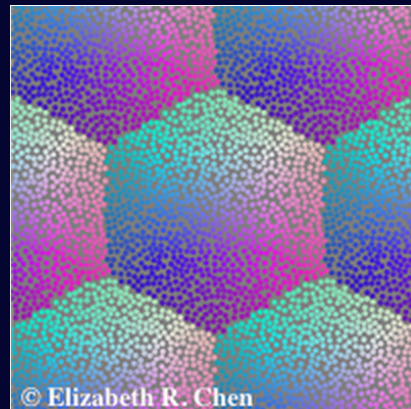


# stochastic effects in hematopoiesis

Jorge M. Pacheco  
*Math – U. Minho*

 atp group

<http://www.ciul.ul.pt/~ATP/>



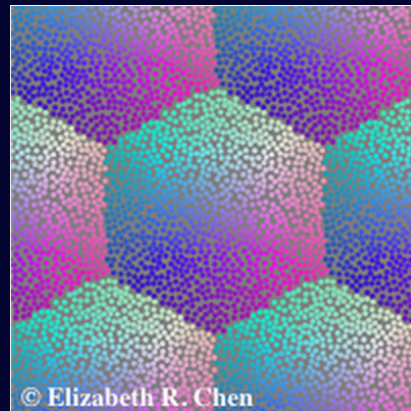
21-FEB-2013

# If cells play dice, can we gamble our way out of cancer ?

Jorge M. Pacheco  
*Math – U. Minho*

 atp group

<http://www.ciul.ul.pt/~ATP/>



21-FEB-2013

# Cancer

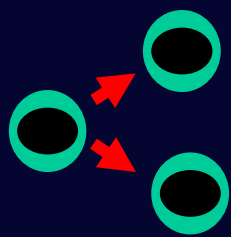
( a poor summary of some of David Dingli's slides )

- ❖ cancer is a consequence of multicellularity
  - ❖ cellular genome is under permanent attack  
(environmental or metabolic genotoxic agents)
  - ❖ DNA replication machinery is not perfect
- } mutations
- ❖ many mutations are neutral
  - ❖ others → malignant transformation → clonal development
  - ❖ impact of mutations:  $\mu$  rate, # cells@risk, cell-lifetime

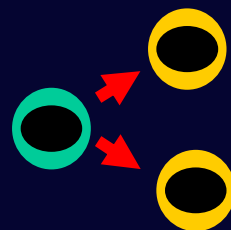
# tissue architecture

- ❖ tissue architecture has evolved
- ❖ most tissue cells have a ↘ lifetime & a ↗ turnover
  - minimize impact of **mutations**
- ❖ many tissues evolved a hierarchical structure
  - tree-like structure
- ❖ at the root of the tree are the tissue-specific stem-cells
  
- ❖ example: **hematopoiesis**
- ❖ **stem-cell** concept was developed in hematopoiesis and has been extended to many other tissues
- ❖ **HSC resilience** relies on ↘ # & ↘ turnover of stem cells

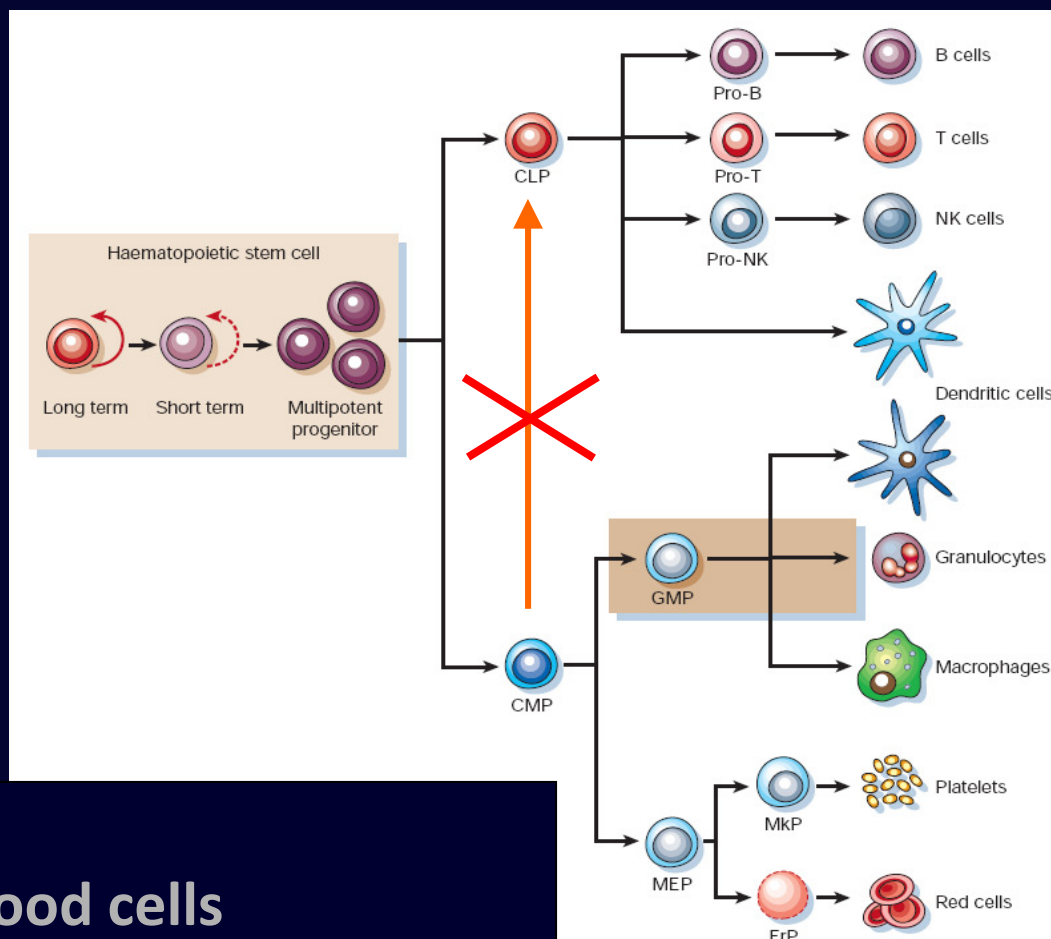
# hematopoietic stem cells (HSC)



*self-renewal* :  
capacity to clone  
themselves

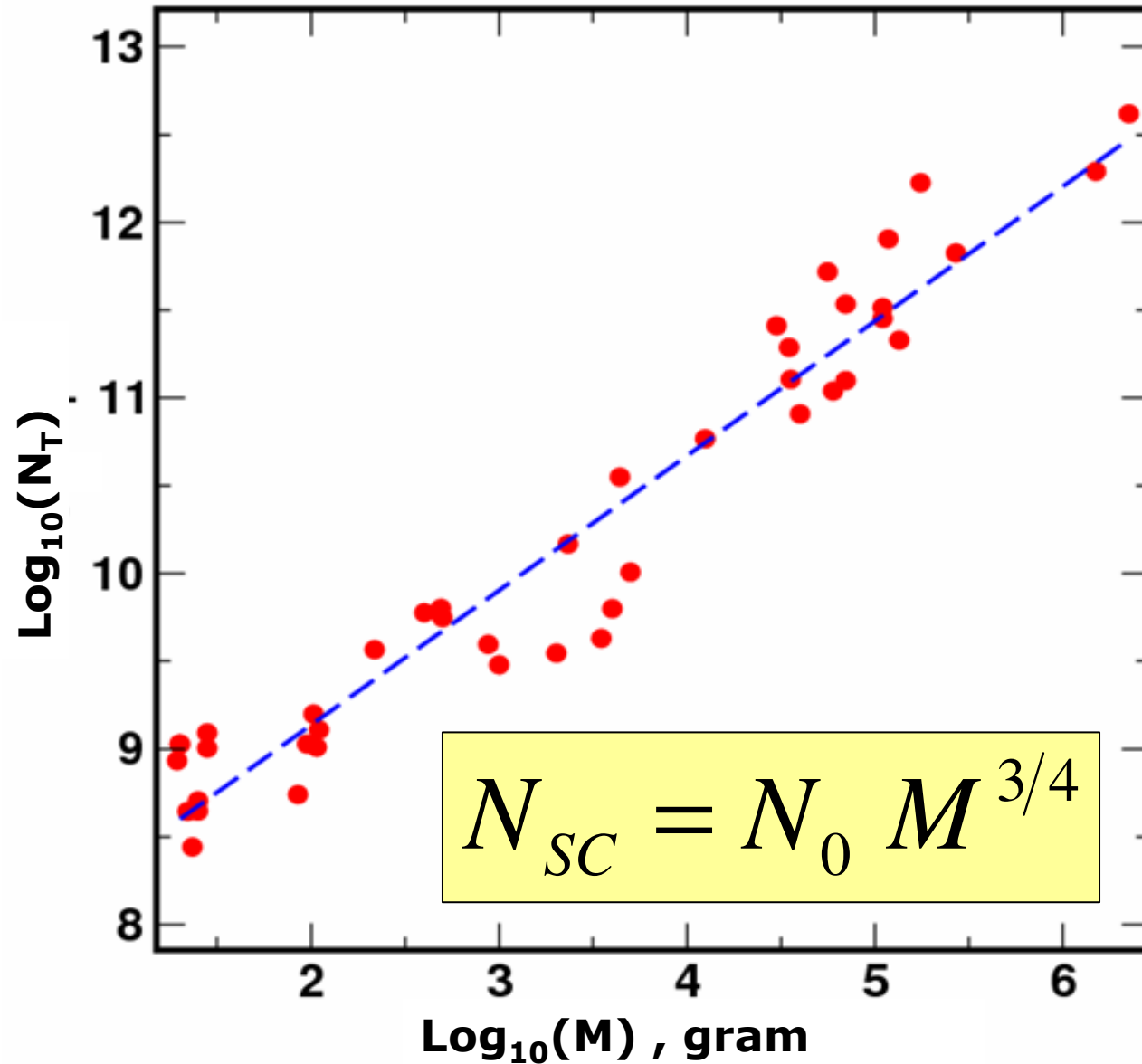


*differentiation* :  
capacity to differentiate  
into all other types of blood cells



*stemness* is a matter of degree – you have to stand at the  
*root* of the *hematopoietic tree*

# allometric scaling of hematopoiesis in land mammals



# allometric scaling of hematopoiesis in land mammals

use experimental estimates for **cats** for calibration ( **fix  $N_0$**  ):  
 under normal conditions,  **$\geq 40$  !** ( Abkowitz et al, Blood, 2002 )

what	model predictions ×	experimental data
<b>HSC in humans</b> cat = 40	<b>385</b>	<b>~400</b> ( Buescher et al, J Clin Invest, 1985 )
<b>rate HSC division</b> cat post-TRX = 8 week <sup>-1</sup>	<b>60 week<sup>-1</sup></b>	<b>~ 52-104 week<sup>-1</sup></b> ( Rufer, et al, J Exp Med, 1999 )
<b>human post-transplant</b> cat = 13	<b>111</b>	<b>~ 116</b> ( Nash et al, Blood, 1988 )
<b>mouse</b>	<b>1</b>	<b>1</b> ( Abkowitz et al, PNAS, 1995 )
<b>rate macaques</b>	<b>23 week<sup>-1</sup></b>	<b>23 week<sup>-1</sup></b> ( Shepherd et al, Blood, 2007 )
<b>rate baboons</b>	<b>36 week<sup>-1</sup></b>	<b>36 week<sup>-1</sup></b> ( Shepherd et al, Blood, 2007 )

## the hematopoietic *tree*

- ❖ in humans  $\sim 400$  HSC divide each once per year
- ❖ *but* : daily output of bone marrow  $\sim 3.5 \times 10^{11}$  cells !!!

*how to explain this enormous amplification given the slow replication rate of HSC ?*

- ❖ one must consider :

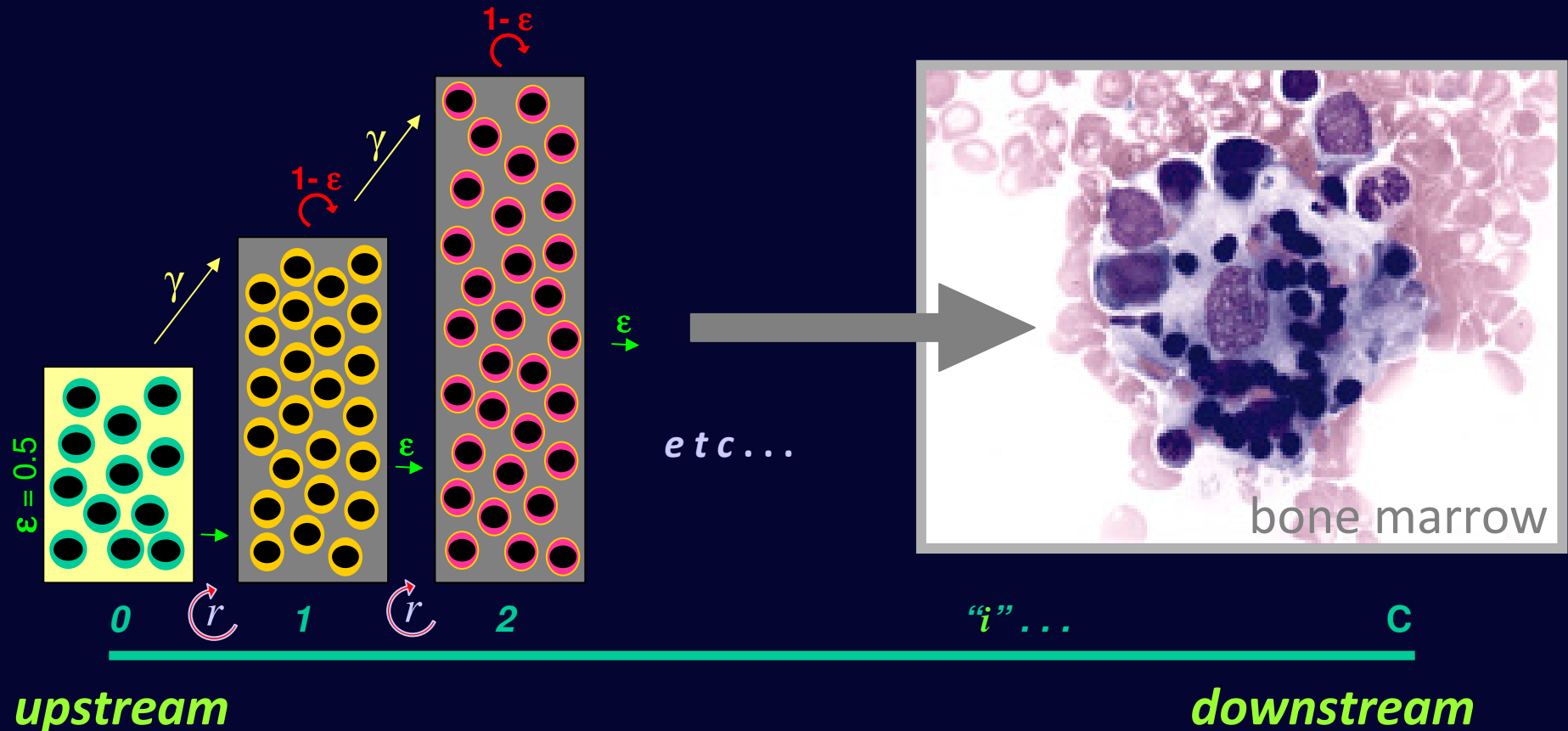


*asymmetric division* : more parameters, see Dingli et al. *PLoS-CB*, 2007



# the hematopoietic tree

- ❖ we consider a **compartmentalized structure** in which **cells from upstream compartments flow into downstream compartments**, under **stationary flux conditions**;



# deterministic dynamics of the hematopoietic tree

$$\dot{N}_i(t) = -d_i N_i(t) + b_{i-1} N_{i-1}(t)$$

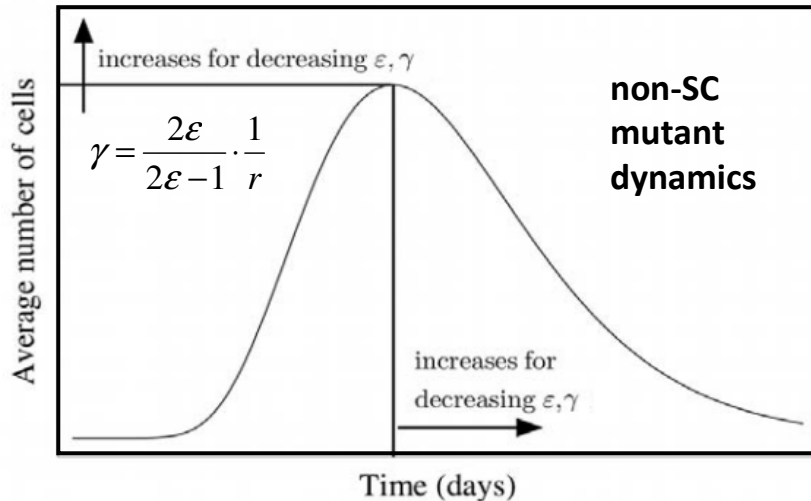
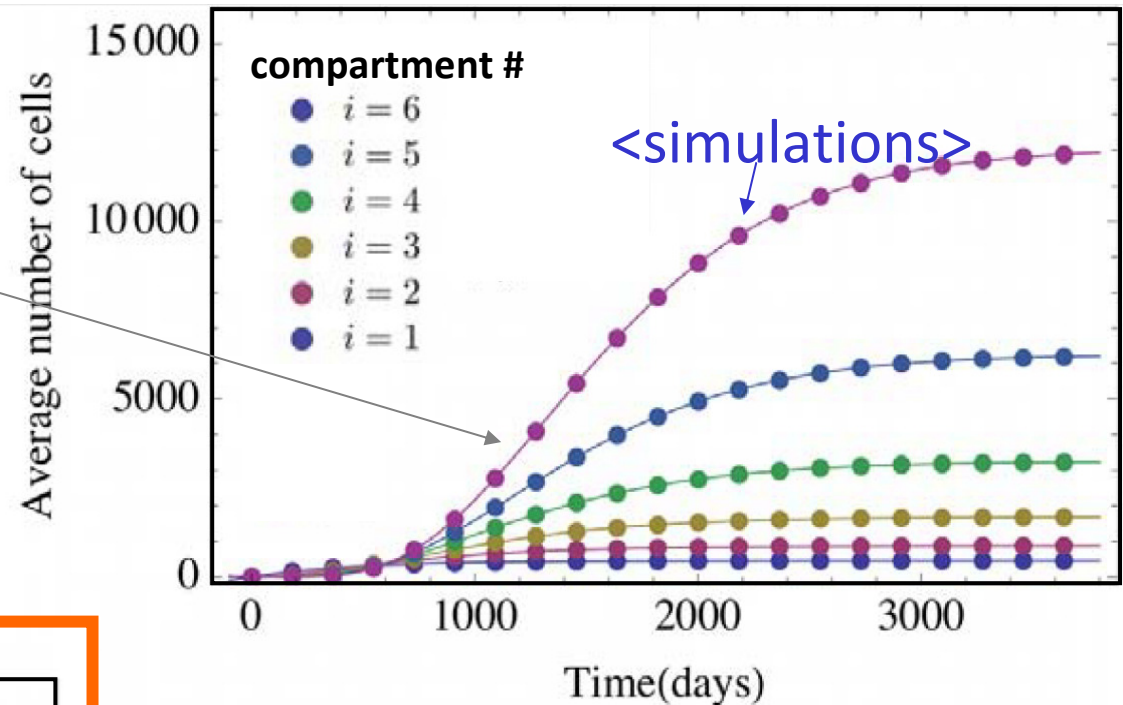
$$d_i = (2\varepsilon - 1)r_i$$

$$b_i = 2\varepsilon r_i$$

$$r_i = r^i r_0$$

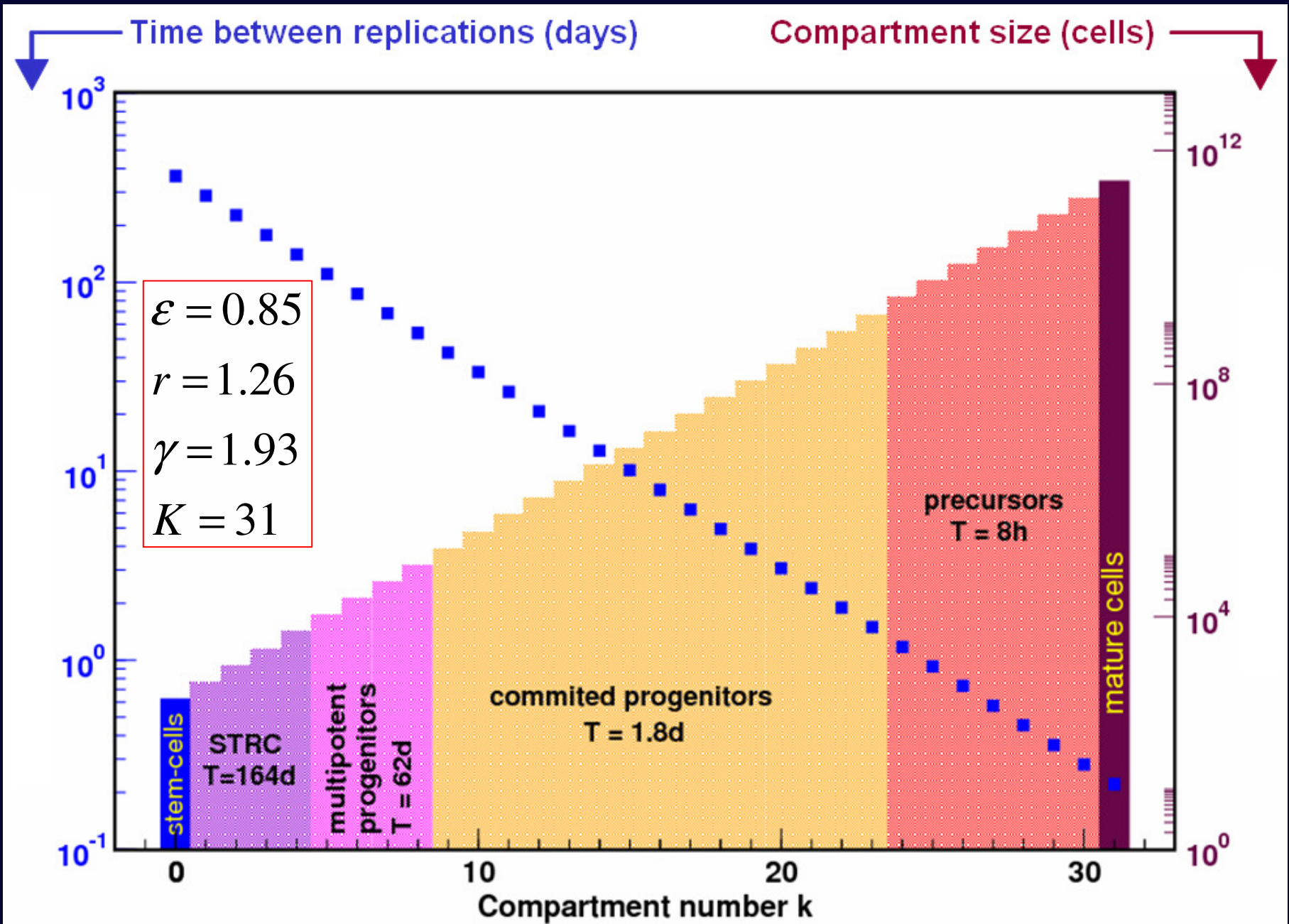
analytic solution

Werner *et al.*, PLoS-CB, 2011

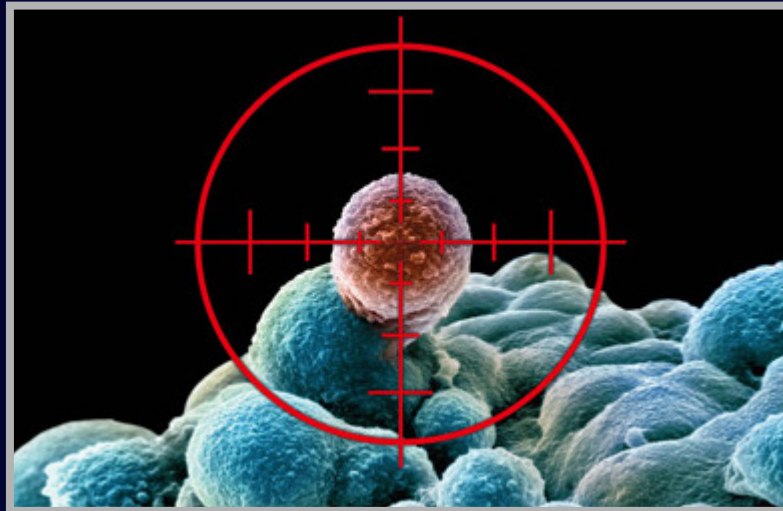


the deterministic model provides numerous insights onto the average dynamics in the hematopoietic tree; for instance, that **even if you kill the LSC**, you may still die from its progeny (ex: CML)

# the hematopoietic *tree*



# DISEASE



The Economist, 13<sup>th</sup> September, 2008  
( article on cancer stem cells )

## scaling relations . . .

❖ number of **HSC** in adult mammals :

$$N_{SC} \approx 16.55 M^{3/4} \quad [ -- ]$$

❖ number of **HSC** during human ontogeny :

$$N_{SC} \approx 5.5 m(t) \quad [ -- ]$$

❖ **HSC** replication rate :

$$r_0 \approx 2.9 M^{-1/4} \quad [ year^{-1} ]$$

❖ average life-span of organism :

$$L \approx 8.6 M^{1/4} \quad [ year ]$$

( [M] = kg )

## simple implications . . .

**Hayflick hypothesis (1961):**

*cells undergo a limited number of divisions during their lifespan*

*from the scaling relations, each cell divides*

$$N \sim \text{rate} \times \text{lifespan} \sim M^{-1/4} \times M^{1/4} \sim M^0$$

*that is, constant & independent of the mammalian species :*

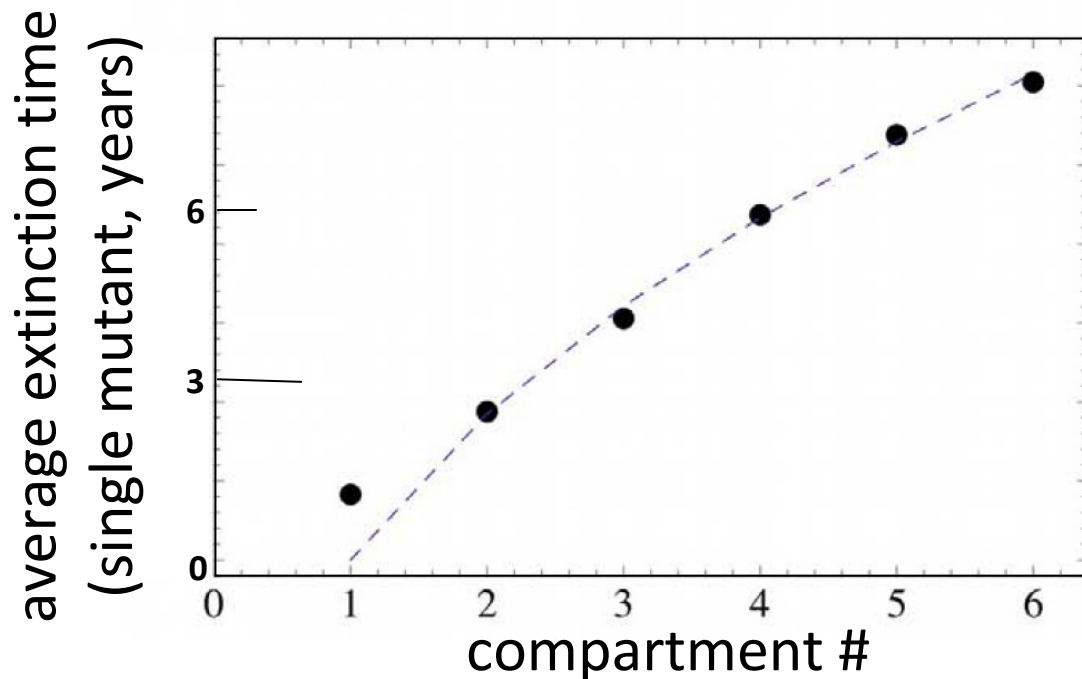
*a mouse-HSC and an elephant-HSC replicate, on average, the same number of times during the ~2-year and the ~70-year lifespans of the mouse and elephant, respectively; humans are the exception, as we live much longer than lifespan estimate.*

## are stochastic effects important ?

- ❖ *in vivo stochastic effects in hematopoiesis were found in 1996*

( Abkowitz et al, Nat. Med. , 1996 )

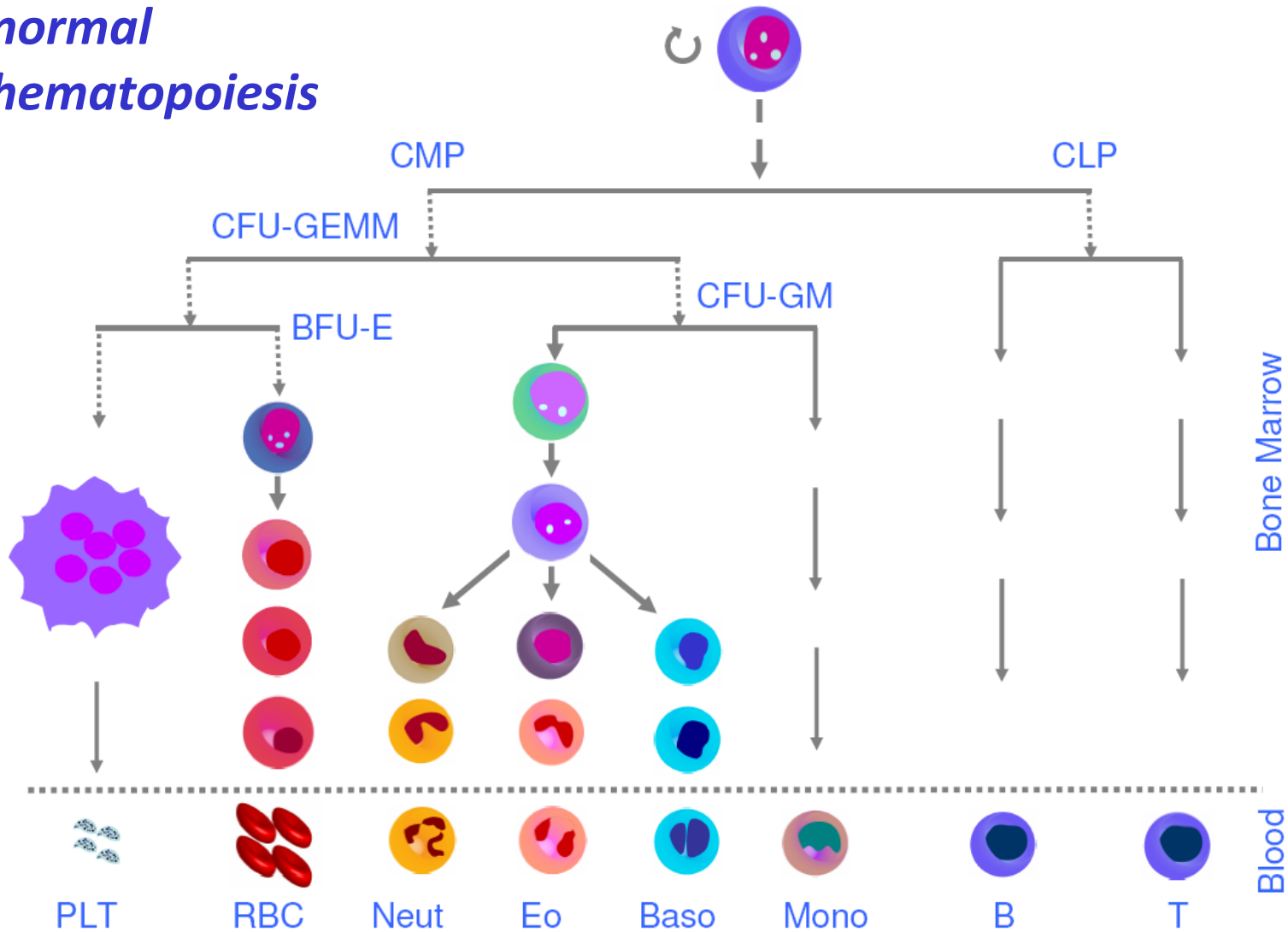
- ❖ *deterministic models (of hematopoiesis) **at best** describe average population dynamics behaviour, and may provide poor descriptions of small cell populations and neutral dynamics, in particular of **HSCs** ; **this may have sizeable impact on disease dynamics***



# trouble

normal :  $10^{-7} < \mu < 10^{-6}$  per cell per replication

## normal hematopoiesis

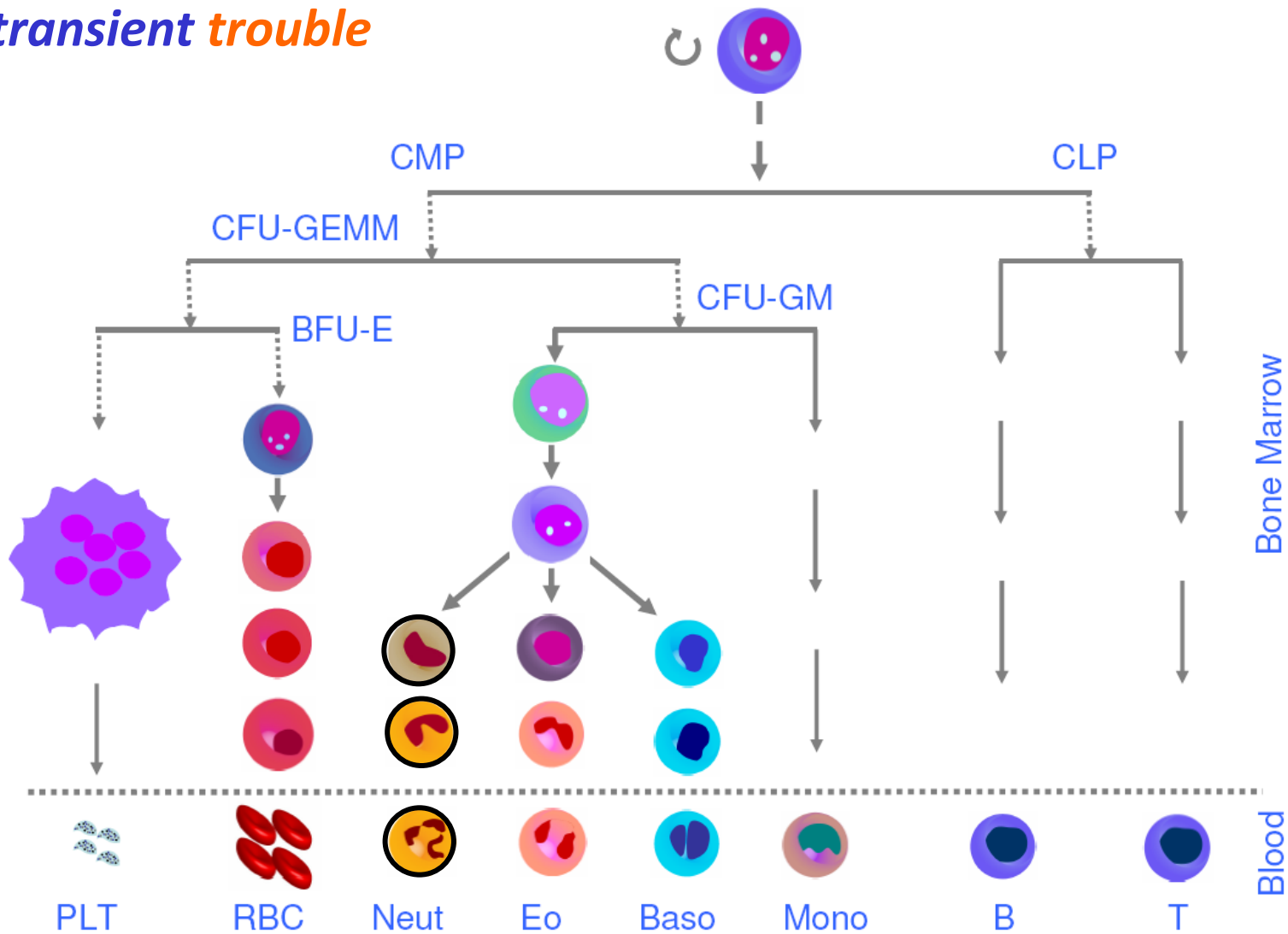




# trouble

normal :  $10^{-7} < \mu < 10^{-6}$  per cell per replication

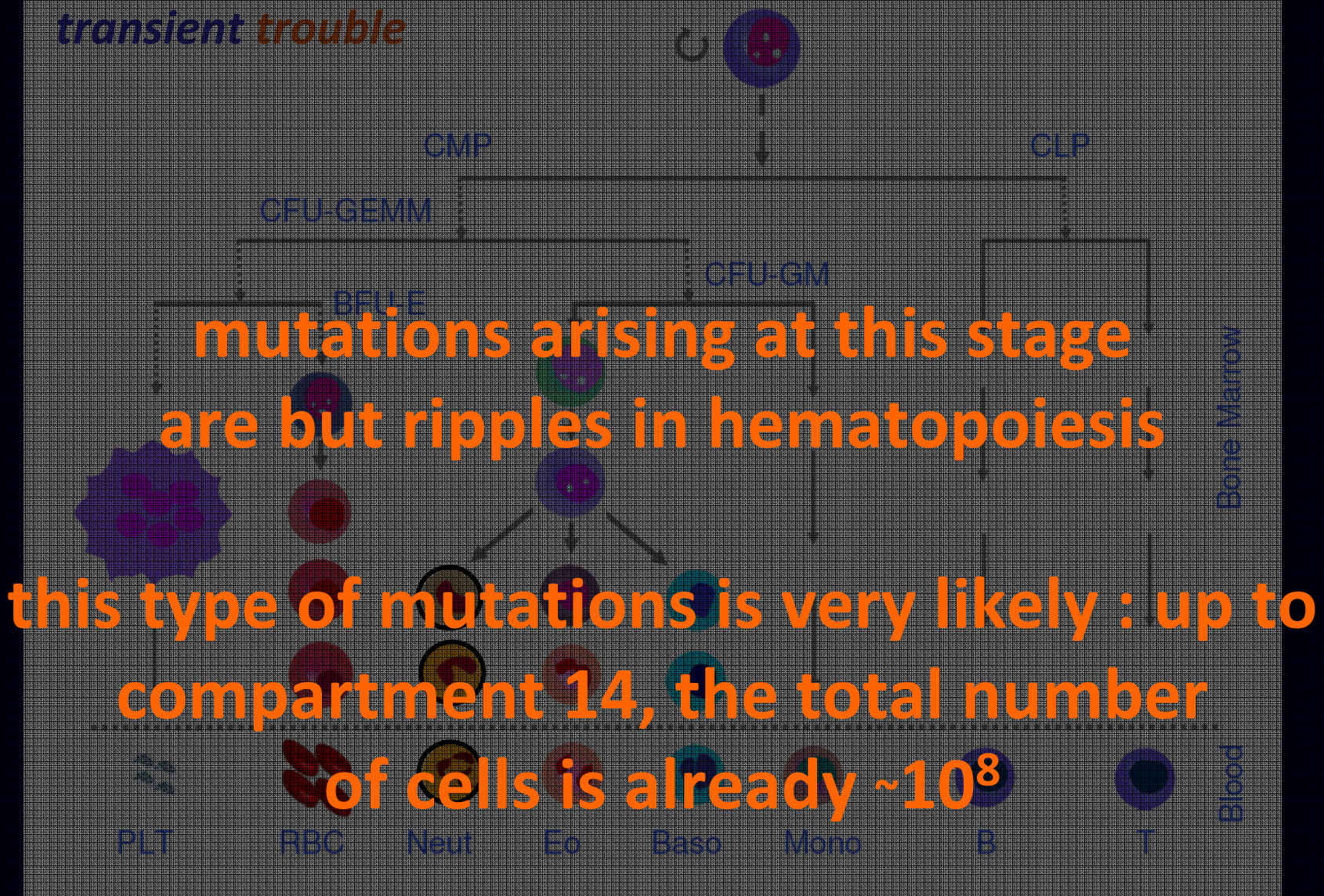
transient trouble



# trouble

normal :  $10^{-7} < \mu < 10^{-6}$  per cell per replication

transient trouble



PLT

RBC

Neut

Eo

Baso

Mono

B

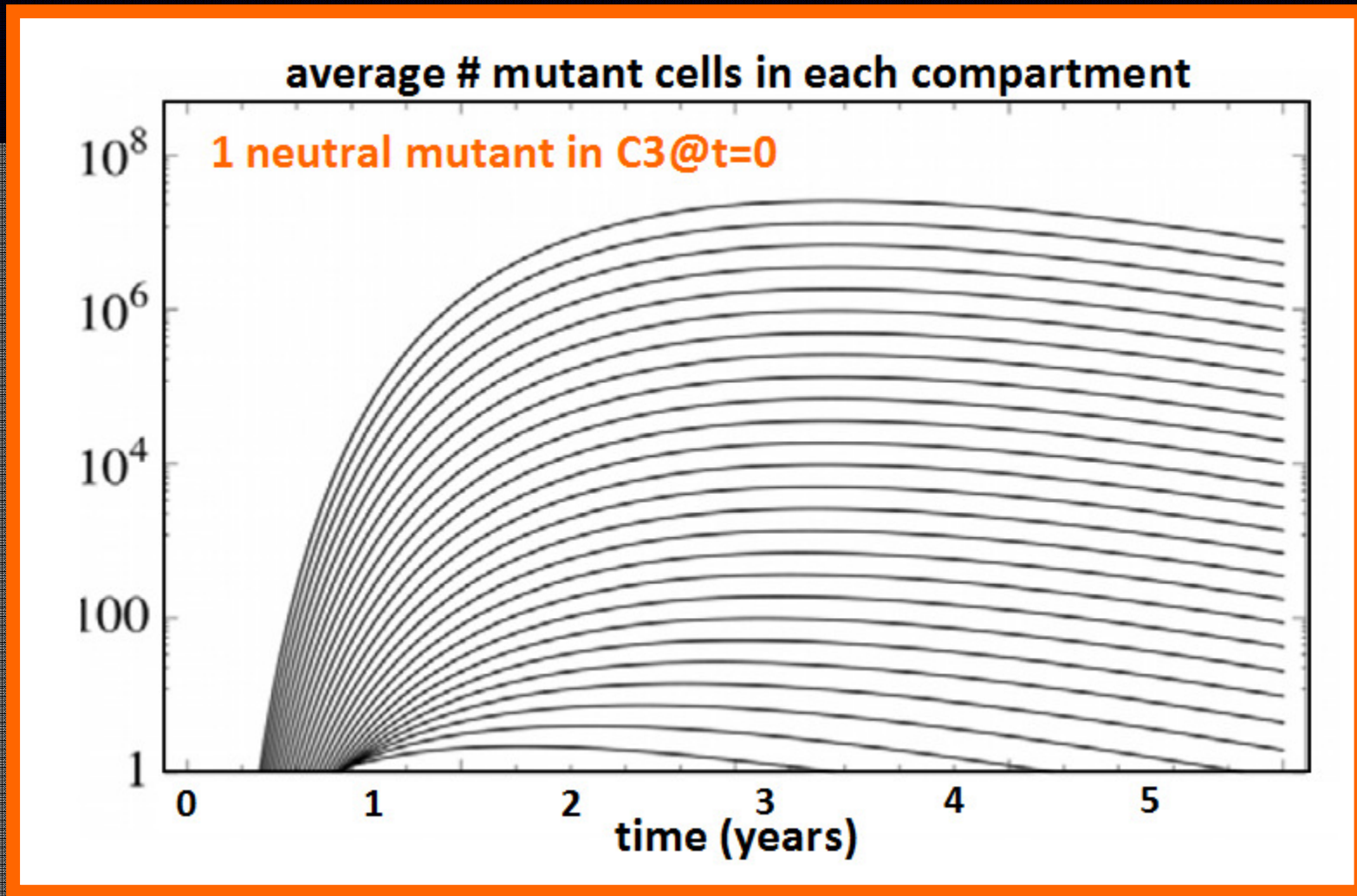
T

Bone Marrow

Blood



trouble



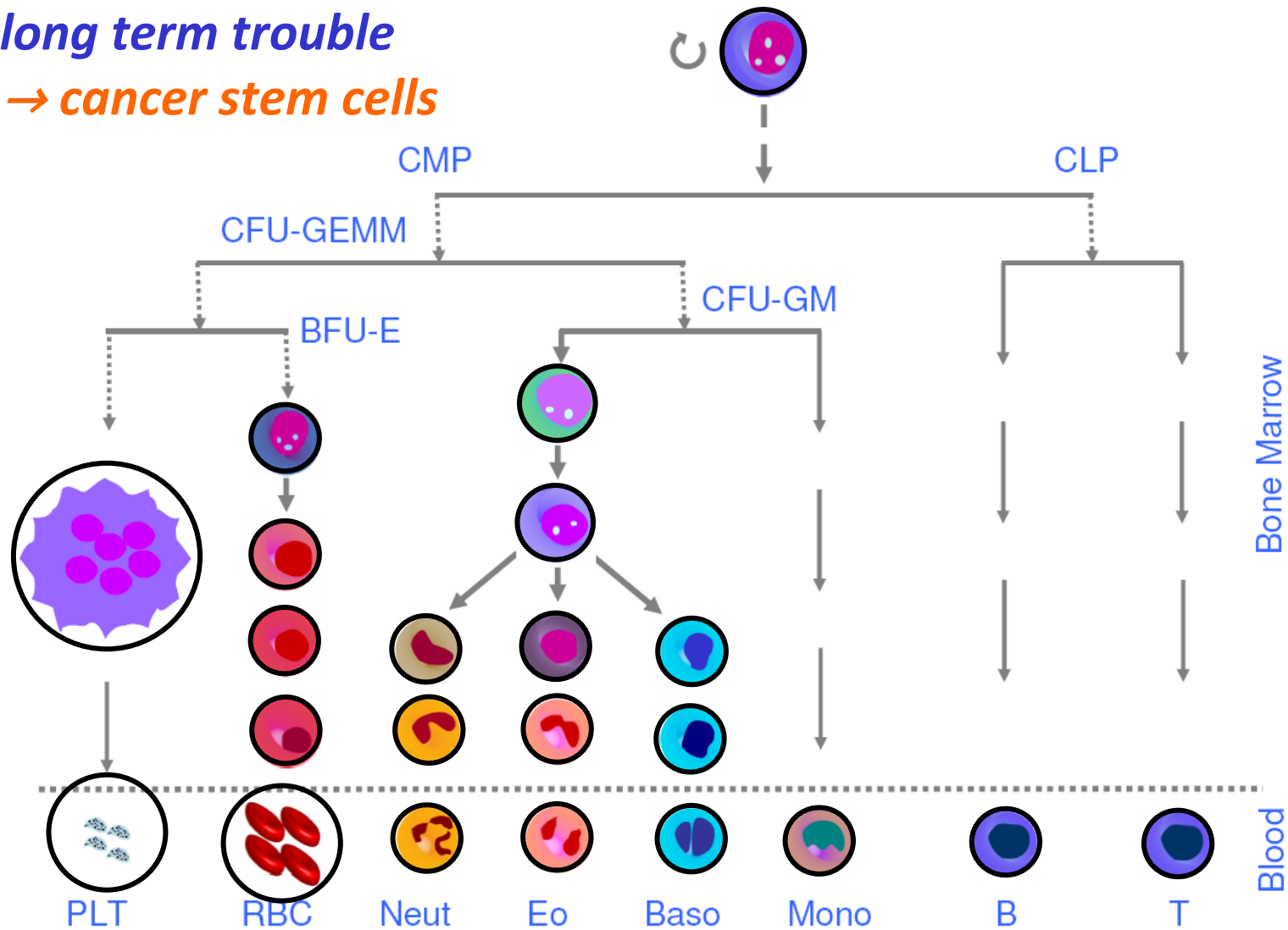
mutations arising in early progenitors may lead to long term trouble but are ultimately washed out (after many years)

PLC FBC Neut LC Baso Mono B T

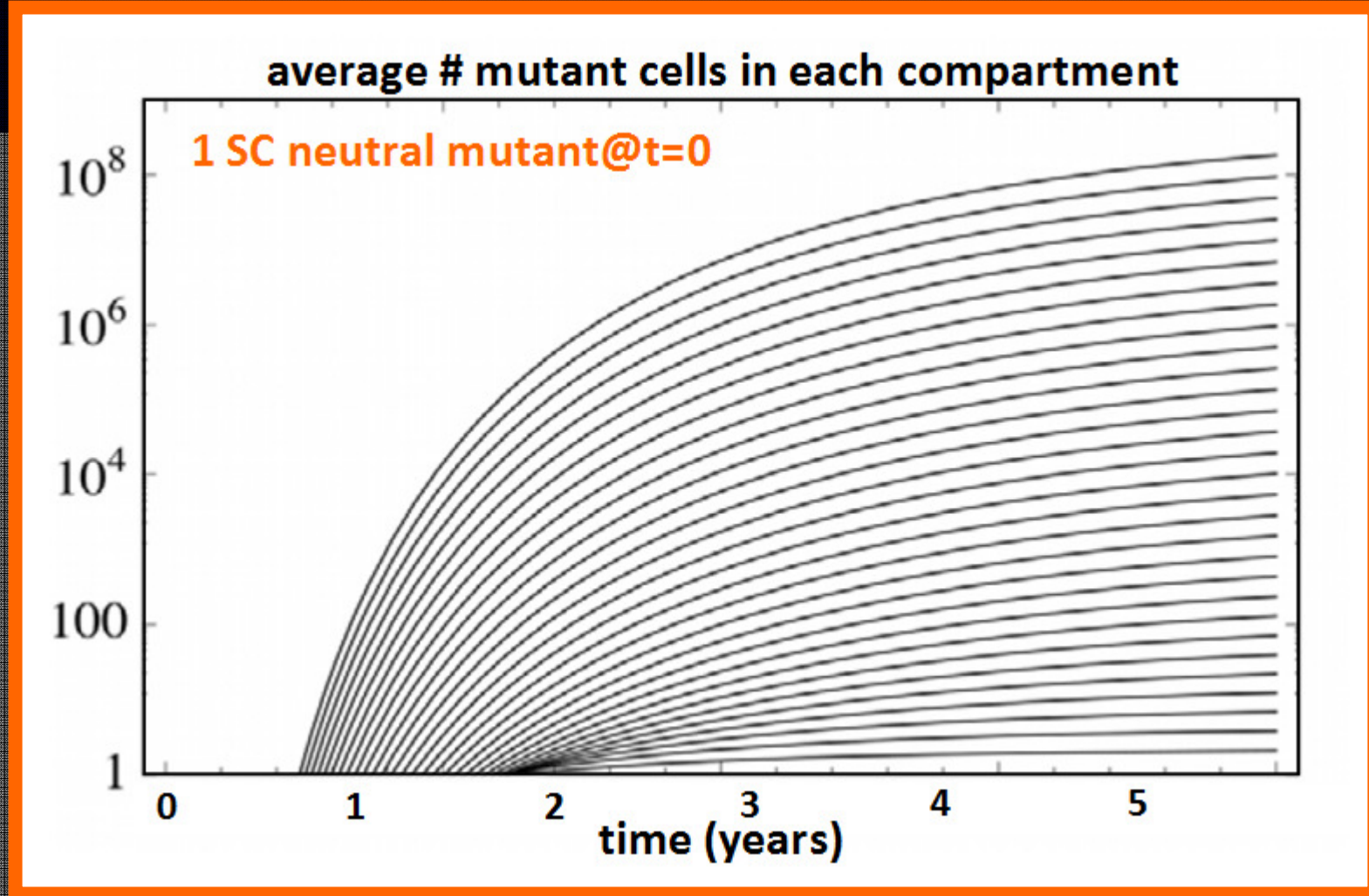
# trouble

normal :  $10^{-7} < \mu < 10^{-6}$  per cell per replication

long term trouble  
→ cancer stem cells

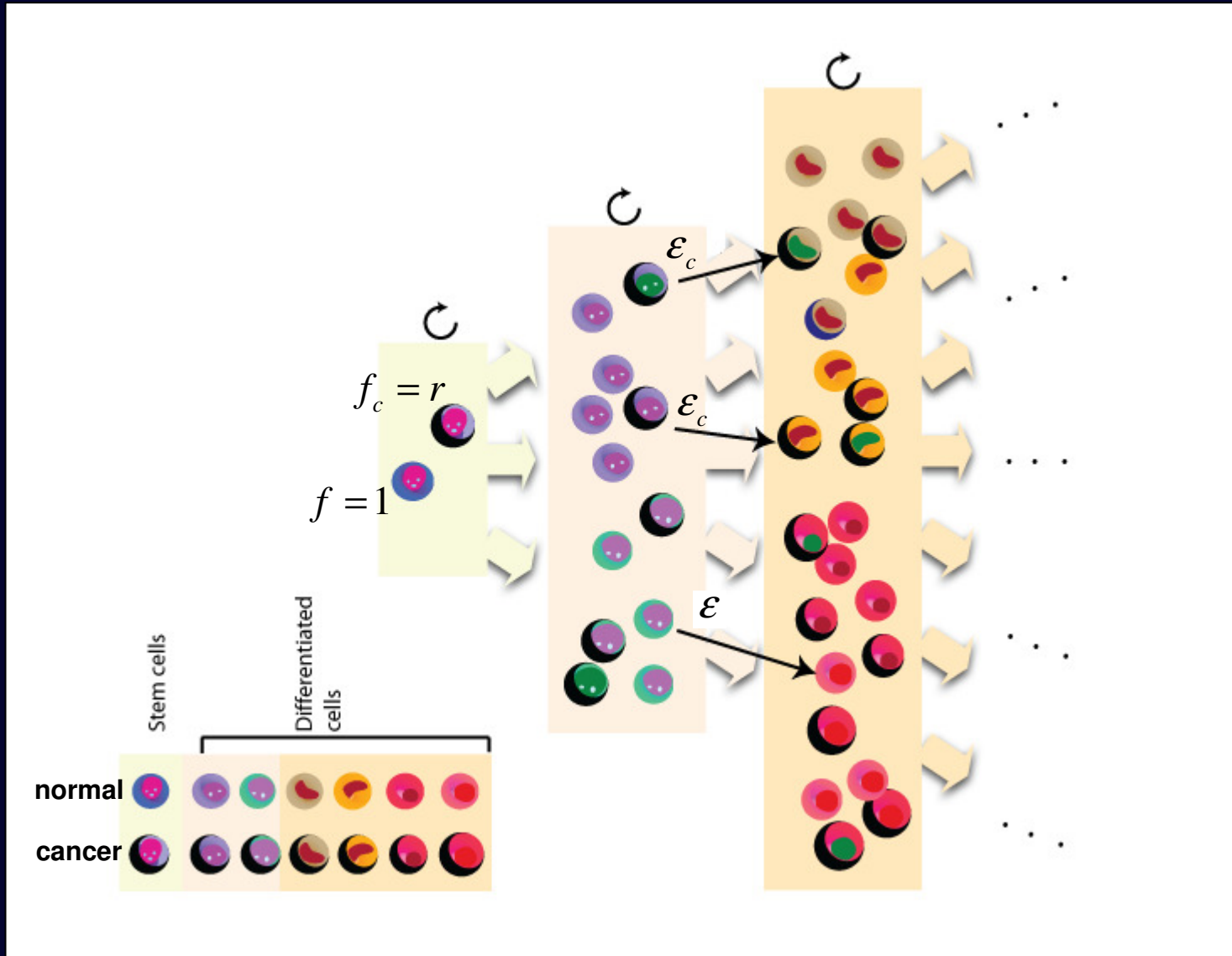


*trouble*



mutations arising in **HSC** lead to potentially permanent trouble as hematopoiesis may evolve toward a new steady state (stochastic effects may change this)

# *troubled* hematopoiesis

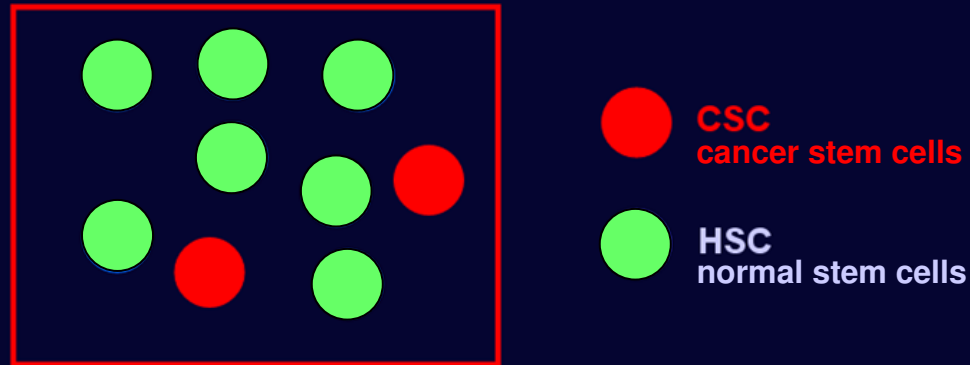


***cancer dynamics becomes a multi-scale ecology of cell competition***

starting upstream with a small number of HSC & CSC and getting downstream into very large numbers of cells of all kinds

# stochastic dynamics of *HSC* (birth-death)

stochastic model for *HSC* :

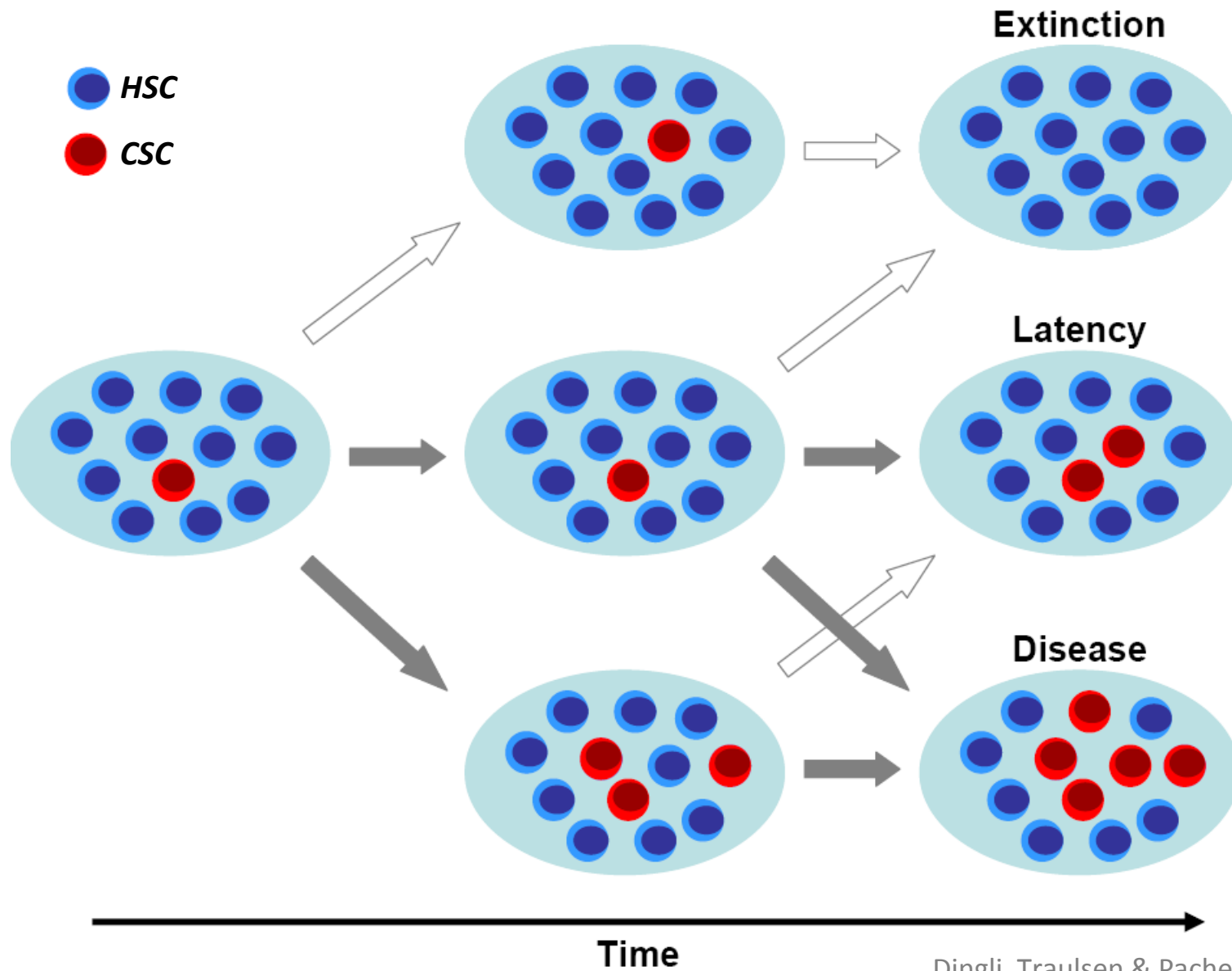


- ❖ SC population remains constant ( $16.55 M^{3/4}$ )
- ❖ HSC divide at normal rate ( $2.9 M^{-1/4}$ )
- ❖ CSC divide at rate  $r \times$  normal, where  $r$  = relative fitness
- ❖ when a cell is selected, gives rise to two new identical cells
- ❖ subsequently, 1 cell is randomly selected for export
- ❖ HSC may suffer *mutations* and transform into CSC



# stochastic dynamics of *HSC*

*several possible scenarios out of this simple process:*



## on the small number of **HSC**

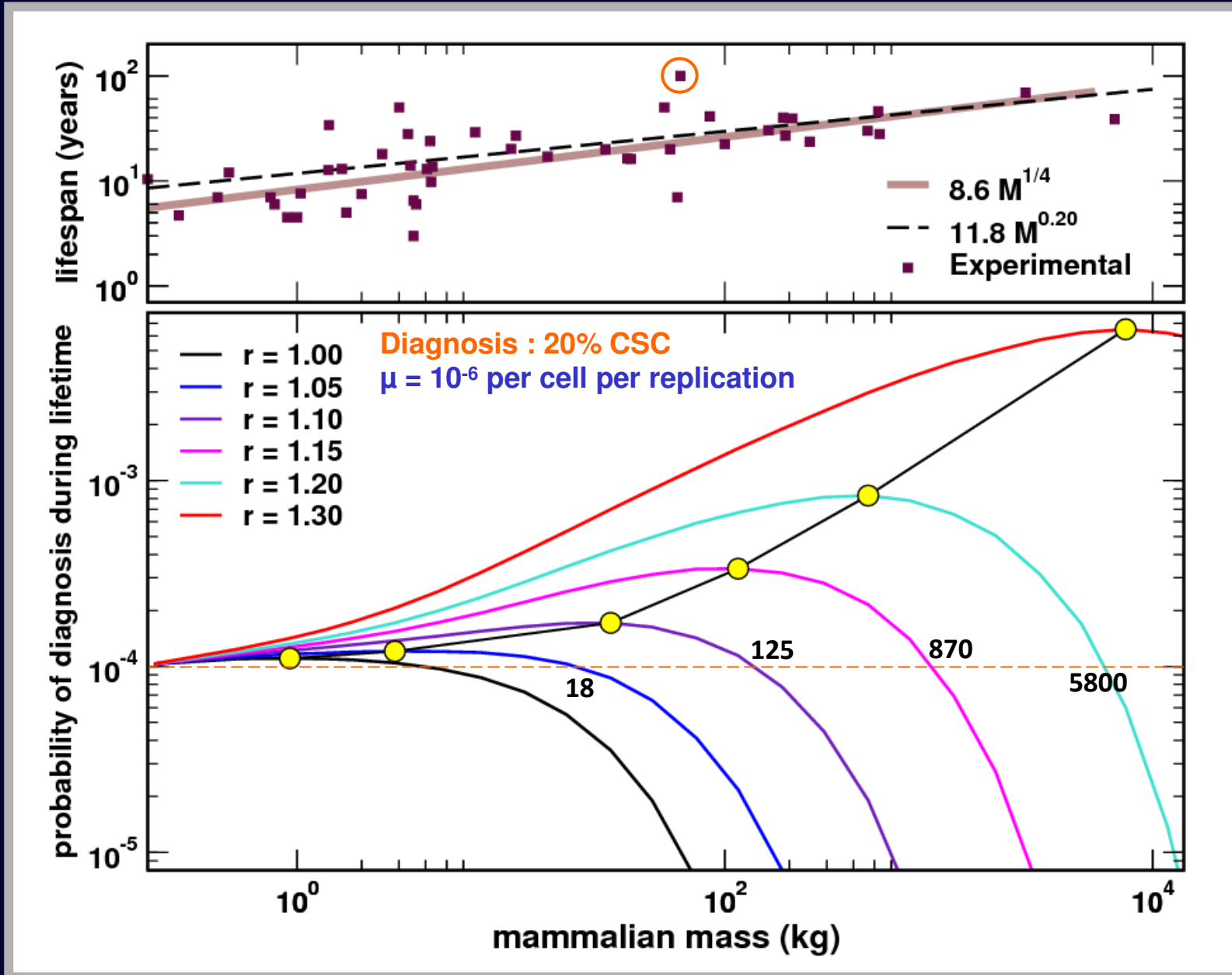
- ❖ *a patient is diagnosed in association with some level of disease burden*
- ❖ *diagnosis can only happen during the lifetime of the organism*
- ❖ *... which means there may be no time for diagnosis to happen*
- ❖ *for an organism with a finite lifetime, in which disease means some threshold is surpassed, selection & mutation play a curious game ...*
- ❖ *if we assume that disease is equally represented in all cell lineages, we may look at dynamics within the **HSC** compartment only ( not always true )*

*the previous model provided **average values***

***stochastic dynamics** → **time distribution functions**  
**probability during lifetime***

# on the small number of *HSC*

❖ is there a good reason for a small *HSC* pool ? ( use scaling for all *M* )



## on the small number of *HSC*

- ❖ *is there a good reason for a small HSC pool ?*
- ❖  *$r$  is usually very difficult to determine experimentally; unfortunately, it is consensual that, in general,  $r$  is large (  $>1.5$  )*
- ❖ *when  $r \sim 1$ , large mammals are more protected than small mammals;*
- ❖ *when  $r > 1.3$ , small mammals are more protected, since the probability for the organism to acquire cancer mutations is minimized;*
- ❖ *a small active HSC pool minimizes the risk of mutations; once mutations occur, the path to full blown disease opens up easily (whenever  $r > 1$ ).*

*how about the probability distribution functions ?*

# stochastic dynamics of *HSC* in Humans

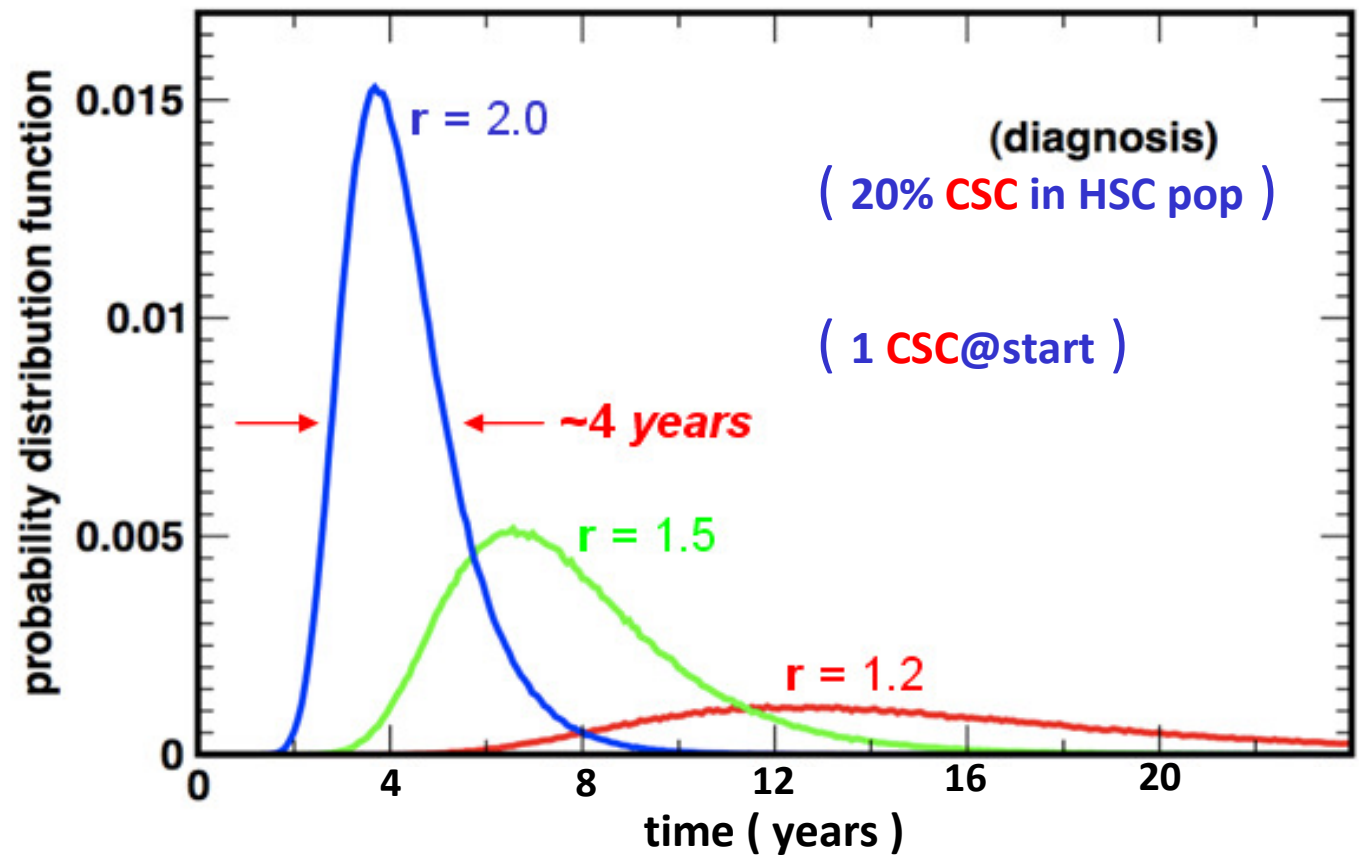
**disease diagnosis** : 20% “blasts” in AML

( acute myeloid leukemia )

10% of plasma cells in MM

( multiple myeloma )

how much time is required  
for a **mutation to develop**  
& give rise to diagnosis of  
a HSC disorder ?



# stochastic dynamics of *HSC* in Humans

**disease diagnosis** : 20% “blasts” in AML

( acute myeloid leukemia )

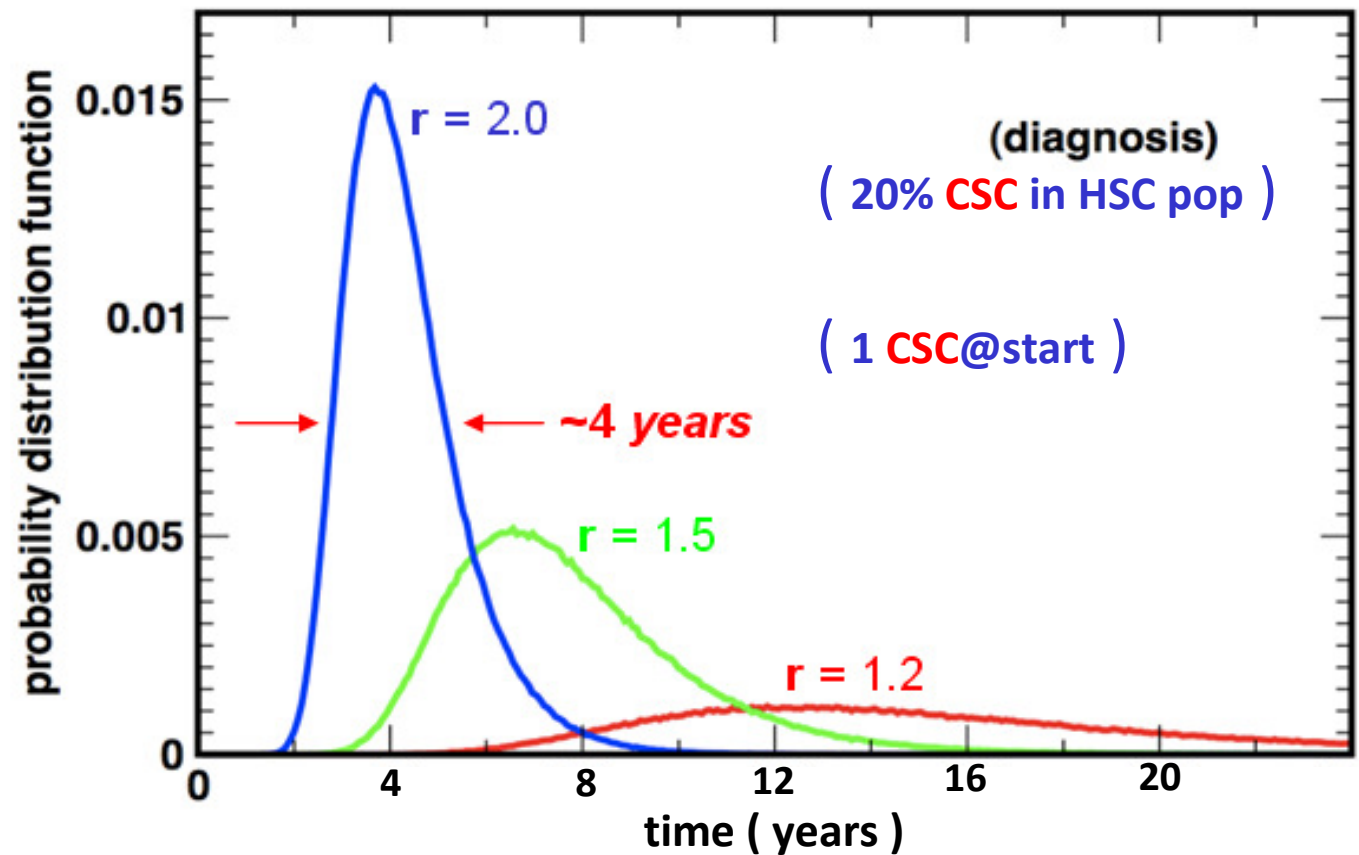
10% of plasma cells in MM

( multiple myeloma )

how much time is required  
for a **mutation to develop**  
& give rise to diagnosis of  
a HSC disorder ?



even for  $r=2$ , the FWHM  
is  $\sim 4$  years !



**stochastic effects at play in specific diseases**

# paroxysmal nocturnal hemoglobinuria

## what is known :

- ❖ rare disease
- ❖ true **HSC** disorder, since it originates in the PIG-A gene of a **HSC**
- ❖ rate of PIG-A gene mutation is known to be normal
- ❖ often BMF is later observed

## conventional wisdom regarding disease development :

- ❖ 1<sup>st</sup> mutation is neutral but a 2<sup>nd</sup> mutation leads to a fitness advantage of PNH cells → disease expansion ( **rare event** )

Dingli, Pacheco & Traulsen, *Physical Review E* 77 (2008) 021915

upper limit for the appearance of a 2<sup>nd</sup> mutation until the first one leads to diagnosis

$$F < \mu N_0 t_M^1 = \mu \frac{N_0^2}{M} \sum_{i=1}^{M-1} \frac{M-i}{N-i} < 10^{-3}$$

- ❖ **relative fitness advantage** of PNH cells due to an **immune attack to normal HSC** → disease expansion

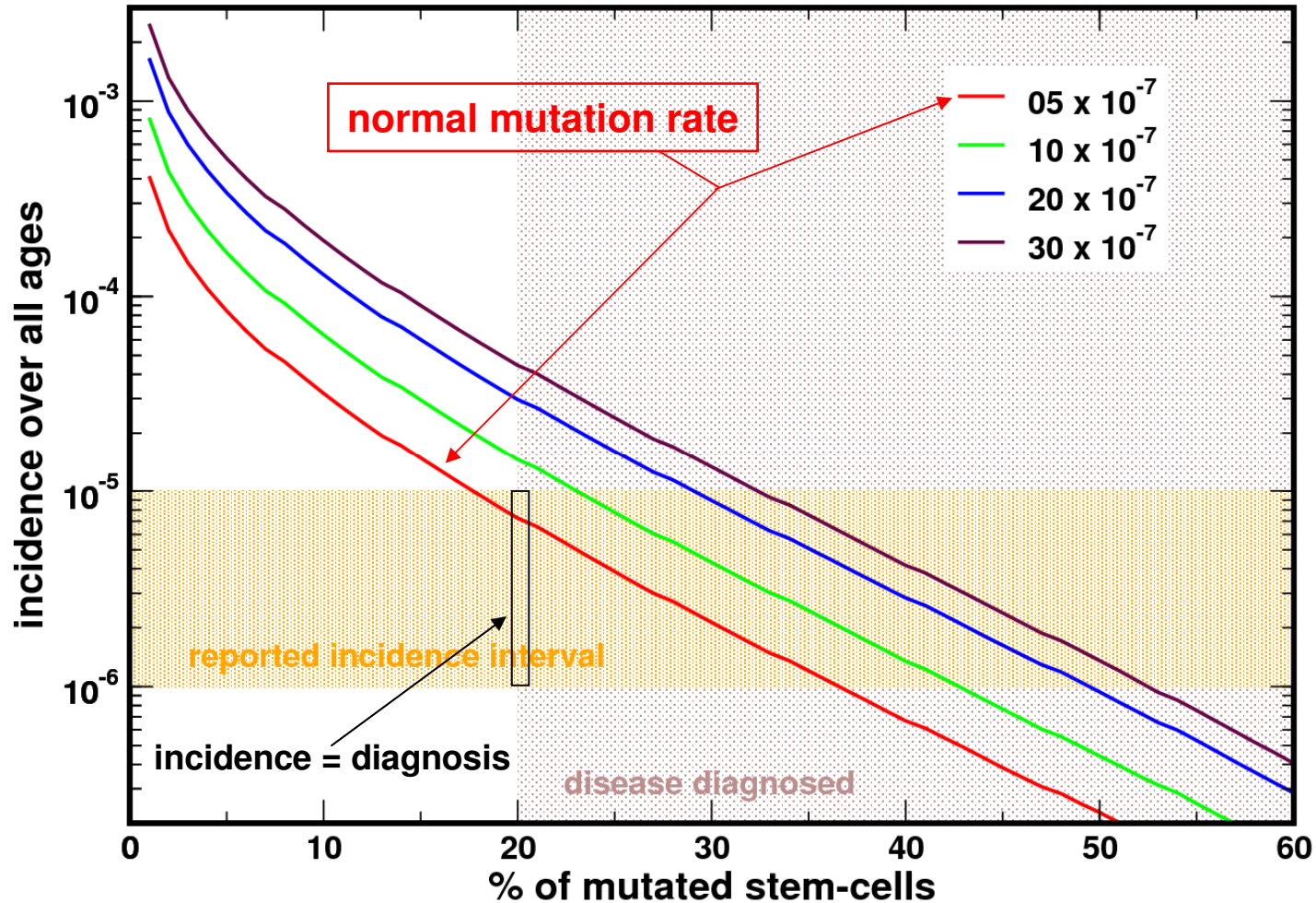


## PNH - model features

### *disease development*

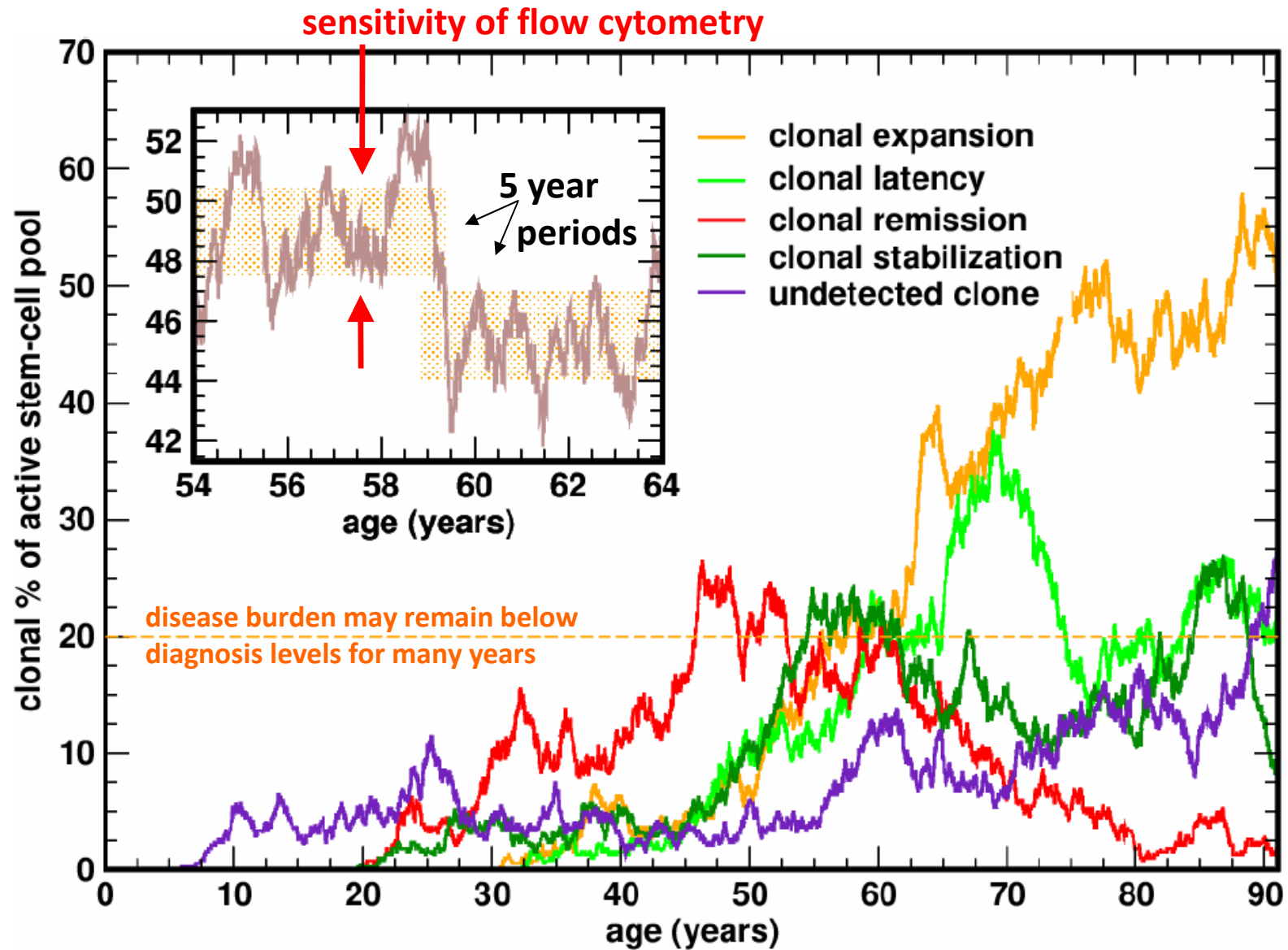
- ❖ use  $N_{sc} = 400$
- ❖ simulate **HSC** activity in virtual USA ( $10^9$  virtual Americans)
- ❖ use normal mutation rate for HSC → PNH transformation
- ❖ assume *neutral drift* ( $r=1$ ) between **HSC** & **PNH** cells
- ❖ fold data with CENSUS 2000 for USA population
- ❖ compare results with incidence data in *USA*

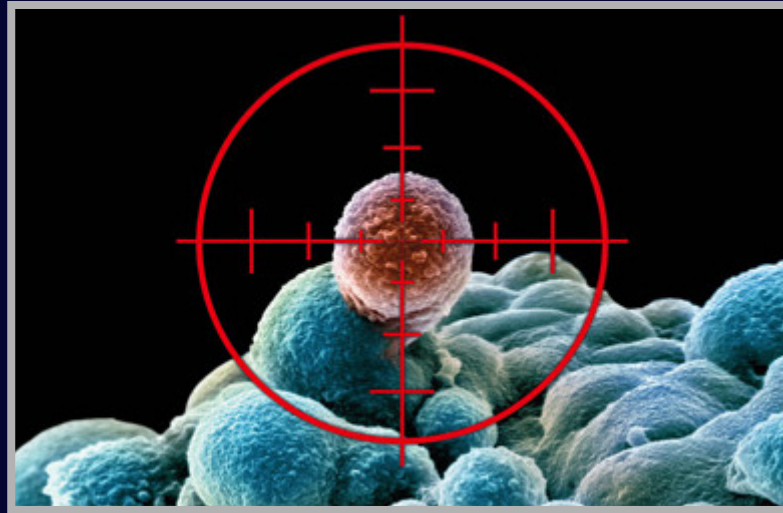
# results



results above (& other results) suggest that **it is not necessary to invoke a relative fitness difference to explain** incidence of PNH

# results – individual history & variable outcomes





*neutral evolution* relies on the stochastic nature of cell behavior, & **PNH** shows us that, likely, many individuals suffer the PIG-A mutation but are never diagnosed PNH, as it is more likely for the mutant to become extinct than to evolve into a clone. This, in turn, suggests that the current way of approaching the (now over) 40-year old war-on-cancer, that is,

**cure = kill-every-single-cancer-cell**

is perhaps not always the best, or sometimes maybe even unnecessary. In fact, profiting from competition through natural selection may turn out to be a more viable strategy.

# Chronic Myeloid Leukemia

## what is known :

- ❖ Hematopoietic stem cell disorder
- ❖ Initial event: Philadelphia chromosome
- ❖ ? HSC are enough to drive chronic phase ?
- ❖ clonal expansion and myeloproliferation
- ❖ stem cell derived but progenitor cell driven
- ❖ *abl*-kinase inhibitors very effective

## CML dynamics

- ❖ Q-RT-PCR data from patients treated with *imatinib*
- ❖ 2 data sets available
  - ❖ Michor *et al*, *Nature*, 2005
  - ❖ Roeder *et al*, *Nature Medicine*, 2006
  - ❖ other data recently available for *nilotinib*
- ❖ **data fitting** ( using deterministic model )

## CML - model features

### *disease development*

- ❖ use existing model of hematopoiesis
- ❖ how to get from HSC origin to progenitor driven disease ?
- ❖ bone marrow expansion  $\rightarrow \epsilon_{\text{CML}} < \epsilon_0$

### *treatment*

- ❖ how does *imatinib/nilotinib* work ?
- ❖ does *imatinib/nilotinib* induce cell death?
- ❖ how many cells are responding to *imatinib/nilotinib* ?

## CML - model constraints

### *disease development*

- ❖ time from initial insult to diagnosis is 3.5 – 6 years
- ❖ progenitor cell expansion >14%
- ❖ total number of active *HSC* is *not* increased
- ❖ daily bone marrow output is ~ 3 x normal

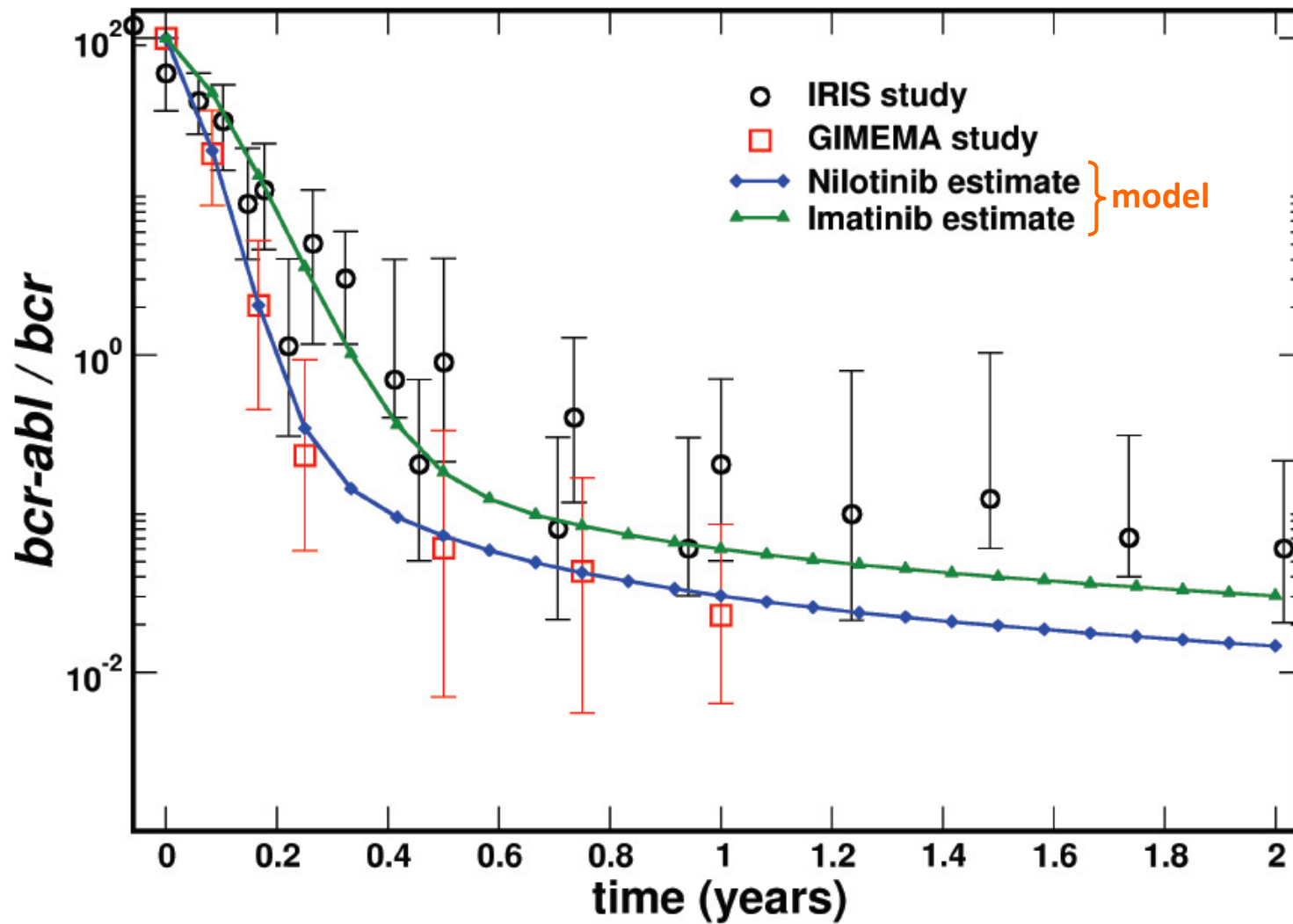
### *treatment*

- ❖ *imatinib/nilotinib* leads to  $\epsilon_{\text{IMAT}} > \epsilon_0 > \epsilon_{\text{CML}}$
- ❖ *imatinib/nilotinib* does not affect *HSC*
- ❖ at any time a fraction *z* of cells responds to *imatinib/nilotinib*

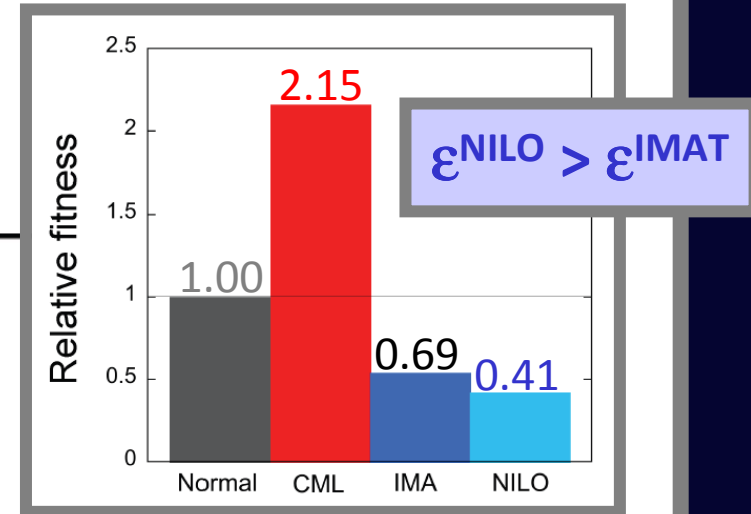
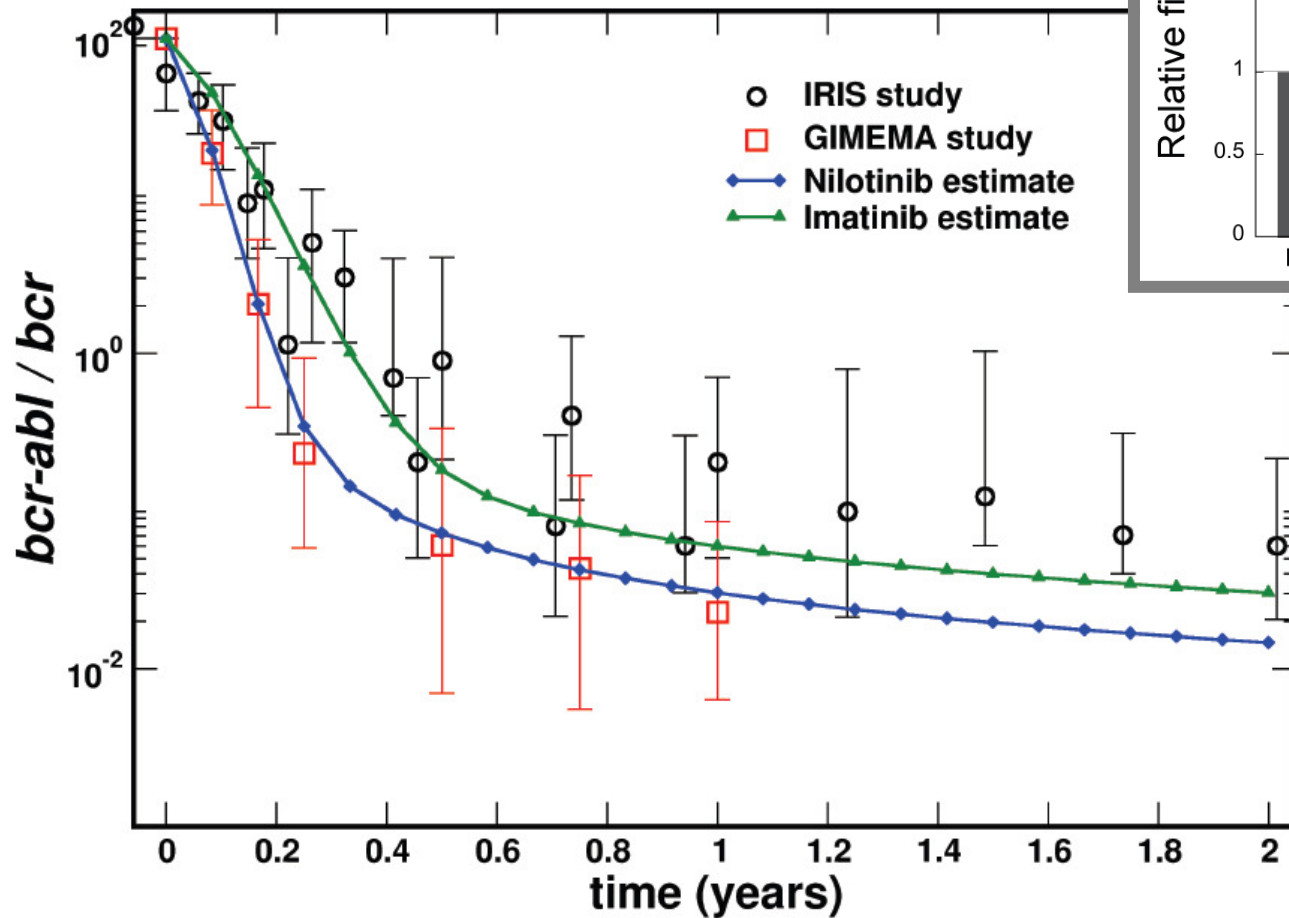
we extract  $\epsilon_{\text{IMAT}} > \epsilon_0 > \epsilon_{\text{CML}}$  & *z* from data ...



# imatinib ⊗ nilotinib



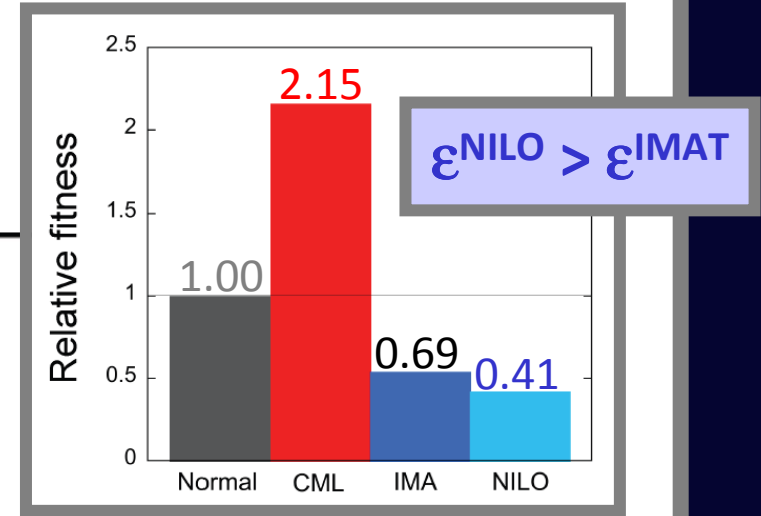
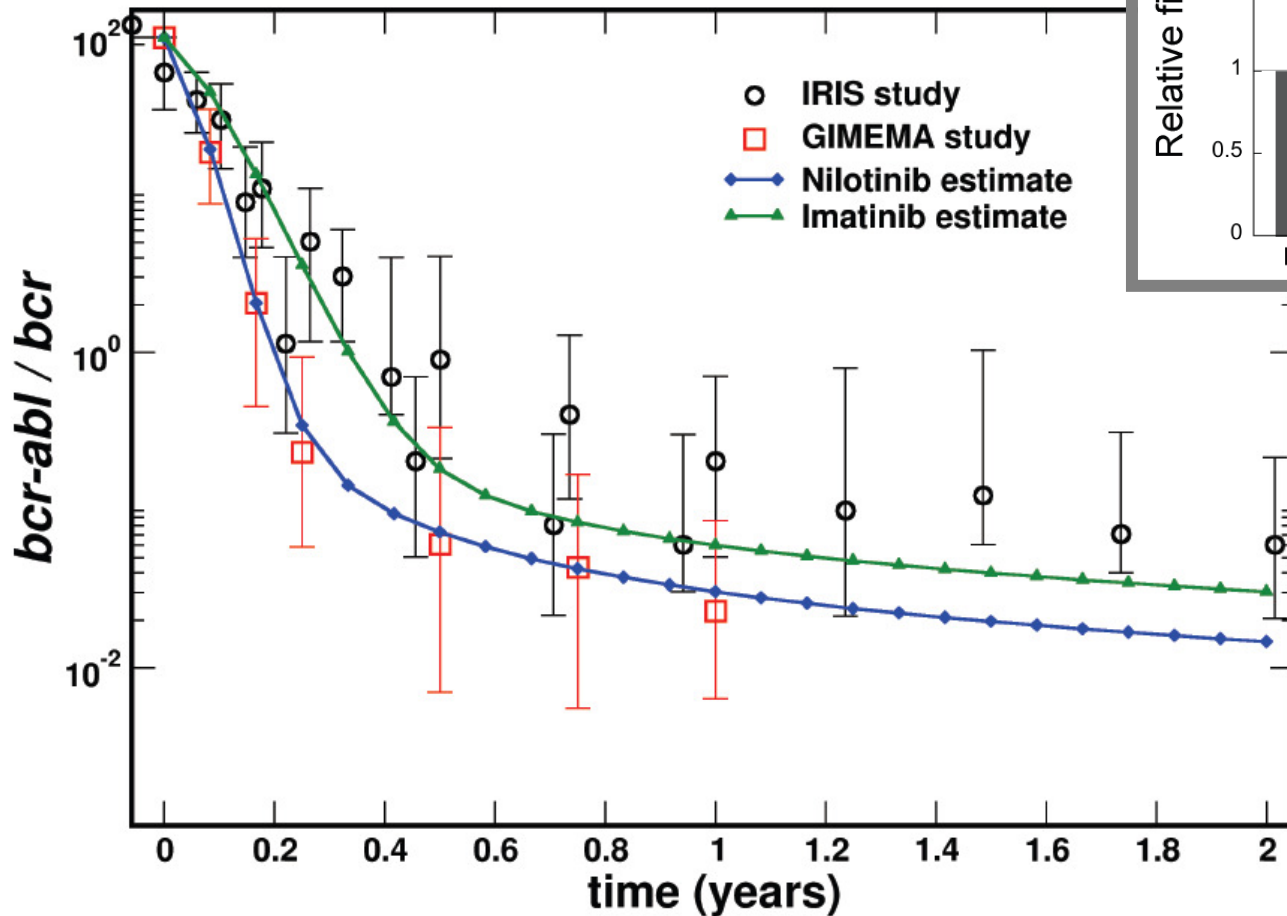
# imatinib ⊗ nilotinib



$$f_j = \frac{\rho_j}{\rho_0} = \frac{1 - \epsilon_j}{1 - \epsilon_0} \frac{\epsilon_0}{\epsilon_j}$$

# imatinib ⊗ nilotinib

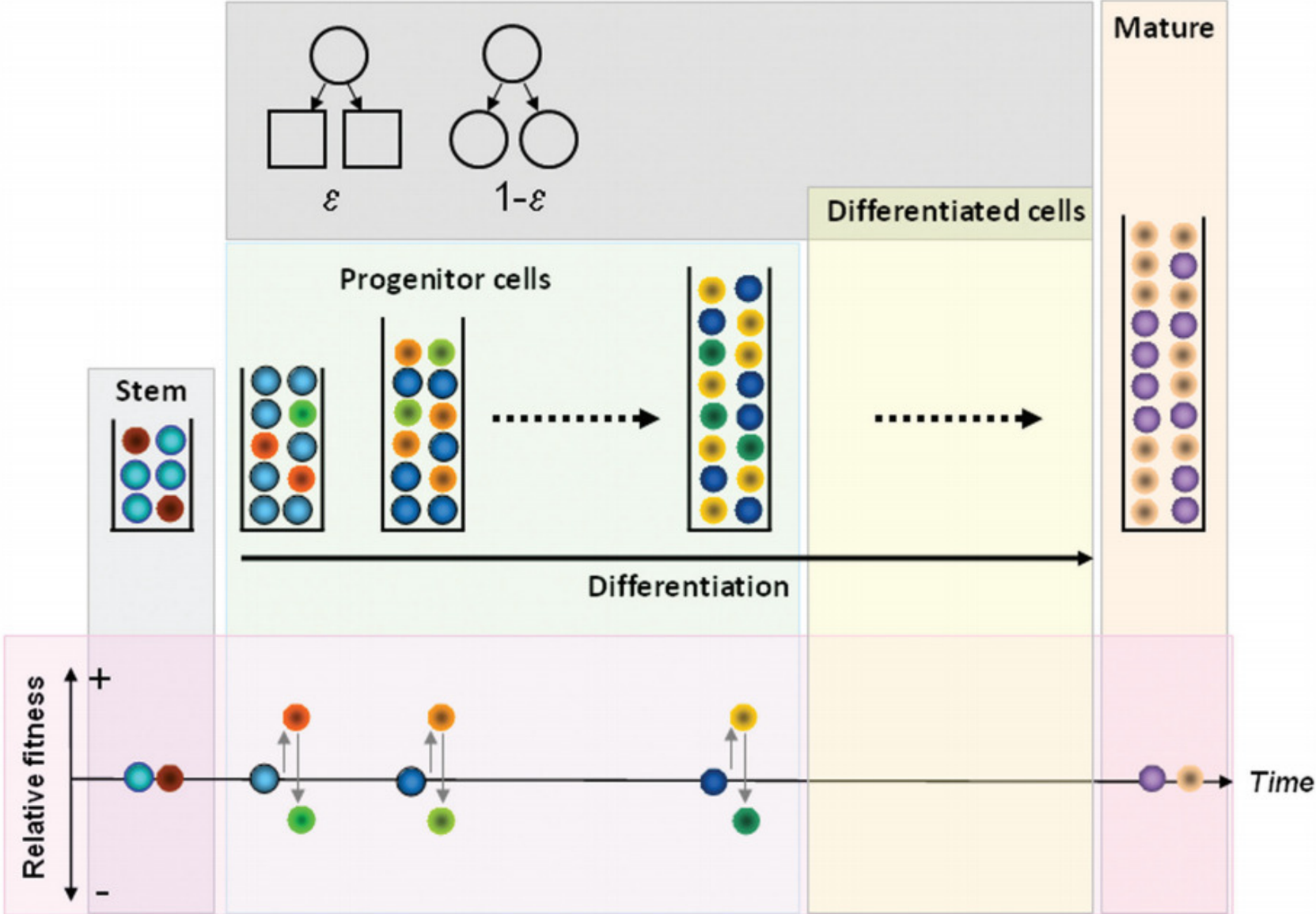
**in vitro** : **no** differences detected  
**in vivo** : **important** differences  
 (ecology of cancer cells is important)



$$f_j = \frac{\rho_j}{\rho_0} = \frac{1 - \epsilon_j}{1 - \epsilon_0} \frac{\epsilon_0}{\epsilon_j}$$

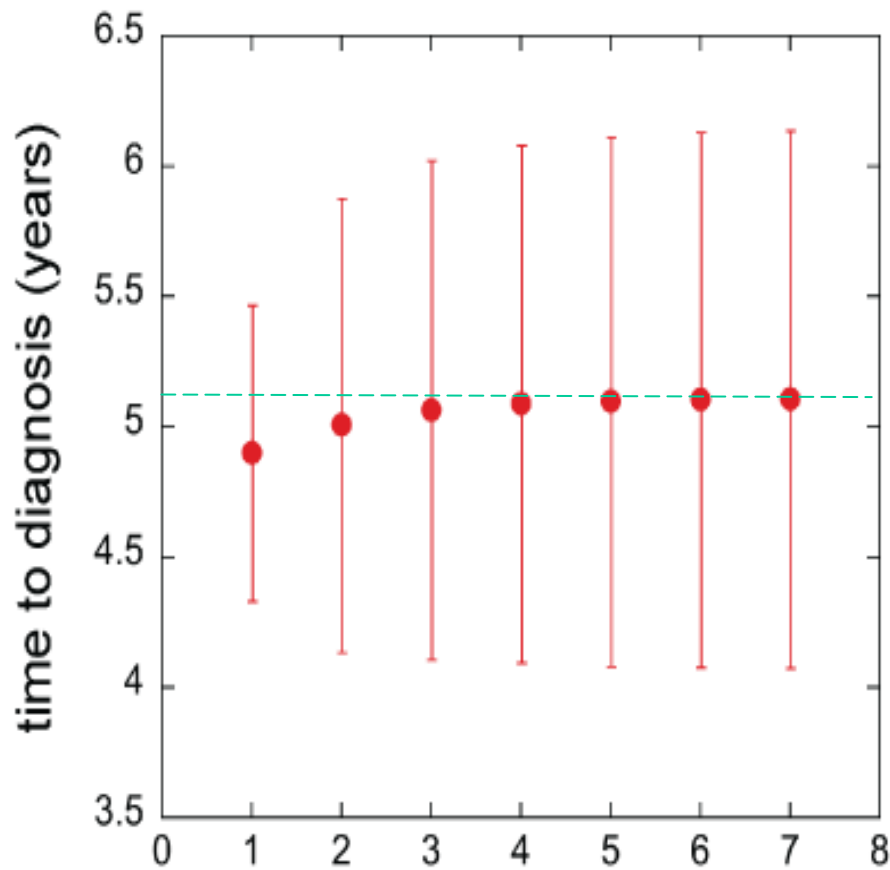
neat way to deduce fitness from data

# evolutionary dynamics of CML

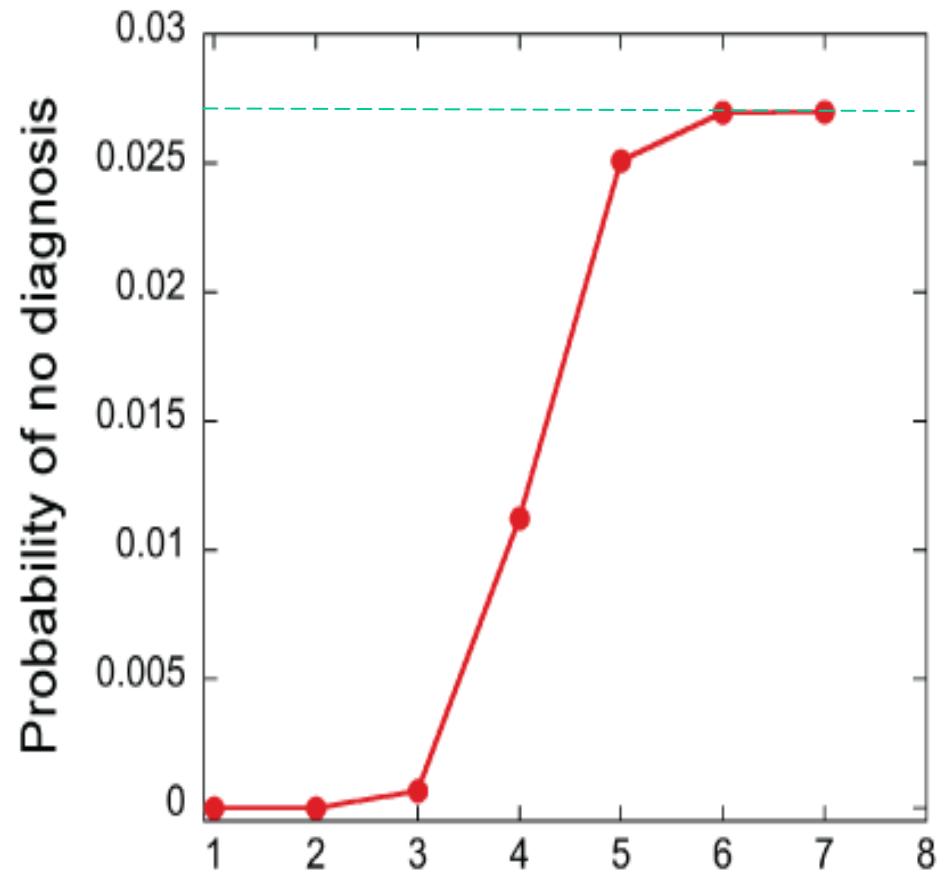


# stochasticity in CML

... stochastic dynamics of  $10^{12}$  cells is unfeasible → **hybrid model**

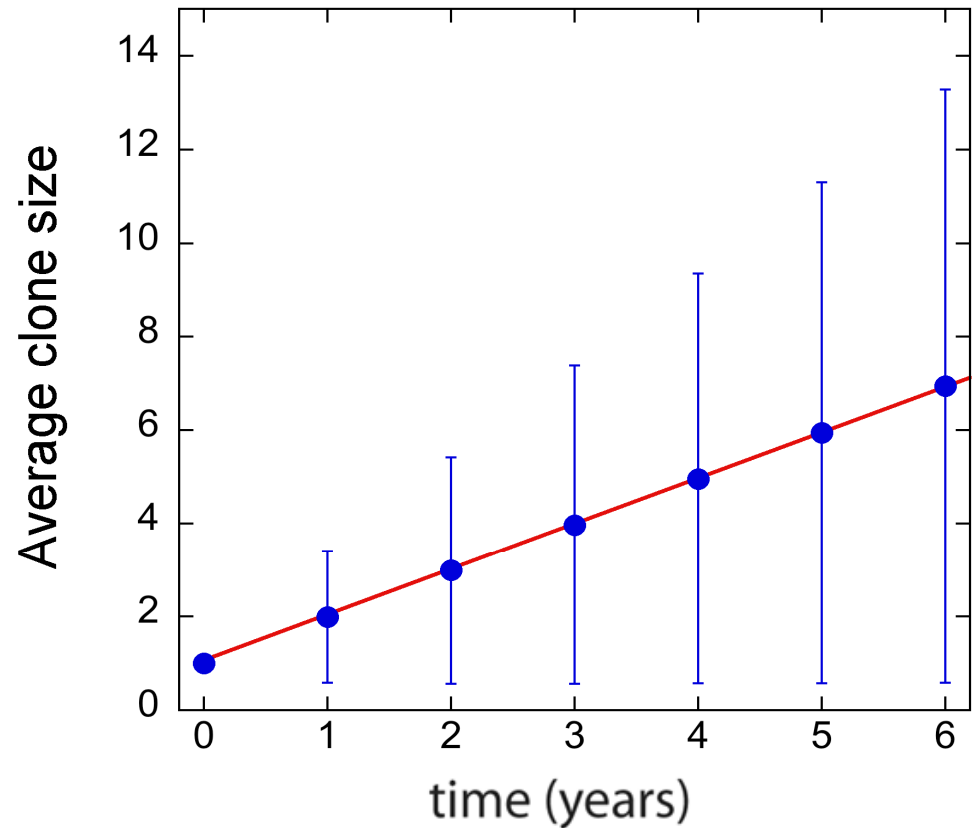
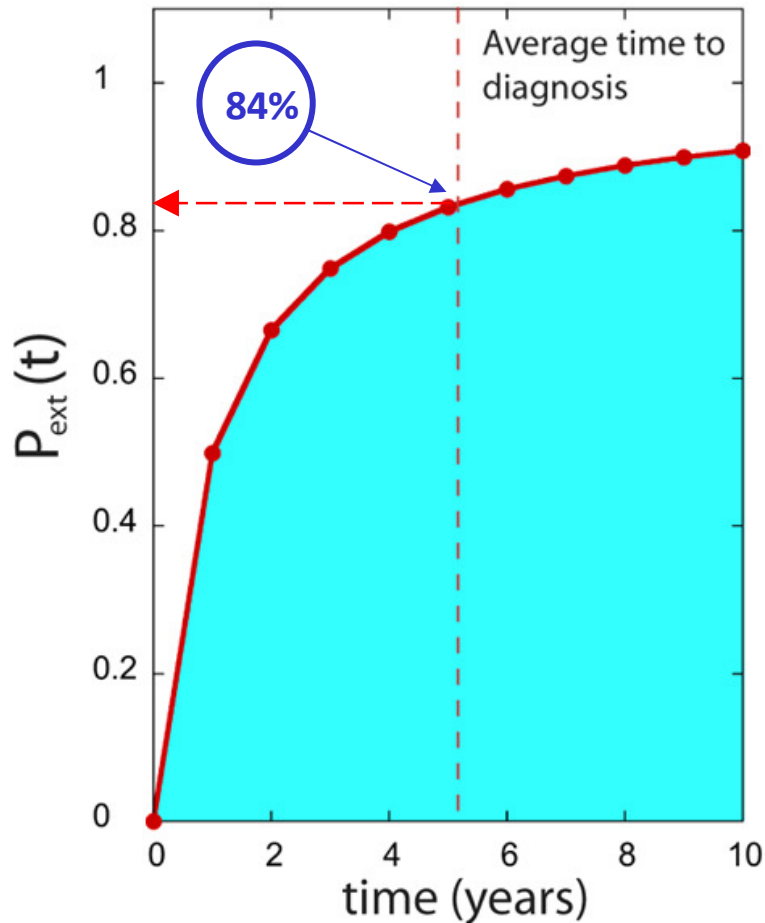


Number of stochastic compartments



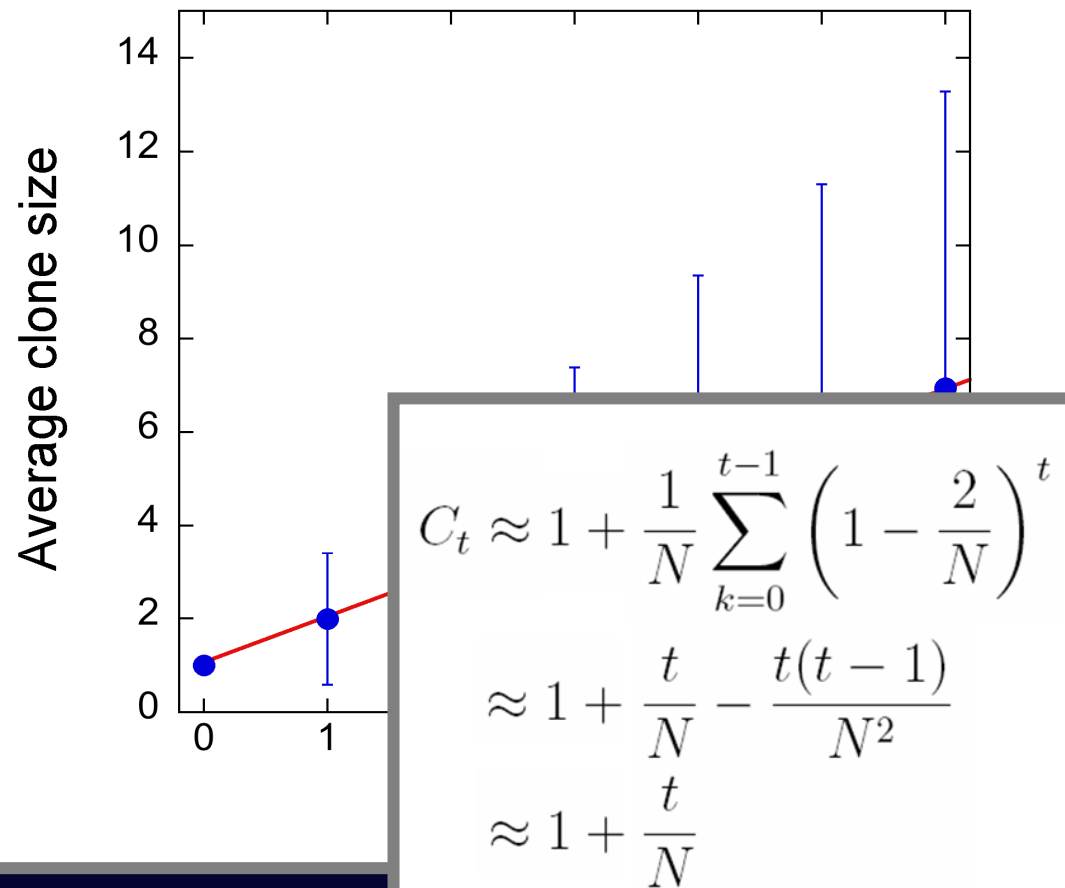
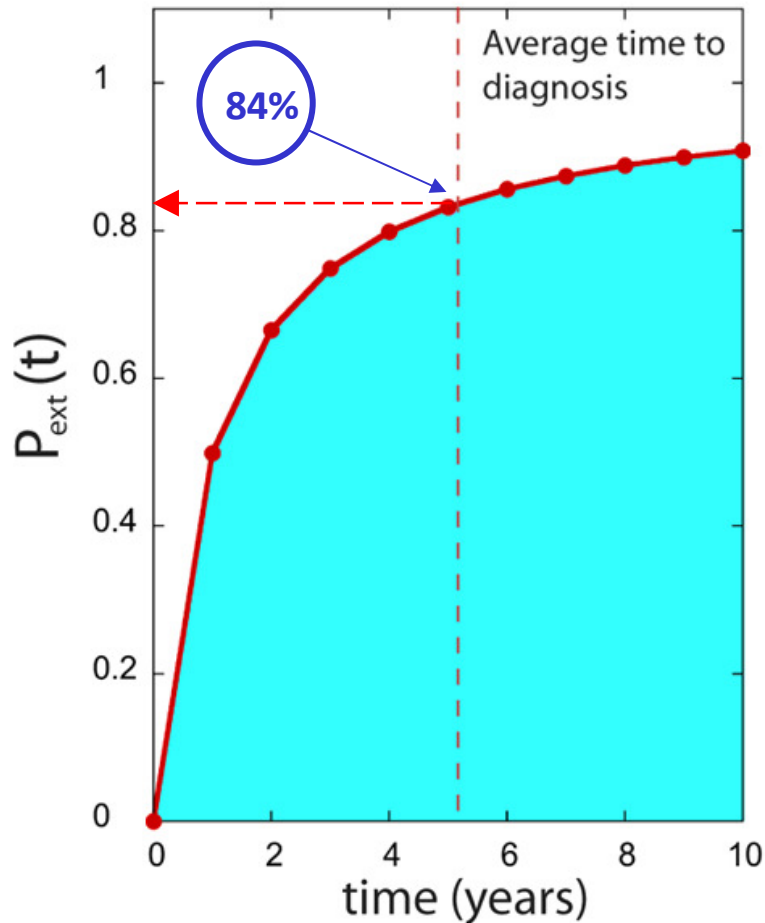
# stochasticity in CML

in **84%** of individuals, **CSC** population goes extinct before diagnosis  
in **16%** of individuals, **CSC** population grows, on average, 1 per year

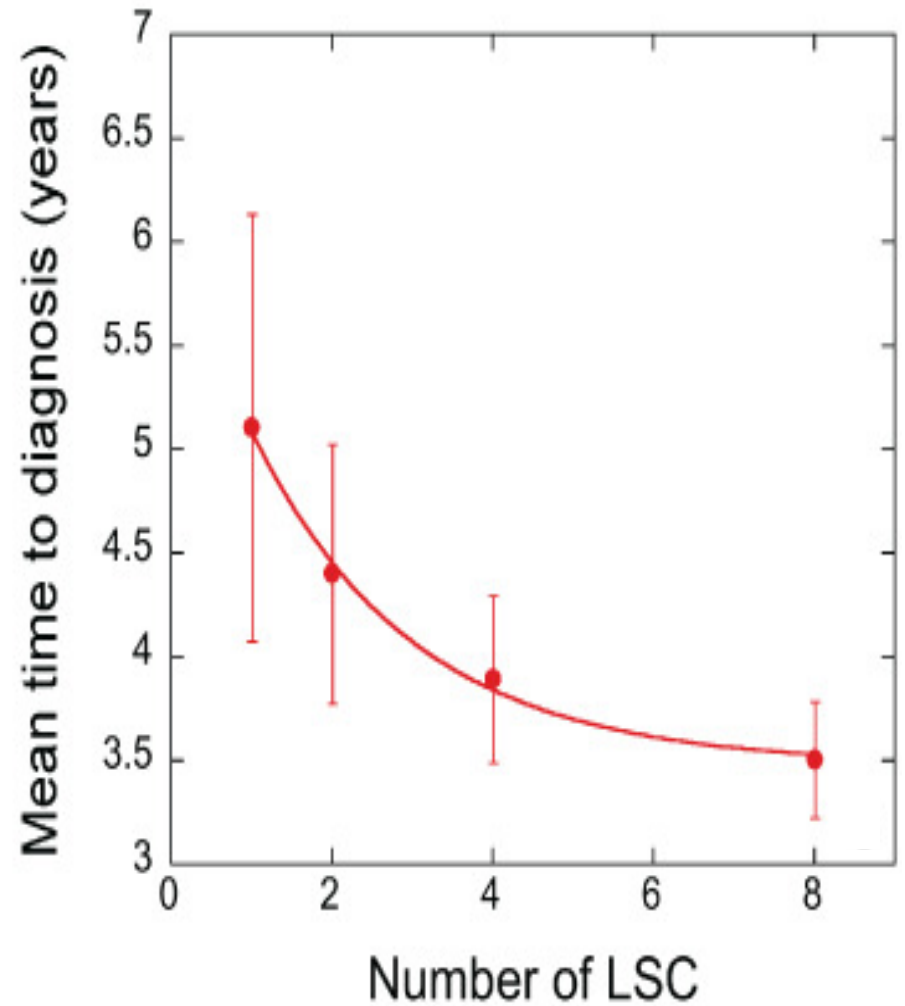
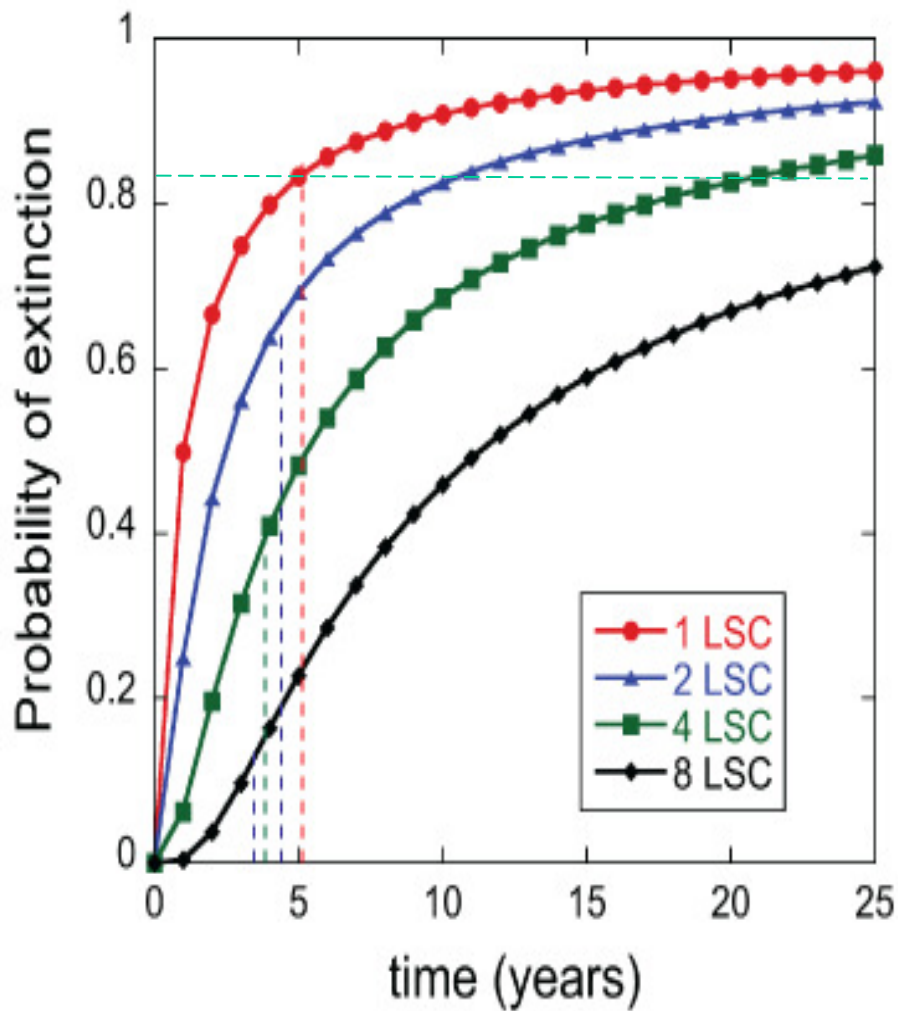


# stochasticity in CML

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# stochasticity in CML

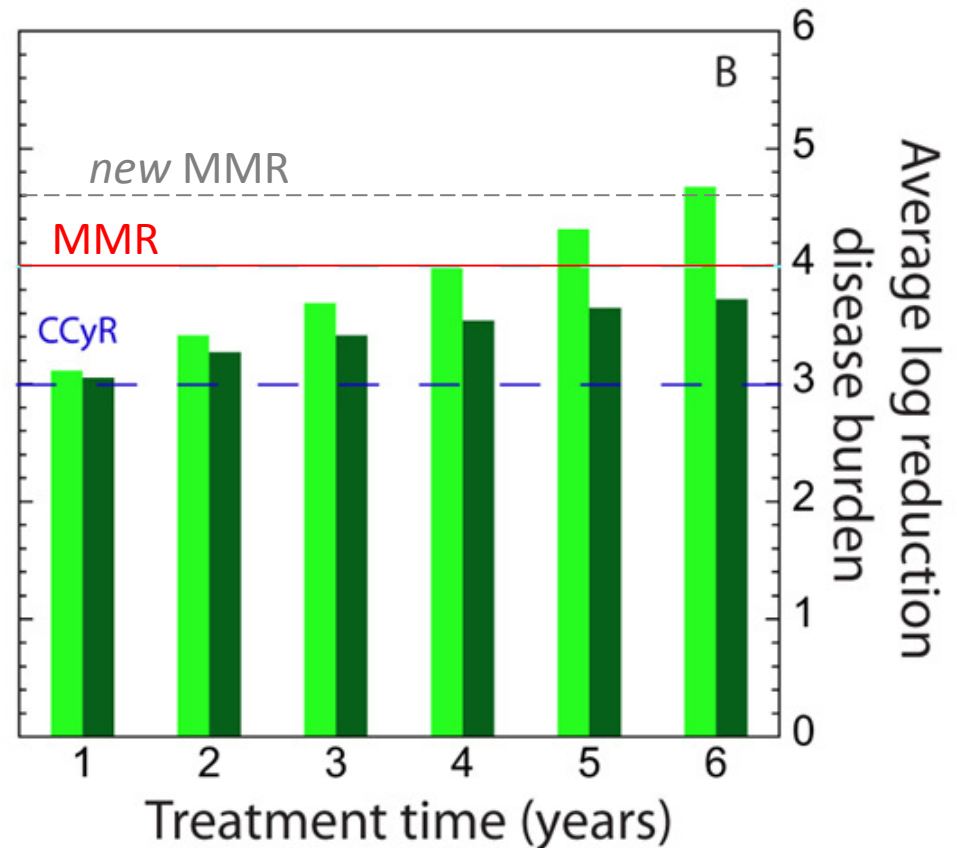
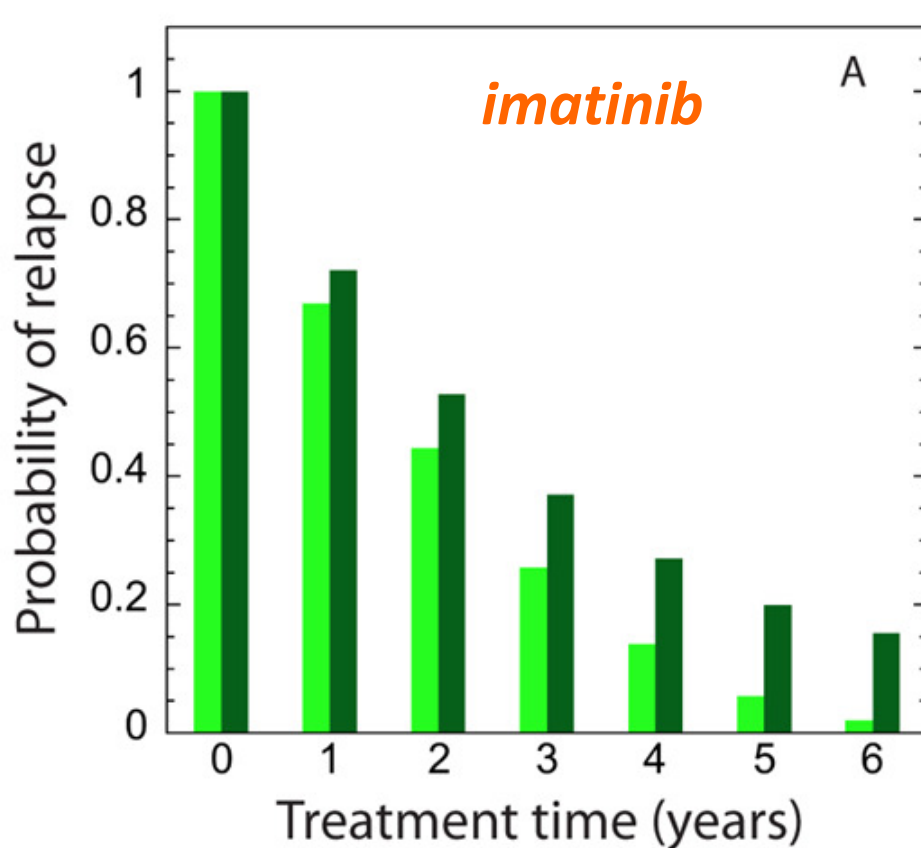




# stochasticity in CML

c1 ■ no LSC @ diagnosis

c2 ■ including 16% patients with LSC @ diagnosis



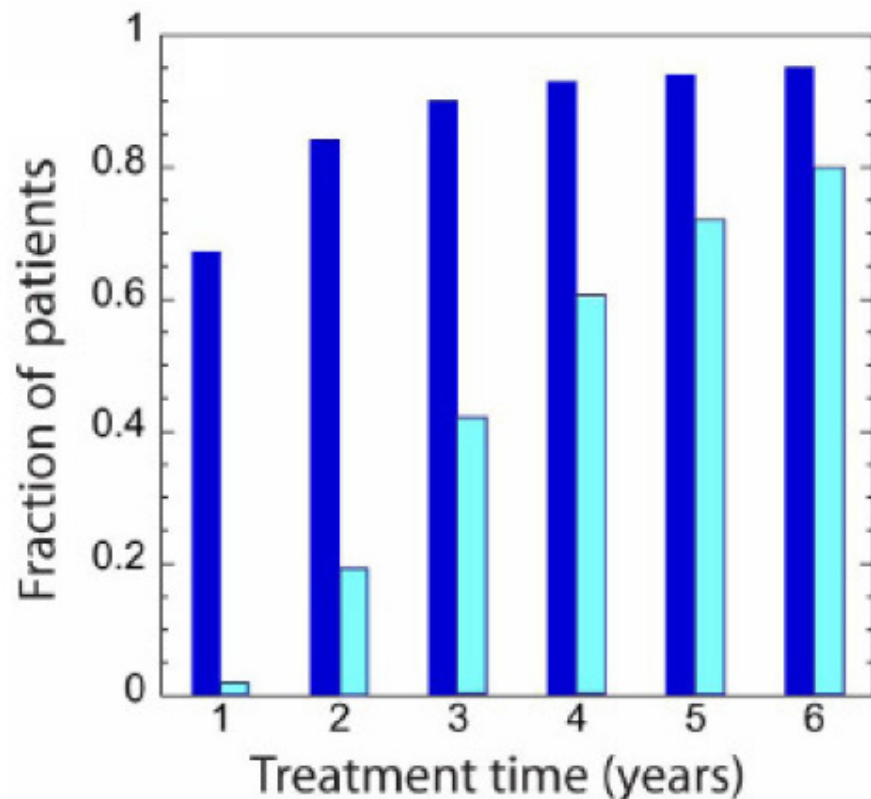
despite **NOT** affecting directly **CSC**,

**imatinib + natural selection** can cure the majority of **CML** patients

ongoing: development of **resistance mutations** . . .

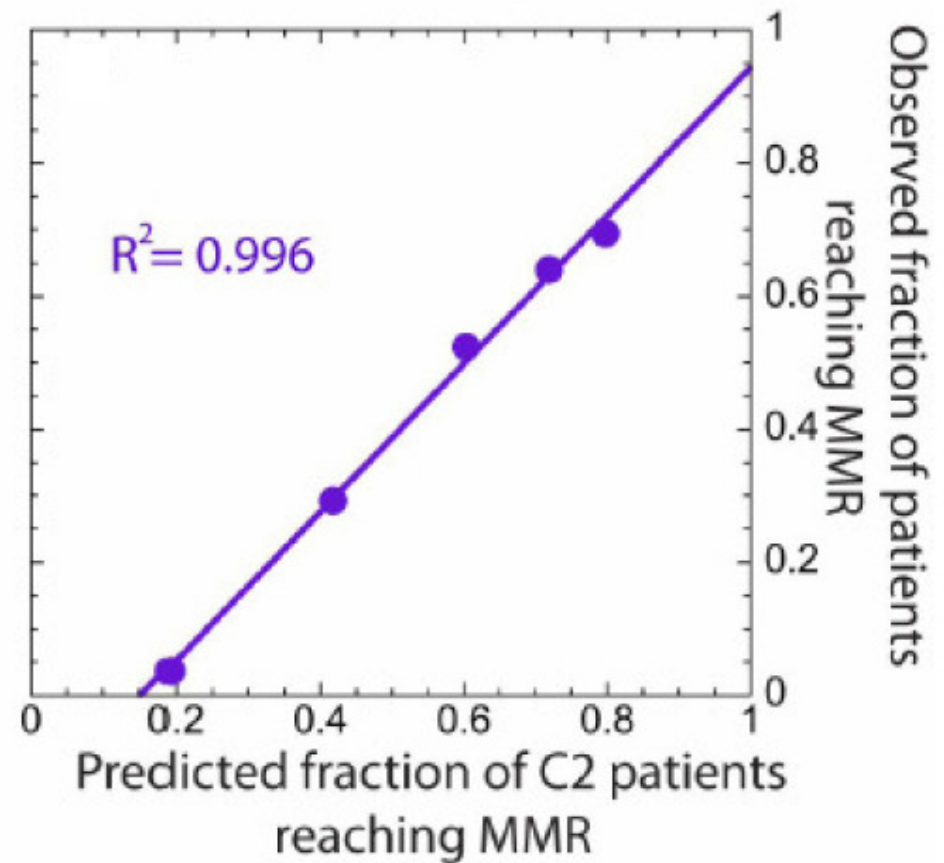
# stochasticity in CML

treatment with TKI-inhibitors helps an individual to stay alive and live his everyday life while natural selection helps him getting rid of the cause of the disease; however, it takes years for one to **gamble his way out of cancer**.



*imatinib*

■ Complete Cytogenetic Response (CCyR)

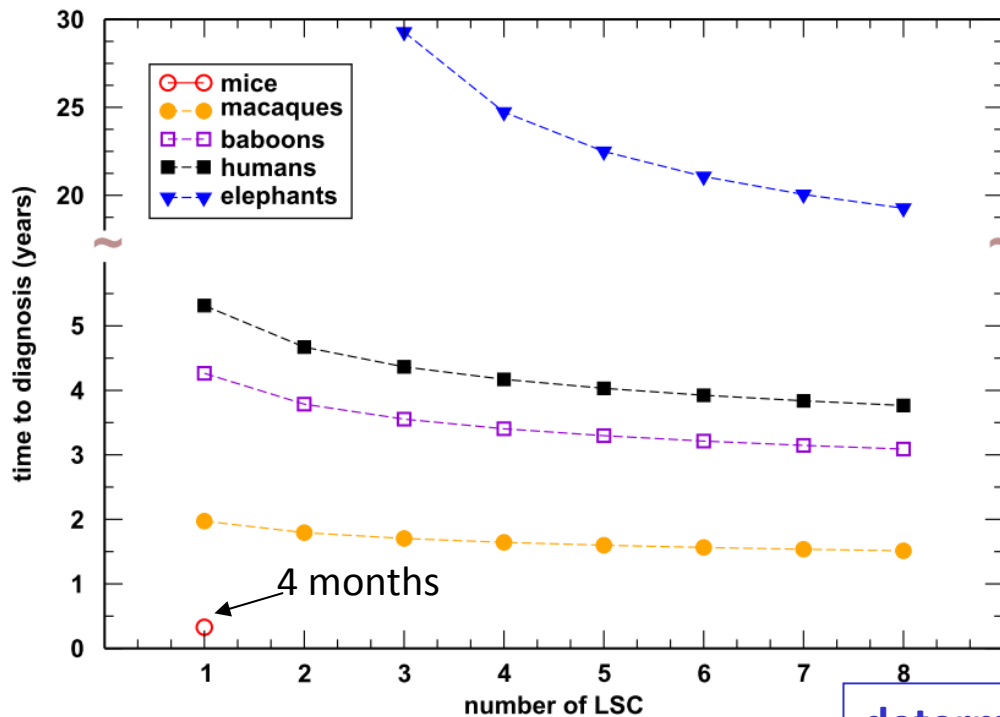


■ Major Molecular Response (MMR)

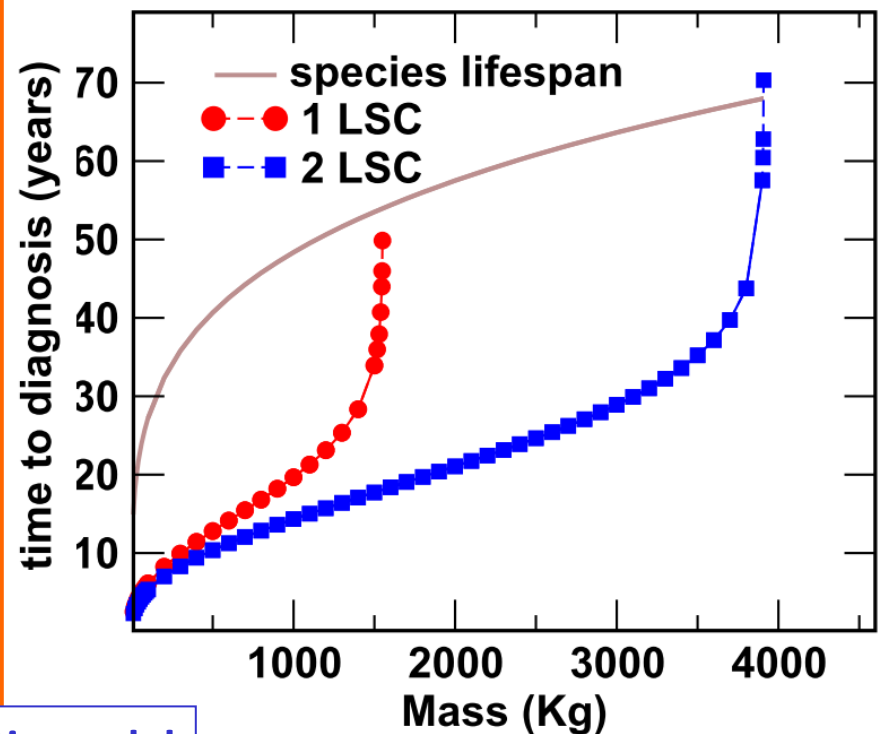
# CML in other mammals

there is no reason, *a priori*, to suppose that what we observe in humans stays with humans; how will CML proceed in other mammals ?

- ❖ **HSC** population remains constant ( $16.55 M^{3/4}$ )
- ❖ **HSC** & **CSC** divide at normal rate ( $2.9 M^{-1/4}$ )
- ❖ how many **CSC** drive (or are required to drive) **CML** in other mammals ?
- ❖ how many compartments will behave stochastically in other mammals ?

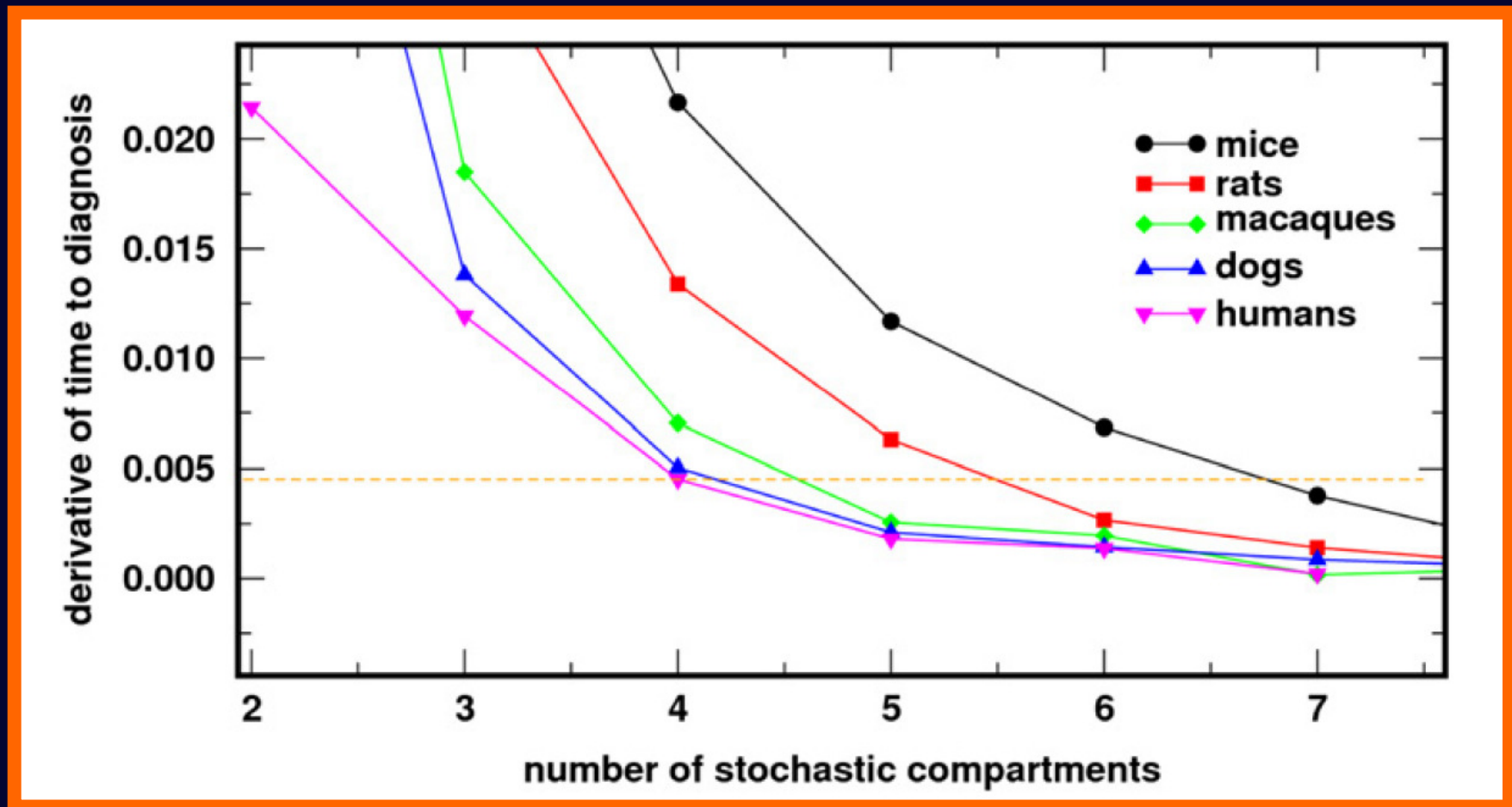


deterministic model



# CML in other mammals

to which extent do stochastic effects remain important in CML on other mammals ?



forward difference formula for the derivative of the TTD as a function of the # of compartments treated stochastically, taking as reference  $K=4$  for humans & 1 CSC@start.

## conclusions

- ❖ *across mammals, hematopoiesis is generated by numbers of **HSC** that may change significantly from species to species, flowing downstream in a multi-compartmental tree-structure in which consecutive compartments interact*
- ❖ *in this simple model, homeostasis is nothing but the stationary solution of the coupled problem.*
- ❖ *this coupled dynamics, together with specific thresholds for disease diagnosis and the finite lifespan of organisms leads to a complex interplay between selection and mutation in hematopoiesis . . .*
- ❖ *. . . where stochastic effects may play an important role and, in some cases, a crucial one.*
- ❖ *in some rare **HSC** diseases (ex: CML), evolutionary dynamics of the disease may favor the patient to get rid of its cause, but this alone may not be enough & treatment may be crucial to keep patient alive*

**END**