







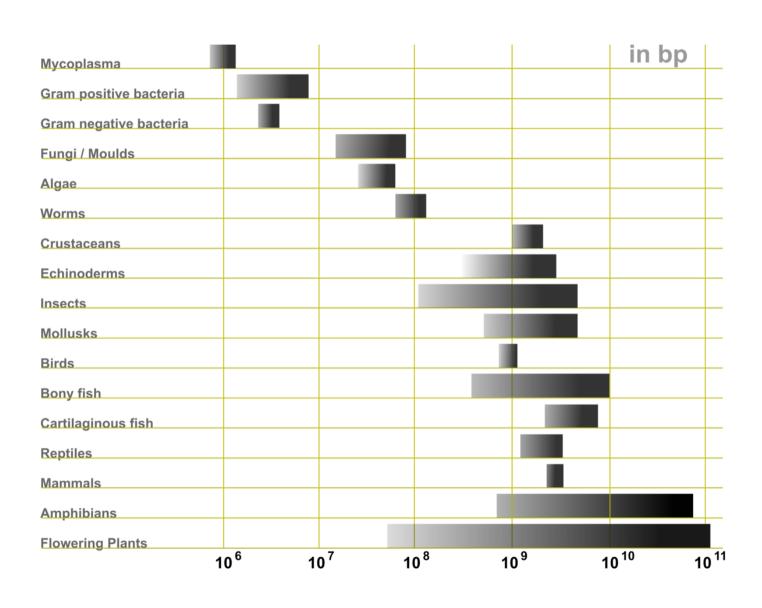
### Genome size evolution: challenging intuition with modelling

Stephan Fischer, Samuel Bernard, Guillaume Beslon, Carole Knibbe

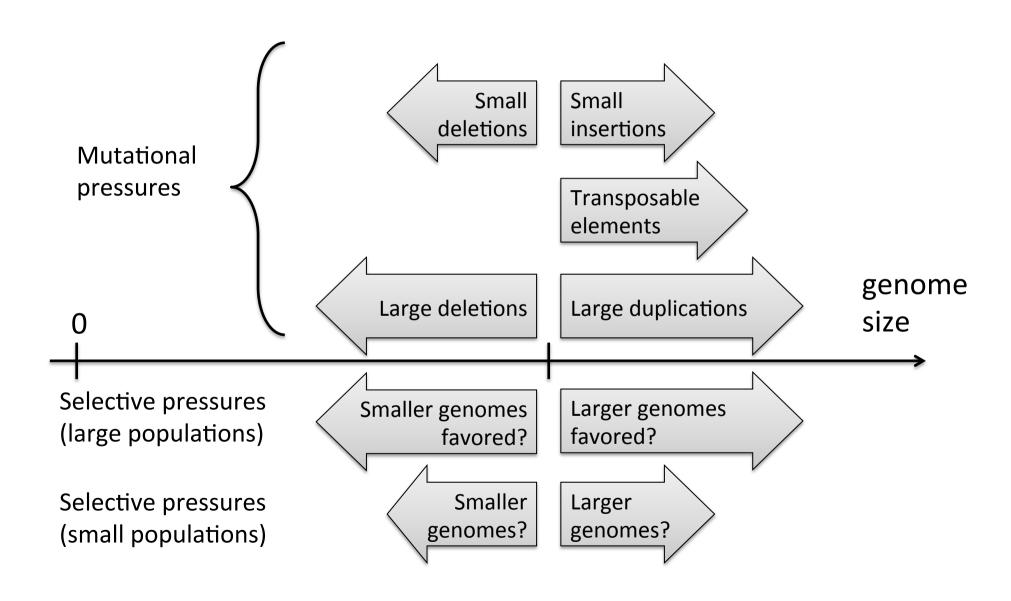
Equipe Inria Beagle, Lyon



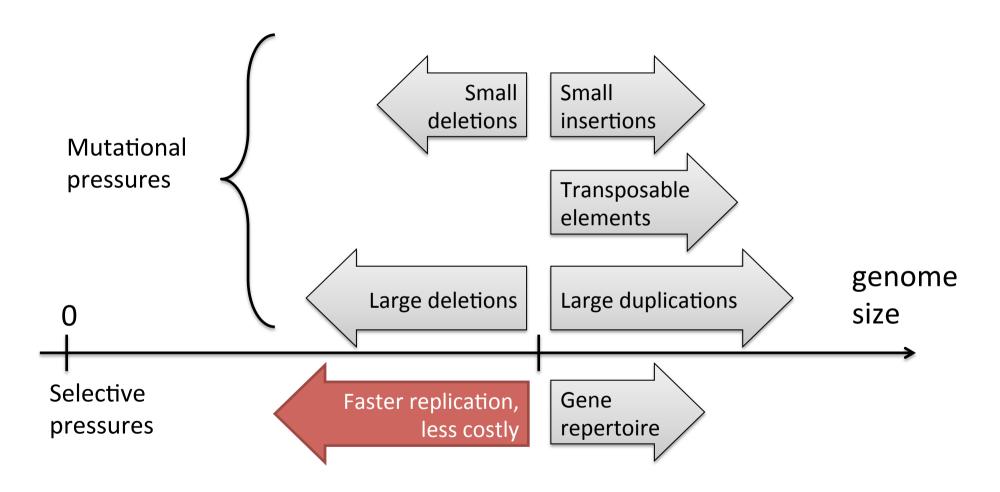
### What determines genome size?



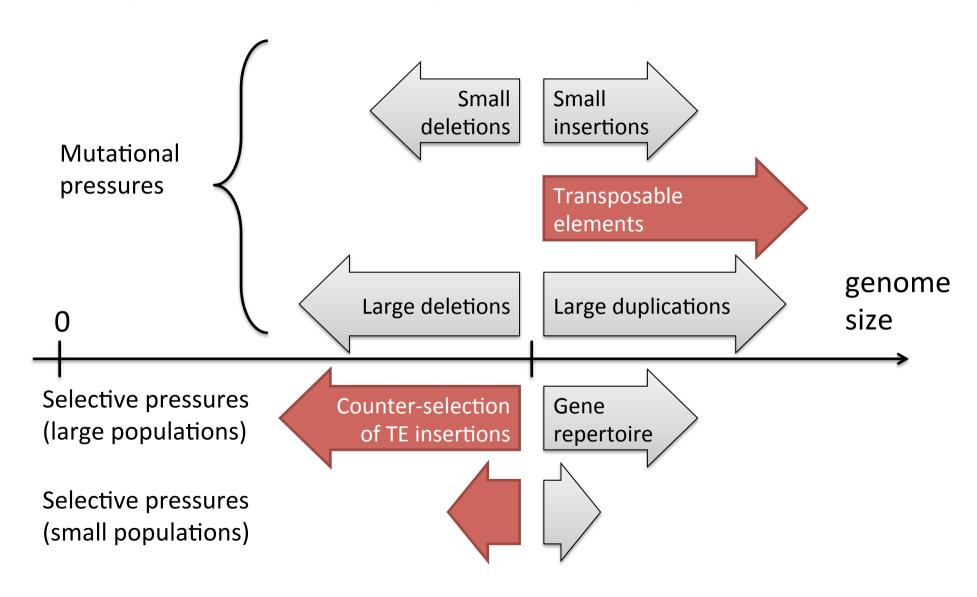
#### The usual « thought experiment »



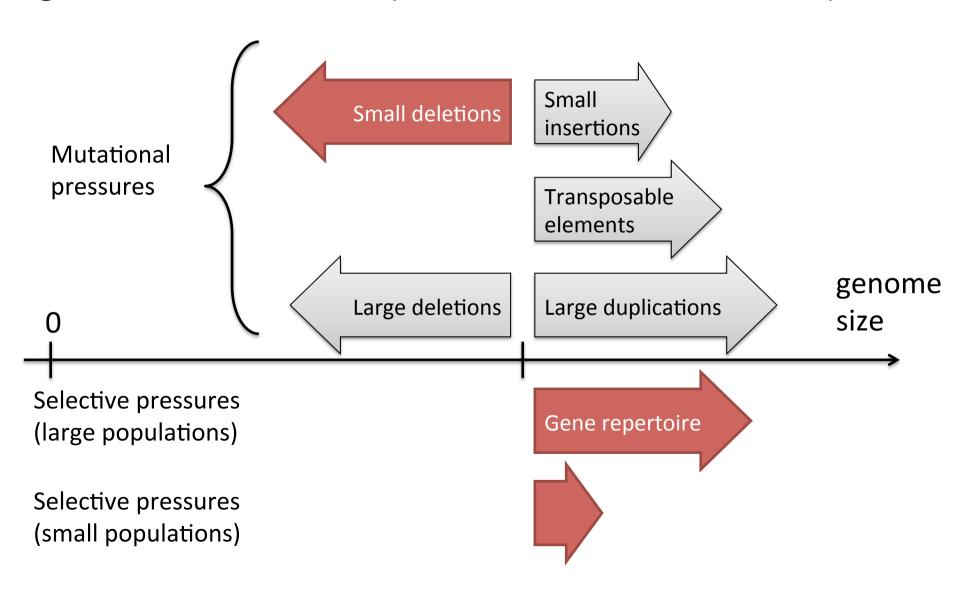
### Theory 1: Shorter is better



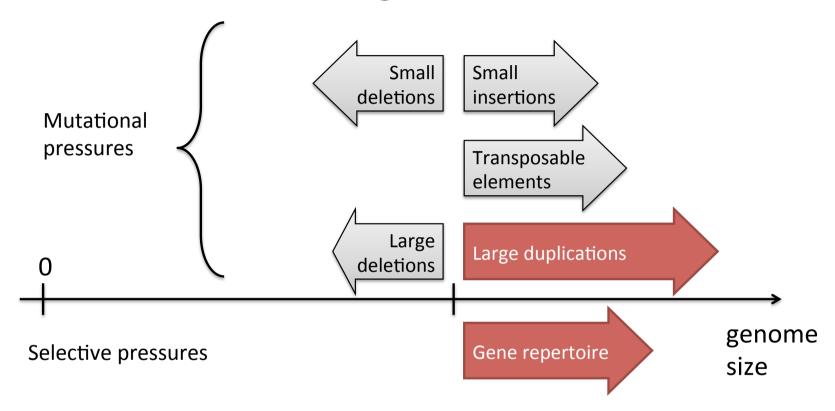
Theory 2: Small populations cannot get rid of transposable elements (Lynch & Conery 2003)



Theory 3: Biases in the small indels mechanisms drive genome size evolution (Petrov 2002, Kuo et al. 2009)



## The question we ask: is the intuitive reasoning correct?



How will genome size evolve if:

- duplications are twice as frequent as deletions,
- transposable elements proliferate,
- and selection systematically favors the highest gene numbers?

# Let's build a minimal model for genome size evolution

#### A minimal model for genome size evolution

Space of all possible genome sizes:  $\mathbb{N}^*$ 

Infinite population, density vector at time t:  $u_t$ 

Transition matrix due to small and large mutations:  $\mathbf{M}_G = ((\mathbf{M}_G)_{ij})_{i,j \in \mathbb{N} \setminus \{0\}}$ 

Evolution in discrete time, without selection (Markov chain):  $u_{t+1} = \nu_t \mathbf{M}_G$ 

#### Possible mutations for a genome of size $s_0$

- Small insertions: + 1 to +  $l_{\it ins}$  bases
- Small deletions: 1 to  $l_{sdel}$  bases (if possible)
- Duplications: + 1 to  $+ s_0$  bases
- Large deletions: 1 to  $s_{\theta}$  bases

The transition probabilities can be defined arbitrarily, but should not depend on the starting size  $s_0$ .

Each possible final state is reached with probability  $1/s_{\theta}$ . (But we will generalize later).



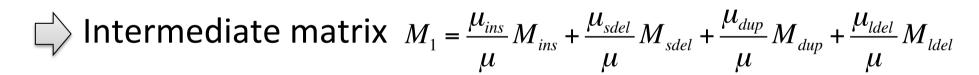
Elementary matrices  $M_{ins}$ ,  $M_{sdel}$ ,  $M_{dup}$ ,  $M_{ldel}$ , where e.g.  $(M_{ins})_{ij}$  is the probability that a genome of initial size i ends up with size j after exactly one small insertion

# Why a uniform distribution for the size of the duplications and deletions?

- Assumption on the underlying mechanism: uniformly distributed breakpoints
- Observations in bacteria: single deletions up to more than 200 kb ≈ 180 genes (Porwollik et al, 2004; Nilsson et al, 2005)
- Observations in humans (Lupski, 2007):
  - in 50% of the cases, the Charcot-Marie-Tooth disease is caused by a 1.4 Mb duplication
  - In 90% of the cases, the Smith-Magenis syndrom is caused by a partial deletion of chromosome 17, spanning from 950 kb to 9 Mb... (9Mb is twice the size of the complete *E. coli* genome)
- And we cannot observe the lethal events, which may be even larger...

# The mutation rates: From the elementary matrices to the full transition matrix $M_{G}$

- 4 mutation rates :  $\mu_{ins}$ ,  $\mu_{sdel}$ ,  $\mu_{ldel}$ ,  $\mu_{dup}$
- Expressed per bp per generation
- Total mutation rate:  $\mu = \mu_{ins} + \mu_{sdel} + \mu_{ldel} + \mu_{dup}$
- Assumption: mutations follow independent Poisson processes, no preferred order



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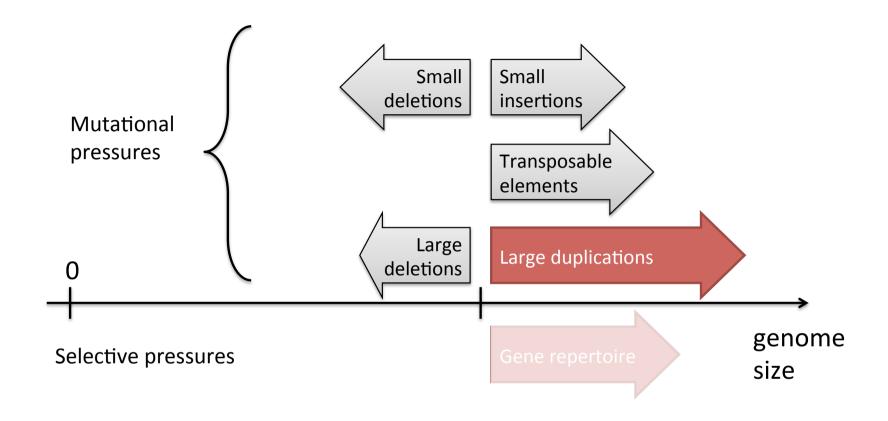
Finally 
$$(M_G)_{ij} = \sum_{n=0}^{+\infty} \frac{e^{-\mu i} (\mu i)^n}{n!} (M_1^n)_{ij}$$

which is the probability that a genome of initial size *i* ends up with size *j* at the end of the reproduction

#### Related models

- Quasispecies models
  - initial model was very general [Eigen, 1971]
  - but most results were obtained for the special case of fixed genome length and point mutations only [Eigen, 1971; Nowak & Schuster, 1989; Barbosa et al., 2012].
- Population genetics models for microsatellite and transposable elements
  - number of elements not bounded [Falush & Iwasa, 1999]
  - additive and multiplicative effects [Stephan, 1987; Falush & Iwasa, 1999]
  - several mutations can occur during the reproduction [Ohta & Kimura, 1981; Stephan, 1987]
  - but no model combines those three features

# What does this model answer to our original question?



#### How will genome size evolve if:

- duplications are twice as frequent as deletions ( $\mu_{dup}=2\mu_{ldel}$ ),
- transposable elements proliferate ( $\mu_{ins} > \mu_{sdel}$ )
- and selection systematically favors the highest gene numbers?

# Result 1: Condition for non-infinite growth without selection

Theorem 2 (Stationary distribution for genome size without selection). If  $(2 \log 2 - 1)\mu_{dup} < \mu_{ldel}$ , then the Markov chain  $(\mathbb{N}^*, \mathbf{M}_G)$  has a unique asymptotic stationary probability vector  $\nu_{\infty}$ . For any initial distribution  $\nu_0$ , the distribution of genome sizes converges to  $\nu_{\infty}$ . Mathematically,

$$\lim_{t \to \infty} \|\nu_0 \mathbf{M}_G^t - \nu_\infty\| = 0$$

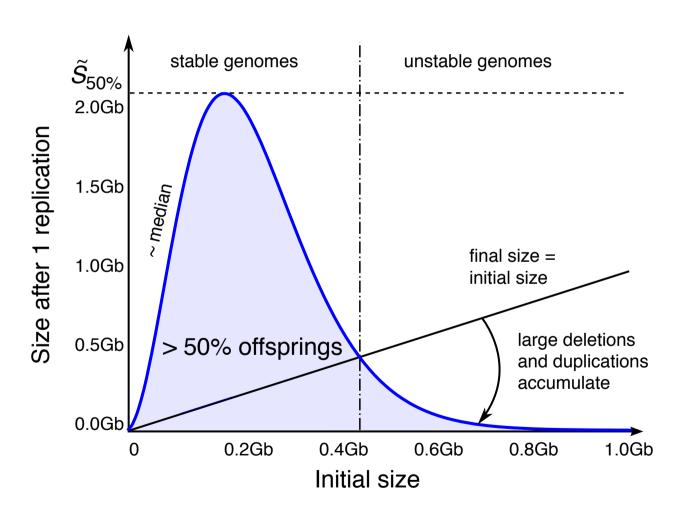
Biologically, the convergence of the distribution implies that, even after a long time of evolution, genome size does not tend to infinity: an arbitrary large part of genomes is located beneath a finite size.

No infinite growth if  $\mu_{dup}$  <  $2.6~\mu_{ldel}$  This condition is independent from the rates of small insertions (eg transposable elements) and small deletions

The proof uses Doeblin's condition.

[Fisher et al., submitted]

### Result 2: Even a caricatural selection cannot push the genomes towards an infinite size



Upper bound for the median of the distribution at t+1

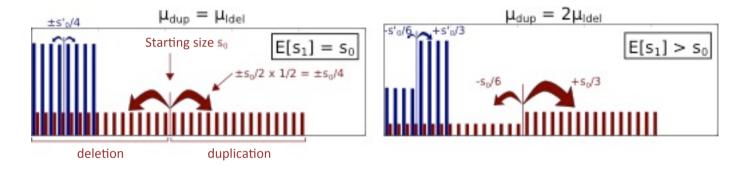
At each generation, >50% of the population is below some threshold

Large genomes can be selected but are too unstable...

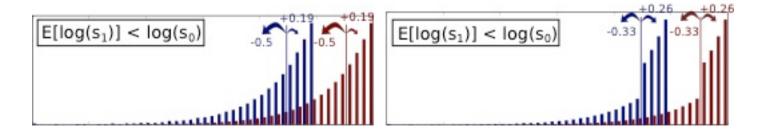
# Which properties underlie these results?

## Property 1: Asymmetry of duplications and deletions in logarithmic scale

Linear scale: apparent symmetry but **no** scale invariance. The behavior is hard to predict intuitively.



Log scale: scale-invariance, no symmetry anymore. A mutational bias towards shrinkage is revealed, even for  $\mu_{dup}=2\mu_{ldel}$ .



## Property 1: Asymmetry of duplications and deletions in logarithmic scale

Property 1. Let  $\Delta(s) = \mathbb{E}[\log(S_{n+1})|S_n = s] - \mathbb{E}[\log(S_n)|S_n = s]$ , the average size of one-mutation jumps in logarithmic scale, starting from s.

- if the (n+1)th mutation is a large deletion,  $\Delta(s) \underset{s \to +\infty}{\longrightarrow} -1$ .
- if the (n+1)th mutation is a duplication,  $\Delta(s) \underset{s \to +\infty}{\longrightarrow} 2 \log 2 1$ .
- if the (n+1)th mutation is an indel,  $\Delta(s) \xrightarrow[s \to +\infty]{} 0$ .

This property is important in the proof of the first result (condition for the existence of a stationary distribution).

## Generalization to non-uniform distributions for the size of duplications and deletions

Corollary 1. (Generalization of Theorem 2) Suppose we have distributions of duplications, large deletions and indels, such that there exists a positive and increasing scaling function f that verifies the following conditions.

For 
$$\Delta(s) = \mathbb{E}\left[f(S_{n+1}) - f(S_n)|S_n = s\right]$$
:

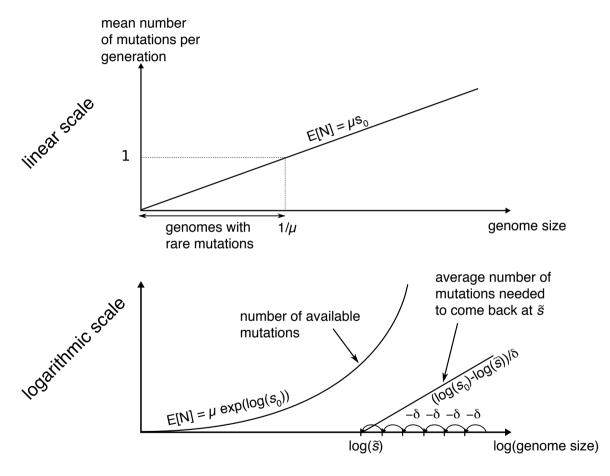
- if the (n+1)th mutation is a deletion,  $\Delta(s) \xrightarrow[s \to +\infty]{} \delta_{ldel}$ .
- if the (n+1)th mutation is a duplication,  $\Delta(s) \underset{s \to +\infty}{\longrightarrow} \delta_{dup}$ .
- if the (n+1)th mutation is an small insertion,  $\Delta(s) \xrightarrow[s \to +\infty]{} \delta_{ins}$ .
- if the (n+1)th mutation is an small deletion,  $\Delta(s) \xrightarrow[s \to +\infty]{} \delta_{sdel}$ .

where  $\delta_{ldel} \leq 0$ ,  $\delta_{dup} \geq 0$ ,  $\delta_{ins} \geq 0$  and  $\delta_{sdel} \leq 0$  are constants among which at least one is nonzero.

Then the Markov chain  $(\mathbb{N}^*, \mathbf{M}_G)$  has a unique stationary probability vector  $\nu_{\infty}$  if

$$\mu_{ldel}\delta_{ldel} + \mu_{dup}\delta_{dup} + \mu_{ins}\delta_{ins} + \mu_{sdel}\delta_{sdel} < 0 \tag{4}$$

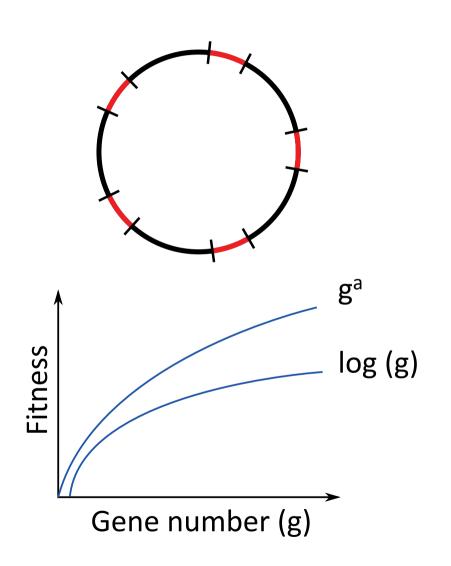
## Property 2: Larger genomes undergo more mutations



This property is important in the proof of the second result (selection cannot overcome the spontaneous mutational dynamics).

# Let's simulate the model with (a rather brutal) selection

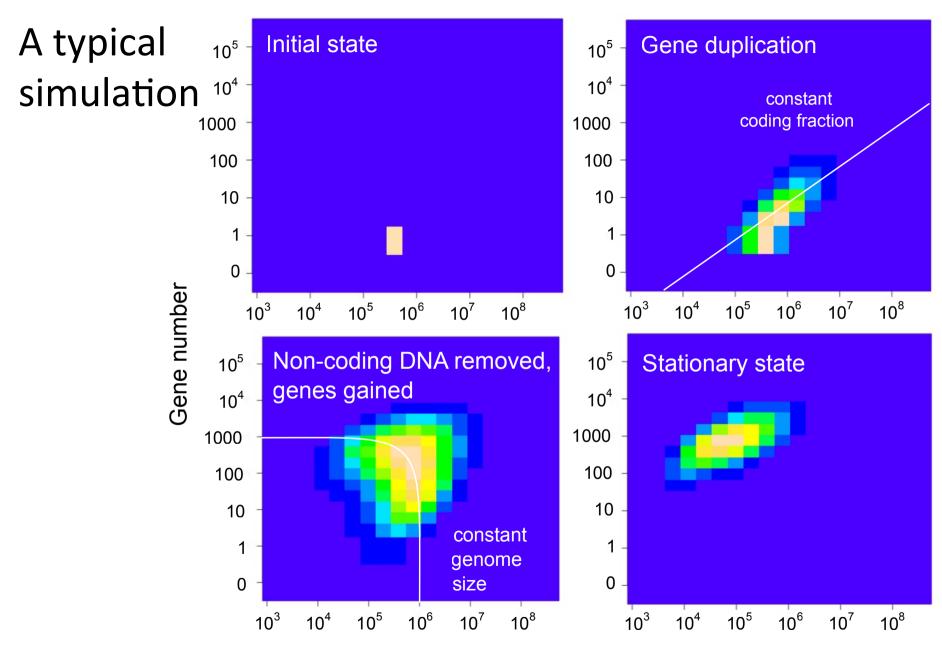
# Introducing fitness: coding versus non-coding DNA



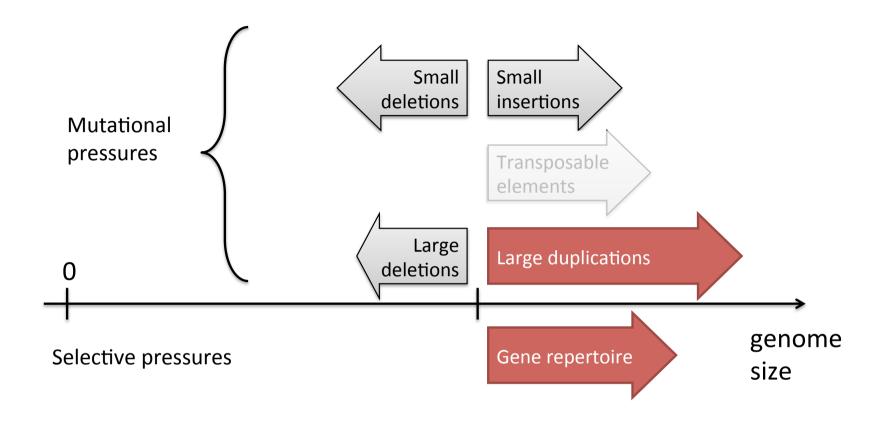
#### Hypotheses:

- circular genome described by its gene number g and the number of non-coding bases,
- Genes all have the same (fixed) length
- Non-coding bases are equally distributed between genes
- The fitness is a monotonically increasing, not bounded, function of g

$$n_{t+1} = n_t \frac{F}{\|n_t F\|} M$$

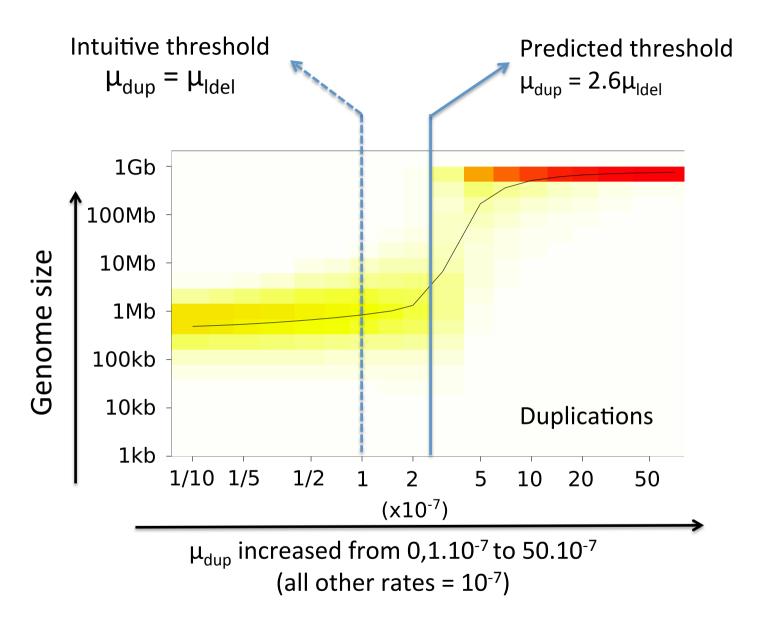


Number of non-coding base pairs

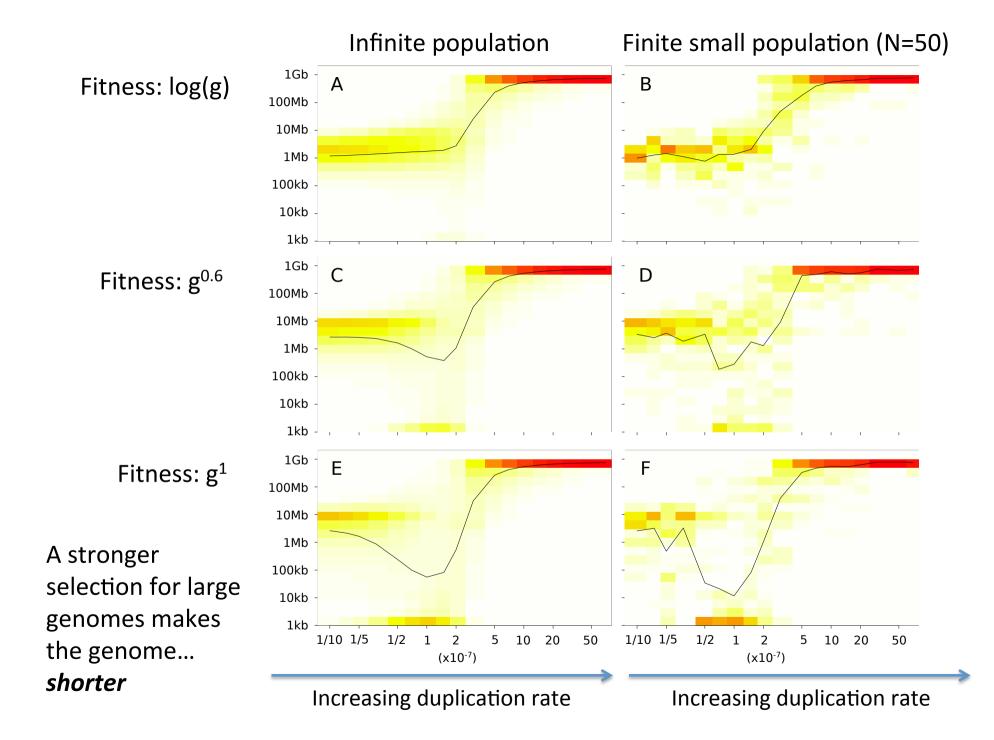


#### How will genome size evolve if :

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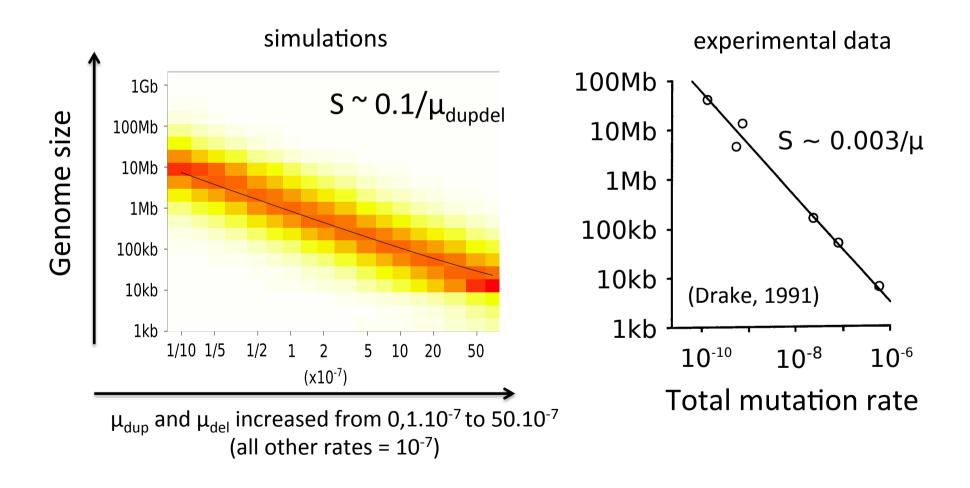
In these runs, fitness was proportional to log(gene number)

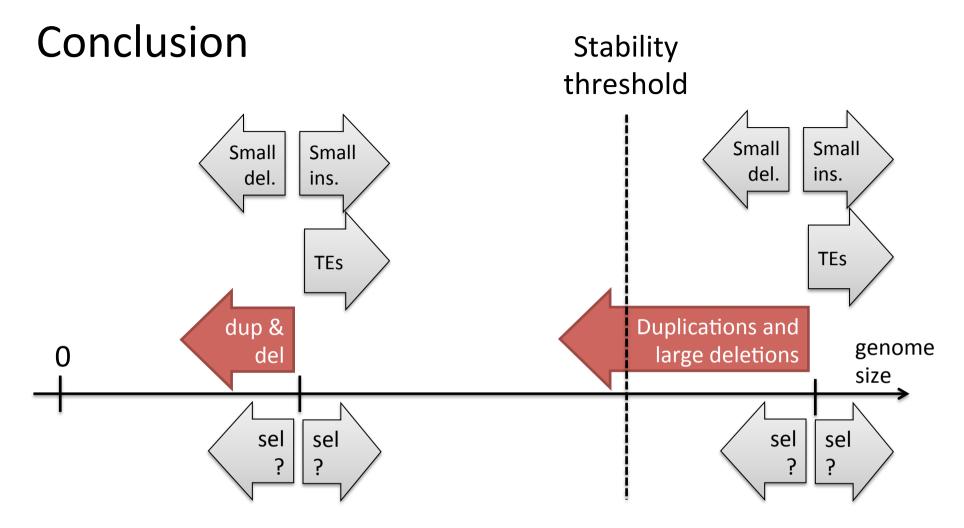


OK, the intuition fails in this scenario...

but in reality, the rates of duplication and deletion do not evolve independently

#### Evolved genome size is inversely proportional to the rate of multiplicative events





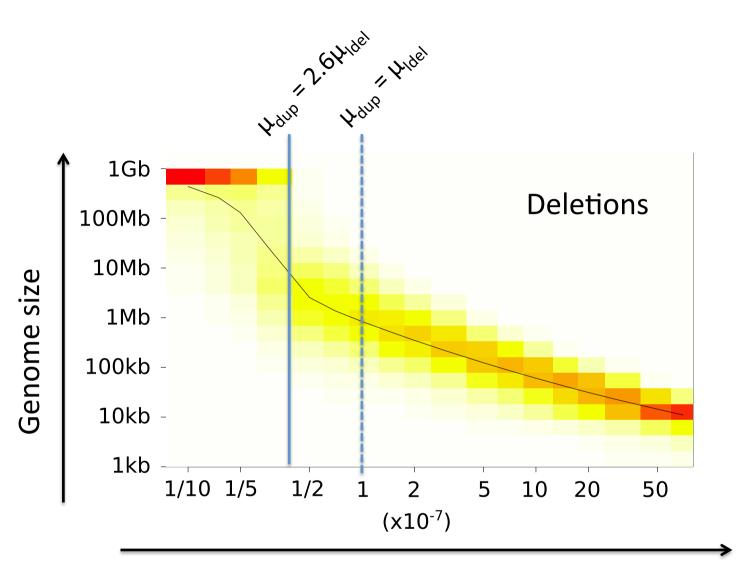
- The instability caused by duplications and deletions prevents genomes from growing above a certain threshold
- Below this threshold, this mutational bias is weaker and other pressures can play a role

#### Perspectives

- Incorporate transposable elements in the simulations too, not just in the formal analysis
- Update after each duplication or deletion the number of other events to be done, not just at each replication
- Try more realistic fitness landscapes, where not every duplication is beneficial and where deletions can be lethal

#### Take-home message

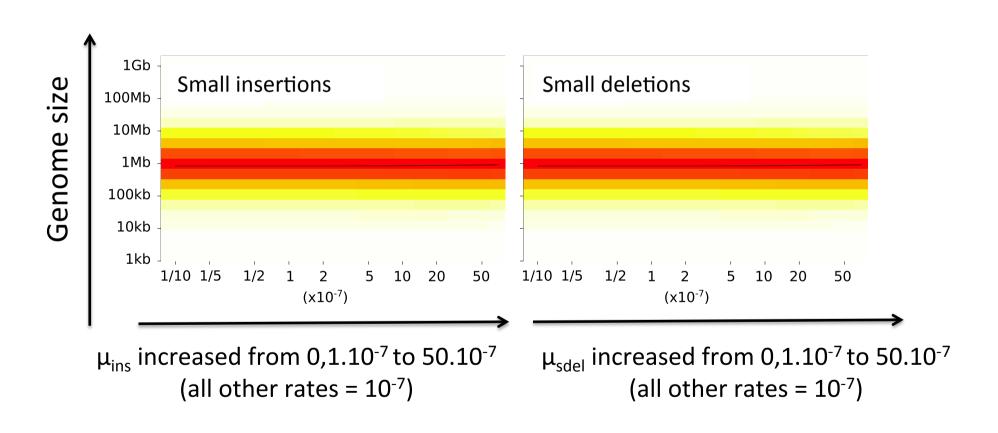
Evolution can be more subtle than we think, do not trust « thought experiments »



 $\mu_{ldel}$  increased from 0,1.10<sup>-7</sup> to 50.10<sup>-7</sup> (all other rates = 10<sup>-7</sup>)

In these runs, fitness was proportional to log(gene number)

#### (Non) effect of small insertions and small deletions on the stationary distribution of genome size



In these runs, fitness was proportional to log(gene number)