - 1 \right) \right]$$

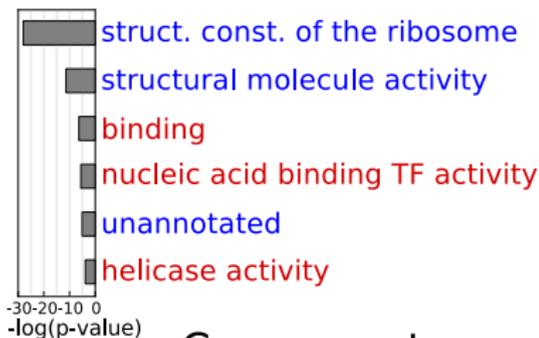




Process



Function



Component

