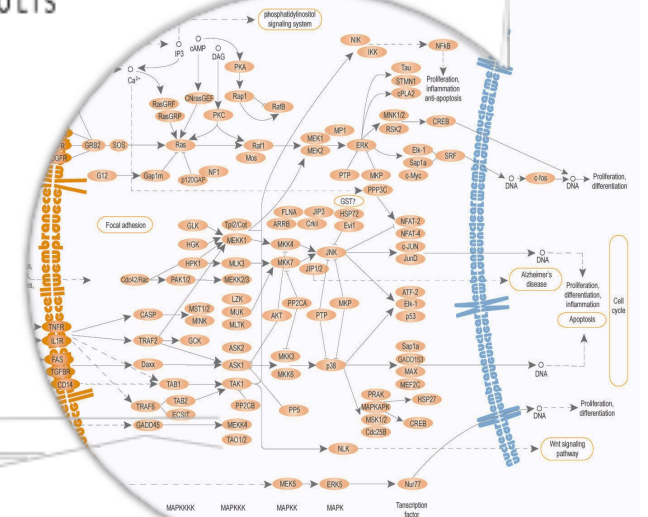


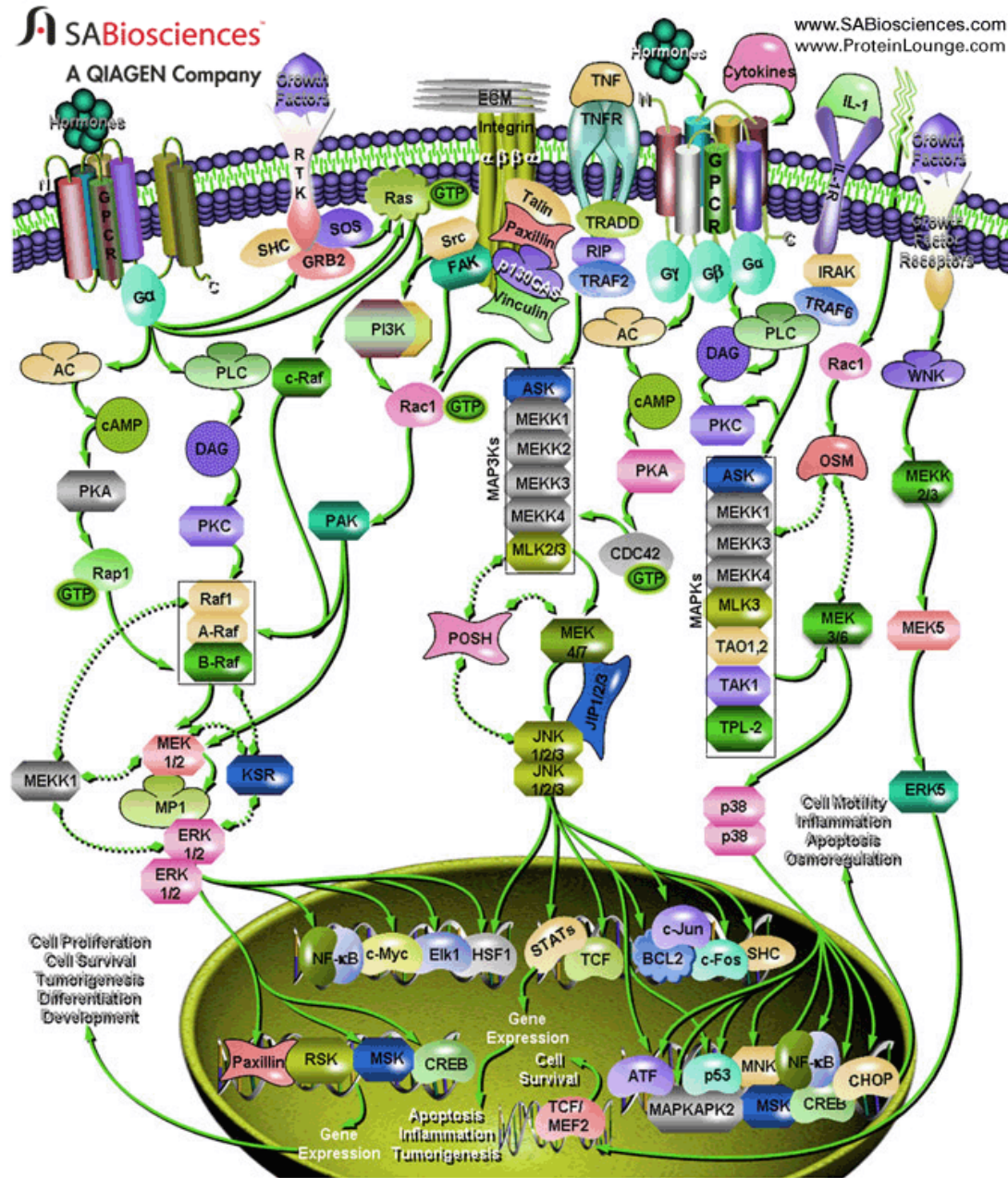


Un grain de sel de la physique non linéaire dans la soupe moléculaire d'une cellule vivante

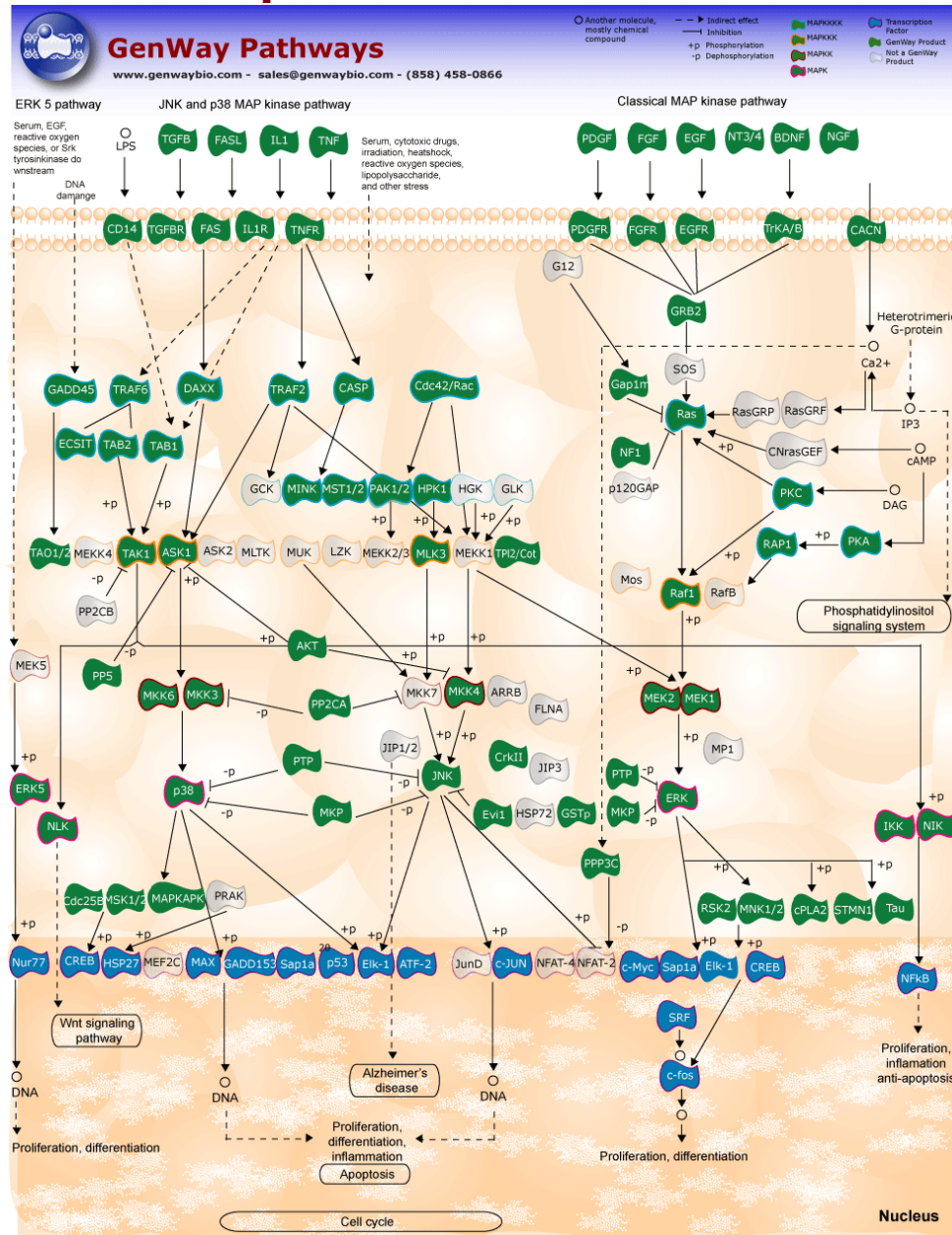
Jacques-Alexandre SEPULCHRE



The molecular soup...

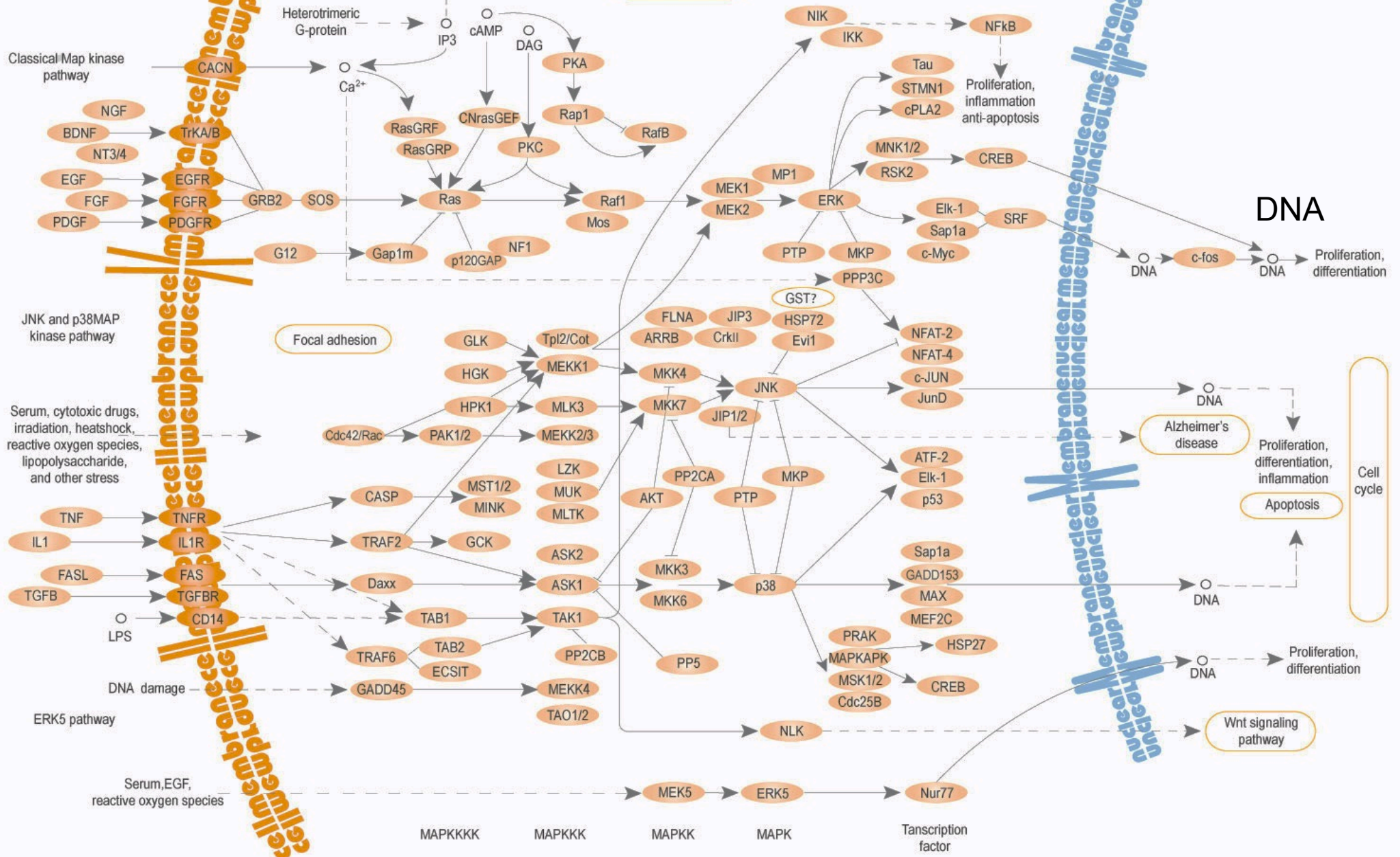


The molecular soup...



The molecular soup

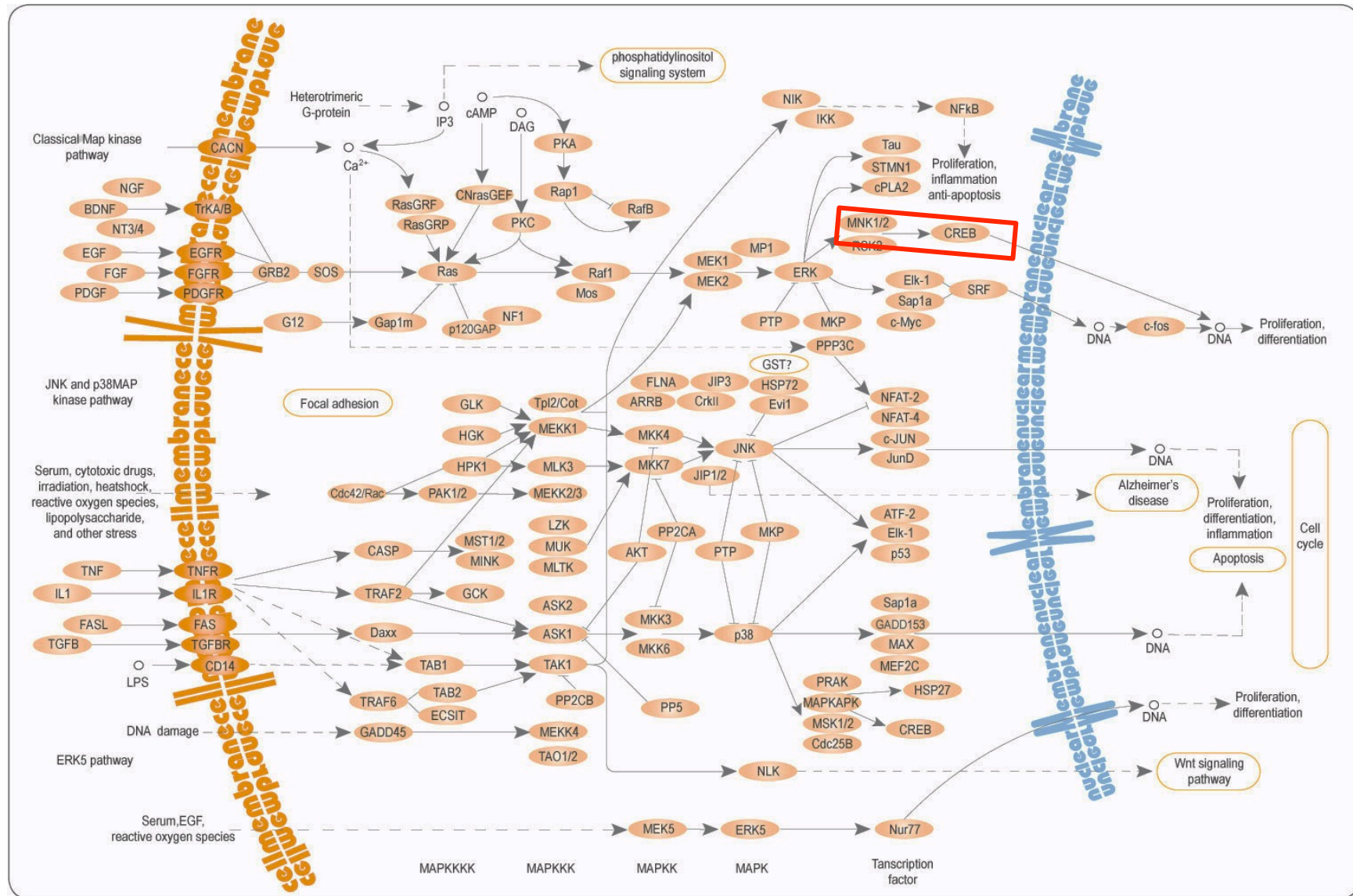
phosphatidylinositol signaling system



Plan

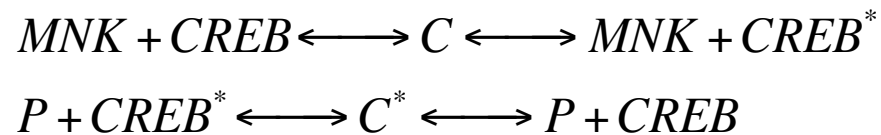
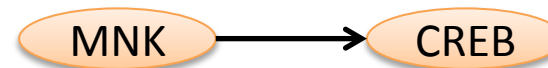
1. Some principles of the modeling of intracellular biochemical kinetics by coupled ODEs
2. Optimal response of genetic expression under periodic stimulations
3. Onset of autonomous periodic oscillations in signaling cascades

1. Some principles of the modeling...



1. Some principles of the modeling...

Enzymatic activation (e.g. phosphorylation)



$$\begin{aligned}
 \frac{d[CREB]}{dt} &= \dots \\
 \frac{d[C]}{dt} &= \dots, \quad \frac{d[C^*]}{dt} = \dots
 \end{aligned}$$

a) Kinetics associated with the *law of mass action* (\rightarrow 3 var.)

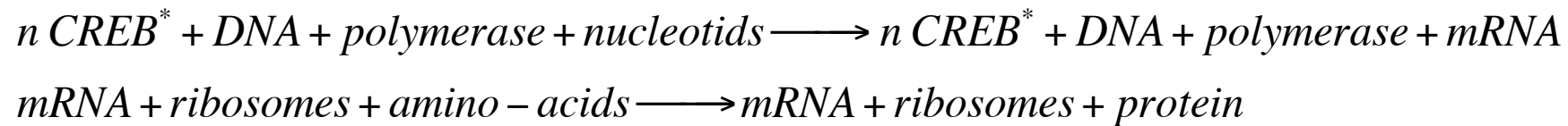
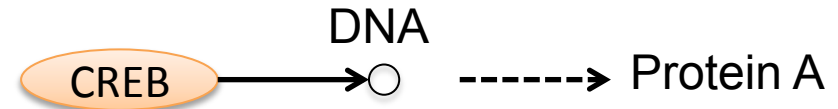
b) Kinetics associated with the Michaelis-Menten approximation (\rightarrow 1 var.)

$$x = \frac{[CREB^*]}{[CREB]_{total}}, \quad \frac{dx}{dt} = k[MNK]_{total} \frac{(1-x)}{K + (1-x)} - k'[P]_{total} \frac{x}{K' + x}$$

Rem : an enzymatic activation involves often a double phosphorylation

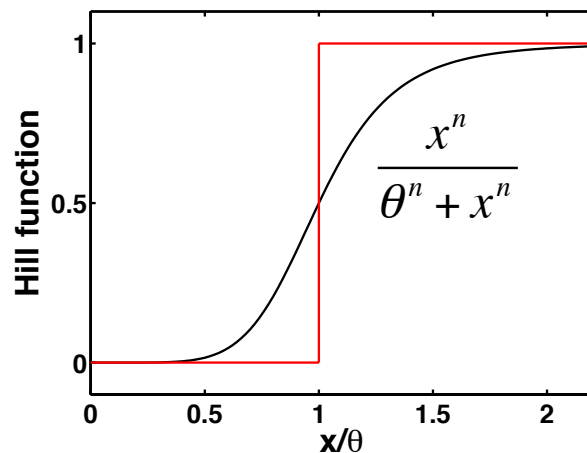
1. Some principles of the modeling...

Activation of a gene by a transcription factor



Phenomenological kinetics with a Hill function :

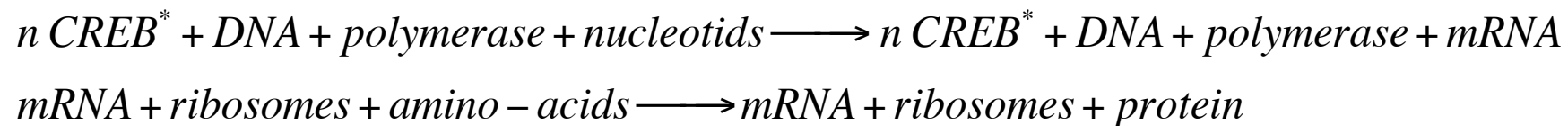
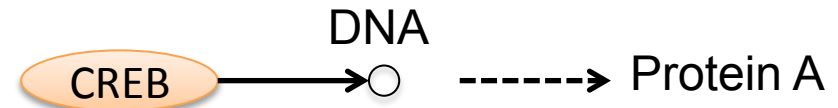
$$\frac{dA}{dt} = \beta \frac{x^n}{\theta^n + x^n} - \alpha A$$



Approximation : $\frac{x}{\theta^n + x^n} \approx H(x - \theta)$

1. Some principles of the modeling...

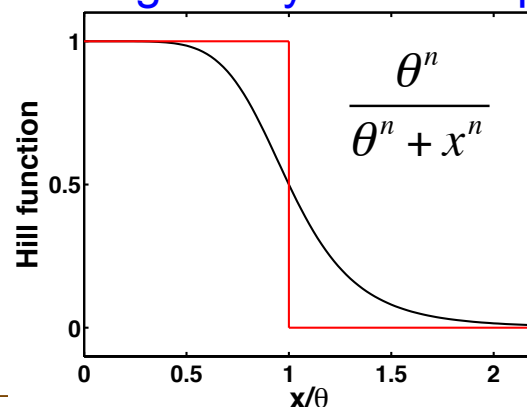
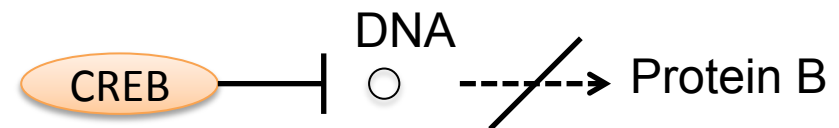
Activation of a gene by a transcription factor



Phenomenological kinetics with a Hill function :

$$\frac{dA}{dt} = \beta \frac{x^n}{\theta^n + x^n} - \alpha A$$

Repression of a gene by a transcription factor:



$$\frac{dB}{dt} = \beta \frac{\theta^n}{\theta^n + x^n} - \alpha B$$

Plan

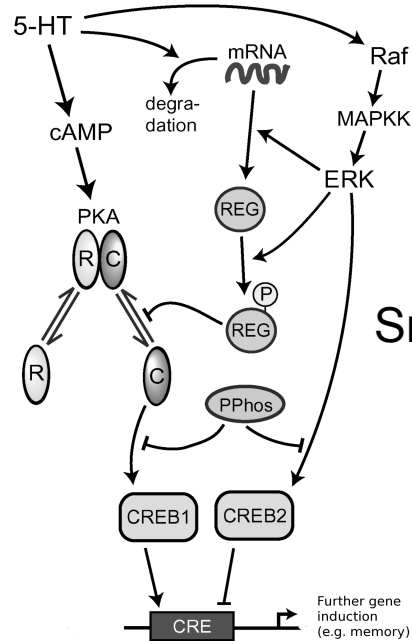
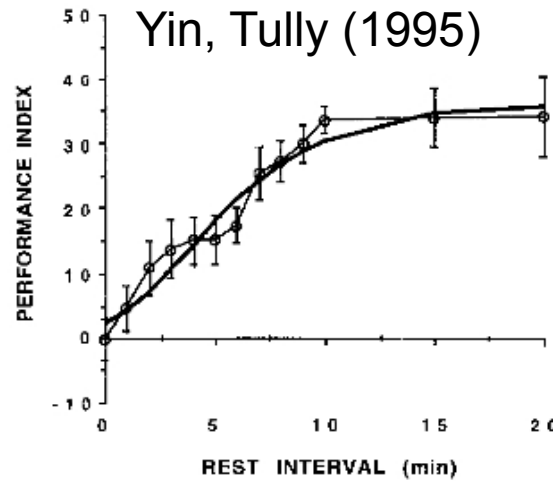
1. Some principles of the modeling of intracellular biochemical kinetics by coupled ODEs
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2. Optimal response of genetic expression under periodic stimulation

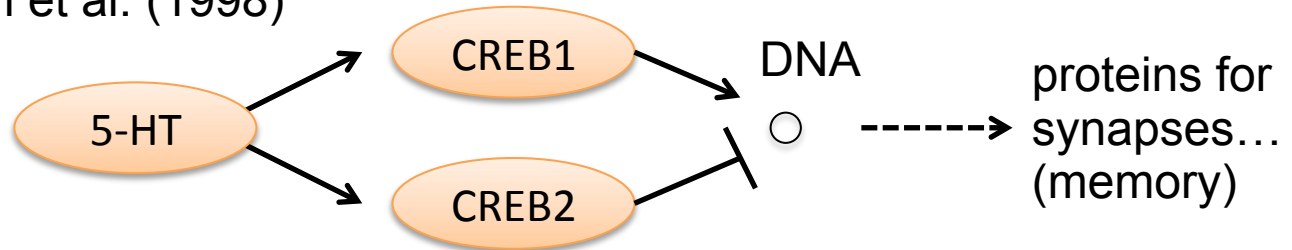
Context : periodic stimulations associated with the formation of memory ... in the drosophila.



B

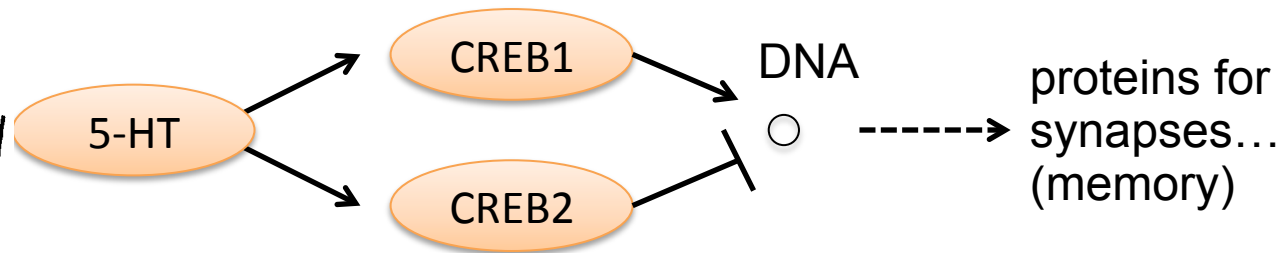
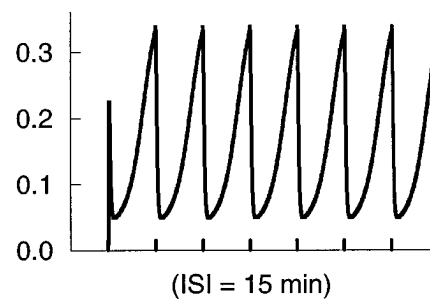


Smolen et al. (1998)



Qualitative modeling (Smolen et al, , *Am J Physiol Cell Physiol*, 1998)

Spaced



$$\frac{d[\text{AP}]}{dt} = k_{A,f}(t) \frac{2A_{\text{tot}} - [\text{AP}]}{2A_{\text{tot}} - [\text{AP}] + K_{A,\text{ph}}} - k_{A,b} \frac{[\text{AP}]}{[\text{AP}] + K_{A,\text{deph}}} \quad (\text{A1})$$

$$\frac{d[\text{AA}]}{dt} = -k_{1,f} [\text{AA}] G_{\text{free}} + k_{1,b} [\text{AAB}] \quad (\text{A2})$$

 where $[\text{AAB}] = A_{\text{tot}} - [\text{AA}]$.

$$r_{\text{Rep}} = k_{\text{Rep}} \frac{[\text{AAB}] [\text{AP}]}{A_{\text{tot}}} \quad (\text{A9})$$

$$\frac{d[\text{Rep}]}{dt} = r_{\text{Rep}} - [\text{Rep}] \quad (\text{A11})$$

$$\frac{d[\text{L1}]}{dt} = \frac{r_{\text{L1}}}{(1 + [\text{Rep}]^2 / K_{\text{Rep}}^2)} - [\text{L1}] \quad (\text{A12})$$

$G_{\text{tot}} = 0.1$	$A_{\text{tot}} = 1.0$	$R_{\text{tot}} = 3.0$
$K_{A,\text{ph}} = 20.0$	$K_{A,\text{deph}} = 20.0$	$k_{A,b} = 0.7 \text{ min}^{-1}$
$k_{R,b} = 7.0 \text{ min}^{-1}$	$k_{1,f} = 10.0 \text{ min}^{-1}$	$k_{1,b} = 10.0 \text{ min}^{-1}$
$K_{R,\text{ph}} = 10.0$	$K_{R,\text{deph}} = 10.0$	$k_{2,f} = 10.0 \text{ min}^{-1}$
$k_{2,b} = 1.0 \text{ min}^{-1}$	ISI varies	$k_{A,\text{max}} = 30 \text{ min}^{-1}$
$k_{R,\text{max}} = 30 \text{ min}^{-1}$	$\tau_1 = 4.0 \text{ min}$	$\tau_2 = 4.0 \text{ min}$

$$r_{R,\text{ph}} = k_{R,f}(t) \frac{2[\text{RR}] + [\text{RRP}]}{2[\text{RR}] + [\text{RRP}] + K_{R,\text{ph}}} \quad (\text{A3})$$

$$r_{R,\text{deph}} = k_{R,b} \frac{2[\text{RRPP}] + [\text{RRP}]}{2[\text{RRPP}] + [\text{RRP}] + K_{R,\text{deph}}} \quad (\text{A4})$$

$$\frac{d[\text{RR}]}{dt} = -r_{R,\text{ph}} \frac{2[\text{RR}]}{2[\text{RR}] + [\text{RRP}]} + r_{R,\text{deph}} \frac{[\text{RRP}]}{2[\text{RRPP}] + [\text{RRP}]} \quad (\text{A5})$$

$$\frac{d[\text{RRP}]}{dt} = r_{R,\text{ph}} \frac{2[\text{RR}]}{2[\text{RR}] + [\text{RRP}]} - r_{R,\text{deph}} \frac{[\text{RRP}]}{2[\text{RRPP}] + [\text{RRP}]} \quad (\text{A6})$$

$$- r_{R,\text{ph}} \frac{[\text{RRP}]}{2[\text{RR}] + [\text{RRP}]}$$

$$+ r_{R,\text{deph}} \frac{2[\text{RRPP}]}{2[\text{RRPP}] + [\text{RRP}]}$$

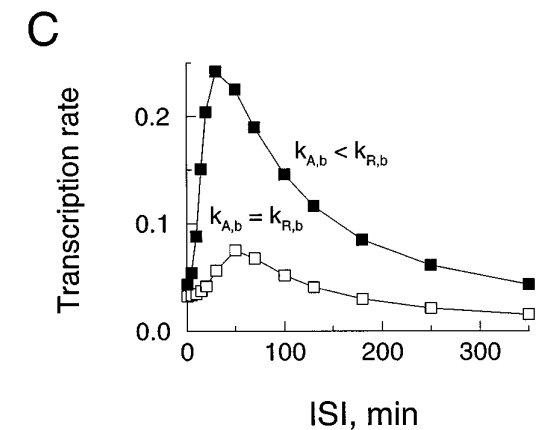
$$\frac{d[\text{RRPP}]}{dt} = r_{R,\text{ph}} \frac{[\text{RRP}]}{2[\text{RR}] + [\text{RRP}]} - r_{R,\text{deph}} \frac{2[\text{RRPP}]}{2[\text{RRPP}] + [\text{RRP}]} \quad (\text{A7})$$

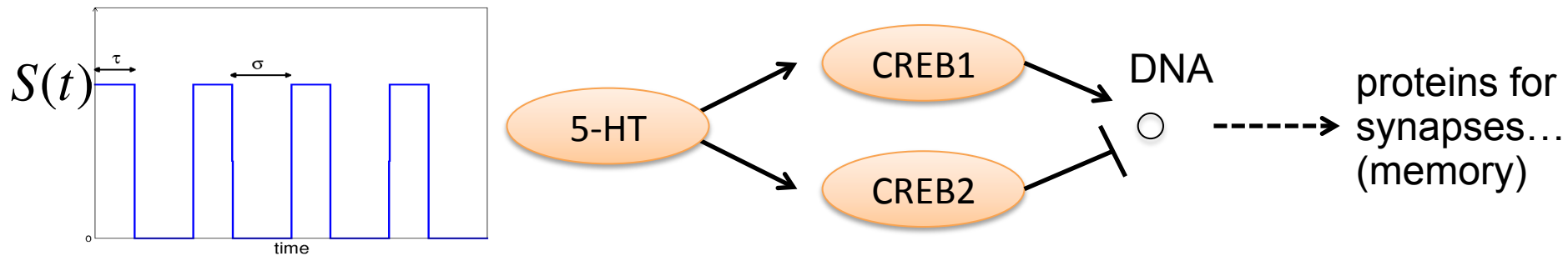
$$- r_{R,\text{deph}} \frac{2[\text{RRPP}]}{2[\text{RRPP}] + [\text{RRP}]}$$

$$- k_{2,f} [\text{RRPP}] G_{\text{free}} + k_{2,b} [\text{RRPPB}]$$

$$\frac{d[\text{RRPPB}]}{dt} = k_{2,f} [\text{RRPP}] G_{\text{free}} - k_{2,b} [\text{RRPPB}] \quad (\text{A8})$$

Result :
« frequency selectivity »
In the gene response



Qualitative modeling (A. Cournac and JAS, *BMC Systems Biol.*, 2009)


$$\dot{x} = k S(t)(x_{tot} - x) - k' x$$

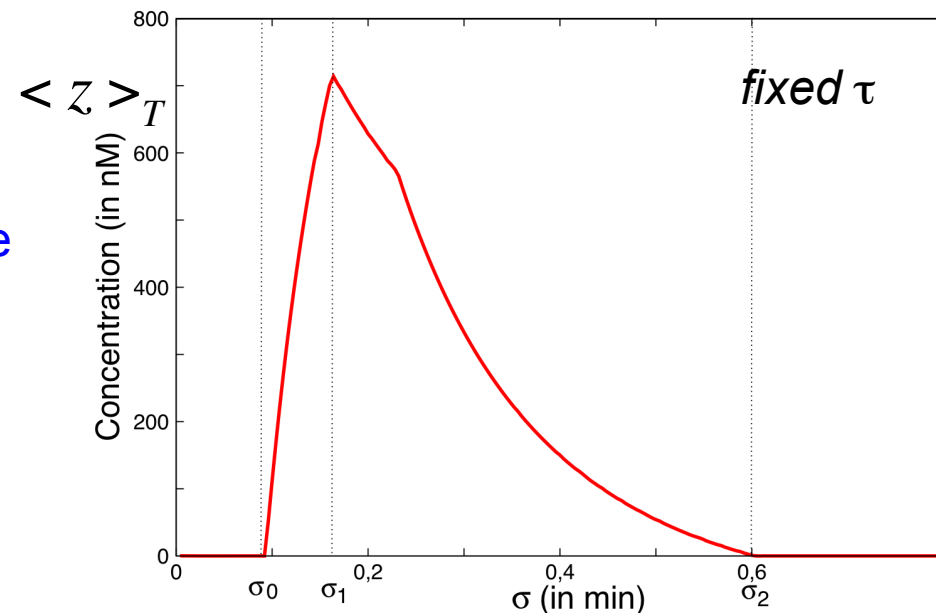
$$\dot{y} = k S(t)(y_{tot} - y) - k' y$$

$$\dot{z} = \beta H(x - \theta_1)H(\theta_2 - y) - \alpha z$$

Result :
interspike selectivity
In the gene response

Some benefits of the simplified modeling:

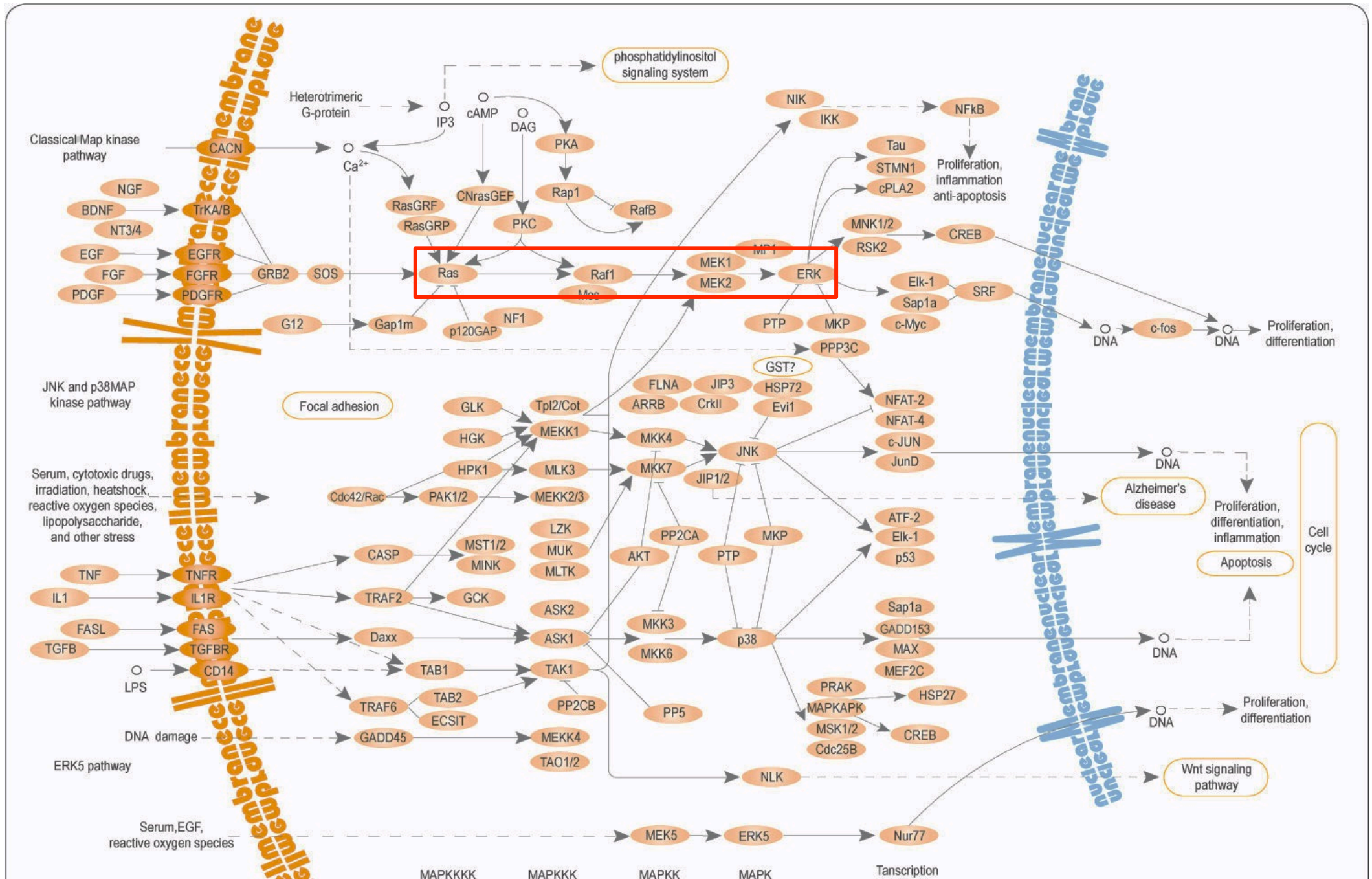
- Analytical calculations are made possible
- Highlight the role of repressor, (and a closed link with the IFFL of U. Alon)
- Highlight the role of the interspike rather than the frequency of the stimulus



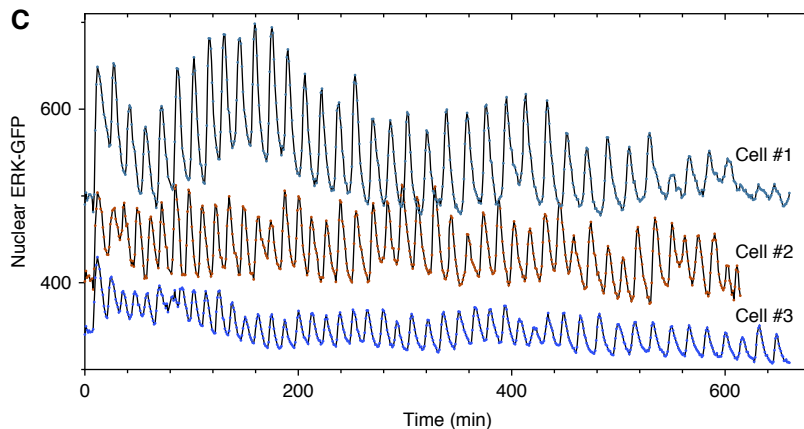
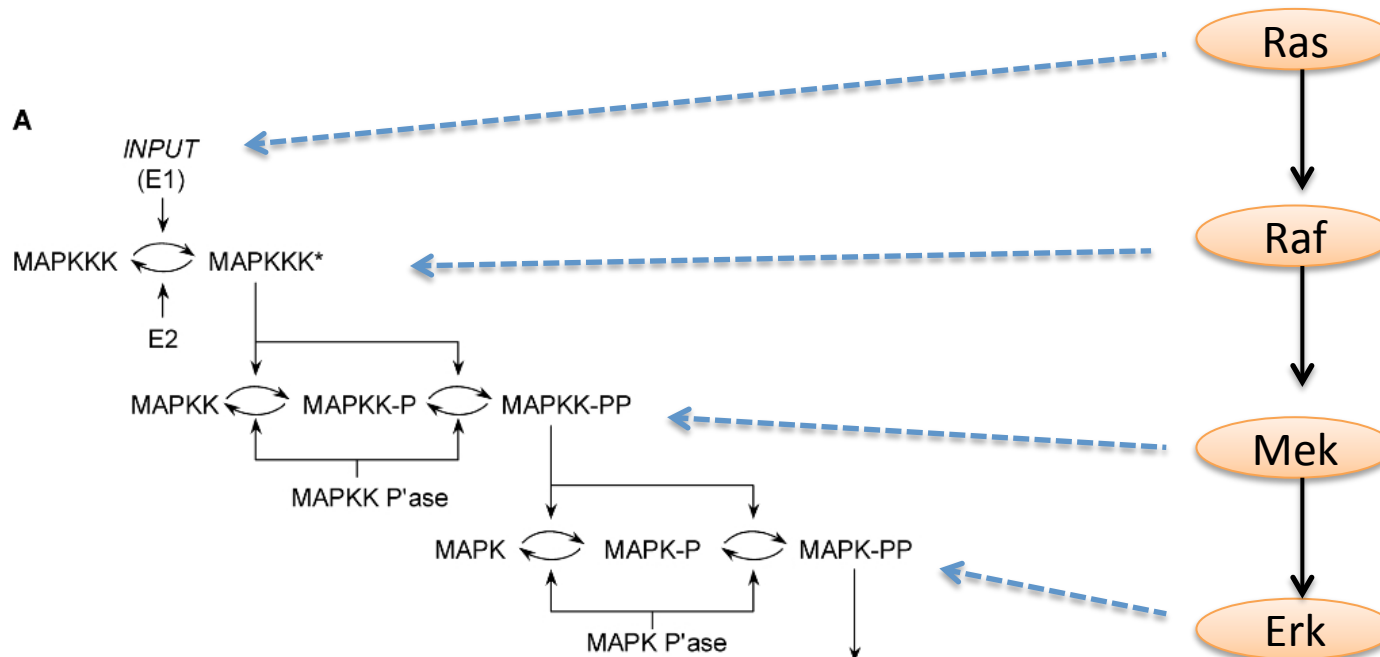
Plan

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3. Periodic oscillations in signaling cascades



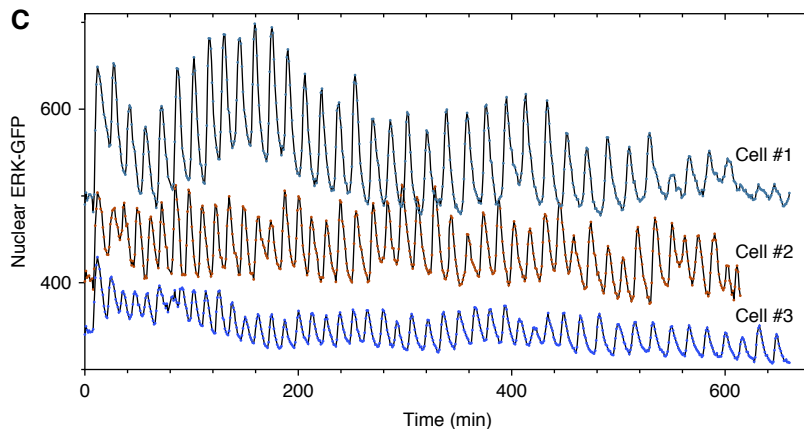
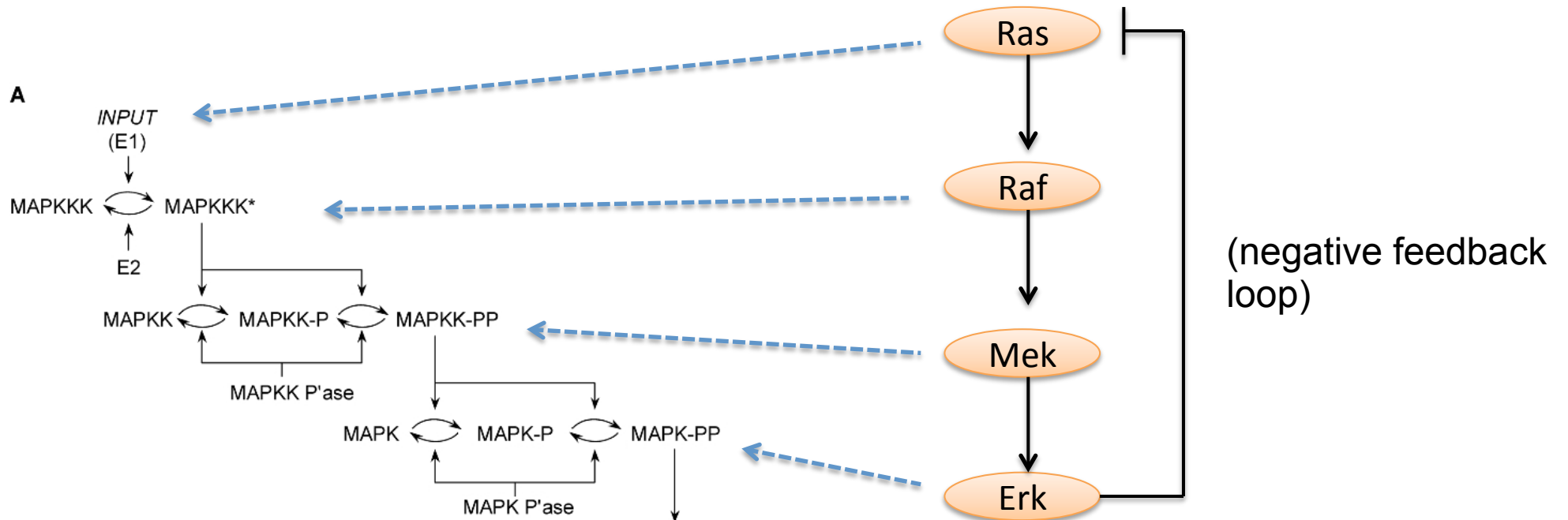
Autonomous oscillations in MAPK cascade



Rapid and sustained nuclear–cytoplasmic ERK oscillations induced by epidermal growth factor

Shankaran et al. Mol. Syst. Biol. 2009.

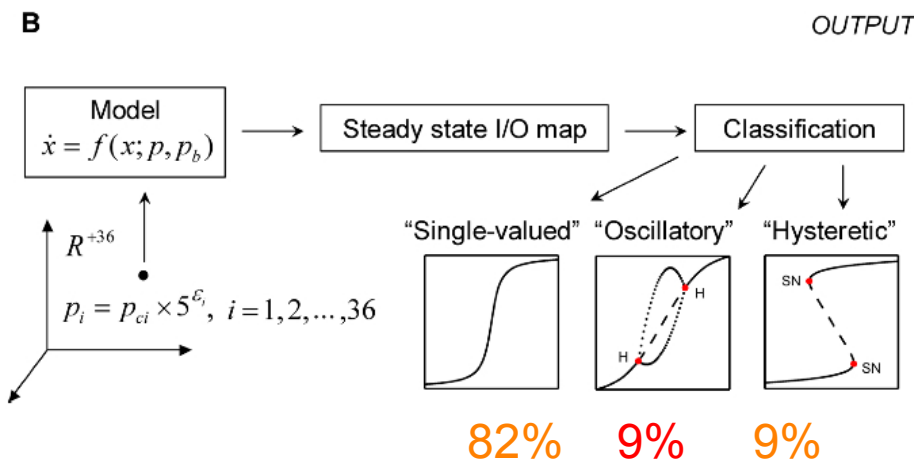
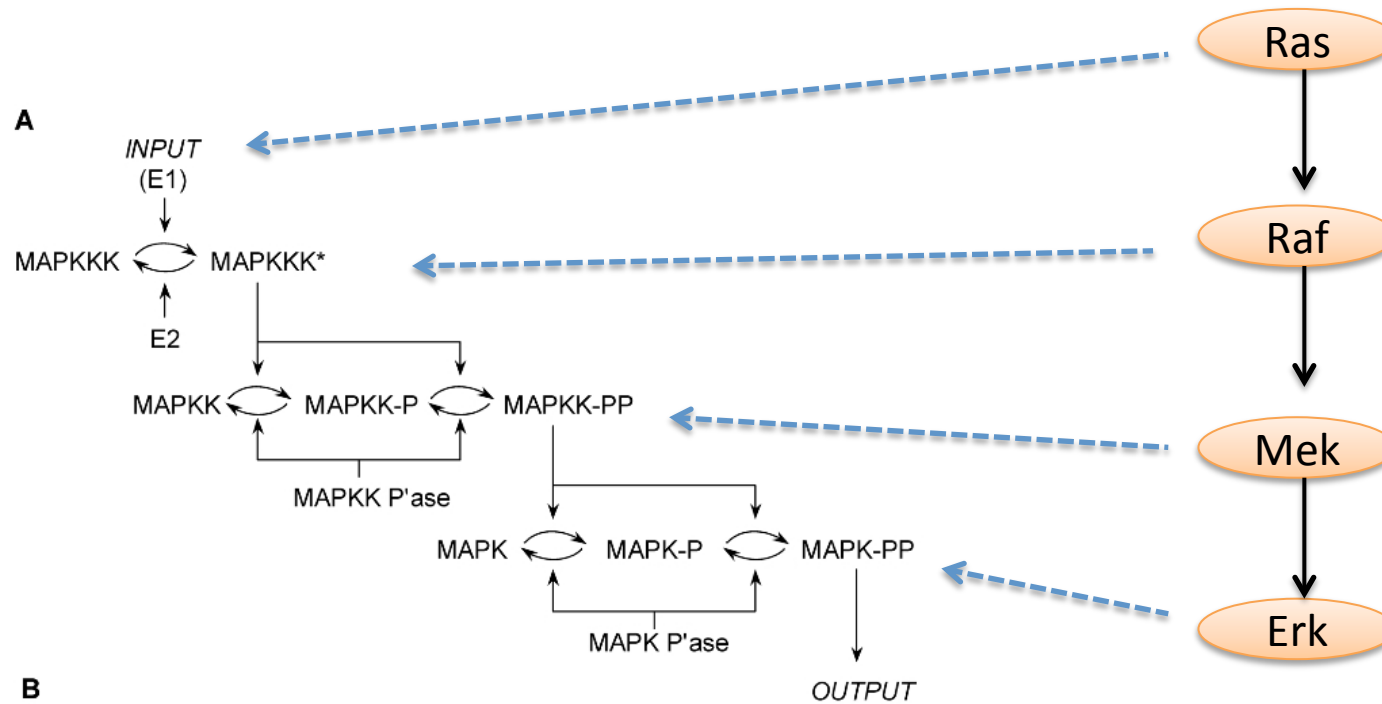
Autonomous oscillations in MAPK cascade



Rapid and sustained nuclear–cytoplasmic ERK oscillations induced by epidermal growth factor

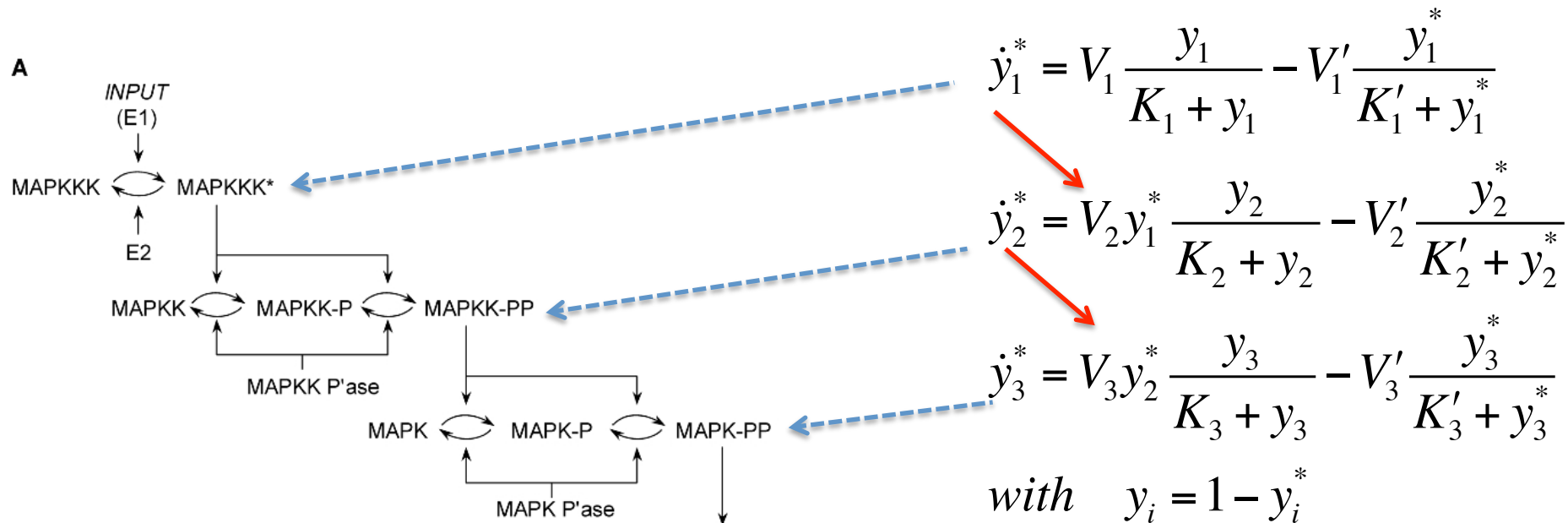
Shankaran et al. Mol. Syst. Biol. 2009.

Autonomous oscillations in MAPK cascade



Qiao et al., (PLoS sept. 2007)

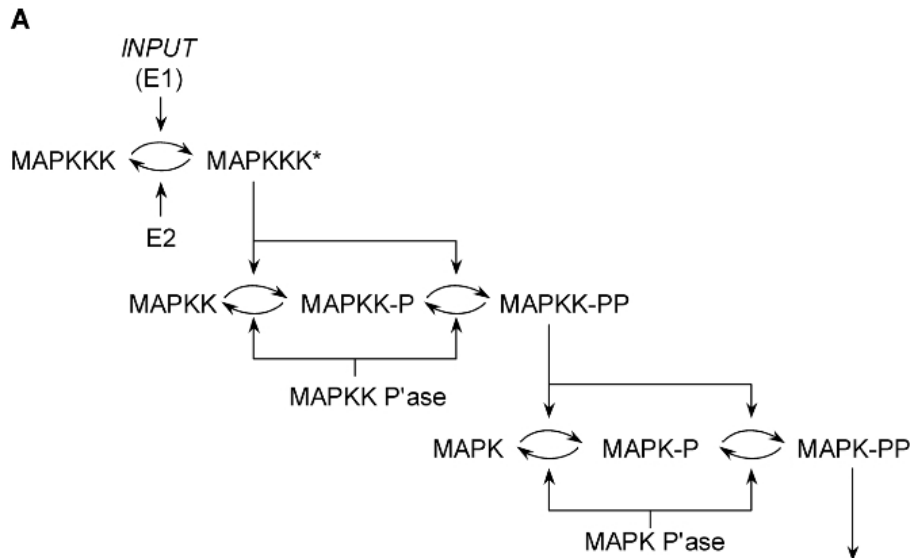
Autonomous oscillations in MAPK cascade



(e.g. Kholodenko, 2000,
 Angeli et al, 2004,
 Giuraniuc et al, 2007)

Autonomous oscillations in MAPK cascade

Results from a rigorous perturbation scheme
(Ventura, JAS, PLoS CB 2008)



$$\dot{x}_1 = V_1 E_1 \frac{y_1}{K_1 + y_1} - V'_1 \frac{x_1}{K'_1 (1 + y_2 / K_2) + x_1}$$

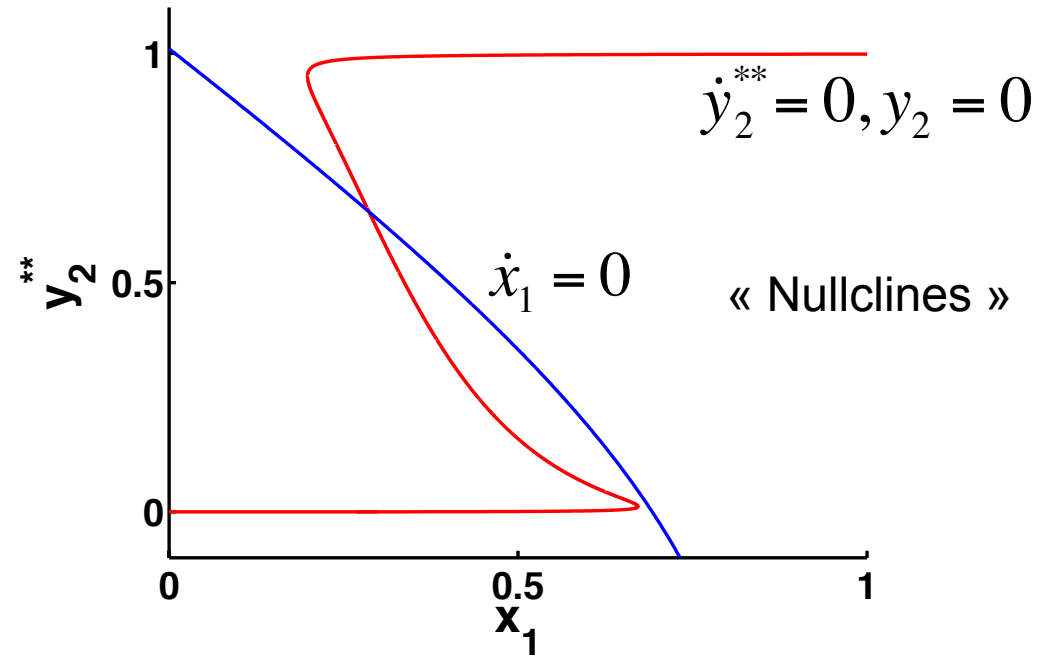
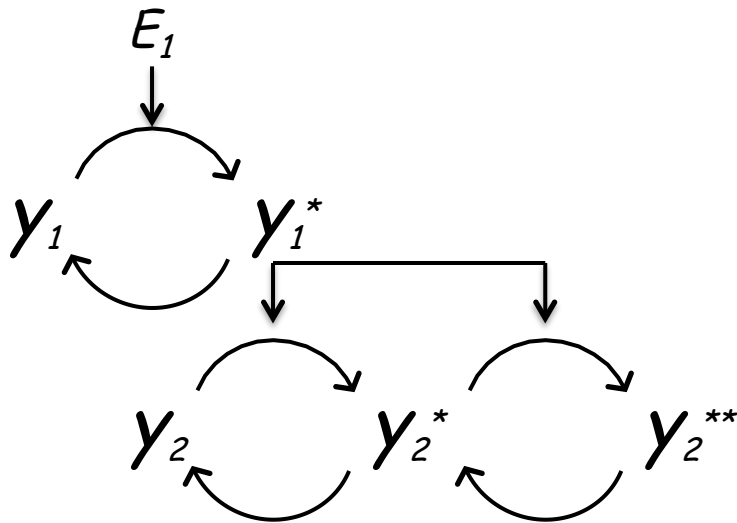
$$\dot{x}_2 = V_2 x_1 \frac{y_2}{K_2 + y_2} - V'_2 \frac{x_2}{K'_2 (1 + y_3 / K_3) + x_2}$$

$$\dot{x}_3 = V_3 x_2 \frac{y_3}{K_3 + y_3} - V'_3 \frac{x_3}{K'_3 + x_3}$$

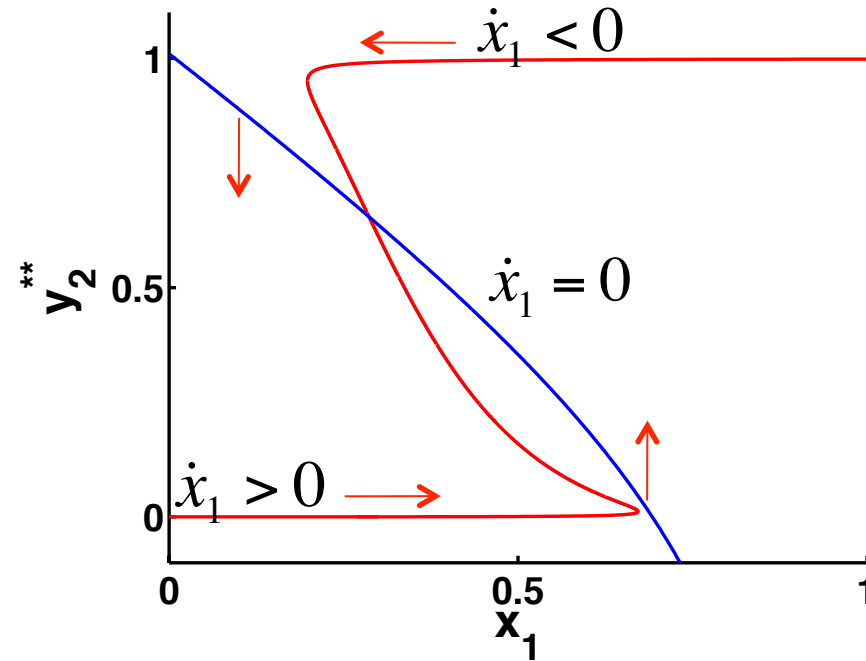
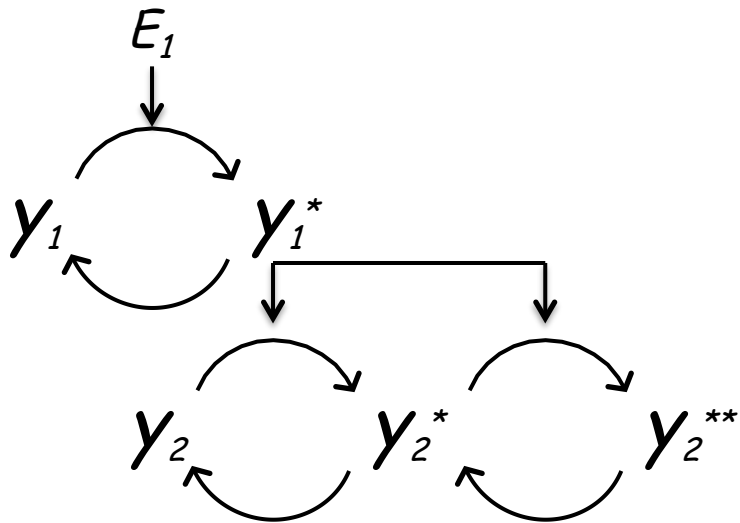
with $y_i = 1 - x_i$ and $x_i = y_i^* (1 + y_{i-1} / K_{i-1})$

There is an intrinsic negative feedback

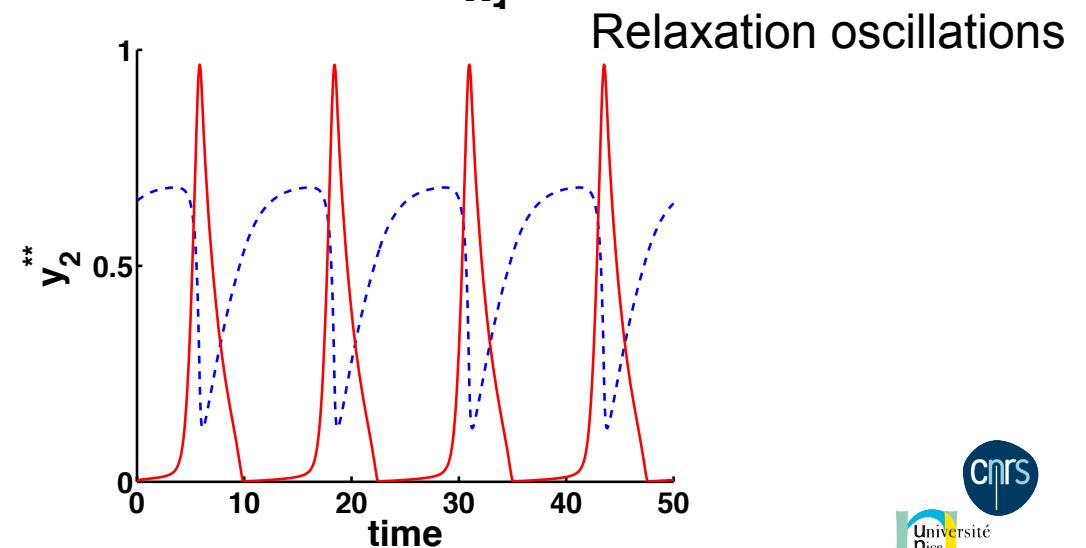
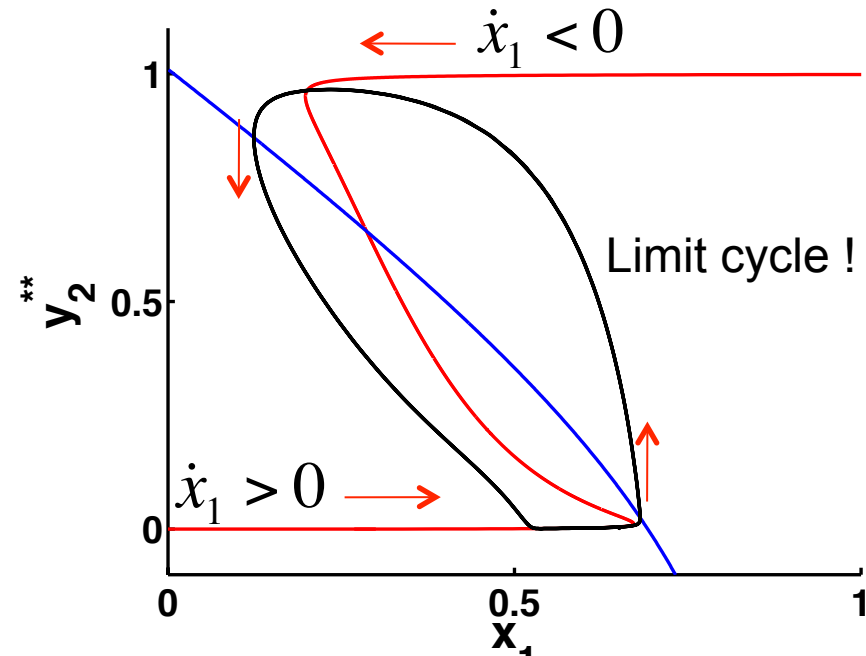
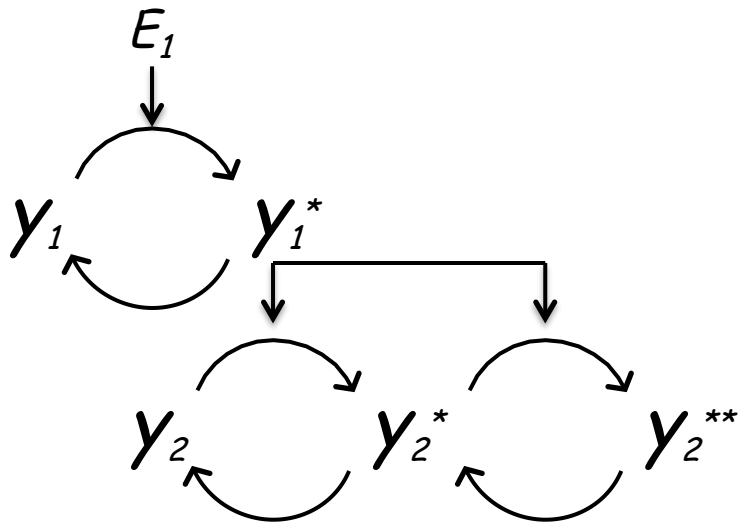
Autonomous oscillations in MAPK cascades



Autonomous oscillations in MAPK cascades



Autonomous oscillations in MAPK cascades



Bifurcations:

- Supercritical Hopf bifurcation
- Saddle-node bifurcation on a cycle

Conclusion : via 2 study cases, we saw that

1. Periodic pulsatile stimulations of gene network motifs can lead to optimal response in the production of proteins. (Possible applications to memory formation)
2. A negative retroactivity exists in signaling cascades. It leads to the possibility of autonomous oscillations in (MAPK) signaling pathways.

We conclude that the « relevant » level of modeling is not absolute:

- Looking for a « minimal » model allows one to better understand (and control) the underlying mechanisms
- This form of reductionism can typically be brought up by physicists !

Acknowledgements

Study case 1

Axel Cournac, LPTMC (postdoc), Univ. Pierre et Marie Curie

Study case 2

Alejandra Ventura, University of Buenos Aires, Argentina (main collaboration)

Sofia Merajver, Cancer Center, University of Michigan