

Mettre un (tout petit) bruit dans les modèles de dynamique des populations

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Inria Modémic

Dans un chémostat

$$\dot{s} = d \cdot (S_{in} - s) - \mu(s)x$$

$$\dot{x} = (\mu(s) - d) \cdot x$$

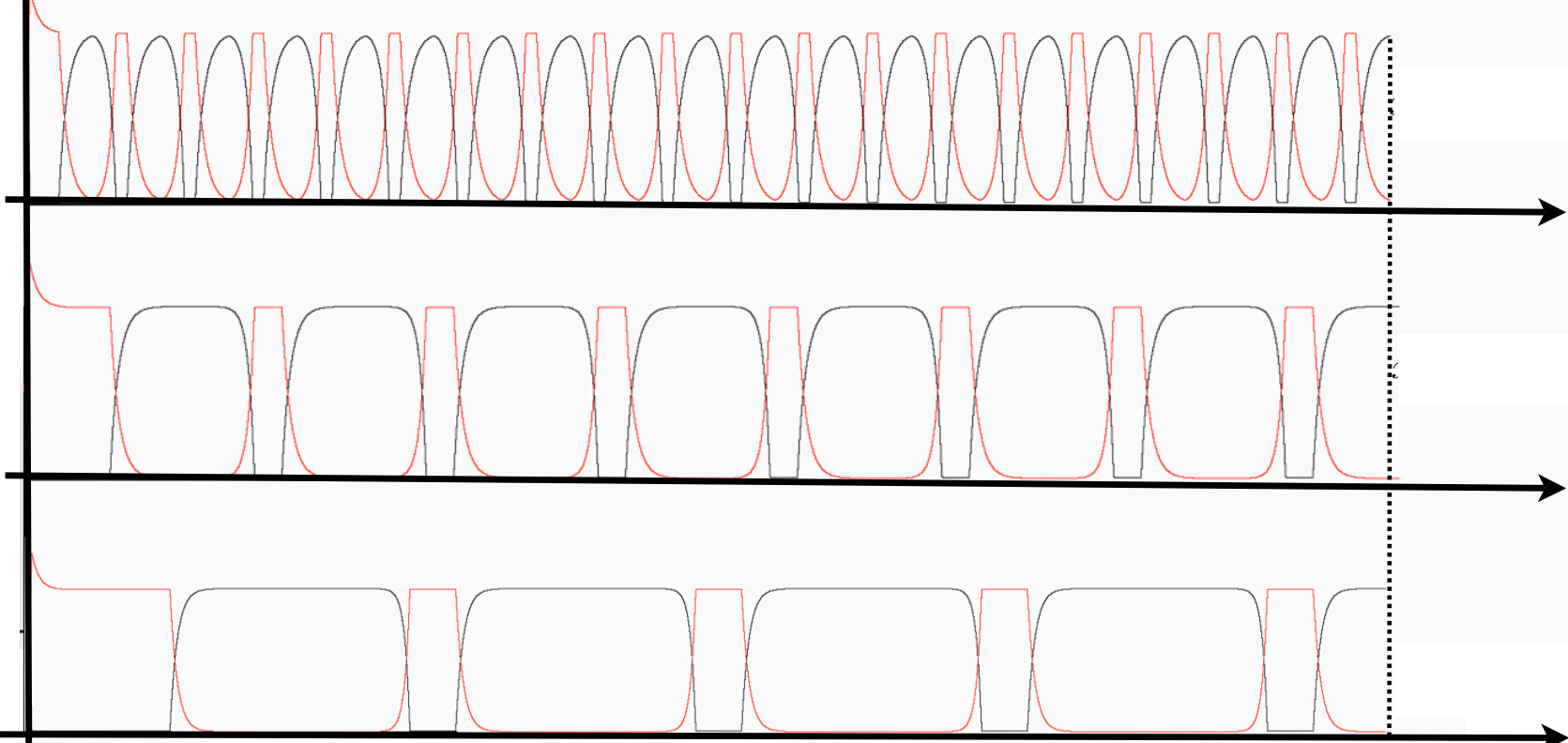
Jour

$$\dot{s} = d \cdot (S_{in} - s) - \cancel{\mu(s)}x$$

$$\dot{x} = \cancel{(\mu(s) - d)} \cdot x$$

Nuit

Dans un chémostat



$$\dot{s} = d \cdot (S_{in} - s) - \mu(s)x$$

$$\dot{x} = (\mu(s) - d) \cdot x$$

$$\dot{s} = d \cdot (S_{in} - s) - \cancel{\mu(s)x}$$

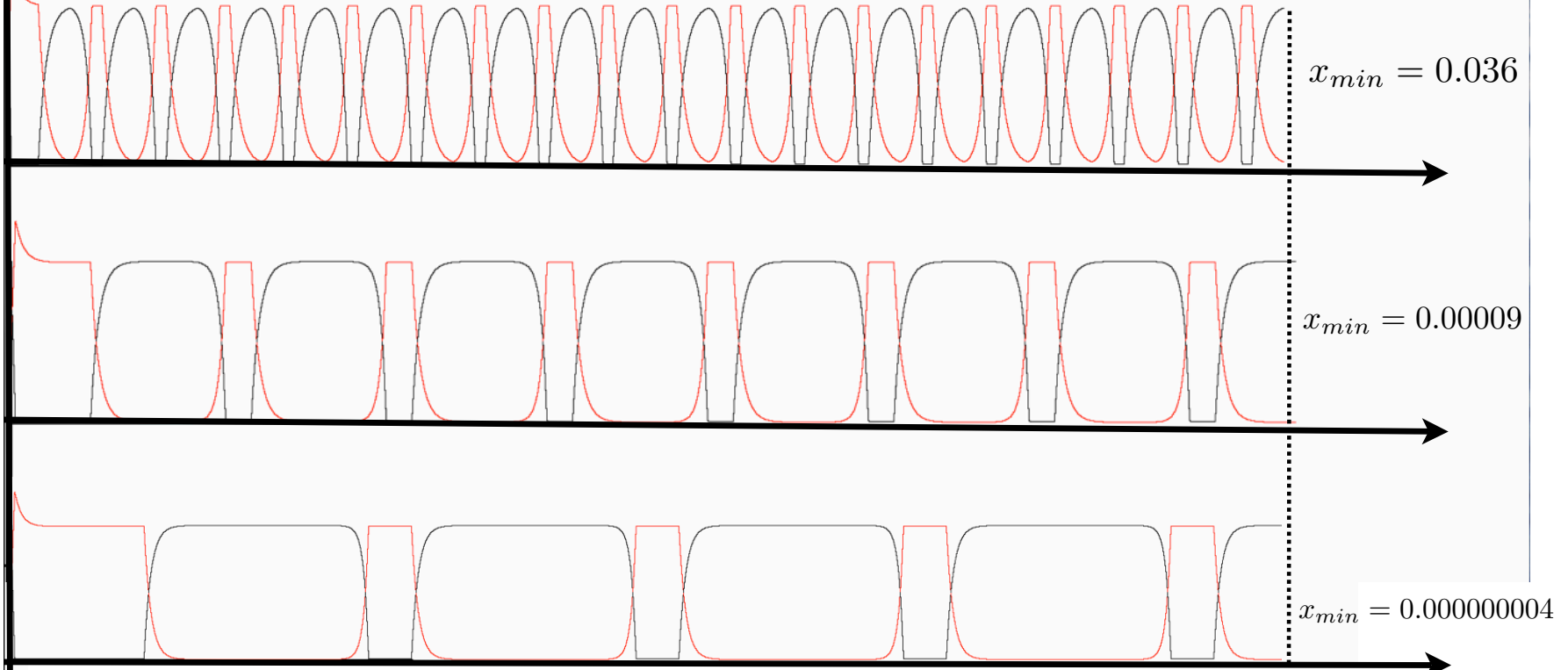
$$\dot{x} = (\cancel{\mu(s)} - d) \cdot x$$

$t = 400$

$$\dot{s} = 0.4 \cdot (2 - s) - \frac{s}{0.02 + s} \cdot x$$

$$\dot{x} = \left(\frac{s}{0.02 + s} - 0.4 \right) \cdot x$$

Dans un chémostat



$$\dot{s} = d \cdot (S_{in} - s) - \mu(s)x$$

$$\dot{x} = (\mu(s) - d) \cdot x$$

$$\dot{s} = d \cdot (S_{in} - s) - \cancel{\mu(s)}x$$

$$\dot{x} = (\cancel{\mu(s)} - d) \cdot x$$

$t = 400$

$$\dot{s} = 0.4 \cdot (2 - s) - \frac{s}{0.02 + s} \cdot x$$

$$\dot{x} = \left(\frac{s}{0.02 + s} - 0.4 \right) \cdot x$$

$t \mapsto x(t)$ ne n'est jamais nul *stricto sensu*.

Zéro machine = 10^{-240}

Mettre dans les programmes :

If $x < \text{seuil}$ then $x := 0$.

Pour les petites populations il faut prendre :

$$x(t) = N(t)$$

où N est un entier

La croissance exponentielle

Proba de naissance

$$P \left(N(t + dt) = \begin{cases} N(t) + 1 \\ N(t) \\ N(t) - 1 \end{cases} \right) = \begin{aligned} &= N(t)pdt + o(dt) \\ &= 1 - N(t)(p + q)dt + o(dt) \\ &= N(t)qdt + o(dt) \end{aligned}$$

Proba de mort

Pour les petites populations il faut prendre :

$$x(t) = N(t)$$

où N est un entier

La croissance exponentielle

$$P \left(N(t + dt) = \begin{cases} N(t) + 1 \\ N(t) \\ N(t) - 1 \end{cases} \right) = \begin{aligned} &= N(t)pdt + o(dt) \\ &= 1 - N(t)(p + q)dt + o(dt) \\ &= N(t)qdt + o(dt) \end{aligned}$$

$$N(t + dt) = N(t) + W \quad W = \begin{pmatrix} +1 \\ 0 \\ -1 \end{pmatrix} \quad \begin{aligned} P(W = +1) &= N(t)pdt + o(dt) \\ P(W = 0) &= 1 - N(t)(p + q)dt + o(dt) \\ P(W = -1) &= N(t)qdt + o(dt) \end{aligned}$$

La croissance exponentielle

$$N(t + dt) = N(t) + W$$

$$W = \begin{pmatrix} +1 \\ 0 \\ -1 \end{pmatrix} \quad \begin{array}{l} P(W = +1) = N(t)pdt + o(dt) \\ P(W = 0) = 1 - N(t)(p + q)dt + o(dt) \\ P(W = -1) = N(t)qdt + o(dt) \end{array}$$

$$E[N(t + dt)] = E[N(t)] + E[N(t)](p - q)dt$$

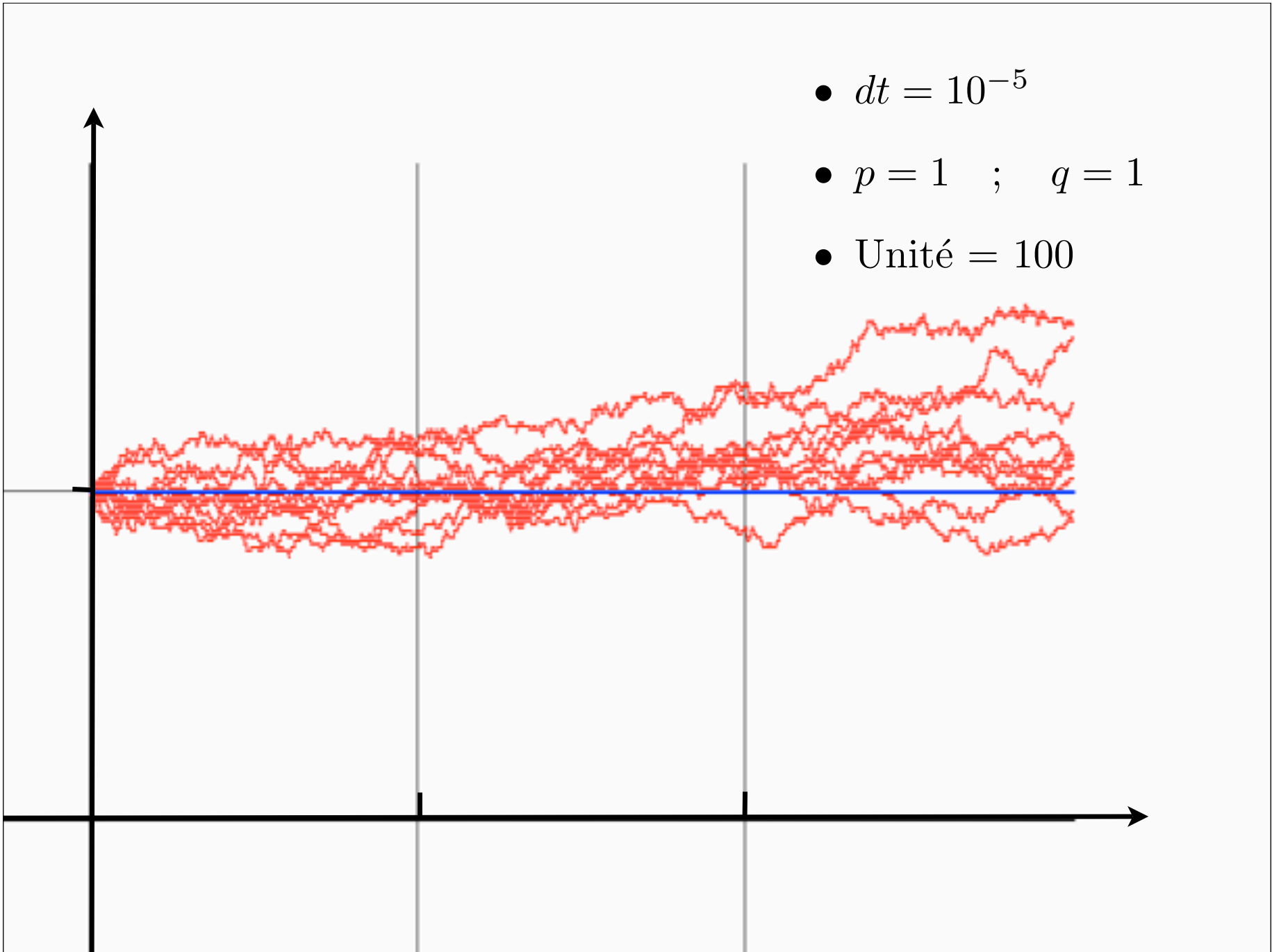
$$m(t) = E[N(t)] \implies m'(t) = (p - q)m(t)$$

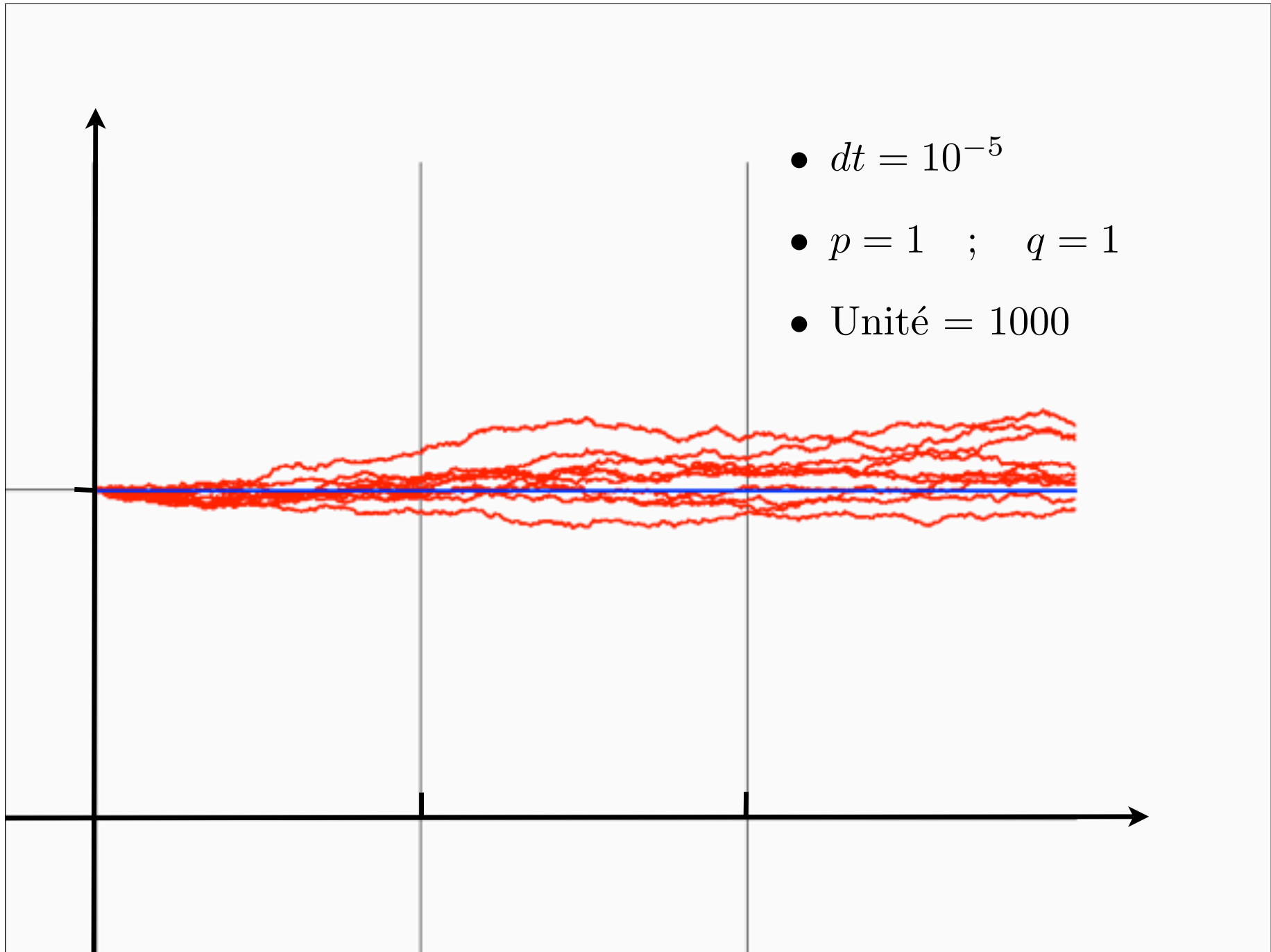
Dépend de la différence des 2 paramètres

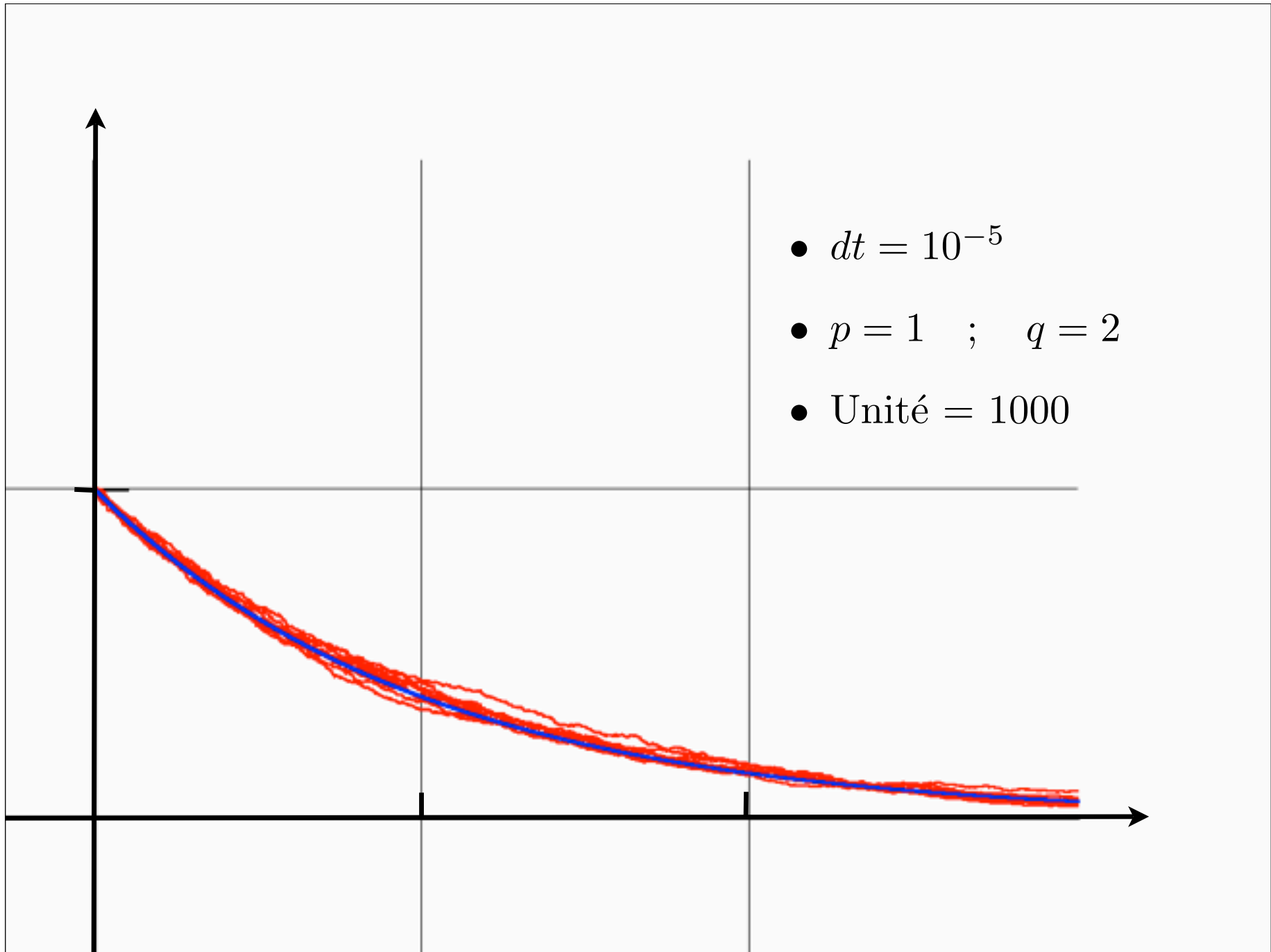
- $dt = 10^{-5}$

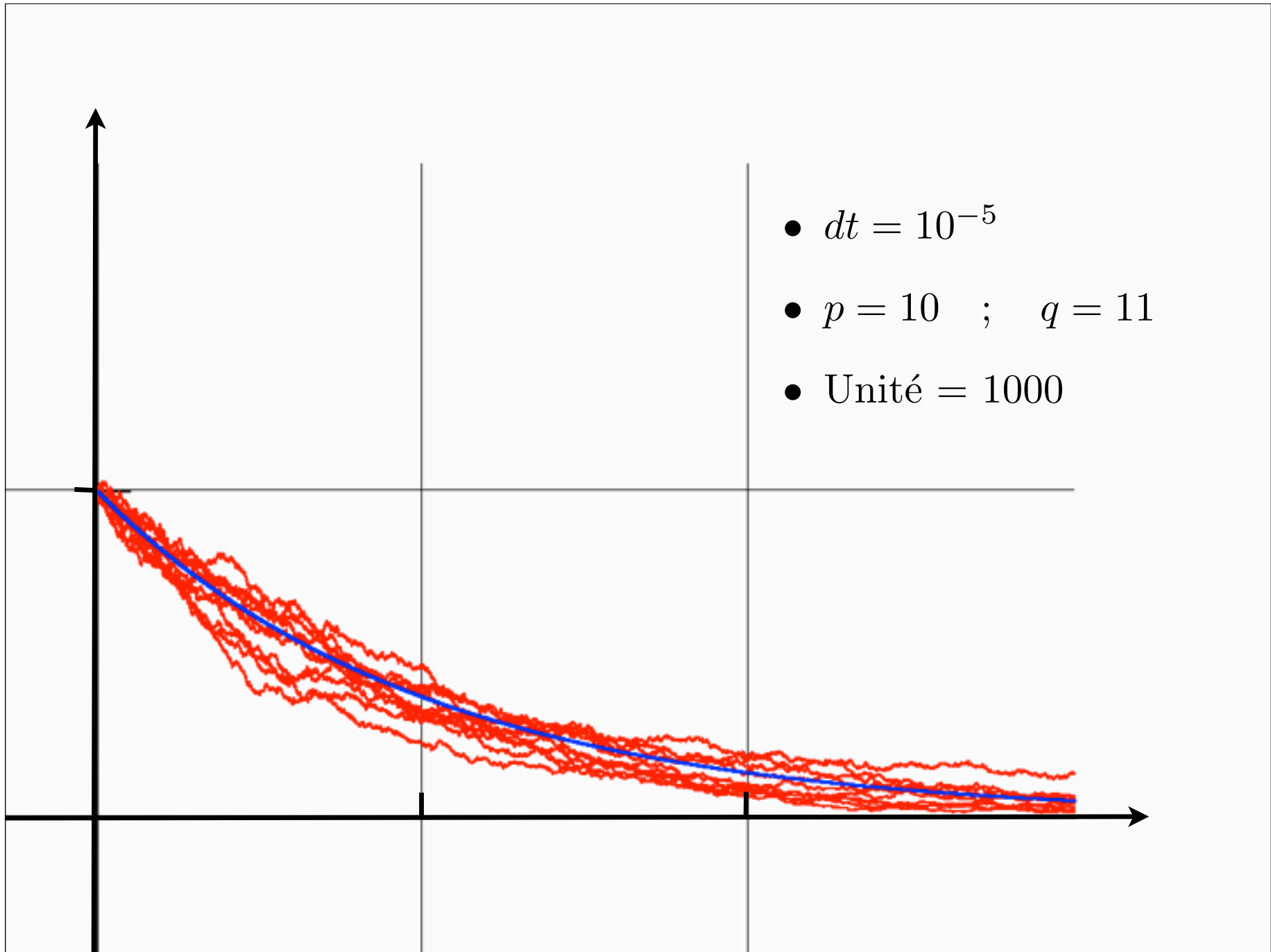
- $p = 1$; $q = 1$

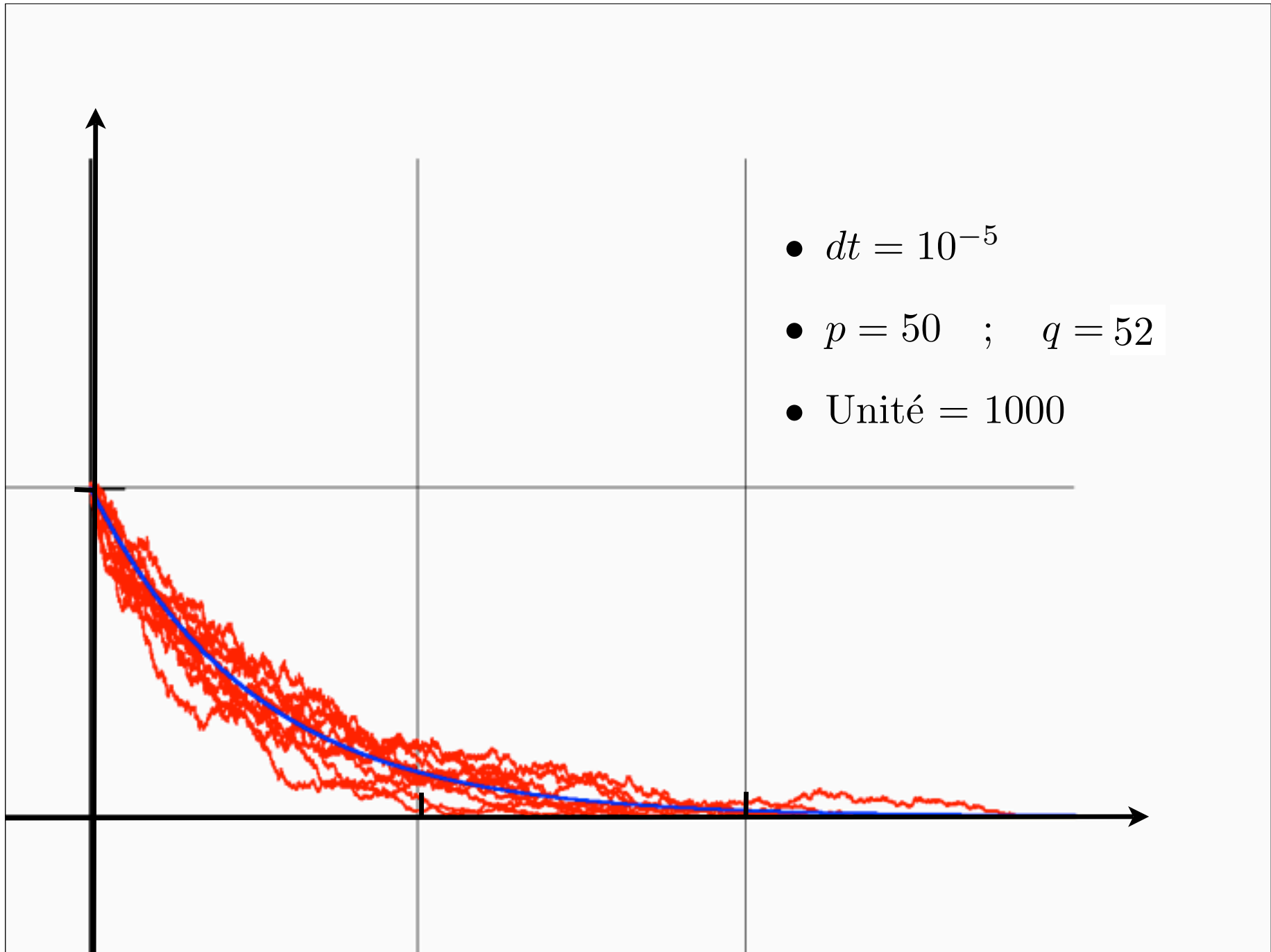
- Unité = 100









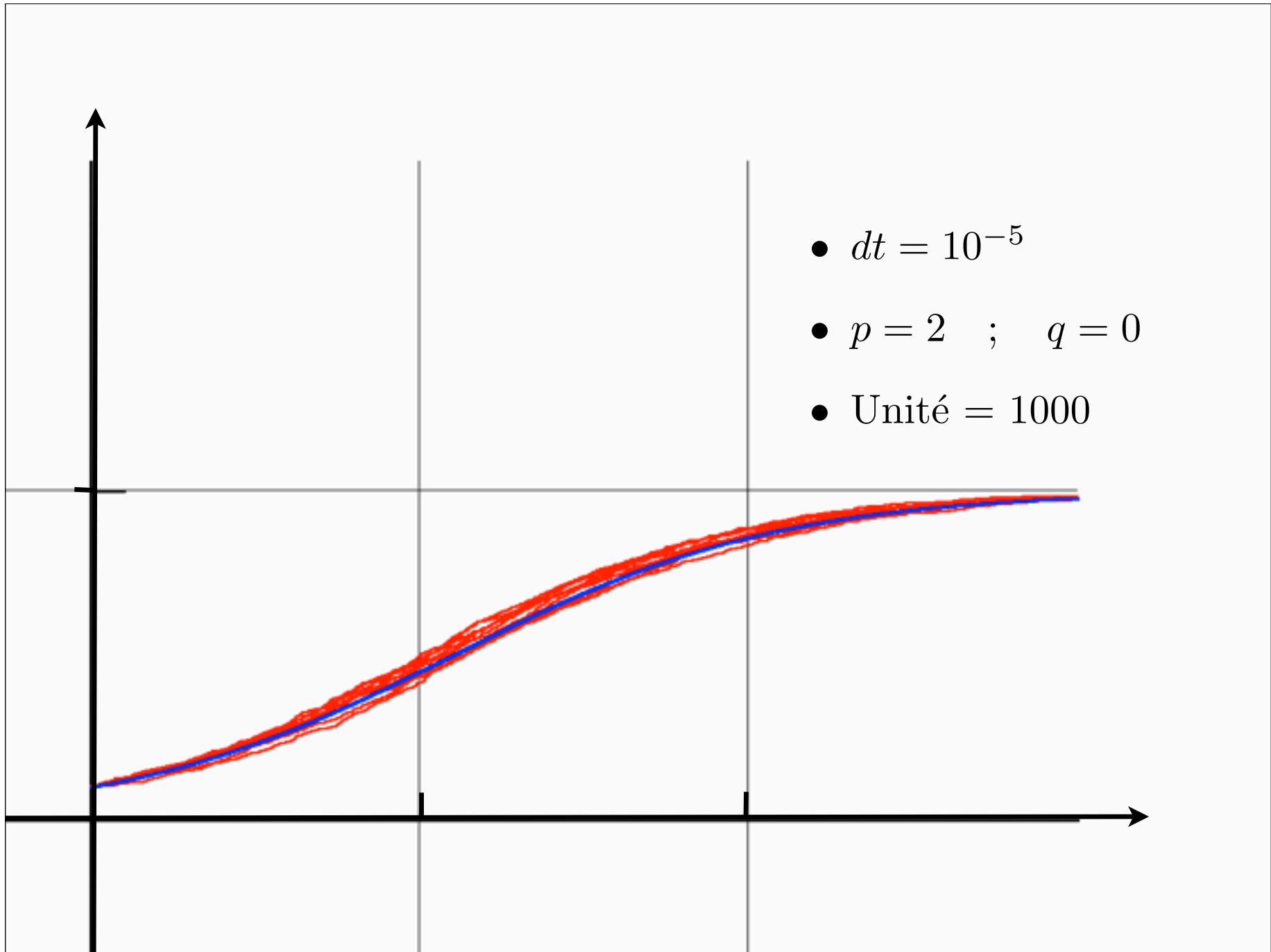


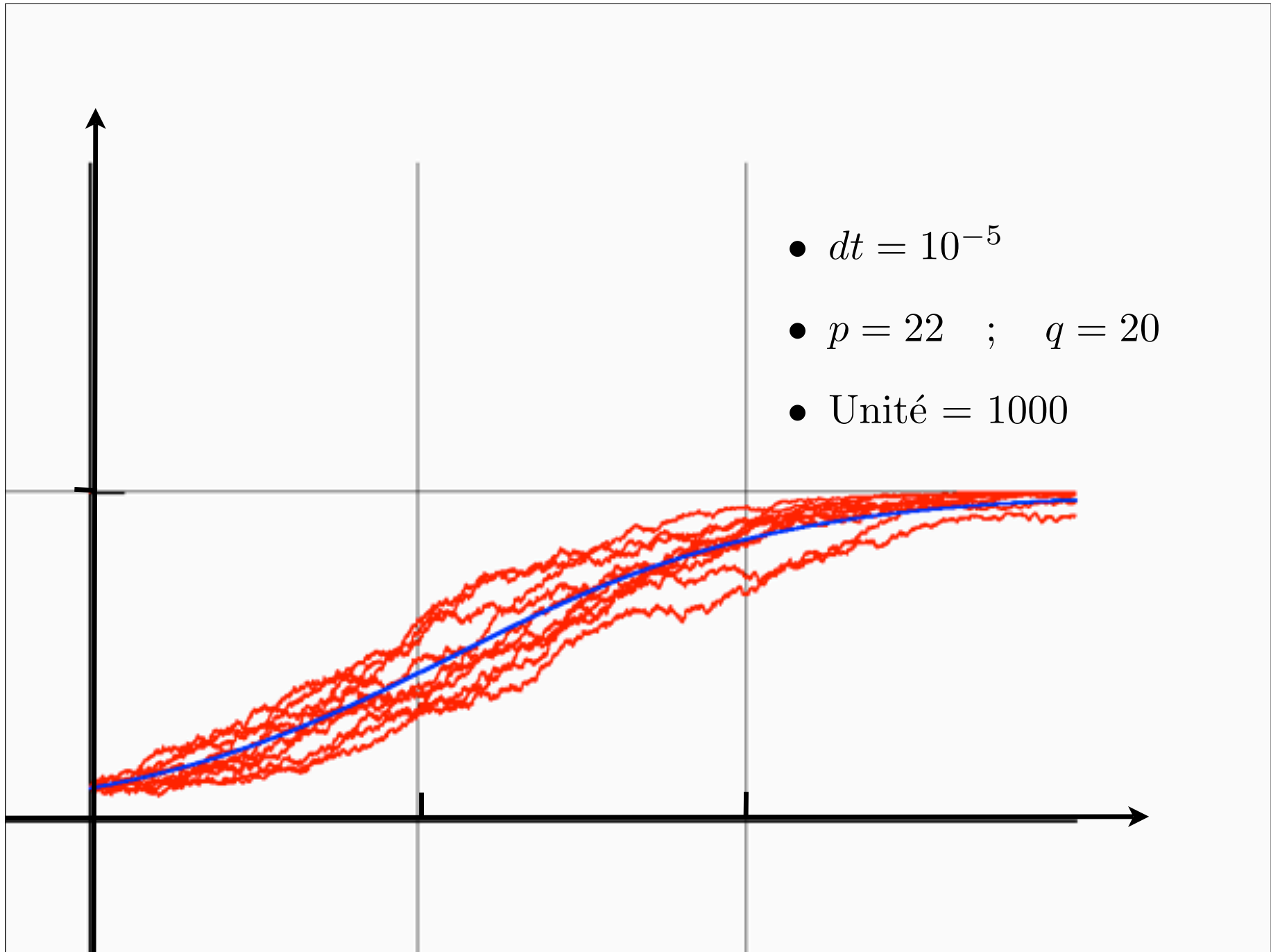
Logistique Vie-et-Mort

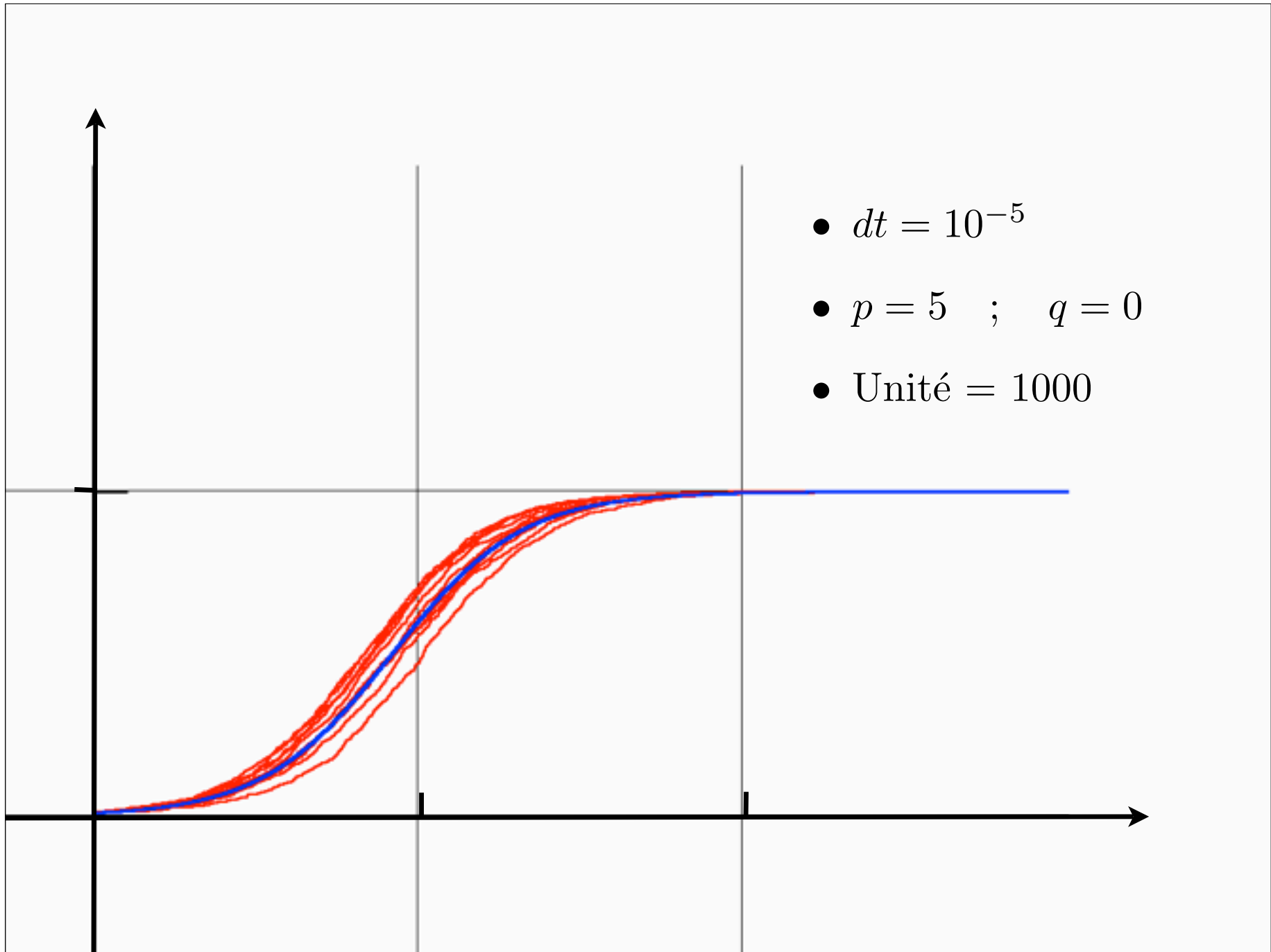
$$\dot{x} = rx(1-x)$$

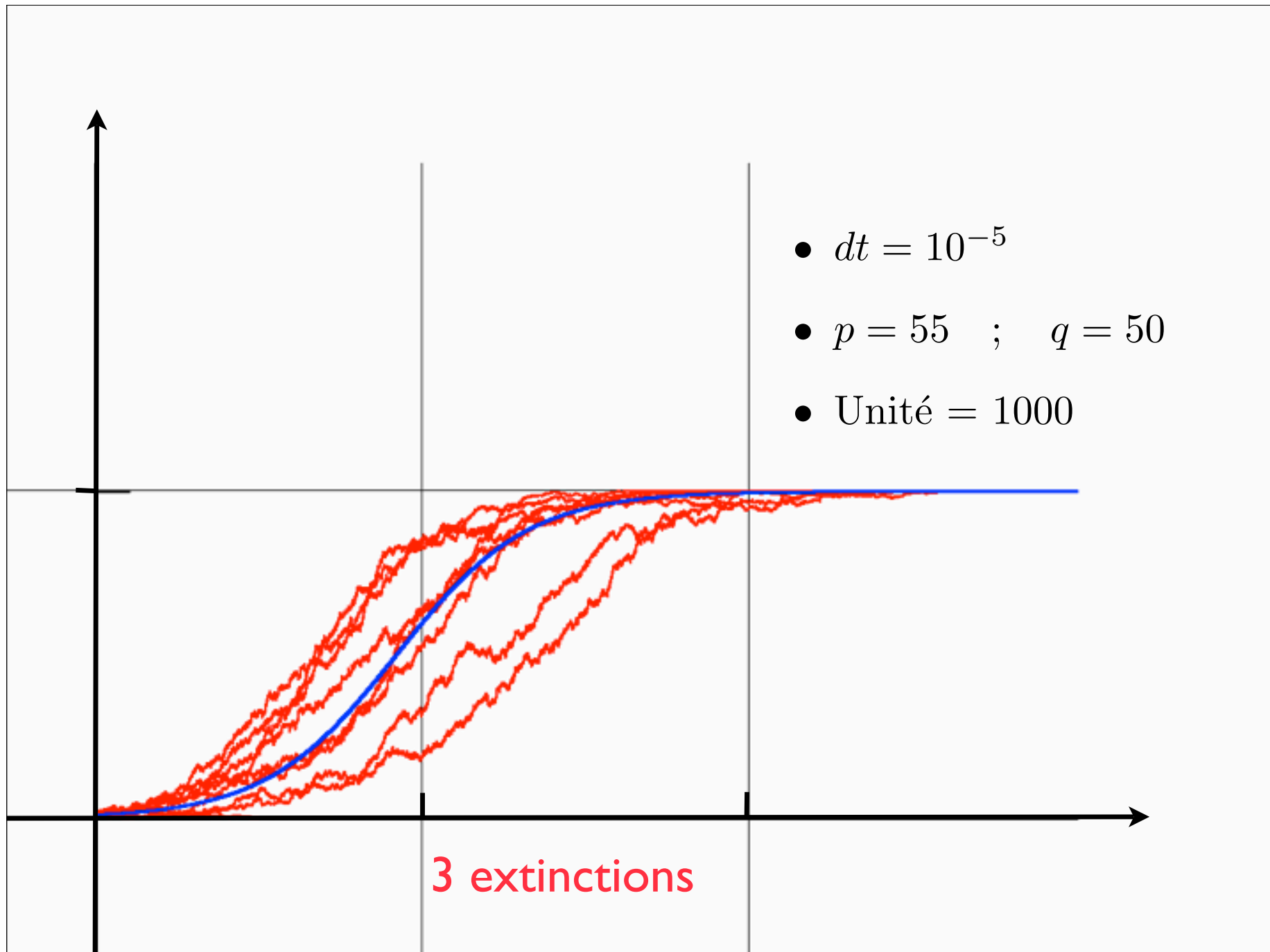
$$P \left(N(t+dt) = \begin{cases} N(t) + 1 \\ N(t) \\ N(t) - 1 \end{cases} \right) \begin{aligned} &\stackrel{=}{=} N(t)pdt + o(dt) \\ &\stackrel{=}{=} 1 - N(t)(p+q)dt + o(dt) \\ &\stackrel{=}{=} N(t)qdt + o(dt) \end{aligned}$$

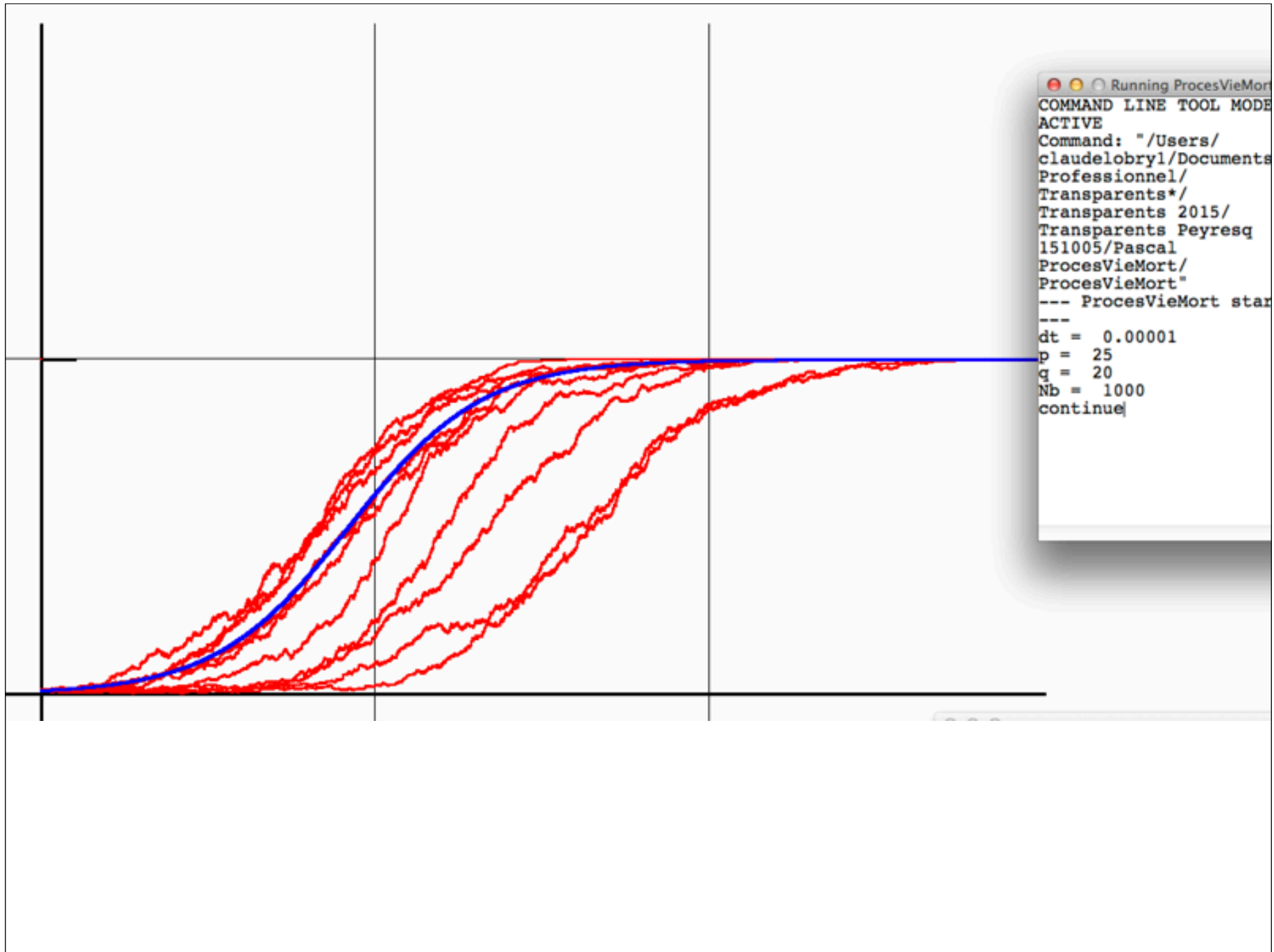
$$P \left(N(t+dt) = \begin{cases} N(t) + 1 \\ N(t) \\ N(t) - 1 \end{cases} \right) \begin{aligned} &= N(t)(1 - N(t)/Nb)pdt + o(dt) \\ &= 1 - N(t)(1 - N(t)/Nb)(p+q)dt + o(dt) \\ &= N(t)(1 - N(t)/Nb)qdt + o(dt) \end{aligned}$$

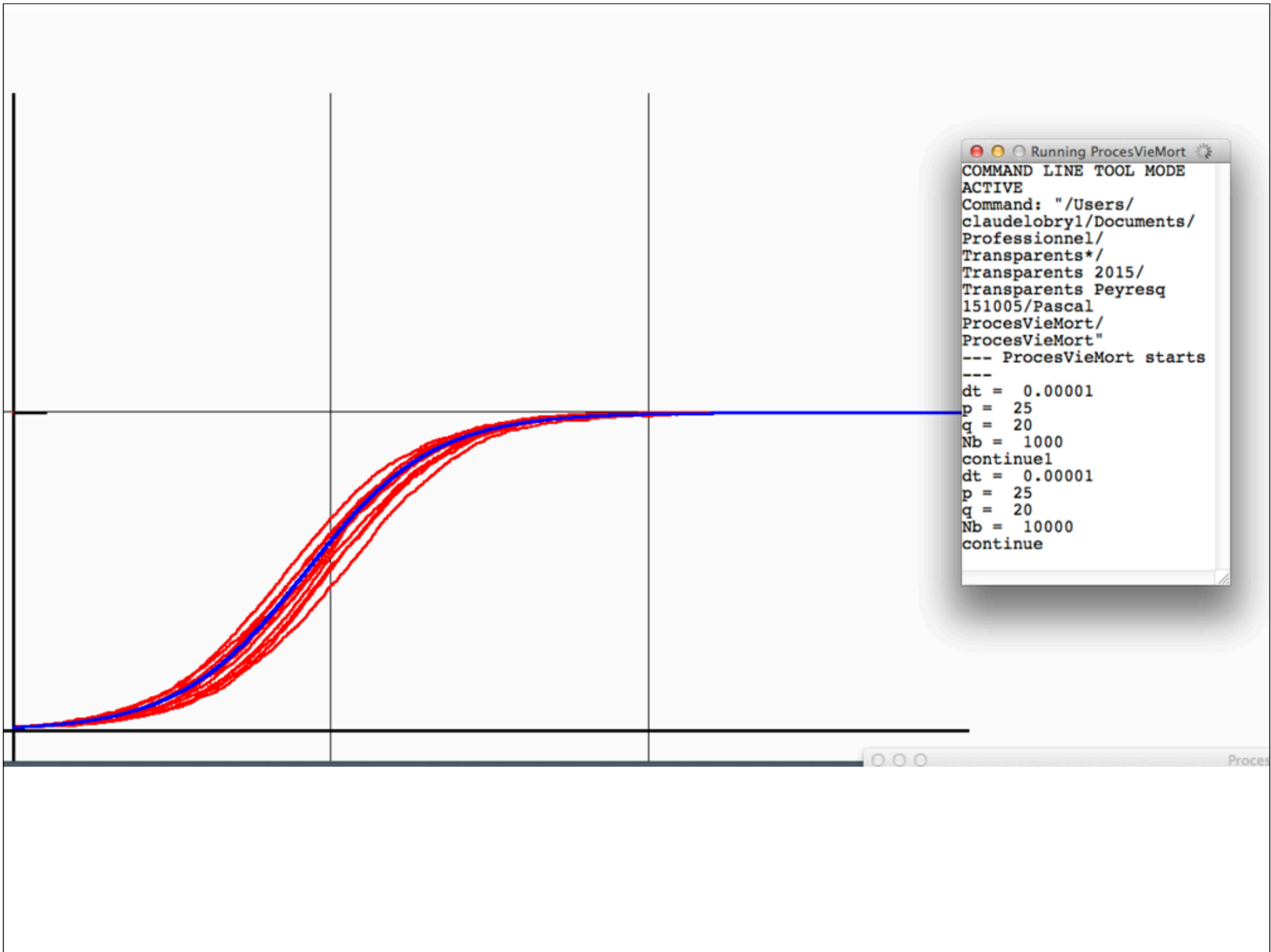






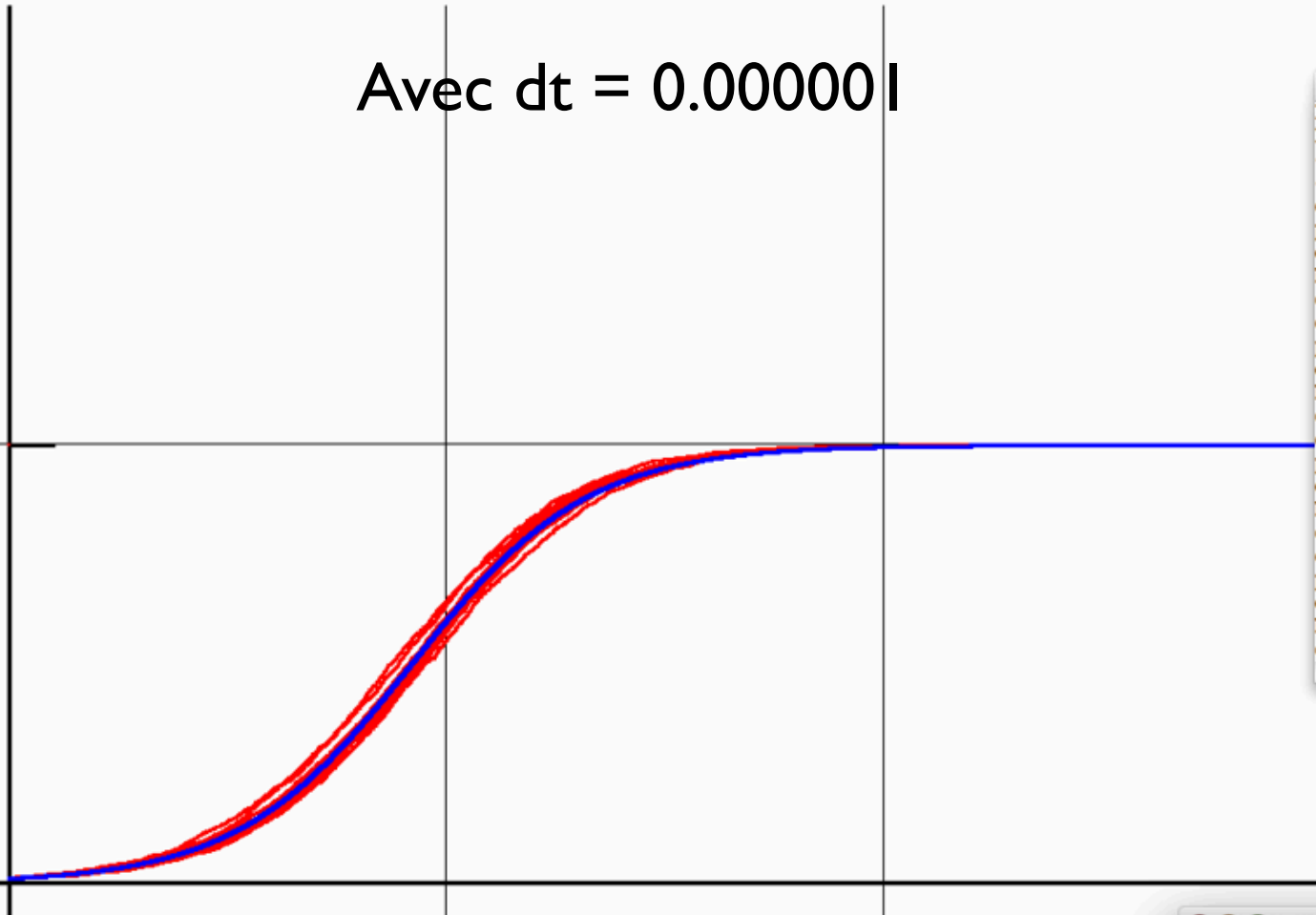






Avec $dt = 0.000001$

```
Running ProcesVieMort/
ProcesVieMort/
ProcesVieMort"
--- ProcesVieMort sta
---
dt = 0.00001
p = 25
q = 20
Nb = 1000
continuel
dt = 0.00001
p = 25
q = 20
Nb = 10000
continuel
dt = 0.00001
p = 25
q = 20
Nb = 20000
continuel
dt = 0.000001
p = 25
q = 20
Nb = 20000
continue
```



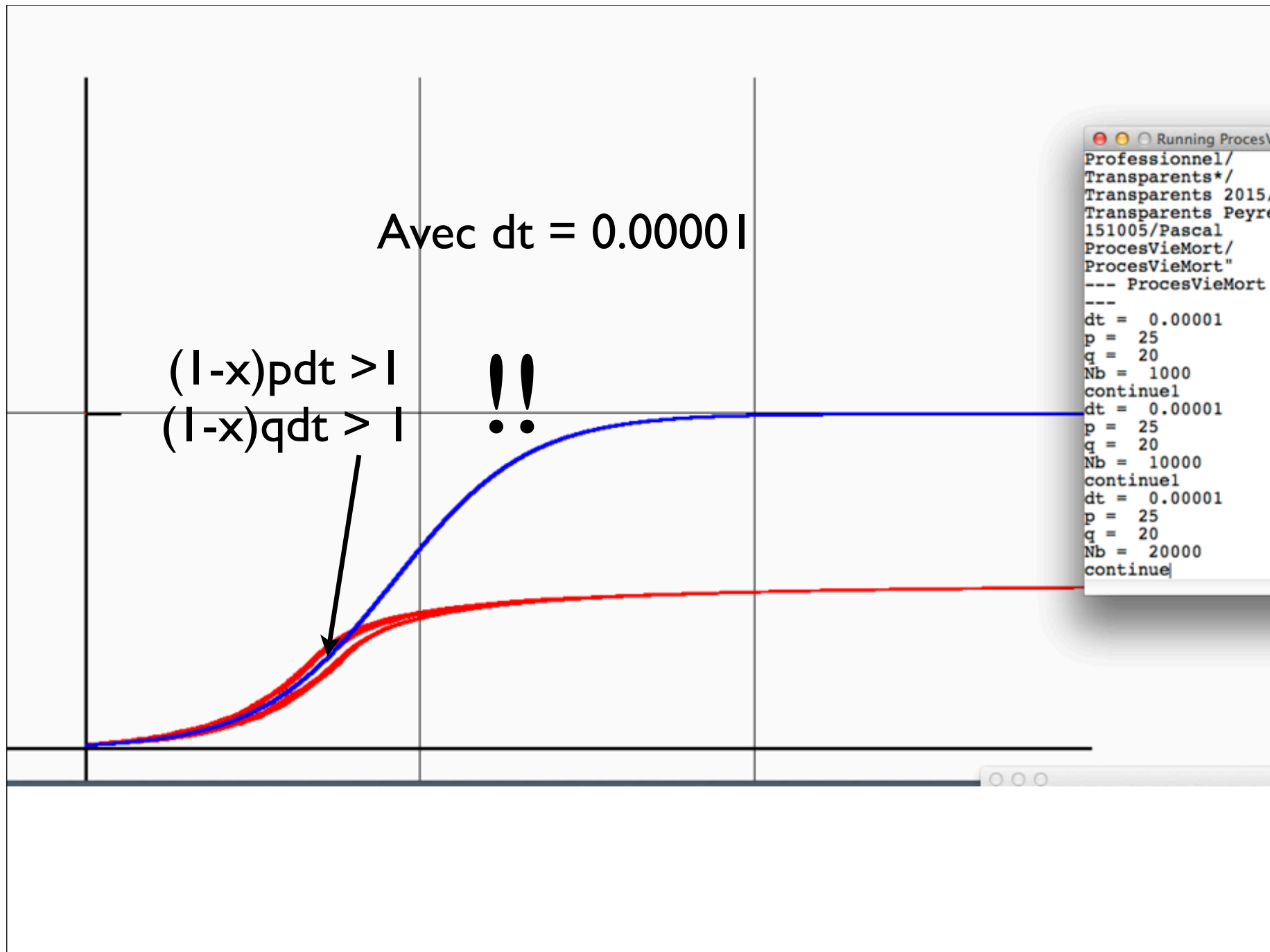
Avec $dt = 0.00001$

$$(1-x)pdt > 1$$

$$(1-x)qdt > 1$$

!!

```
Running ProcesV
Professionnel/
Transparents*/
Transparents 2015,
Transparents Peyre
151005/Pascal
ProcesVieMort/
ProcesVieMort"
--- ProcesVieMort
---
dt = 0.00001
p = 25
q = 20
Nb = 1000
continue|
dt = 0.00001
p = 25
q = 20
Nb = 10000
continue|
dt = 0.00001
p = 25
q = 20
Nb = 20000
continue|
```

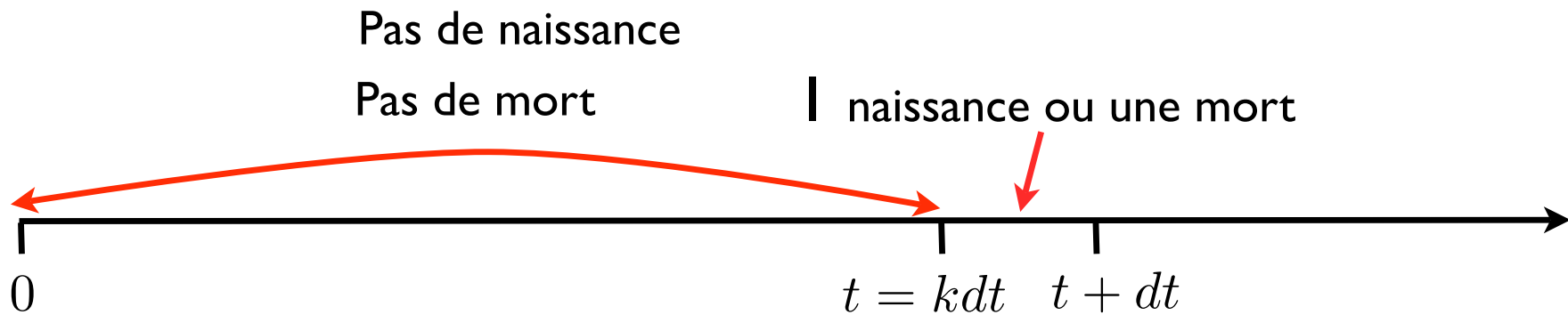


On transforme “vie et mort” en un processus
équivalent dit de
Feller-Gillespie

$$P \left(N(t + dt) = \begin{cases} N(t) + 1 \\ N(t) \\ N(t) - 1 \end{cases} \right) = \begin{aligned} &= N(t)pdt + o(dt) \\ &= 1 - N(t)(p + q)dt + o(dt) \\ &= N(t)qdt + o(dt) \end{aligned}$$

T v.a. “ attente du prochain évènement “

$$\text{Prochain évènement à l'instant } t = P(T \in [t; t + dt])$$



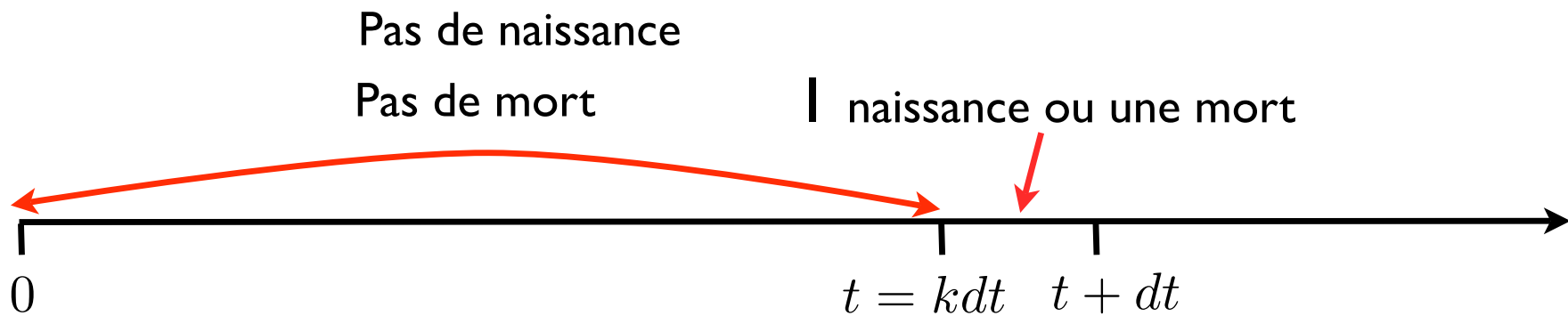
$$P(t \in [t, t + dt]) \approx \left(1 - \frac{N(t)(p + q)dt}{.} \right)^k \cdot N(t)(p + q)dt$$

$$P(T \in [t, t + dt]) = \int_t^{t+dt} N(t)(p + q) e^{-N(t)(p+q)s} ds$$

$$P \left(N(t + dt) = \begin{cases} N(t) + 1 \\ N(t) \\ N(t) - 1 \end{cases} \right) = \begin{aligned} &= N(t)pdt + o(dt) \\ &= 1 - N(t)(p + q)dt + o(dt) \\ &= N(t)qdt + o(dt) \end{aligned}$$

T v.a. “ attente du prochain évènement “

$$\text{Prochain évènement à l'instant } t = P(T \in [t; t + dt])$$



$$P(t \in [t, t + dt]) \approx \left(1 - \frac{N(t)(p + q) \cdot t}{k} \right)^k \cdot N(t)(p + q)dt$$

$$P(T \in [t, t + dt]) \approx \int_t^{t+dt} N(t)(p + q) e^{-N(t)(p+q)s} ds$$

$$P(T \in [t, t + dt]) = \int_t^{t+dt} N(t)(p + q) e^{-N(t)(p+q)s} ds$$

à la limite (ou pour dt i.p.) T suit une loi exponentielle de paramètre

$$\lambda = N(t)(p + q)$$

$$E[T] = \frac{1}{\lambda} = \frac{1}{N(t)(p + q)}$$

$$\sigma^2(t) = \frac{1}{\lambda^2} = \frac{1}{(N(t)(p + q))^2}$$

Processus de Feller-Gillespie

C'est un algorithme :

1. $t \leftarrow 0$;
2. $N \leftarrow N(0)$;
3. On tire T selon une loi exponentielle de paramètre $N(0)(p + q)$;
4. $W \leftarrow +1$ ou -1 selon $p/(p + q)$ et $q/(p + q)$;
5. $t \leftarrow T$;
6. $\leftarrow T + W$
7. On recommence...

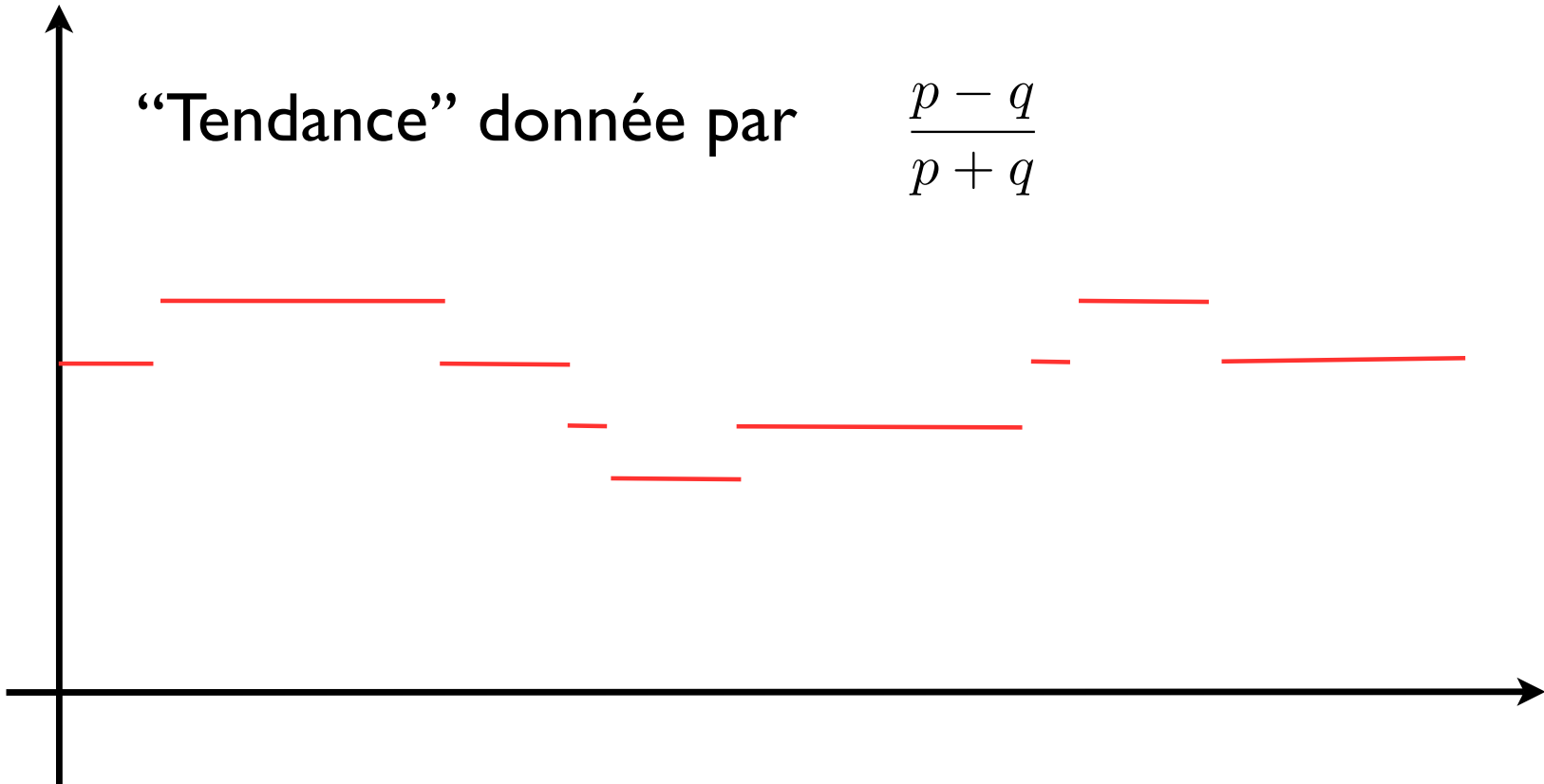
Processus de Feller-Gillespie

Temps moyen entre deux
évènements de l'ordre de
"Fréquence"

$$\frac{1}{N(t)(p+q)}$$
$$\approx N(t)(p+q)$$

"Tendance" donnée par

$$\frac{p-q}{p+q}$$



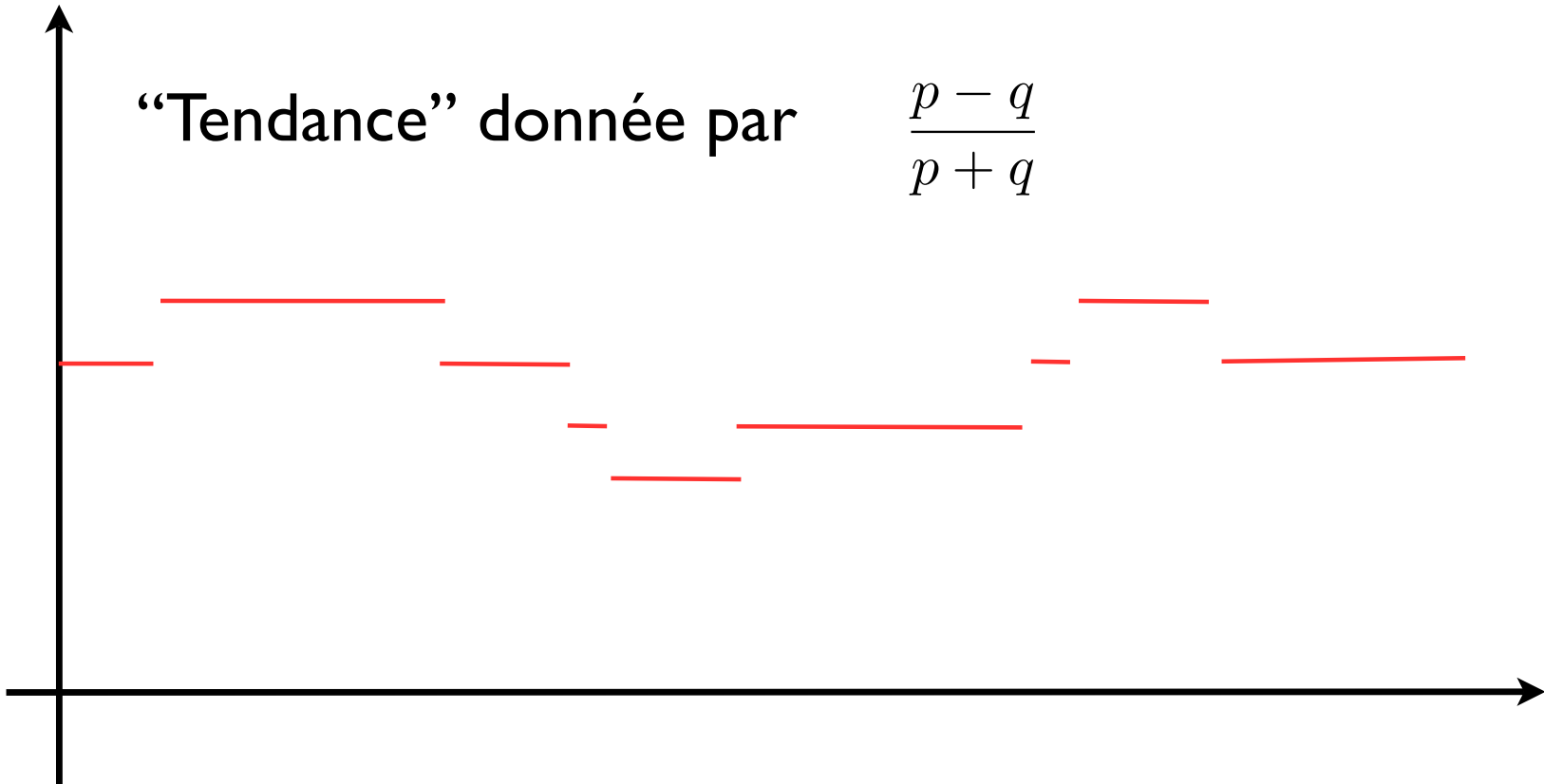
Processus de Feller-Gillespie

Temps moyen entre deux
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$$\frac{1}{N(t)(p+q)}$$
$$\approx N(t)(p+q)$$

"Tendance" donnée par

$$\frac{p-q}{p+q}$$



Processus de Feller-Gillespie

Rien n'empêche p et q de dépendre de $N(t)$ et t

C'est un algorithme : On se donne $p(N)$ et $q(N)$

1. $t \leftarrow 0$;

$$(N(0)(p(N(0), 0)q(N(0), 0))$$

2. $N \leftarrow N(0)$;

3. On tire T selon une loi exponentielle de paramètre ~~$N(0)(p + q)$~~ ;

4. $W \leftarrow +1$ ou -1 selon ~~$p/(p + q)$~~ et $q/(p + q)$;

5. $t \leftarrow T$;

6. $\leftarrow T + W$

$$\frac{p(N(0), 0)}{p(N(0), 0) + q(N(0), 0)}$$

7. On recommence...

Approximation diffusion

Pour $N(t)$ grand le temps de simulation est trop grand

1. $x(t) = \frac{N(t)}{\omega}$: Une unité de x représente ω individus.
2. On se donne $p = p(x, t)$; $q = q(x, t)$
3. On évalue l'accroissement de x sur une durée Δt telle que

$$N(t)(p + q)\delta t = \omega x(t)\Delta t$$

soit grand (de l'ordre de 10^3).

1. $\#$ = “nombre (aléatoire) d'évènements sur” $[t, t + \Delta t[$
2. $x(t)$ est approximativement constant sur $[t, t + \Delta t[$ si Δt n'est pas trop grand.
3. $N(t + \Delta t) - N(t) = \sum_{i=1}^{\#} W_i$; $P(W_i = 1) = \frac{p}{p+q}$; $P(W_i = -1) = \frac{q}{p+q}$
4. comme $\#$ est grand $W = \sum_{i=1}^{\#} W_i$ est approximativement une gaussienne.
5. $E[W] = \# \frac{p-q}{p+q}$; $\sigma^2(W) = \# \frac{4pq}{(p+q)^2} \implies W \approx \# \frac{p-q}{p+q} + \sqrt{\#} \sqrt{\frac{4pq}{(p+q)^2}} \mathcal{N}(0, 1)$
6. “en moyenne” $\# = N(t)(p+q)\Delta t = \omega x(t)(p+q)\Delta t$
7. $x(t + \Delta t) - x(t) = x(t)(p-q)\Delta t + \frac{\sqrt{\Delta t}}{\sqrt{\omega}} \sqrt{\frac{4x(t)p(x(t),t)q(x(t),t)}{p(x(t),t)+q(x(t),t)}} \mathcal{N}(0, 1)$

- $x(t + \Delta t) - x(t) = x(t)(p - q)\Delta t + \frac{\sqrt{\Delta t}}{\sqrt{\omega}} \sqrt{\frac{4x(t)p(x(t),t)q(x(t),t)}{p(x(t),t)+q(x(t),t)}} \mathcal{N}(0, 1)$
- $x(t + \Delta t) = x(t) + f(x(t))\Delta t + \frac{\sqrt{\Delta t}}{\sqrt{\omega}} g(x(t))\mathcal{N}(0, 1)$

Processus de diffusion

Schéma d'Euler + (petit) bruit en $\sqrt{\Delta t}$

Processus de diffusion

Exemple de processus
de diffusion

$$x(t + dt) = x(t) + dt f(x(t)) + \sigma(x(t)) \sqrt{dt} \mathcal{N}(0, 1)$$

Pour respecter la positivité :

$$x(t + dt) = \max \left\{ x(t) + dt f(x(t)) + \sigma(x(t)) \sqrt{dt} \mathcal{N}(0, 1) ; 0 \right\}$$

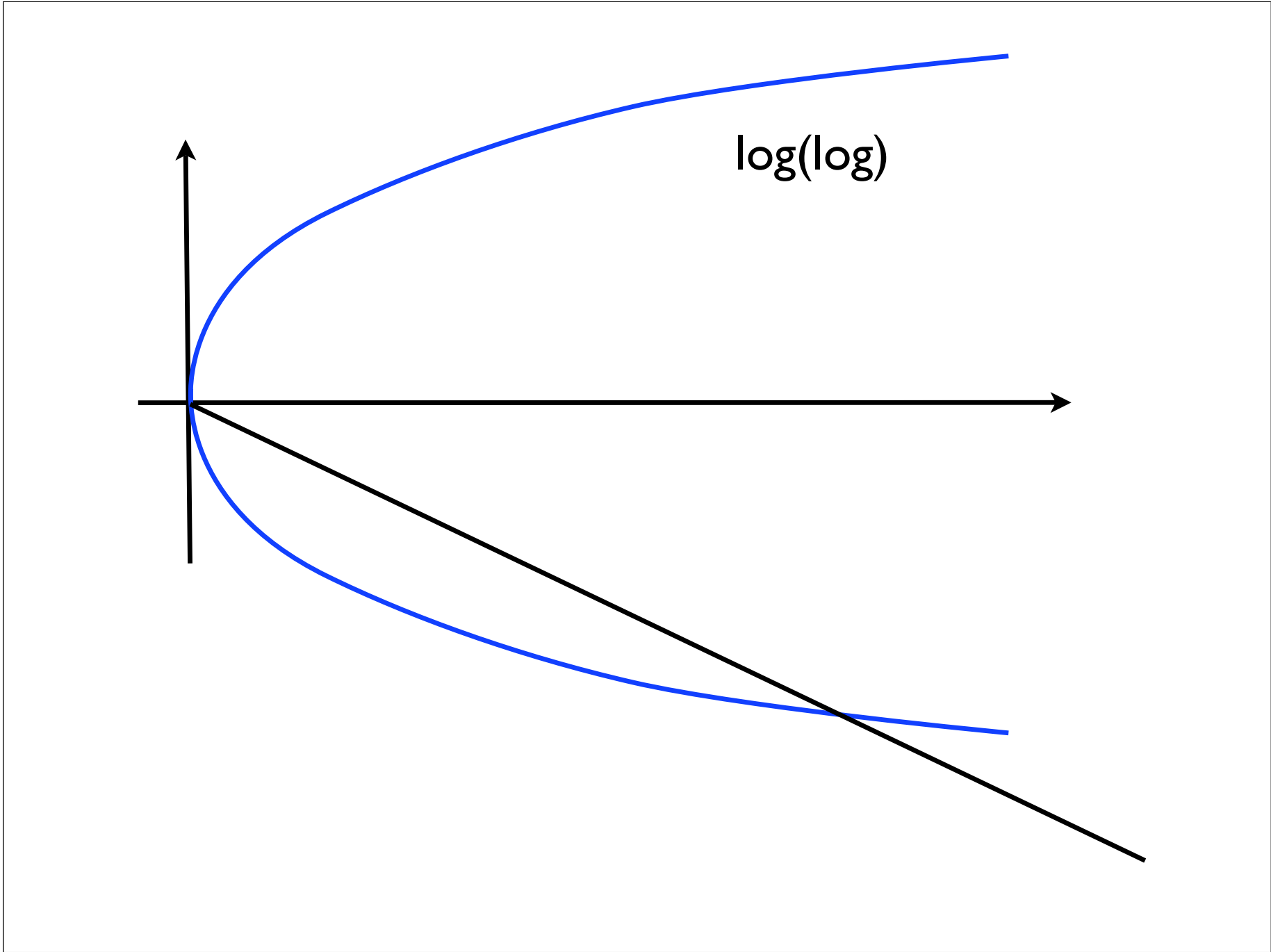
La croissance exponentielle “diffusion”

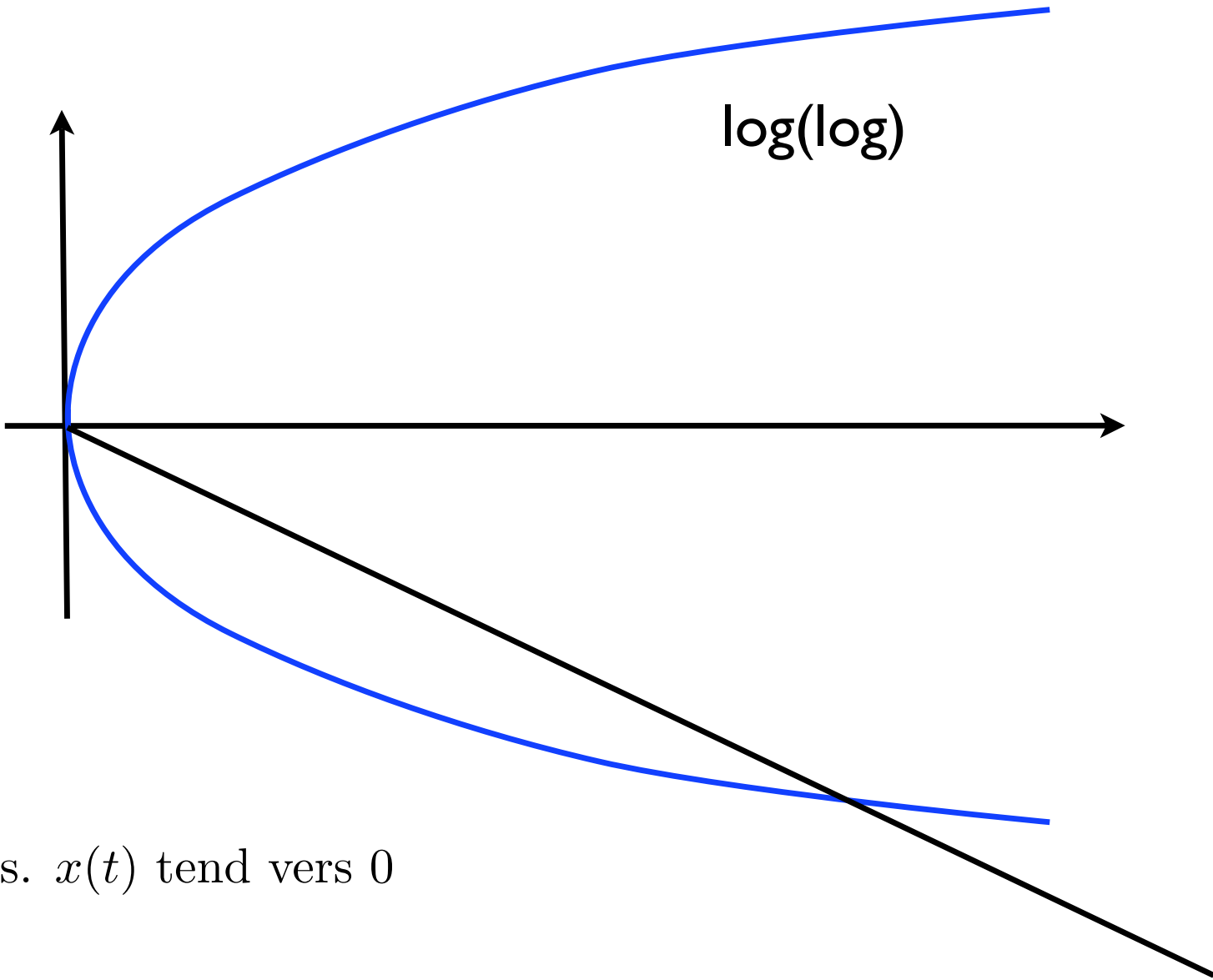
$$x(t + dt) = x(t) + dt r x(t) + (\pm) \sigma x(t) \sqrt{dt}$$

$$y = \ln(x)$$

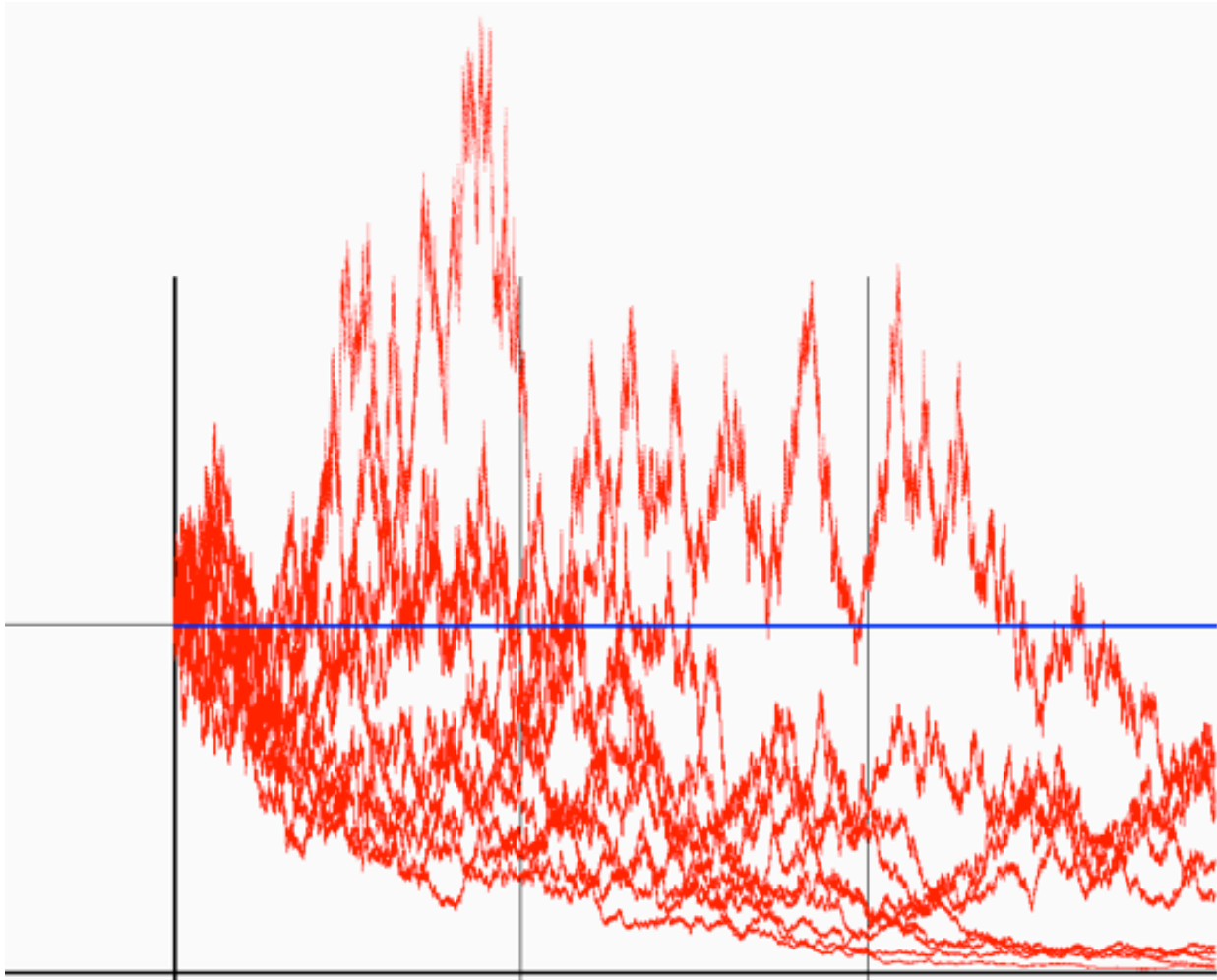
$$y(t + dt) = y(t) + (r dt + (\pm) \sigma \sqrt{dt}) - \frac{1}{2} (r dt + (\pm) \sigma \sqrt{dt})^2$$

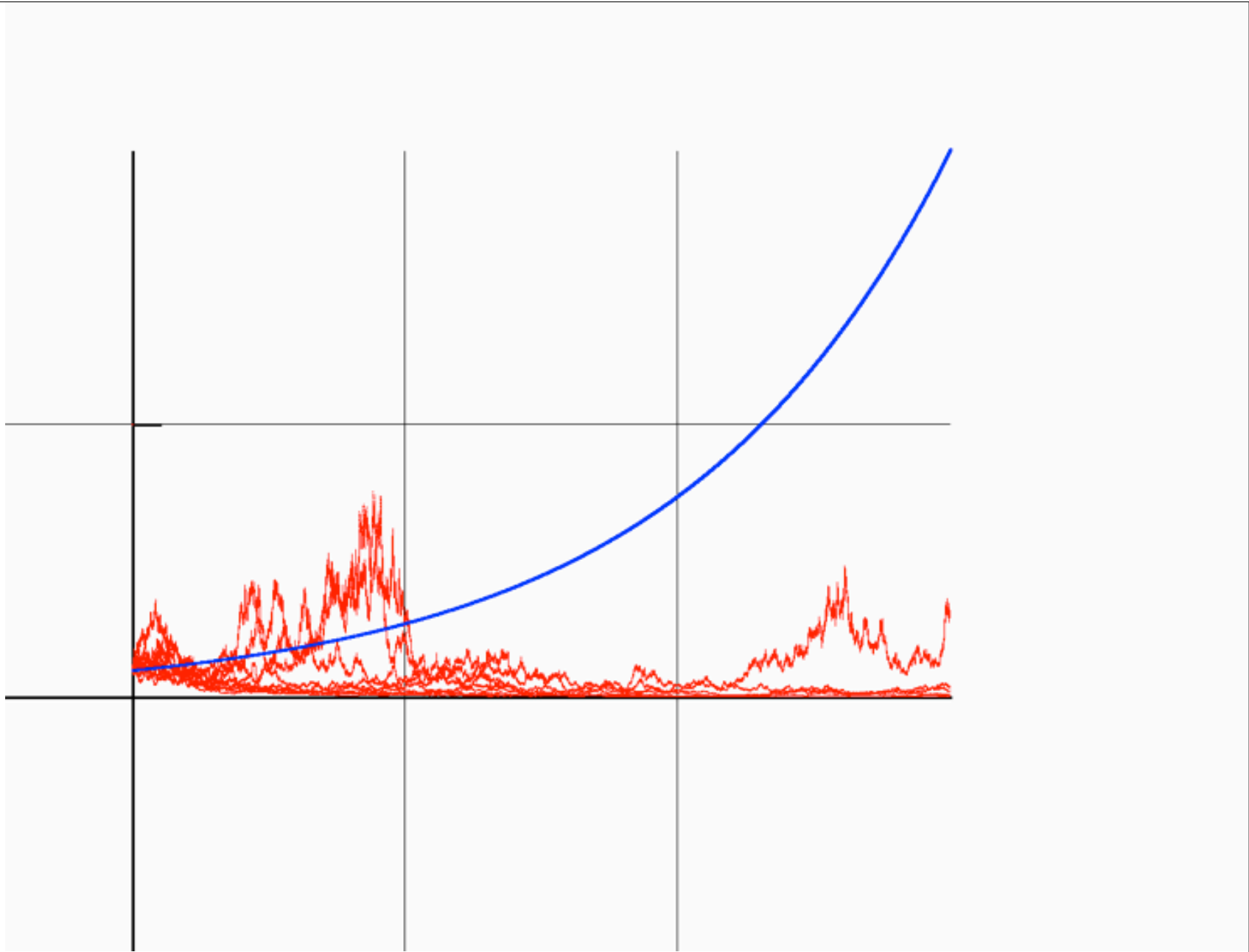
$$y(t + dt) = y(t) + (r - \sigma^2 / 2) dt + (\pm) \sigma \sqrt{dt} + \dots$$





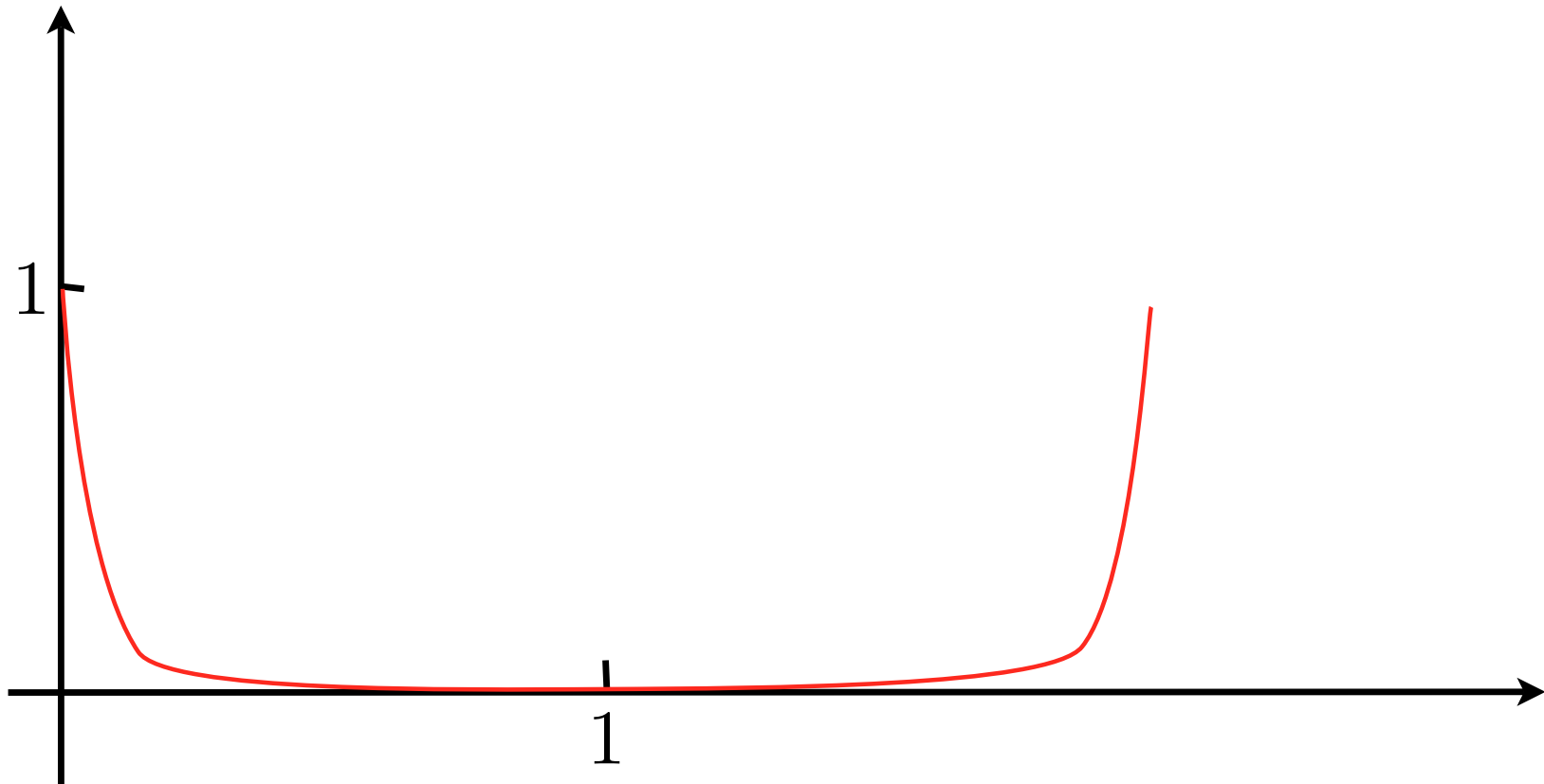
p.s. $x(t)$ tend vers 0

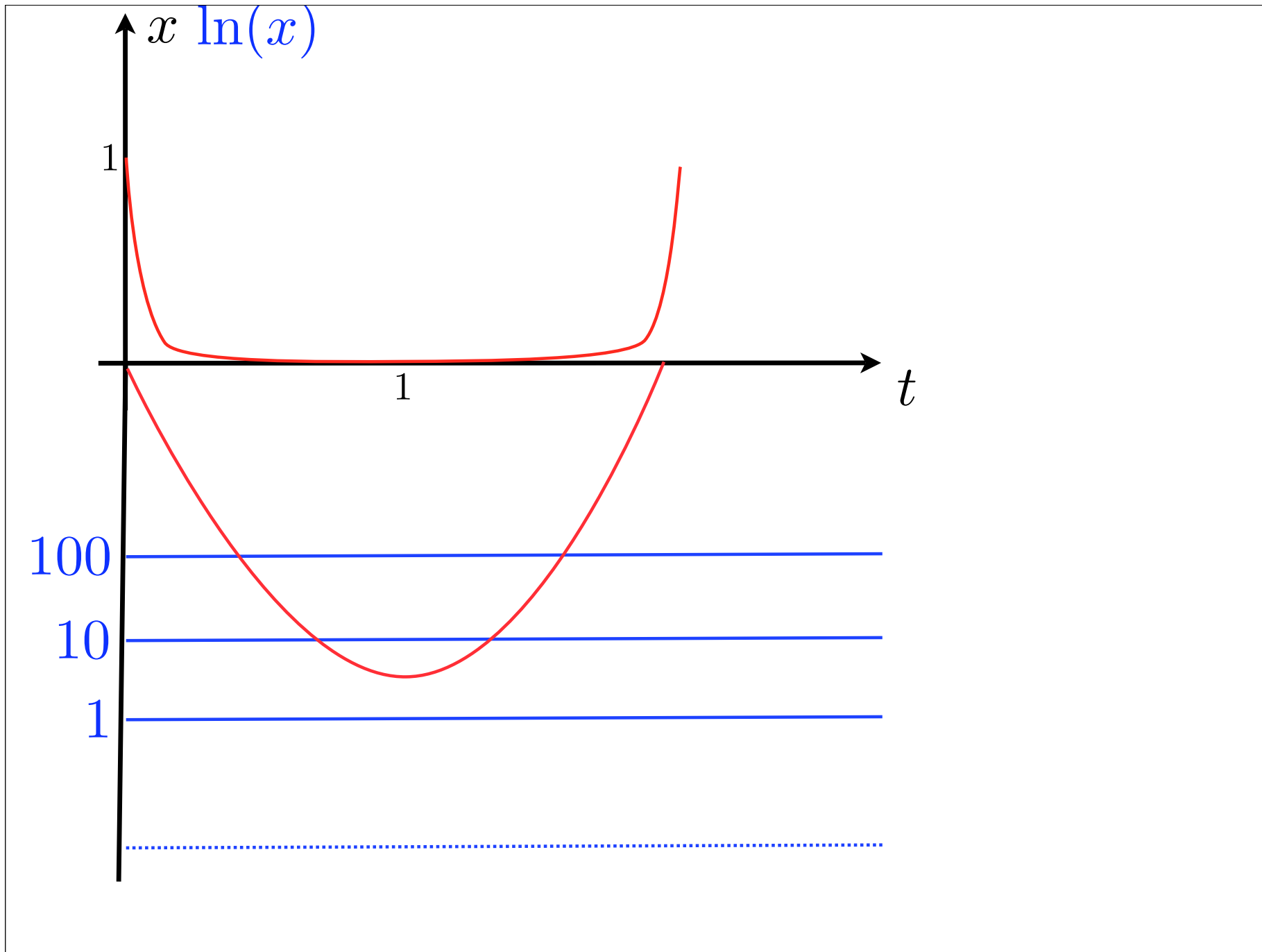


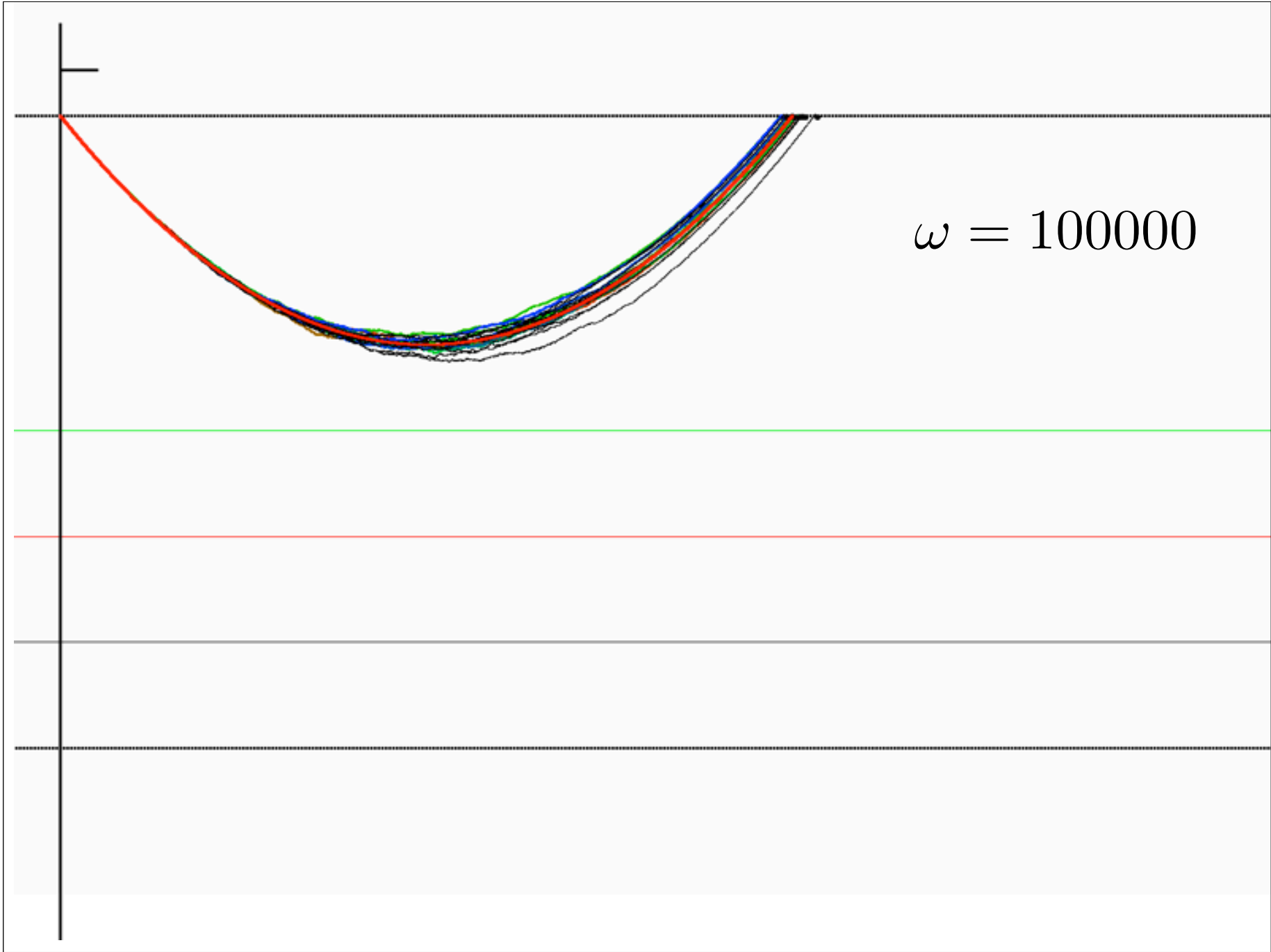


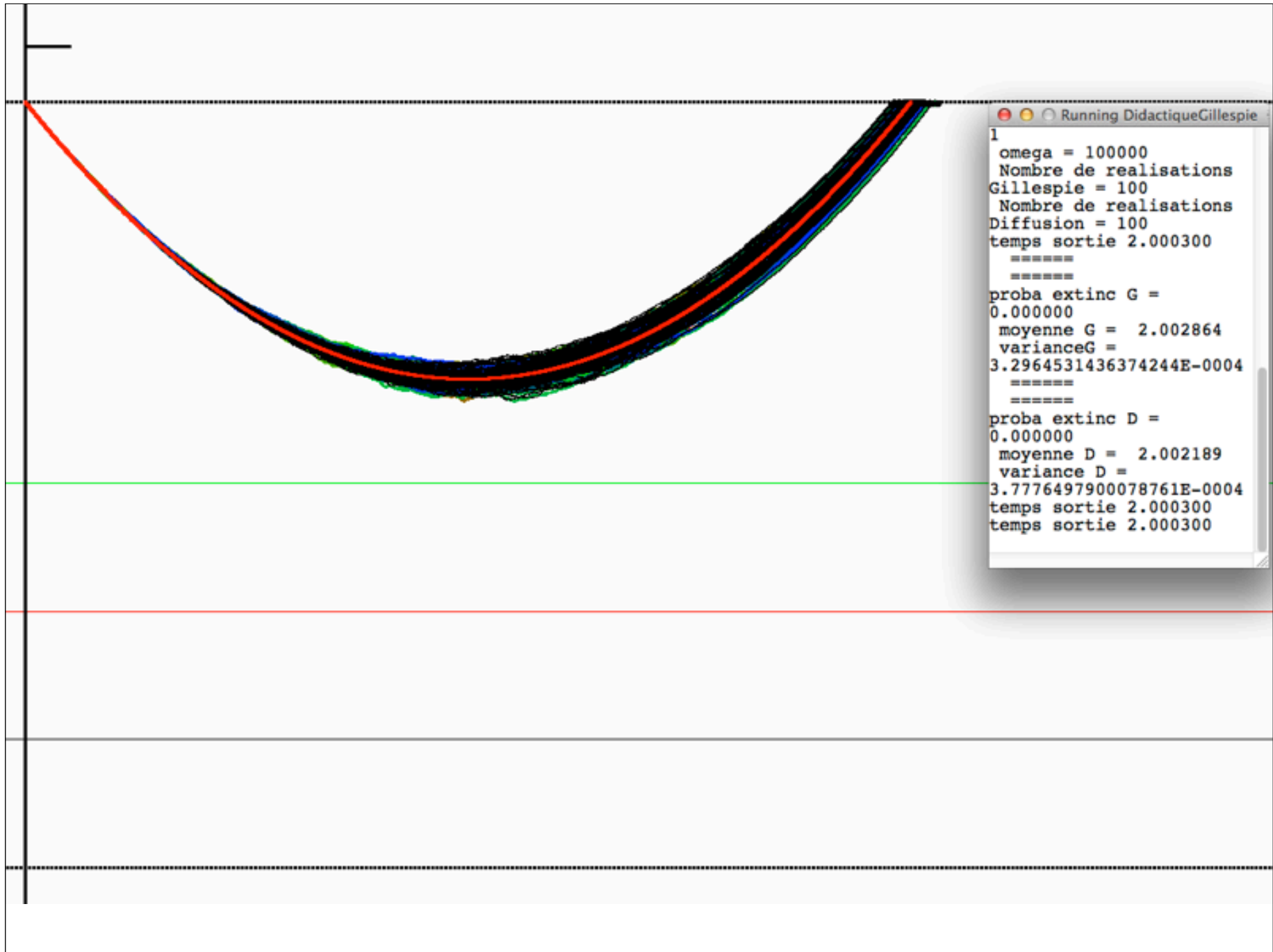
Expériences numériques

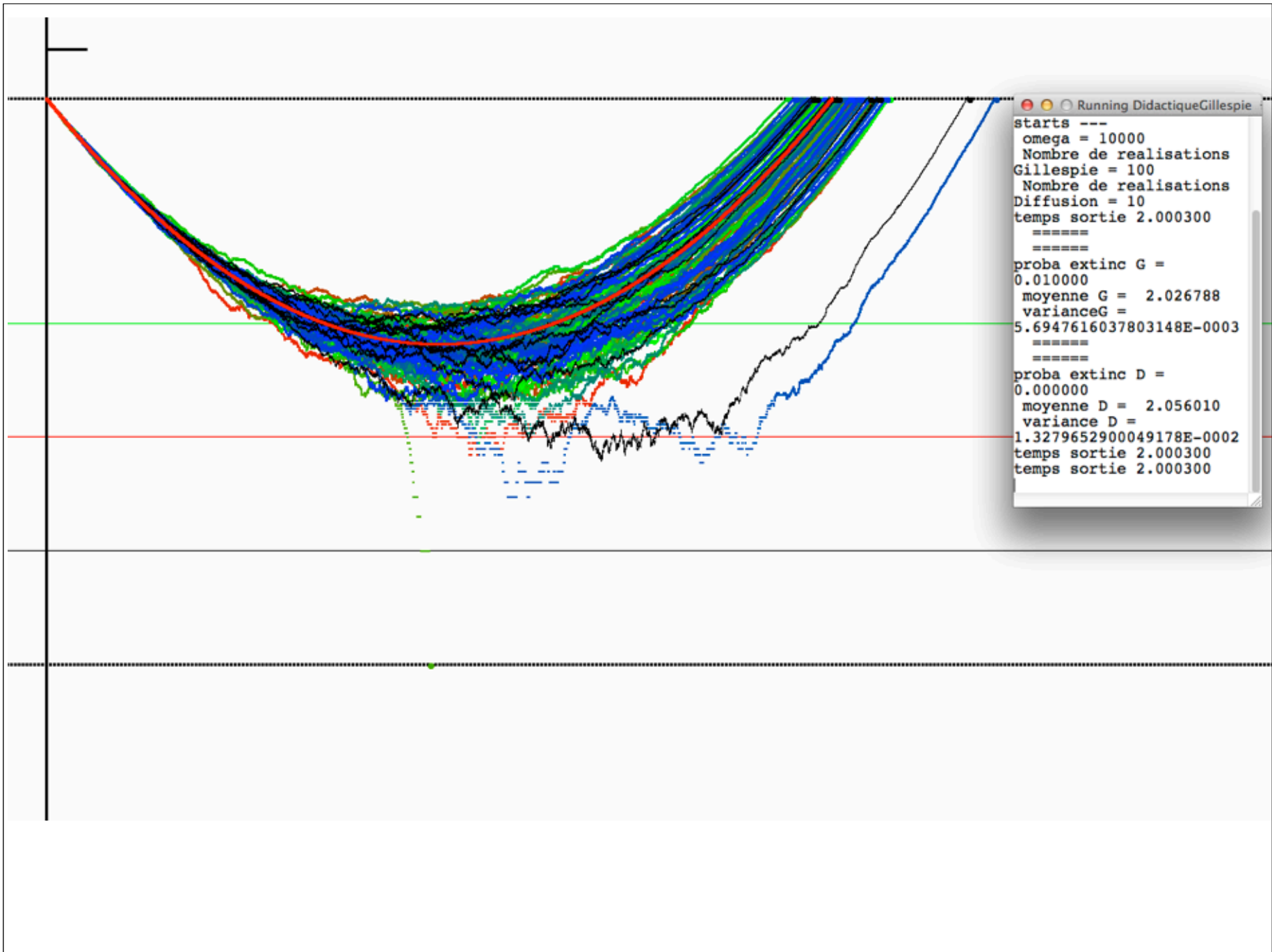
- $p(t) = 10(1 + t/2)$; $q(t) = 10(2 - t/2)$
- Approximation déterministe : $\dot{x}(t) = 10(-1 + t)x(t)$

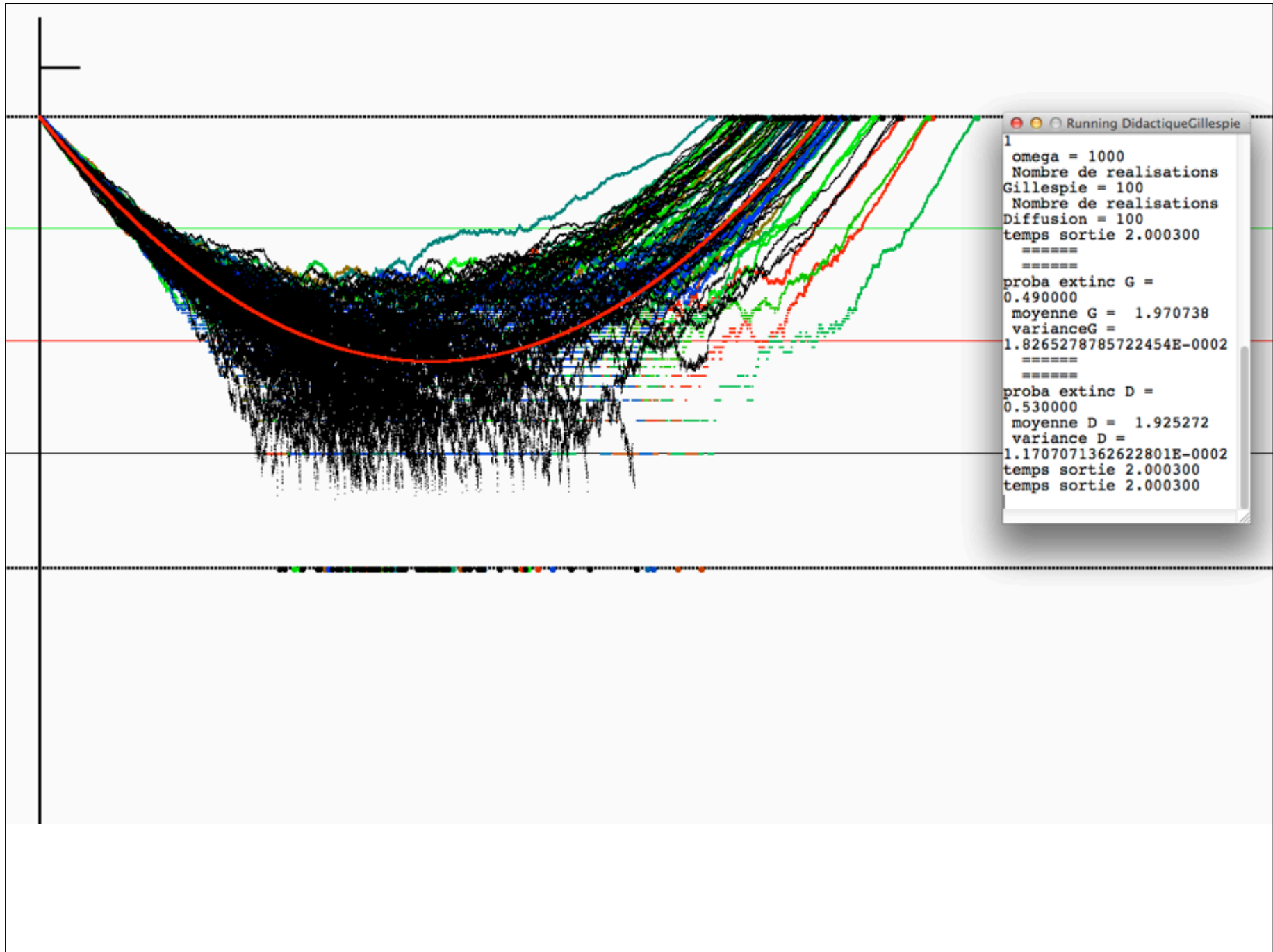


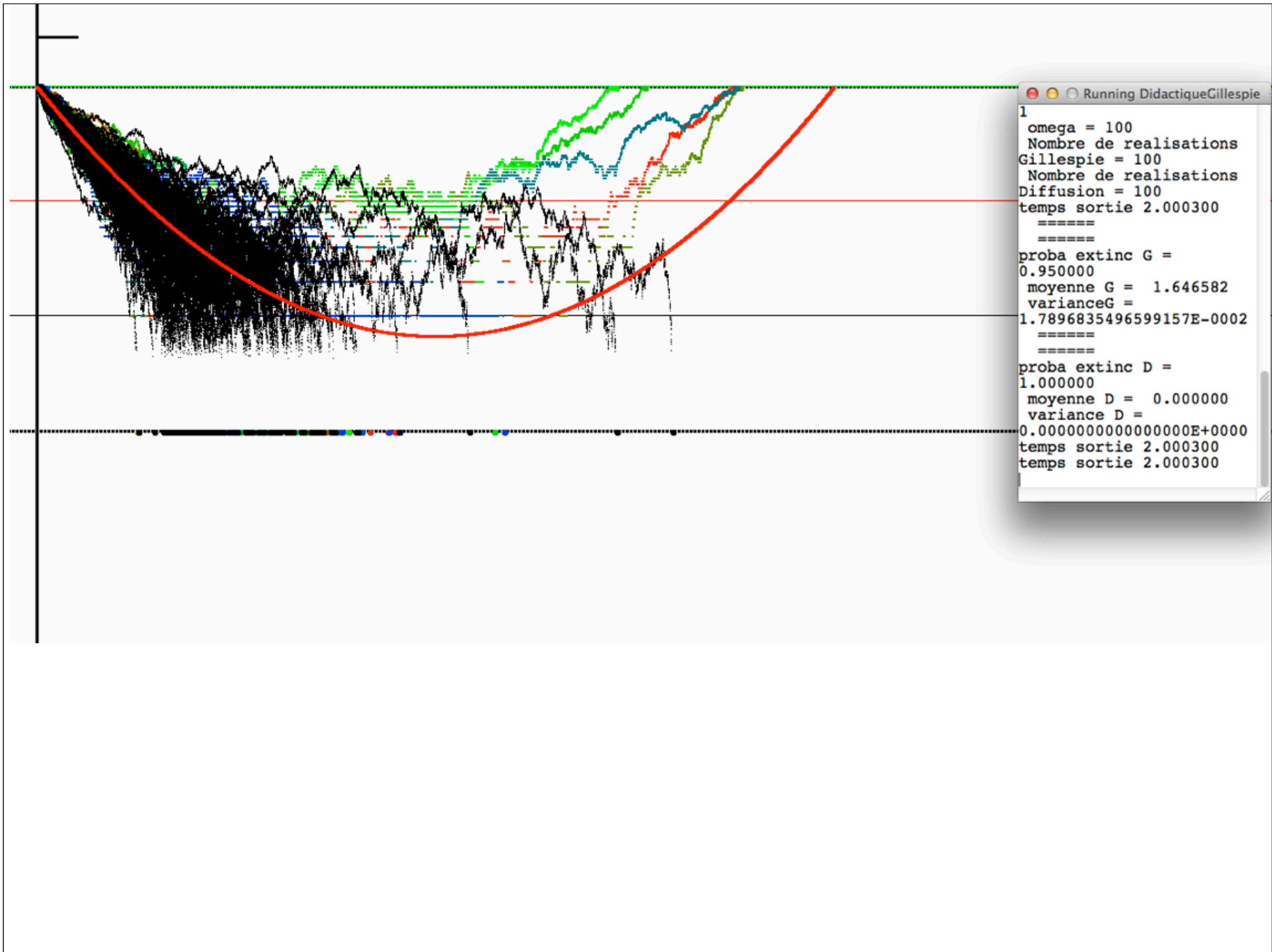


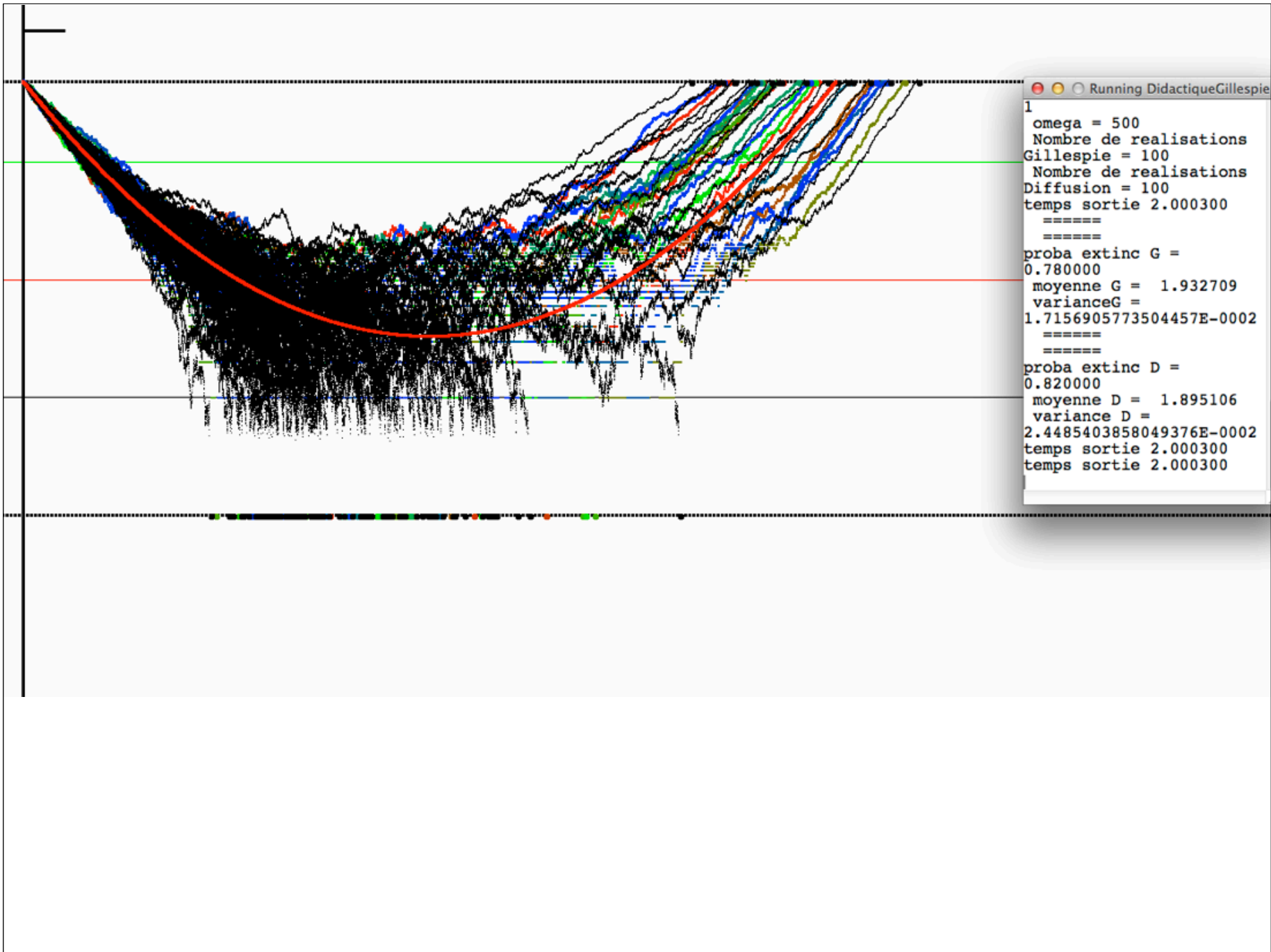


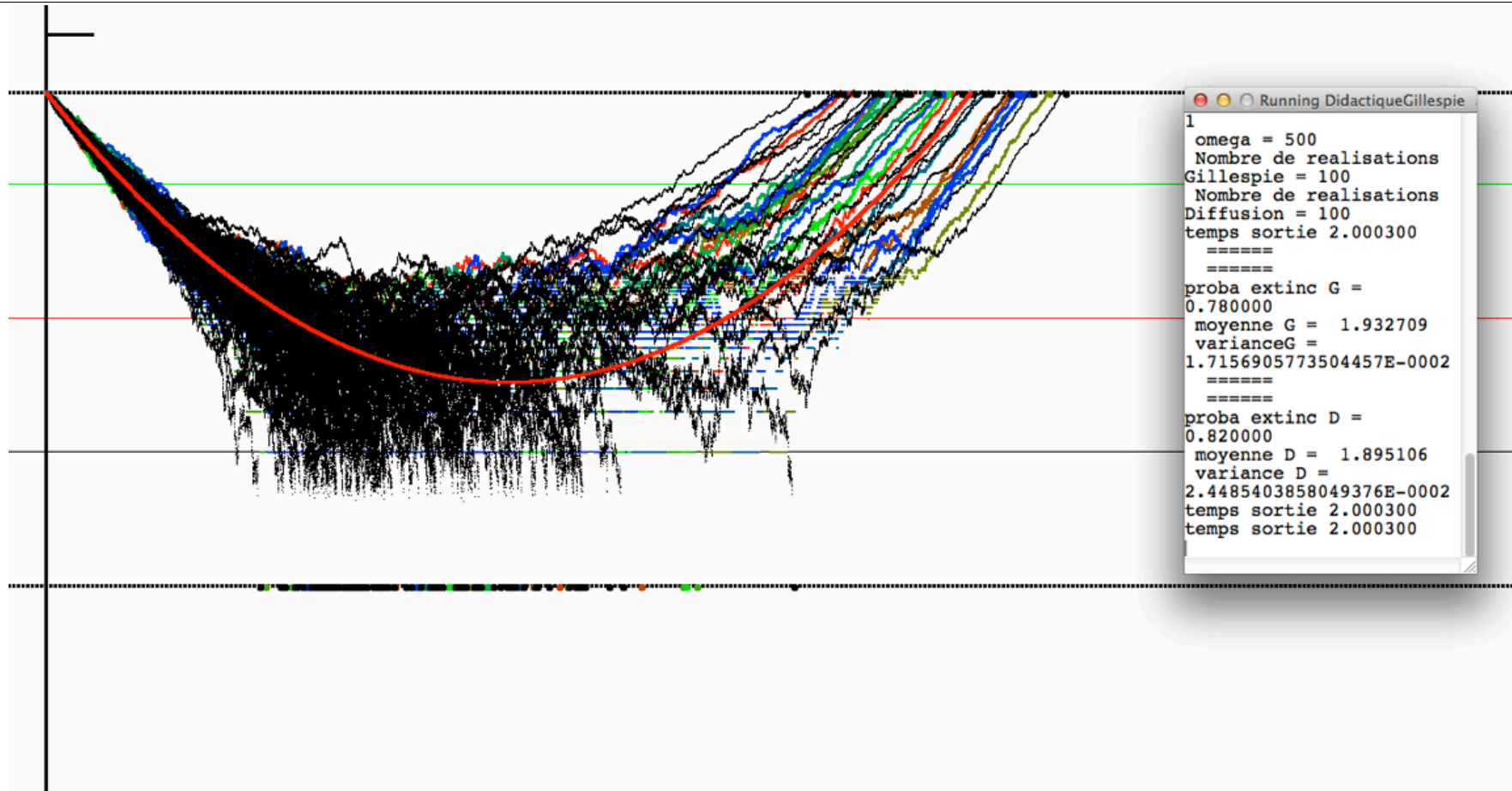




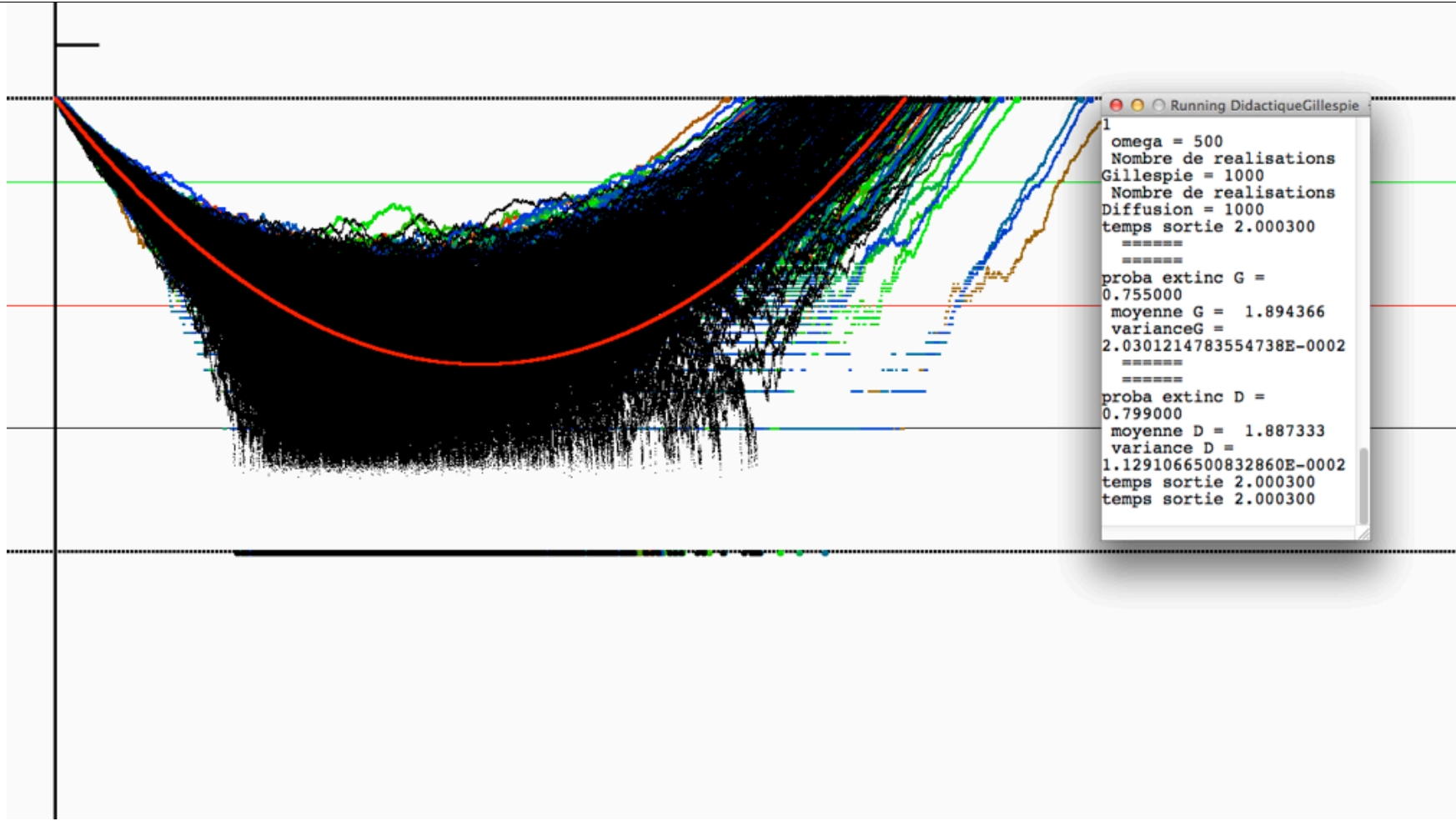




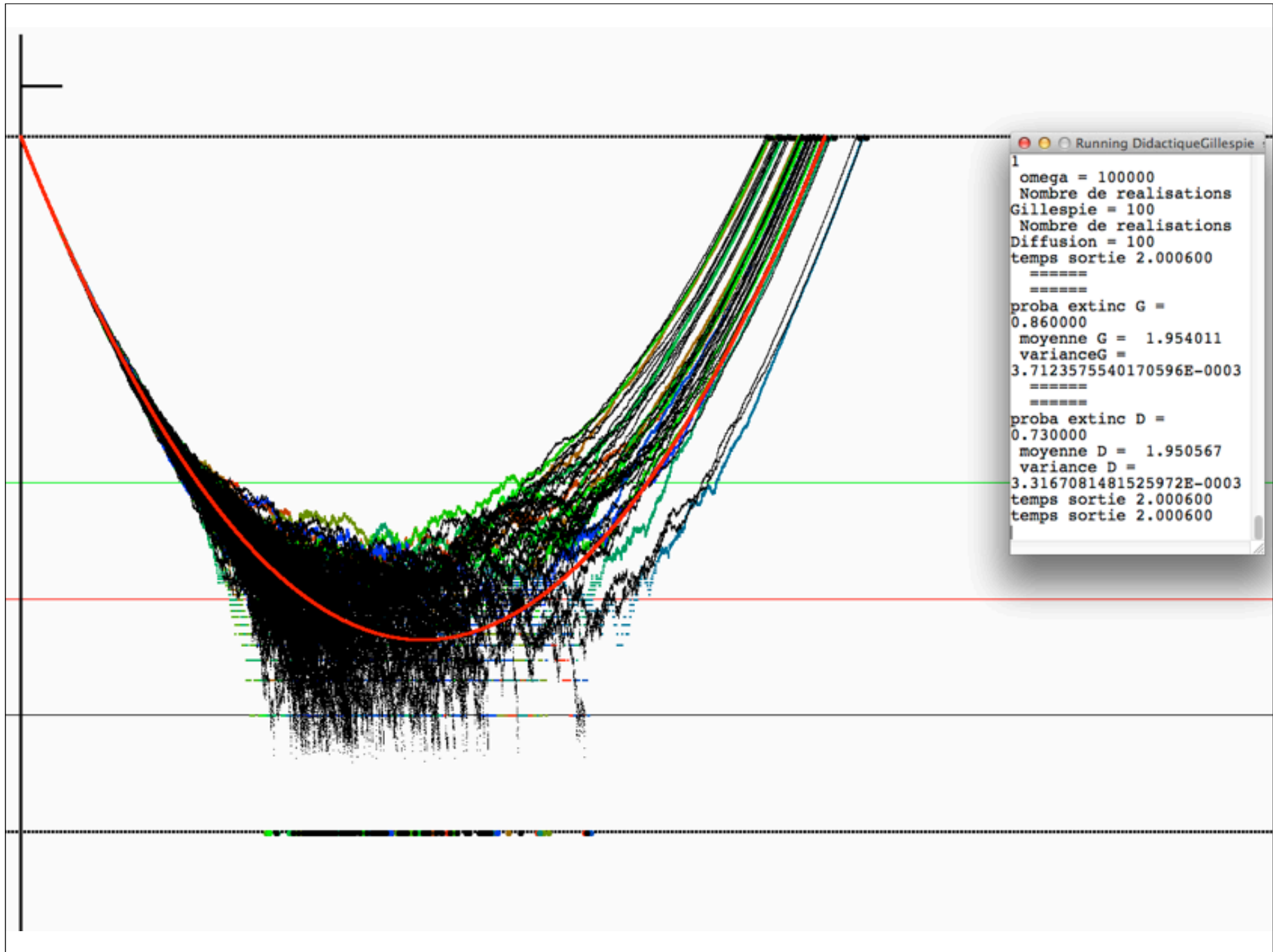


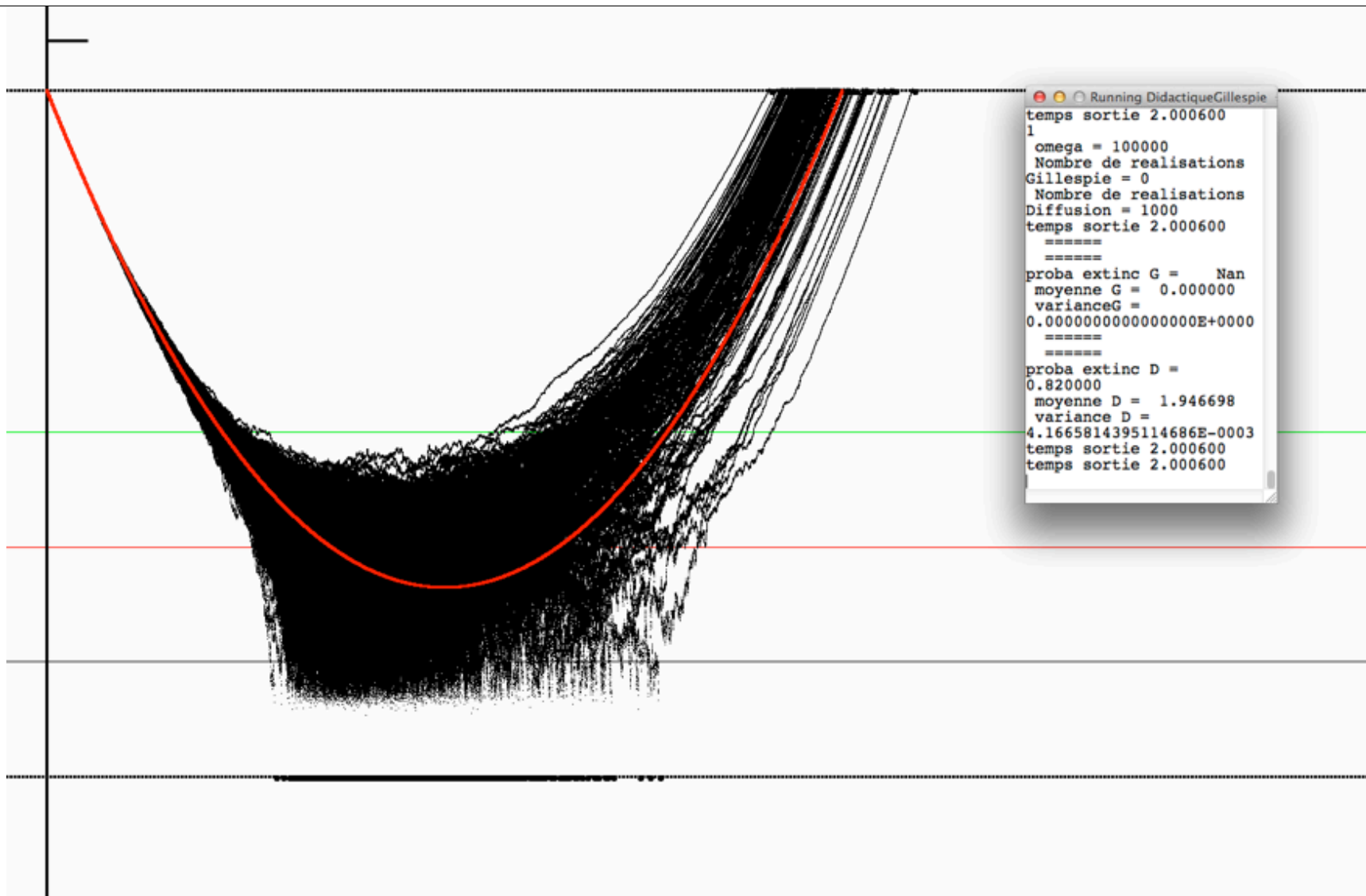


```
Running DidactiqueGillespie
1
omega = 500
Nombre de realisations
Gillespie = 100
Nombre de realisations
Diffusion = 100
temps sortie 2.000300
=====
proba extinc G =
0.780000
moyenne G = 1.932709
varianceG =
1.7156905773504457E-0002
=====
proba extinc D =
0.820000
moyenne D = 1.895106
variance D =
2.4485403858049376E-0002
temps sortie 2.000300
temps sortie 2.000300
```

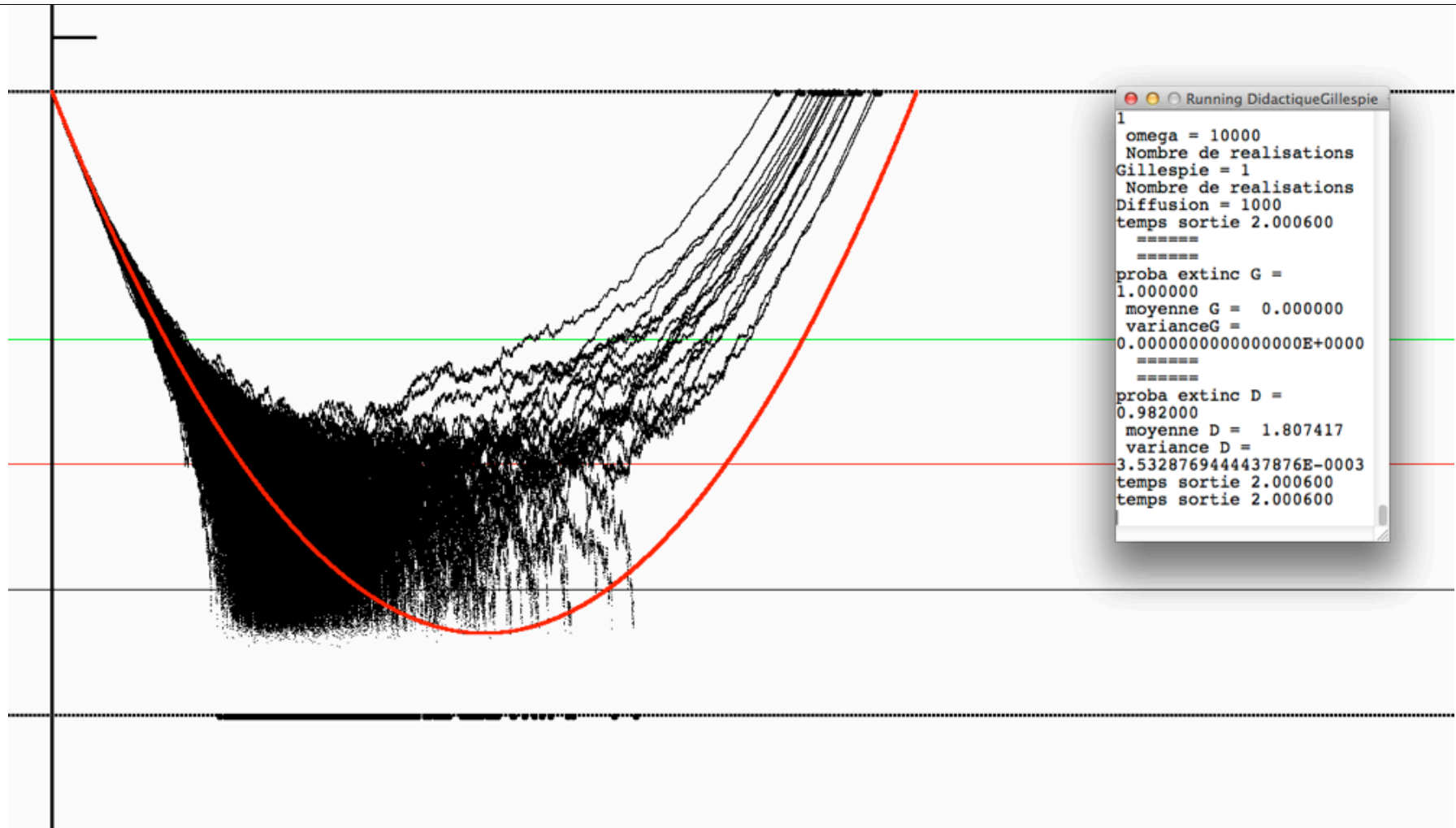



```
Running DidactiqueGillespie
1
omega = 500
Nombre de realisations
Gillespie = 1000
Nombre de realisations
Diffusion = 1000
temps sortie 2.000300
=====
proba extinc G =
0.755000
moyenne G = 1.894366
varianceG =
2.0301214783554738E-0002
=====
proba extinc D =
0.799000
moyenne D = 1.887333
variance D =
1.1291066500832860E-0002
temps sortie 2.000300
temps sortie 2.000300
```

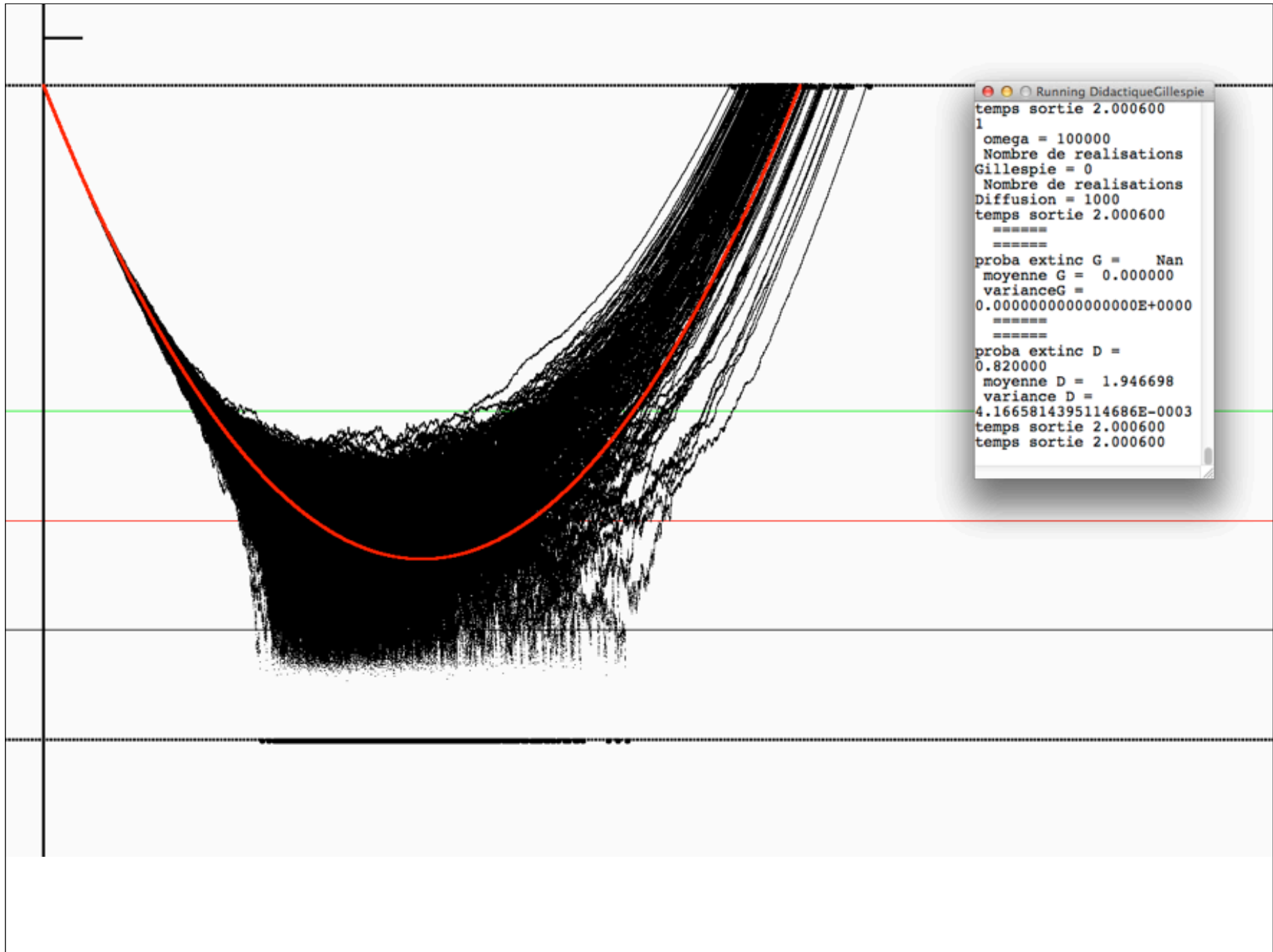




```
Running DidactiqueGillespie
temps sortie 2.000600
1
omega = 100000
Nombre de realisations
Gillespie = 0
Nombre de realisations
Diffusion = 1000
temps sortie 2.000600
=====
proba extinc G = Nan
moyenne G = 0.000000
varianceG =
0.0000000000000000E+0000
=====
proba extinc D =
0.820000
moyenne D = 1.946698
variance D =
4.1665814395114686E-0003
temps sortie 2.000600
temps sortie 2.000600
```



```
Running DidactiqueGillespie
1
omega = 10000
Nombre de realisations
Gillespie = 1
Nombre de realisations
Diffusion = 1000
temps sortie 2.000600
=====
proba extinc G =
1.000000
moyenne G = 0.000000
varianceG =
0.000000000000000000E+0000
=====
proba extinc D =
0.982000
moyenne D = 1.807417
variance D =
3.5328769444437876E-0003
temps sortie 2.000600
temps sortie 2.000600
```



Un modèle proie-prédateur

$$\begin{cases} \frac{dx}{dt} = \frac{1}{\varepsilon}[f(x) - \mu(x)y] \\ \frac{dy}{dt} = (\mu(x) - m)y \end{cases} \quad \begin{array}{l} - f(x) = \frac{1}{2}x(2-x) \\ - \mu(x) = \frac{x}{0.4+x} \\ - \varepsilon = 0.02 \end{array}$$

$$\begin{cases} \frac{dx}{dt} = \frac{1}{\varepsilon}[f(x) - \mu(x)y] \\ \frac{dy}{dt} = (\mu(x) - m)y \end{cases} \quad \begin{aligned} - f(x) &= \frac{1}{2}x(2-x) \\ - \mu(x) &= \frac{x}{0.4+x} \\ - \varepsilon &= 0.02 \end{aligned}$$

$$x(t+dt) - x(t) \approx dt \frac{1}{\varepsilon} [f(x(t)) - \mu(x(t))y(t)] + \sqrt{dt \frac{4}{\omega \varepsilon} \frac{f(x(t))\mu(x(t))y(t)}{(f(x(t)) + \mu(x(t))y(t))}} W_t$$

$$y(t+dt) - y(t) \approx dt [\mu(x(t)) - m]y(t) - \sqrt{dt \frac{\varepsilon}{\omega} \frac{f(x(t))\mu(x(t))y(t)}{(f(x(t)) + \mu(x(t))y(t))}} W_t$$

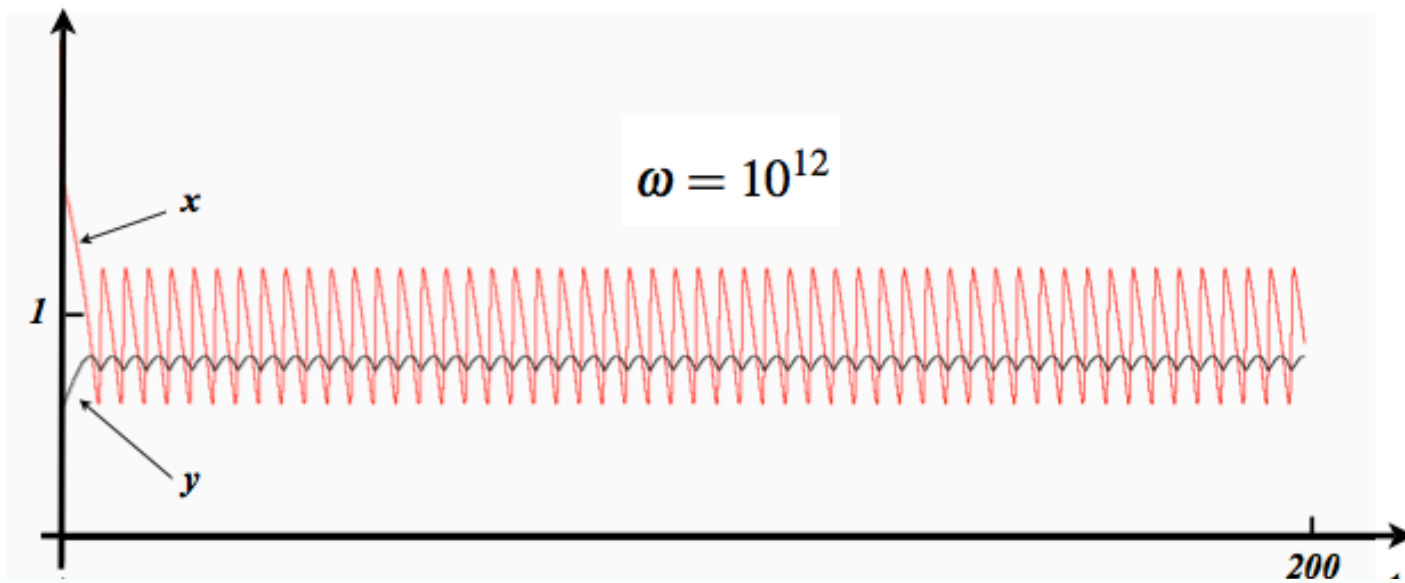
$$\begin{cases} \frac{dx}{dt} = \frac{1}{\varepsilon}[f(x) - \mu(x)y] \\ \frac{dy}{dt} = (\mu(x) - m)y \end{cases} \quad \begin{aligned} - f(x) &= \frac{1}{2}x(2-x) \\ - \mu(x) &= \frac{x}{0.4+x} \\ - \varepsilon &= 0.02 \end{aligned}$$

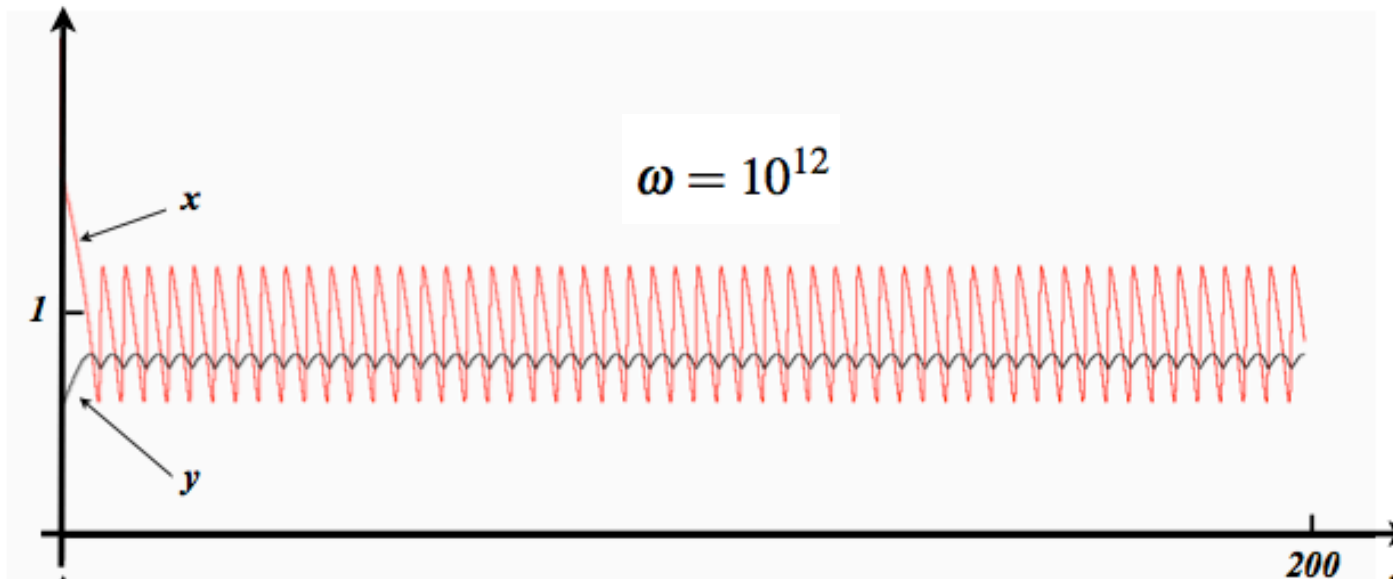
$$x(t+dt) - x(t) \approx dt \frac{1}{\varepsilon} [f(x(t)) - \mu(x(t))y(t)] + \sqrt{dt \frac{4}{\omega \varepsilon} \frac{f(x(t))\mu(x(t))y(t)}{(f(x(t)) + \mu(x(t))y(t))}} W_t$$

$$y(t+dt) - y(t) \approx dt [\mu(x(t)) - m]y(t) - \sqrt{dt \frac{\varepsilon}{\omega} \frac{f(x(t))\mu(x(t))y(t)}{(f(x(t)) + \mu(x(t))y(t))}} W_t$$

1 unité de x = ω individus

Simulations

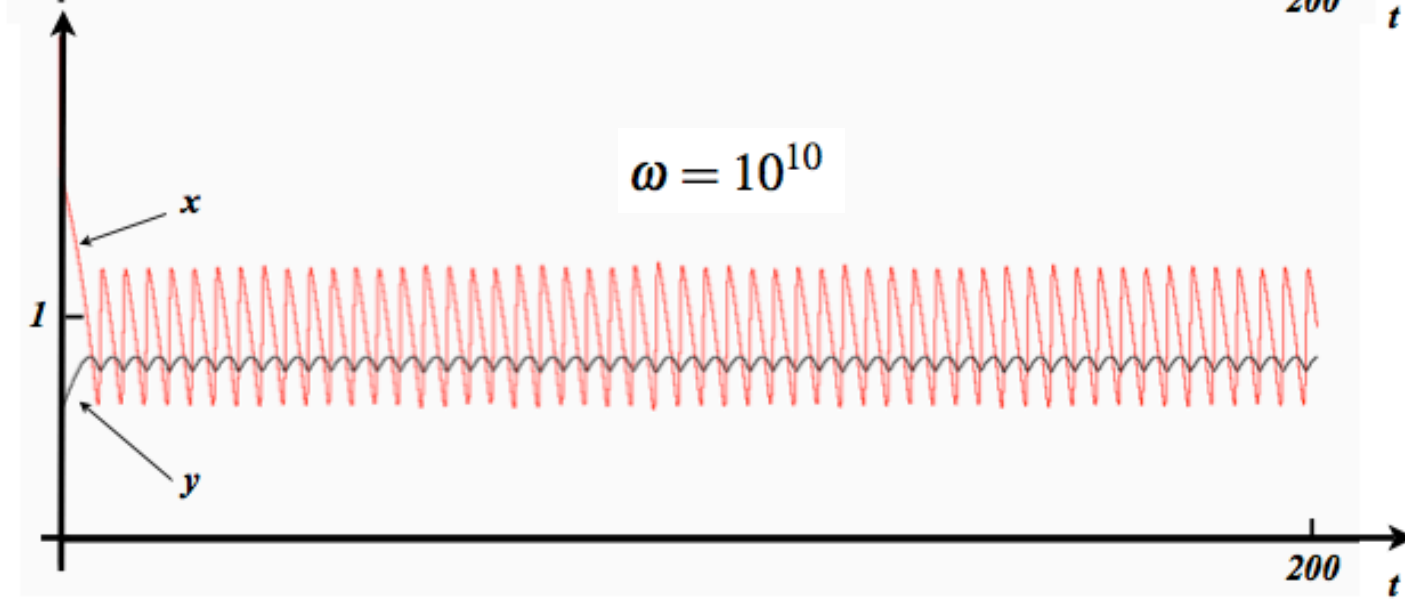
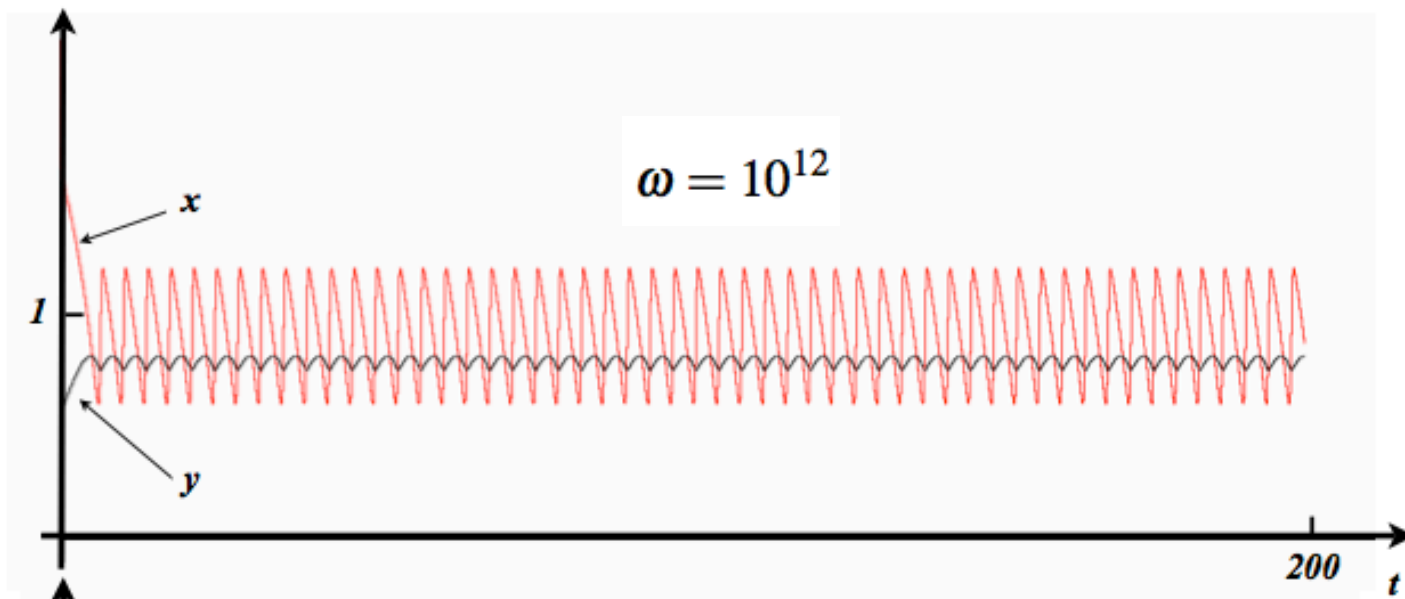


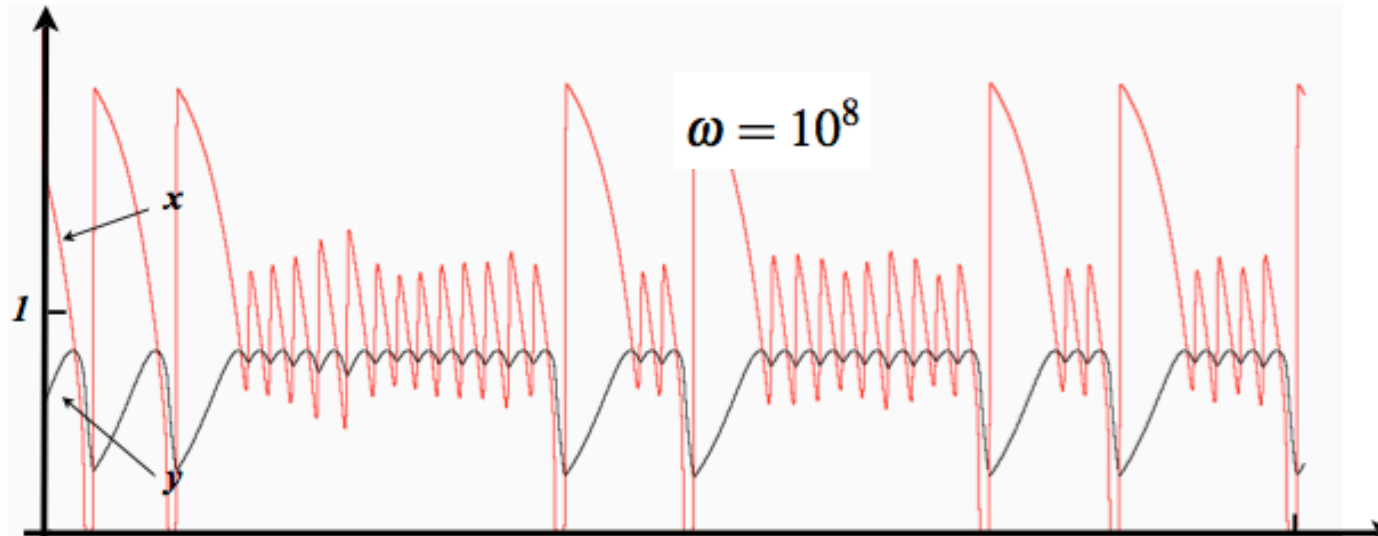


Ce n'est pas une surprise

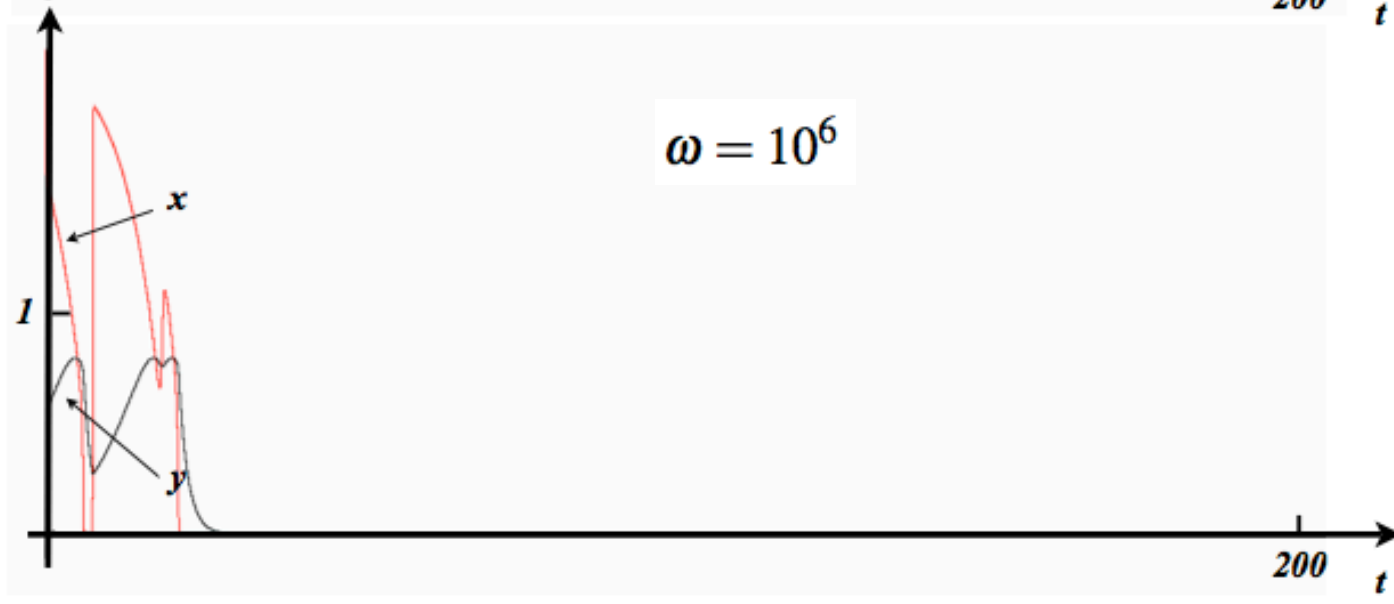
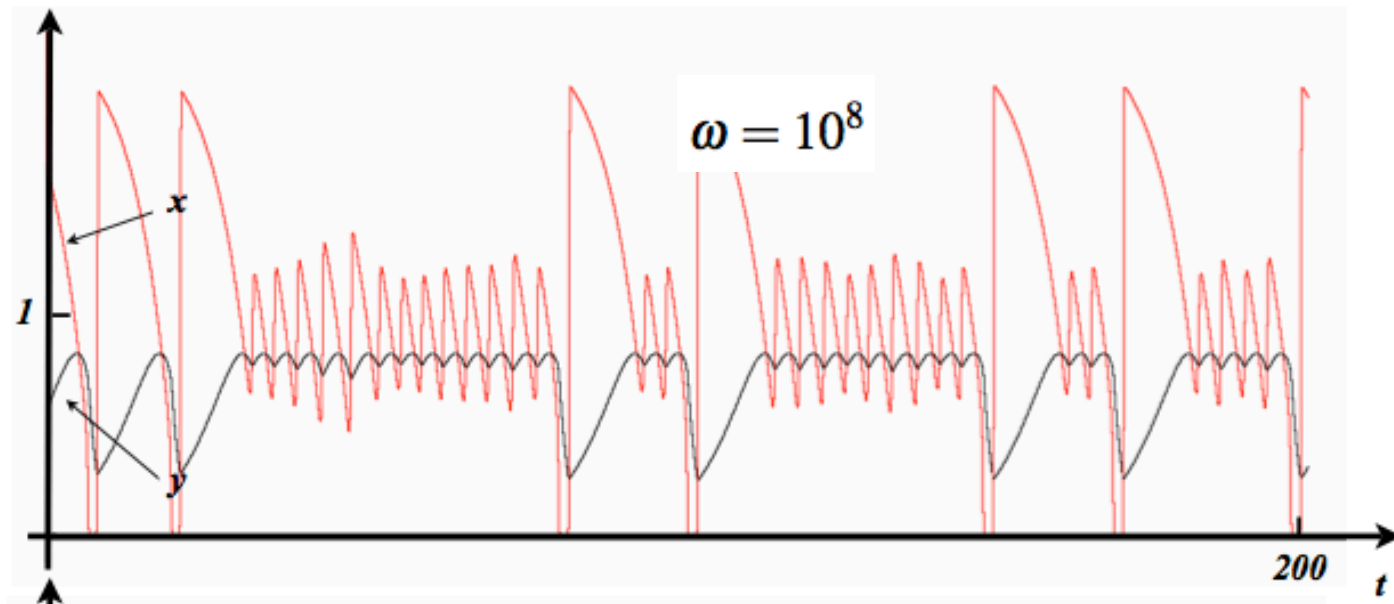
$$\begin{cases} \frac{dx}{dt} = \frac{1}{\varepsilon} [f(x) - \mu(x)y] \\ \frac{dy}{dt} = (\mu(x) - m)y \end{cases}$$

+bruit très (très) petit





C'est une (grosse surprise) car le nombre d'individus est toujours très grand.



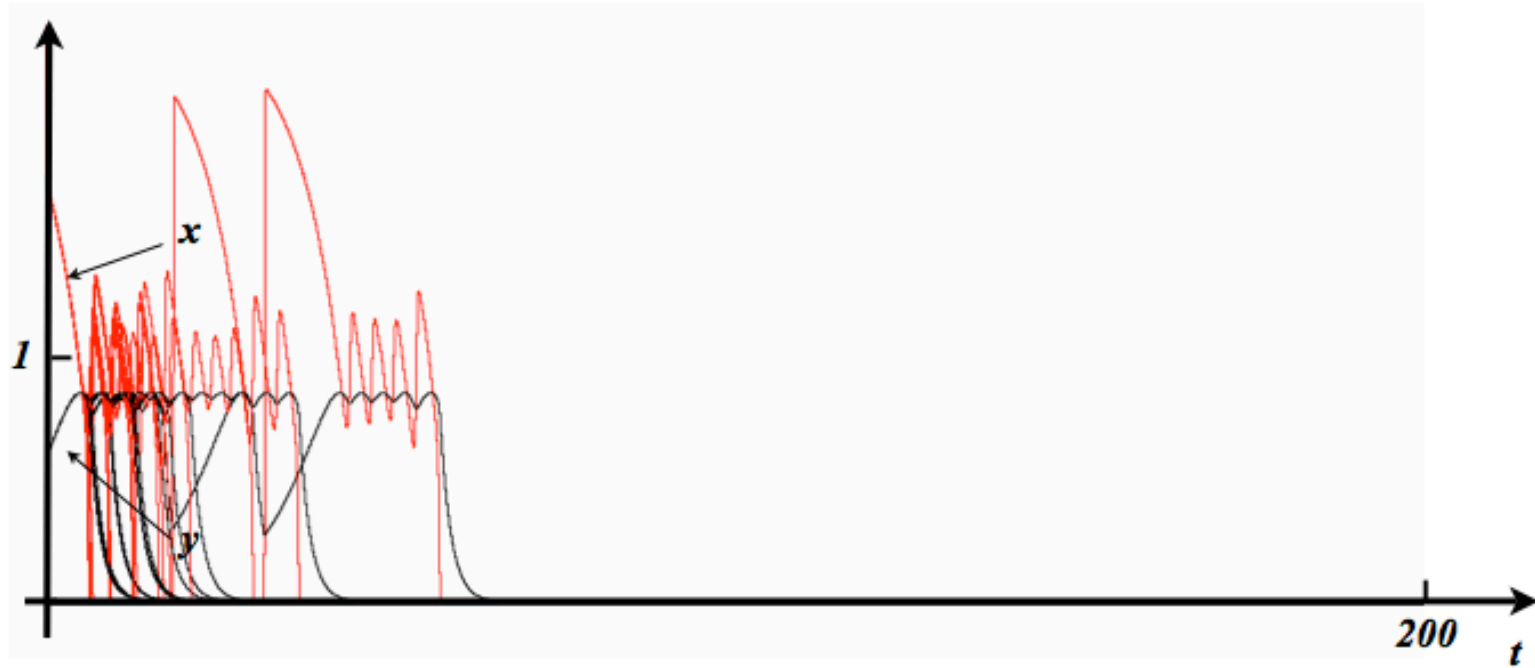


FIG. 3 – Twenty runs with $\omega = 10^6$

ω	$E[T]$	$\sigma(T)$	$P(T \leq 1000)$
10^5	25.8	3.3	1
10^6	32	9.5	1
10^7	143	118	1
$1.5 \cdot 10^7$	410	346	0.95
$1.6 \cdot 10^7$			0.82
$1.7 \cdot 10^7$			0.70
$2 \cdot 10^7$			0.55
$3 \cdot 10^7$			0.06
10^8			0

Probabilité empirique d'extinction en fonction de ω
(1000 réalisations)

Extinction fixée à 1000 individus

$$x(t+dt) - x(t) \approx dt \frac{1}{\varepsilon} [f(x(t)) - \mu(x(t)) y(t)] + \sqrt{dt \frac{4}{\omega \varepsilon} \frac{f(x(t)) \mu(x(t)) y(t)}{(f(x(t)) + \mu(x(t)) y(t))}} W_t$$

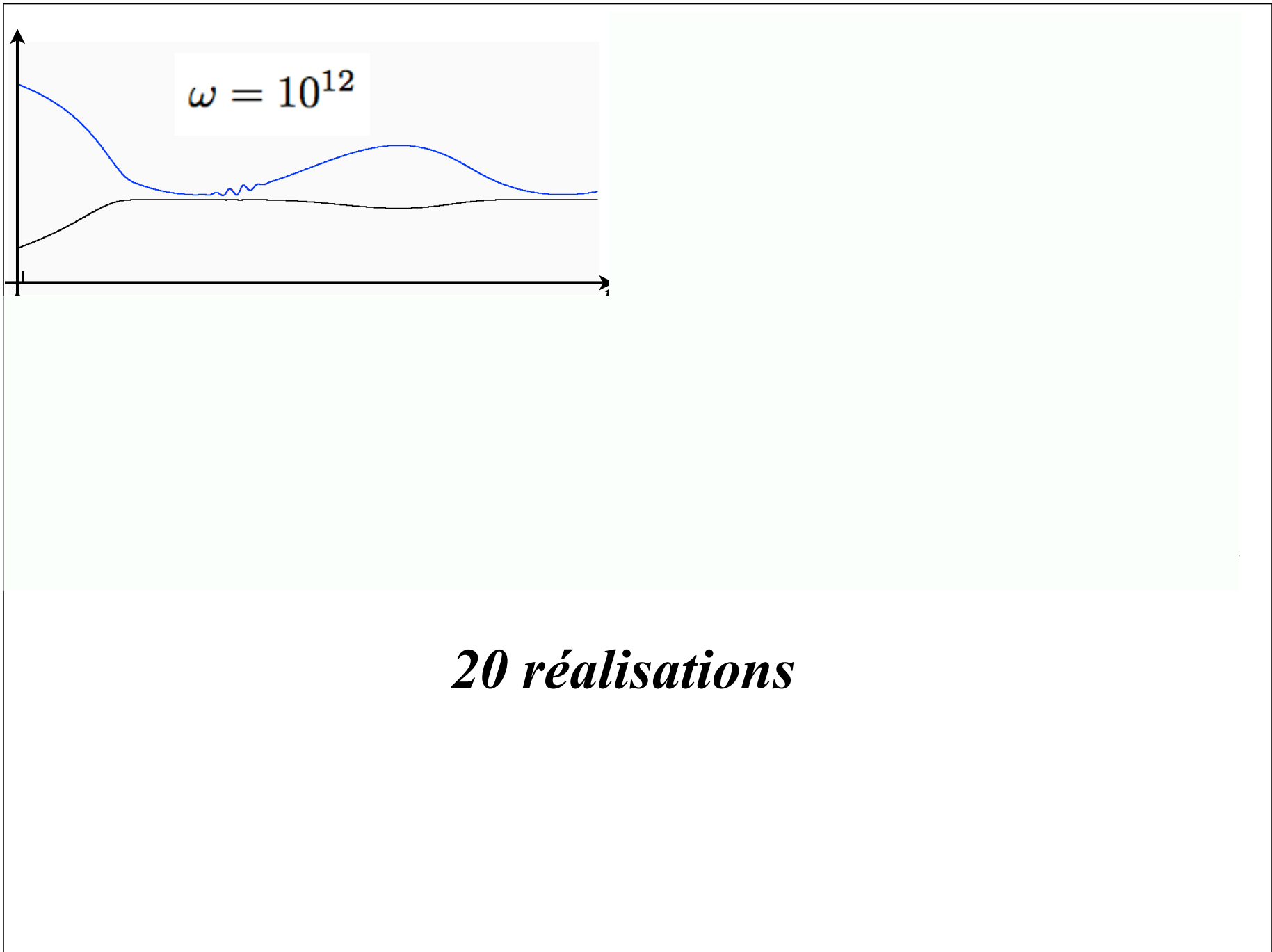
$$y(t+dt) - y(t) \approx dt [\mu(x(t)) - m] y(t) + \sqrt{dt \frac{\varepsilon}{\omega} \frac{f(x(t)) \mu(x(t)) y(t)}{(f(x(t)) + \mu(x(t)) y(t))}} W_t$$

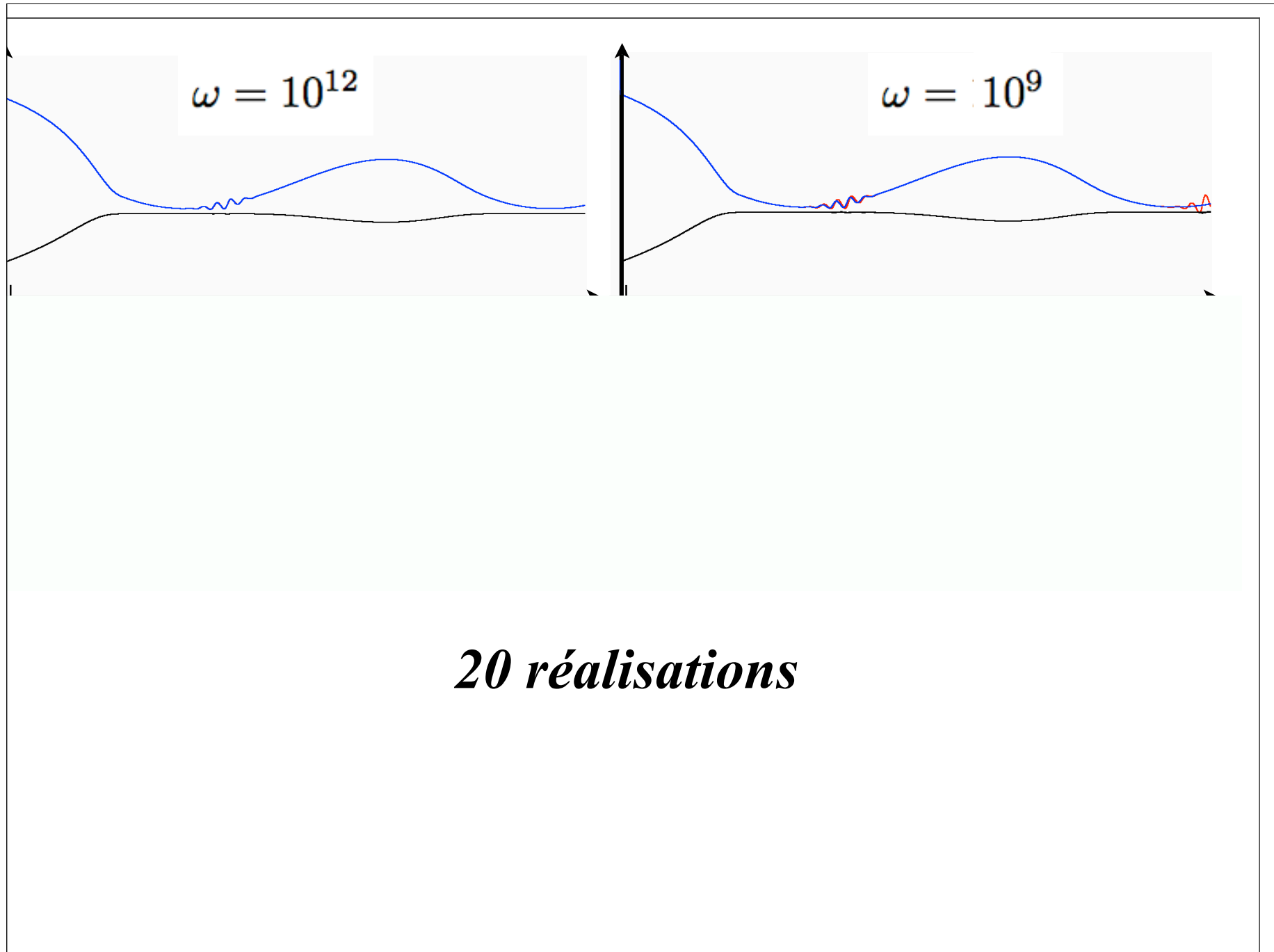
↓

$$m(t) = a + b \cos(r t)$$

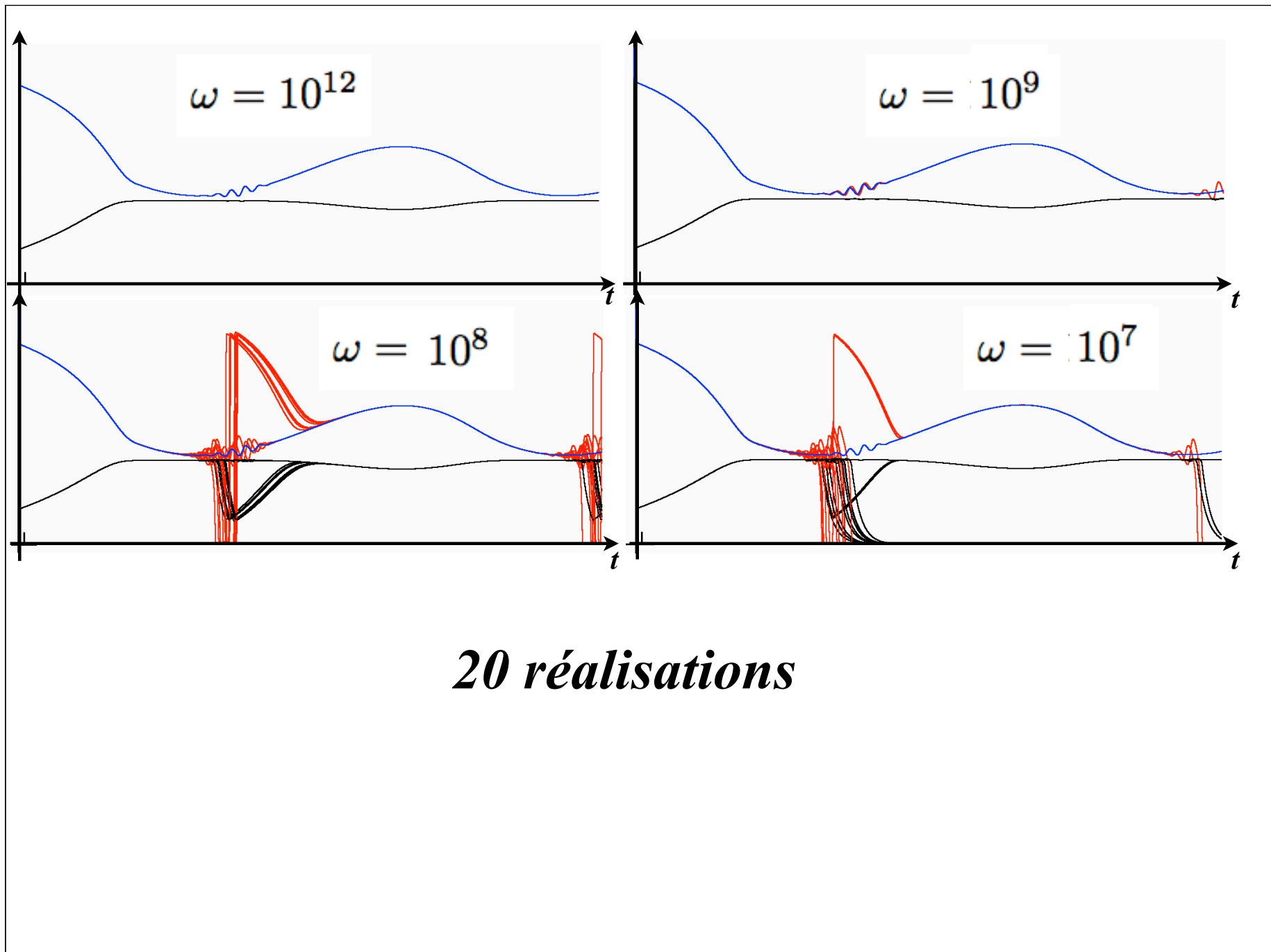
Simulations : **En rouge le modèle complet**

On “repassse”, en bleu, avec la partie déterministe seule





20 réalisations



Explication

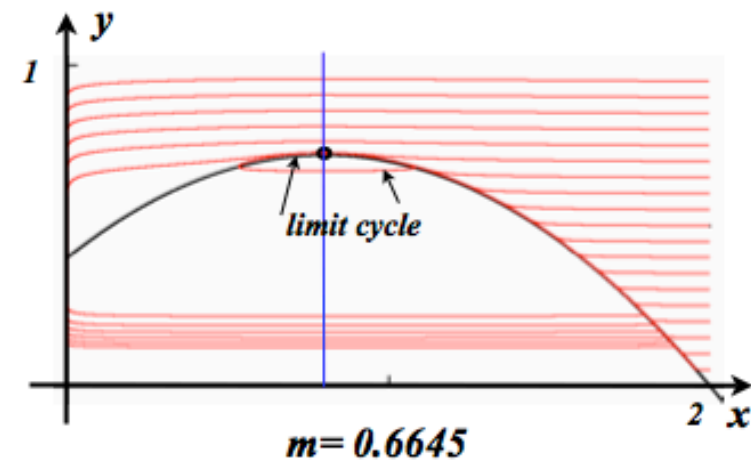
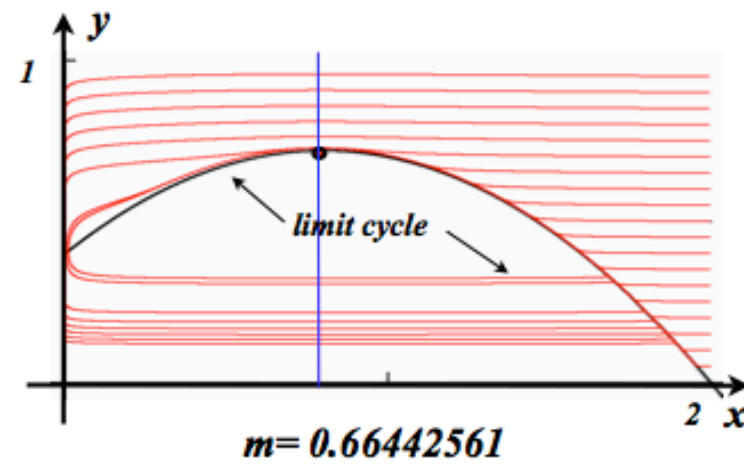
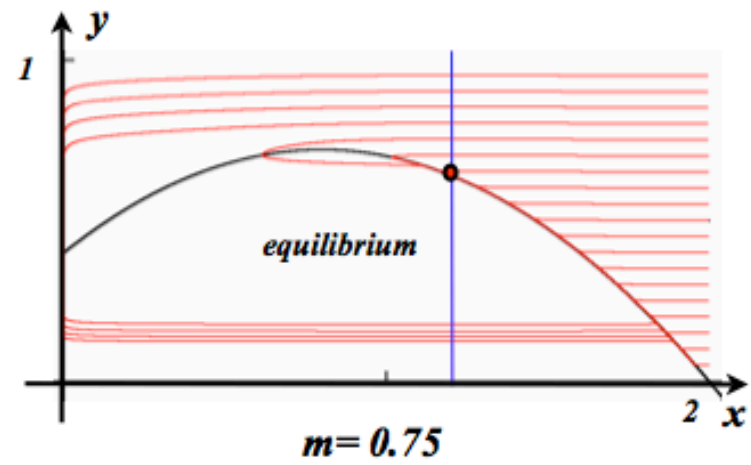
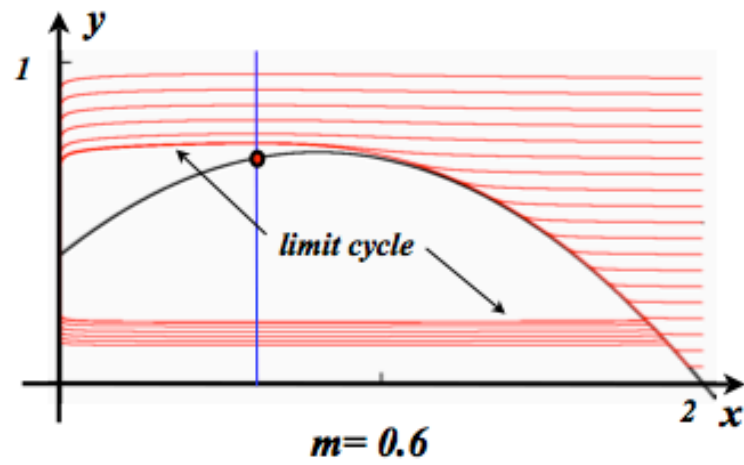


FIG. 6 – Phase portrait of system (7) for different values of m

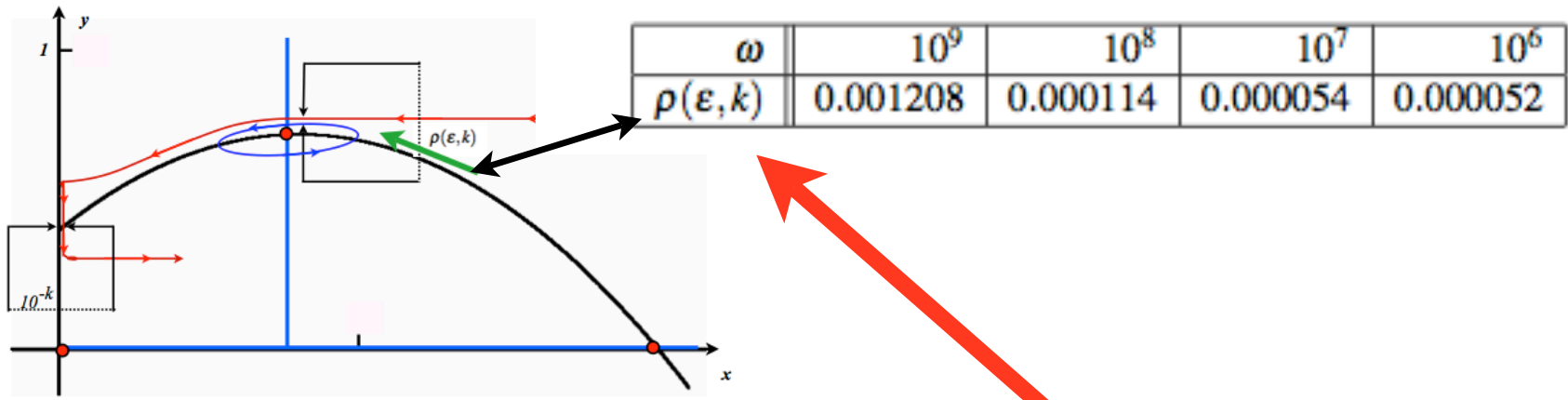


FIG. 9 – The “safety funnel”

$$x(t + dt) - x(t) \approx dt \frac{1}{\epsilon} [f(x(t)) - m\mu(x(t))y(t)] + \sqrt{dt \frac{4}{\omega\epsilon} \frac{f(x(t))\mu(x(t))y(t)}{(f(x(t)) + \mu(x(t))y(t))}} W_t$$

$$y(t + dt) - y(t) \approx dt [\mu(x(t)) - m]y(t) + \sqrt{dt \frac{\epsilon}{\omega} \frac{f(x(t))\mu(x(t))y(t)}{(f(x(t)) + \mu(x(t))y(t))}} W_t$$

