



# Epistasis: Evolutionary Consequences and Metabolic Origins

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# What fraction of possibilities has life explored?



Daniel Fisher

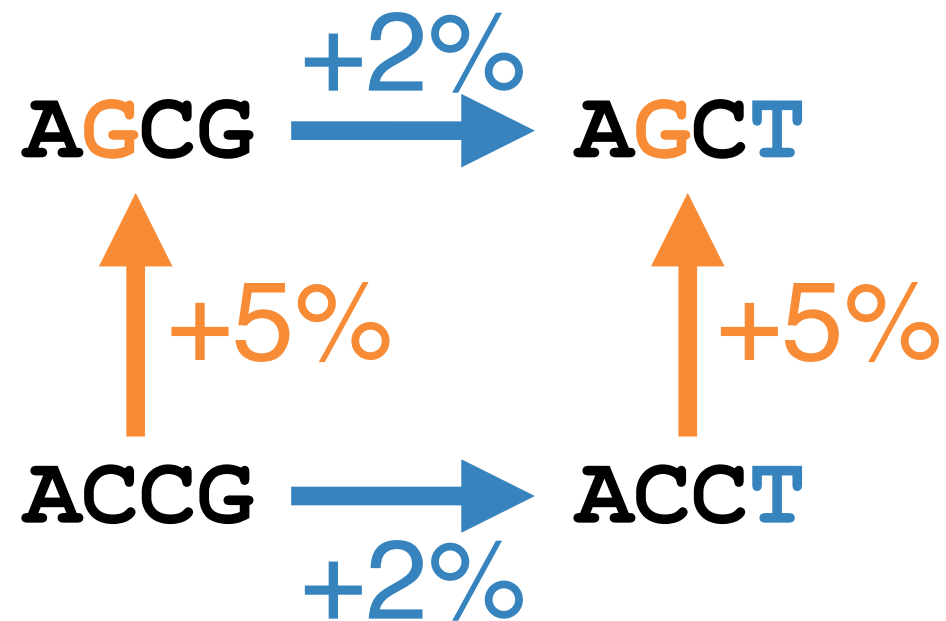
Number of bacteria on Earth  
 **$10^{30}$**

Number of cell divisions  
since origin of life  
 $10^{30} \times (3.5 \times 10^9) \times (2.6 \times 10^4) \approx \mathbf{10^{44}}$

Number of genomes explored by life  
 $10^{44} \times 10^{-4} \times 10^6 \approx \mathbf{10^{46}}$

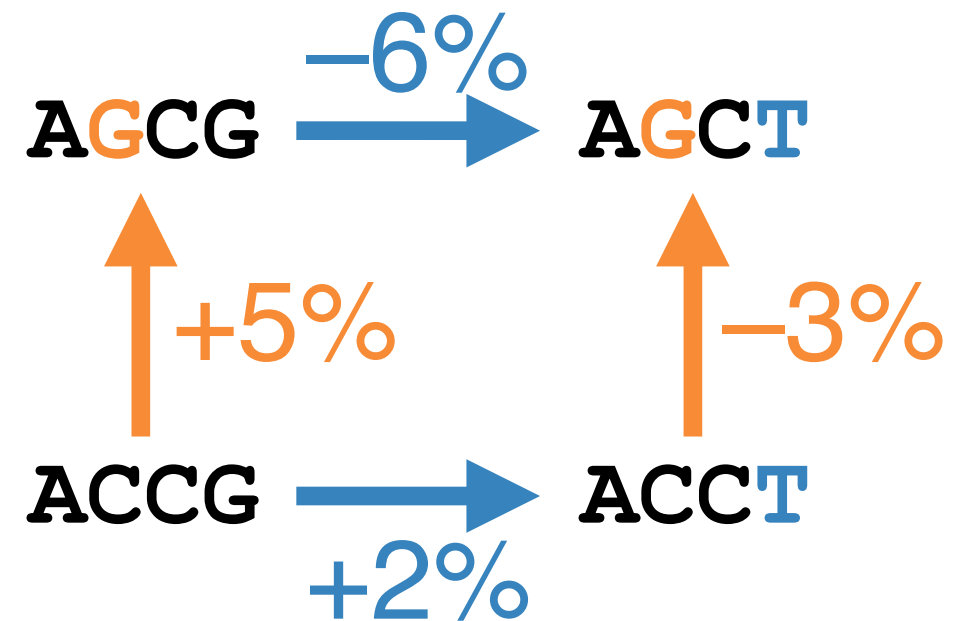
Number of possible genomes  
 $4^{1,000,000} \approx \mathbf{10^{600,000}}$

# Do evolutionary outcomes depend on identity and order of mutations?



Effects of mutations are  
context-independent  
(no epistasis)

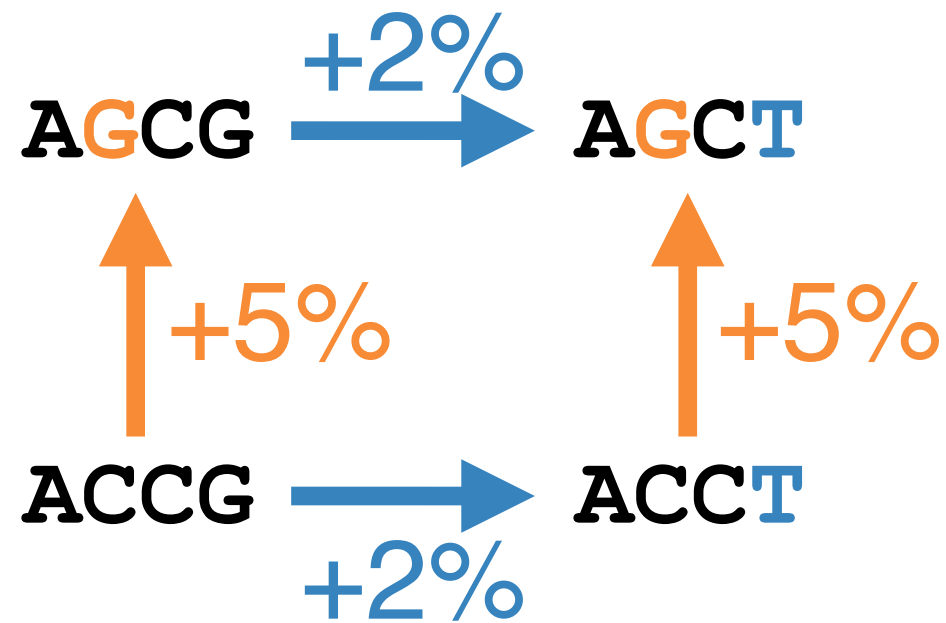
One fittest genotype



Effects of mutations are  
context-dependent  
(epistasis)

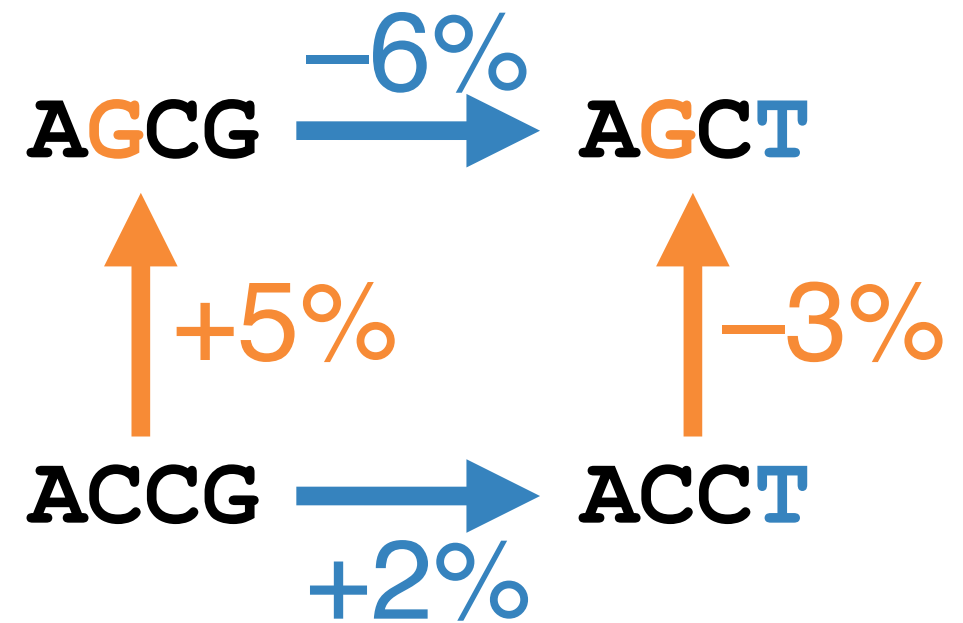
Potentially multiple  
“fitness peaks”

# Do evolutionary outcomes depend on identity and order of mutations?



All mutational paths lead to the same final genotype

no historical contingency  
smooth fitness landscape

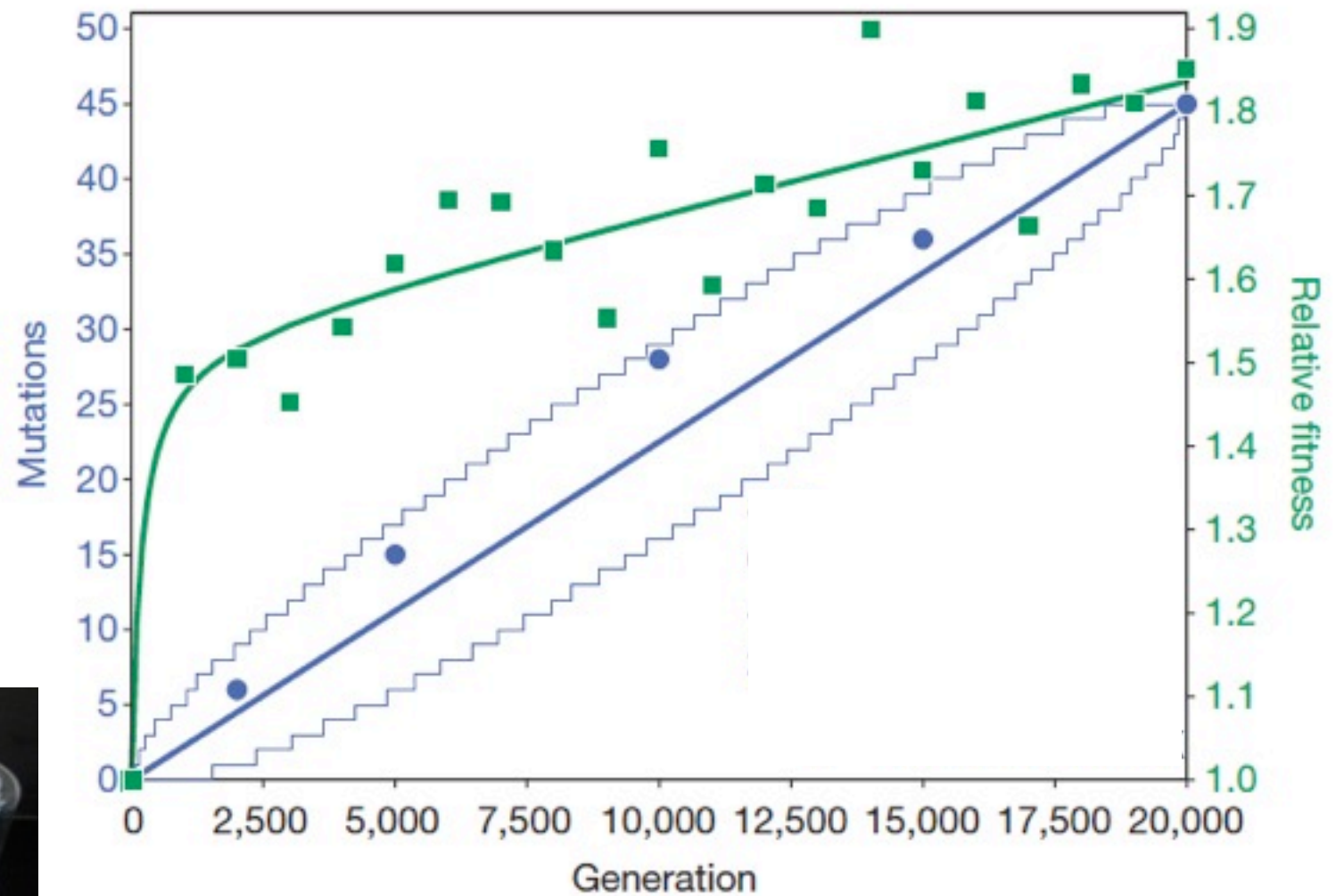


Evolutionary outcomes may be path-dependent

historical contingency  
rugged fitness landscape

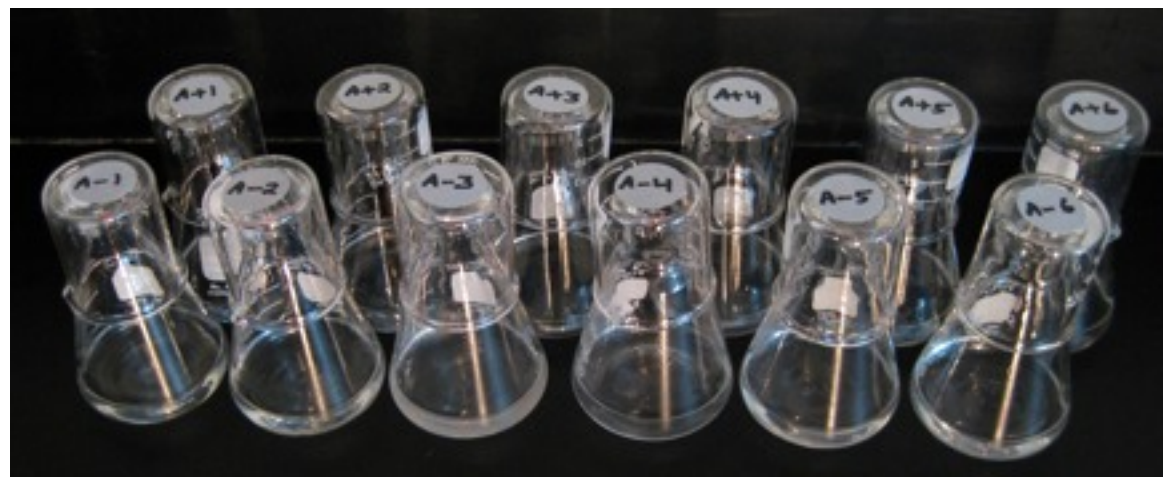
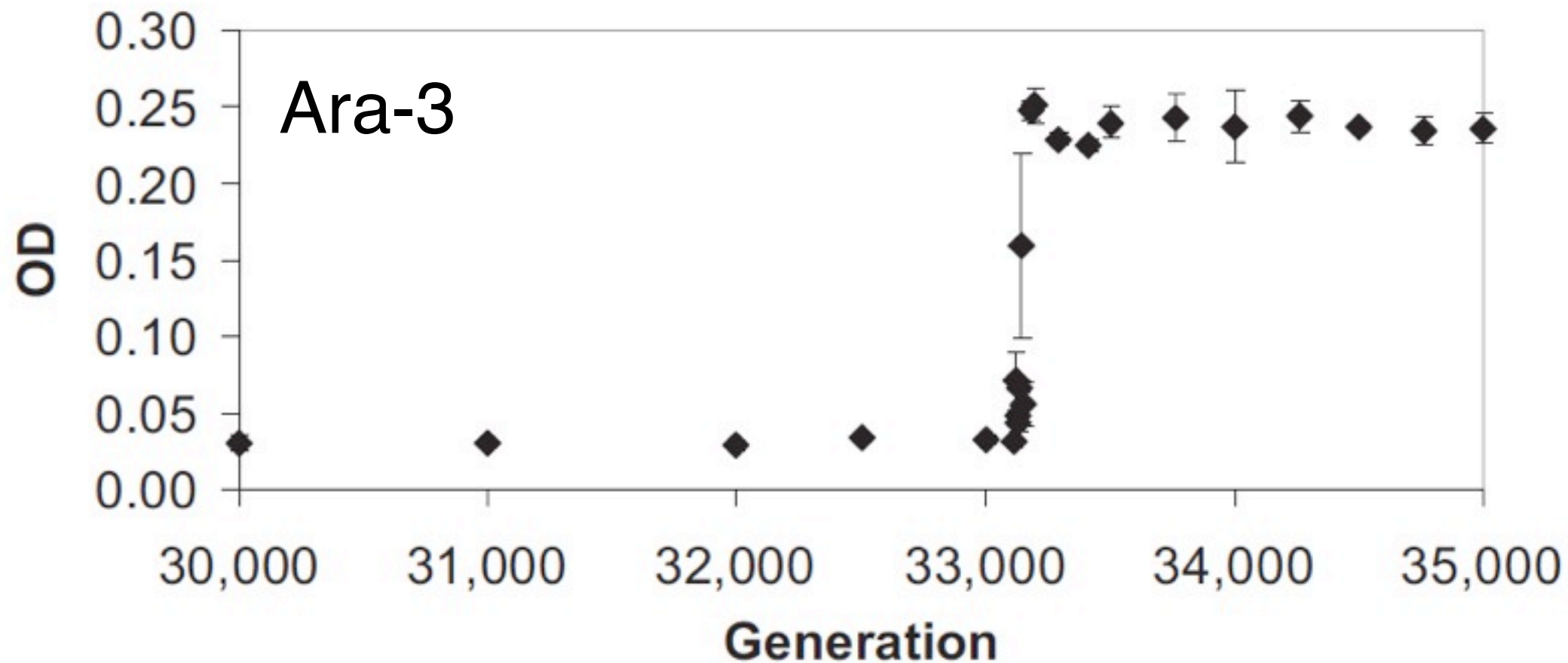


# Evidence for historical contingency due to epistasis



Richard Lenski's LTEE

# Evidence for historical contingency due to epistasis



Cit++ phenotype:

ability to metabolize citrate in the presence of oxygen

Richard Lenski's LTEE



# Evidence for historical contingency due to epistasis

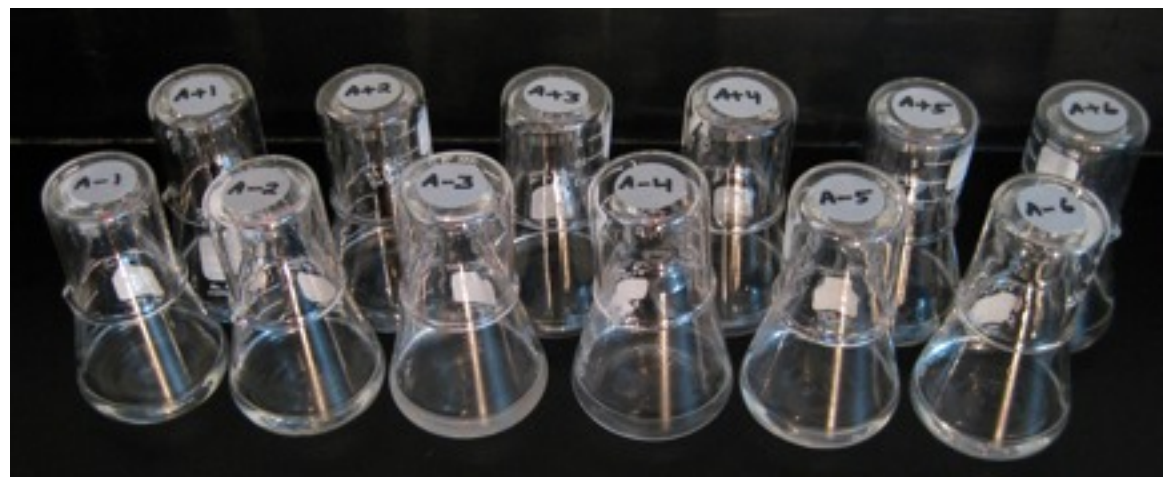


Mutations involved  
in Cit<sup>++</sup> phenotype

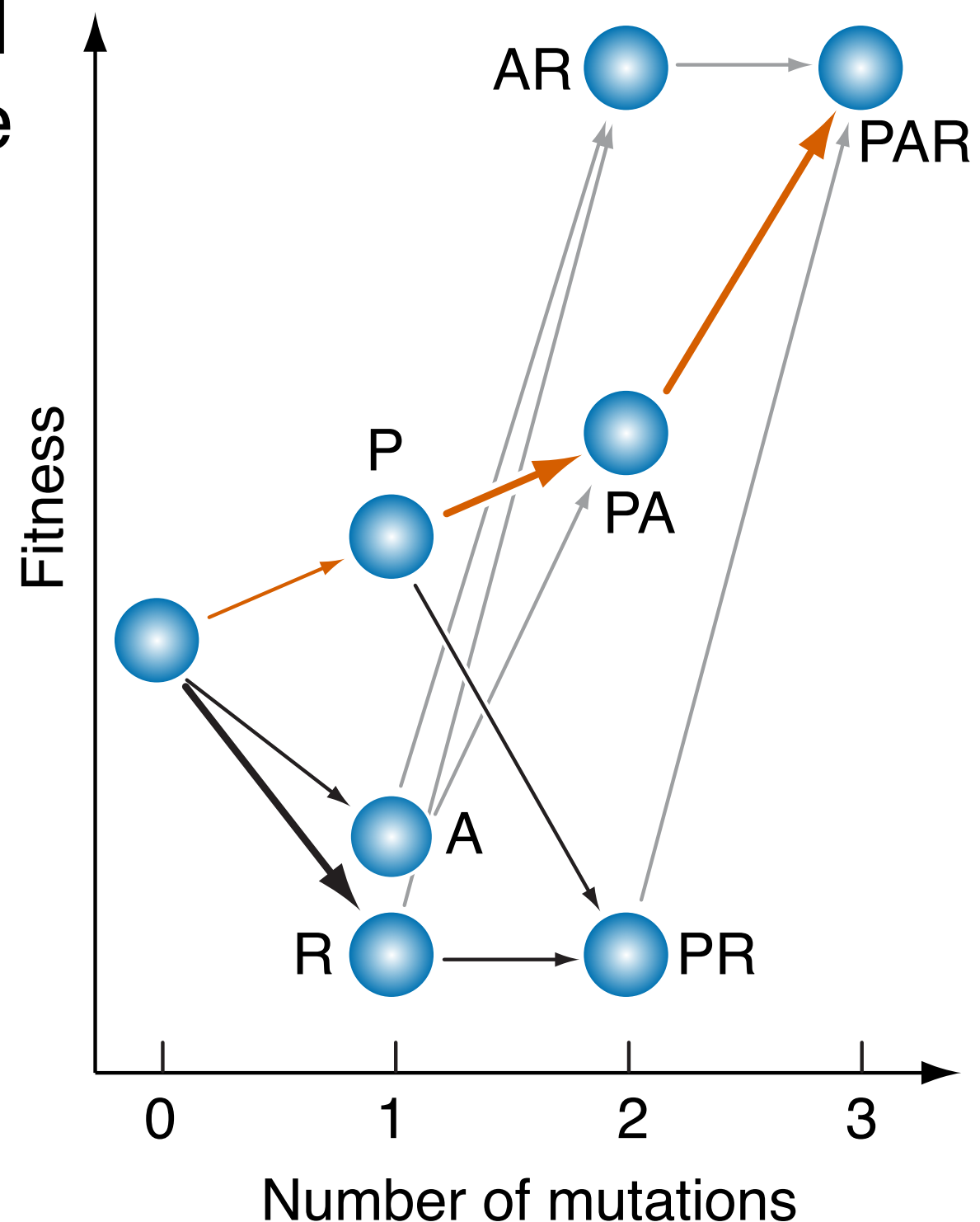
● Potentiating

● Actualizing

● Refining



Richard Lenski's LTEE



# Talk outline

## 1. How typical is historical contingency?

How does adaptation depend on the initial genotype?



Michael Desai



Dan Rice



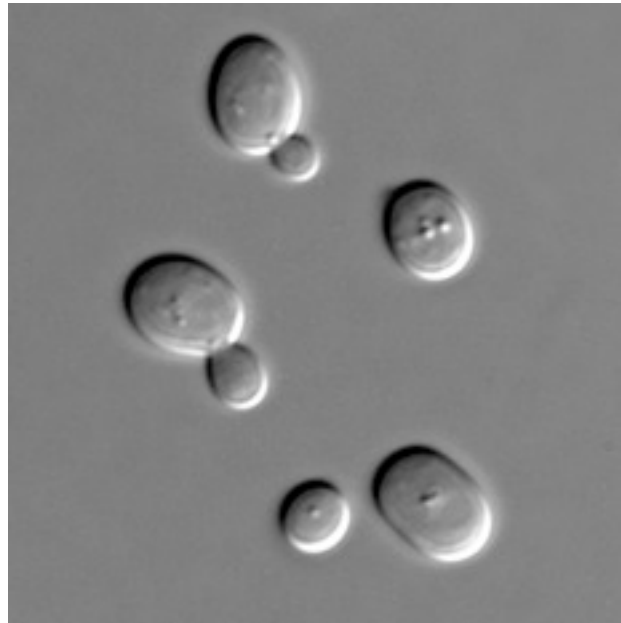
Elizabeth Jerison

## 2. What is the metabolic basis of epistasis?

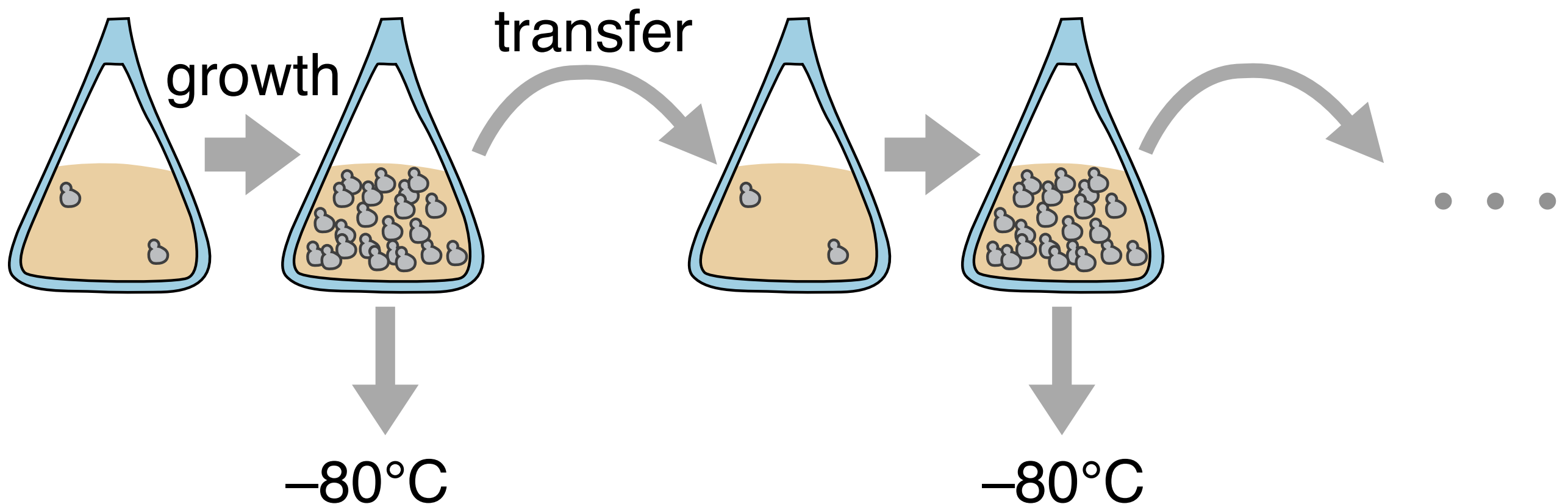
What kinds of epistasis should we expect to observe?



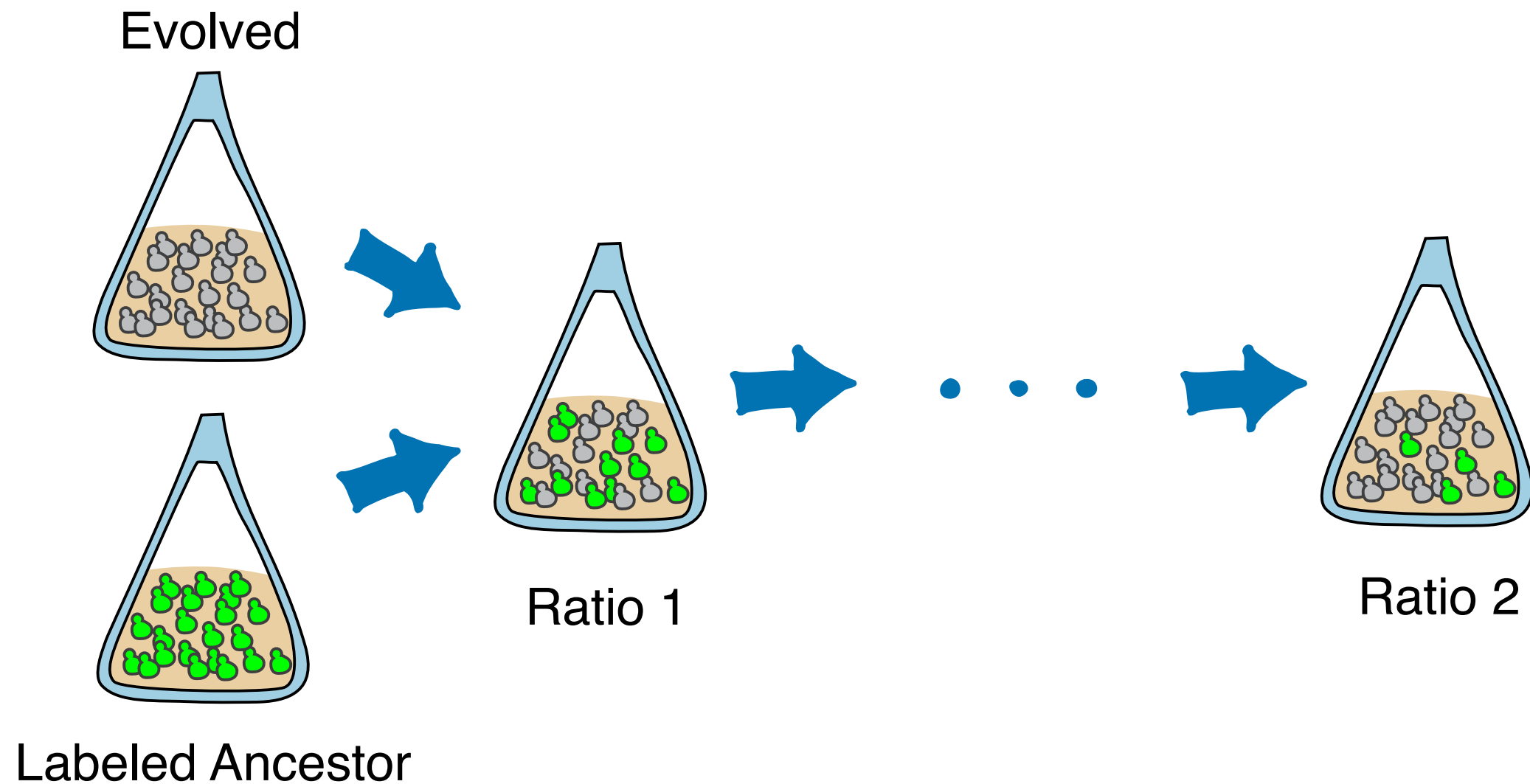
# The basics of evolution experiments



*Saccharomyces cerevisiae*

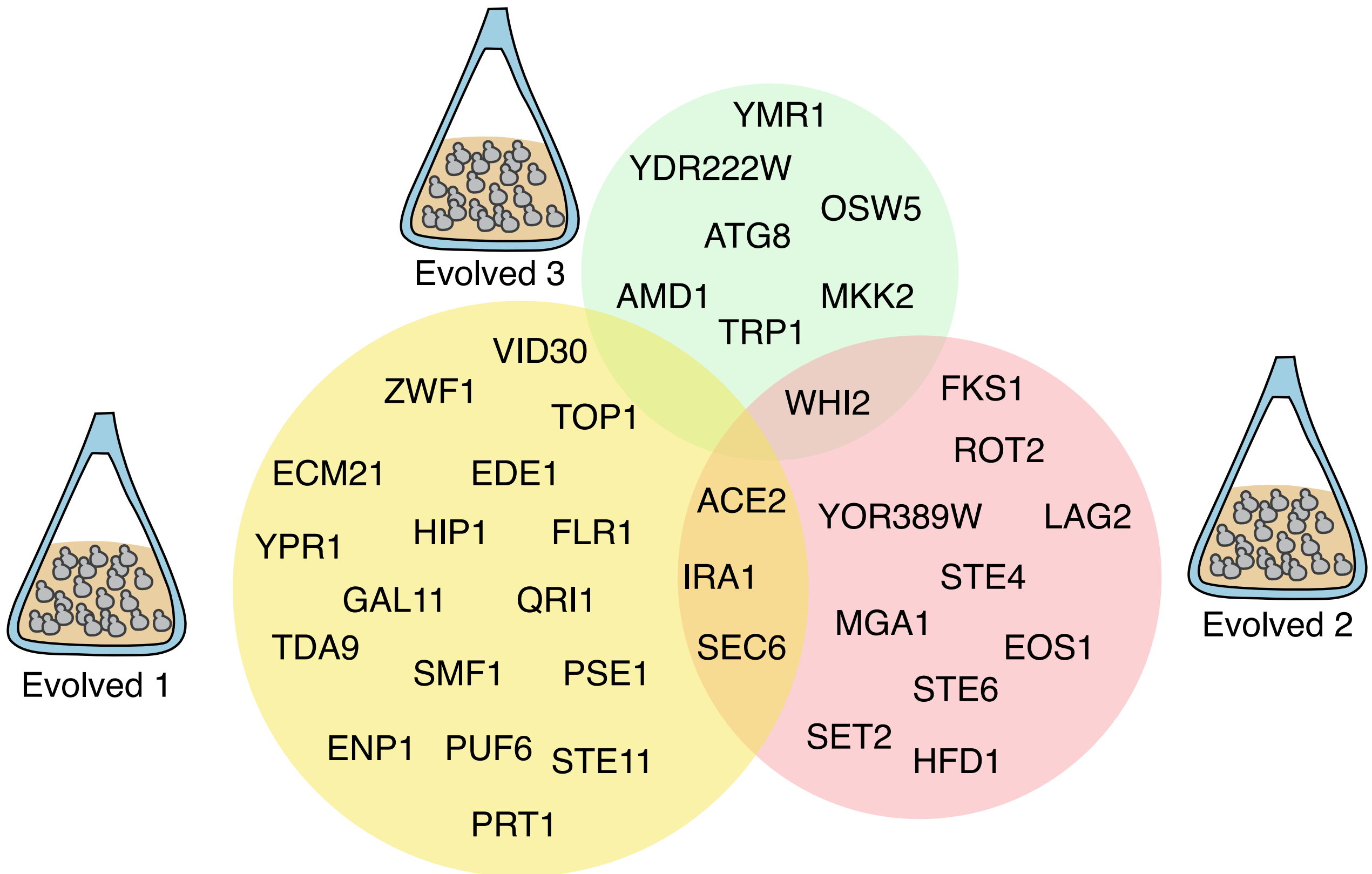


# Measure fitness by direct competition with ancestor



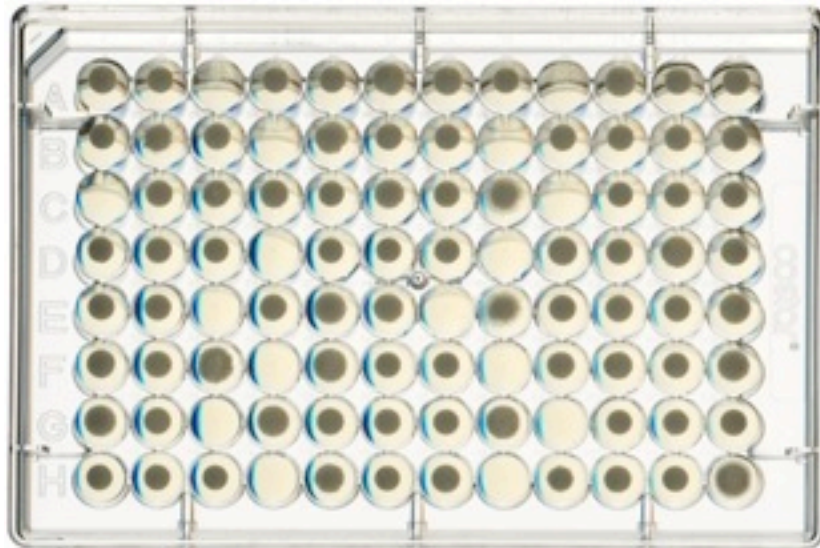
$$\text{Fitness} \propto \ln(\text{Ratio 2}) - \ln(\text{Ratio 1})$$

# Identify adaptive mutations by parallel evolution in replicate lines





# Experimental evolution in hundreds of parallel populations



Maintain



Transfer



Measure fitness

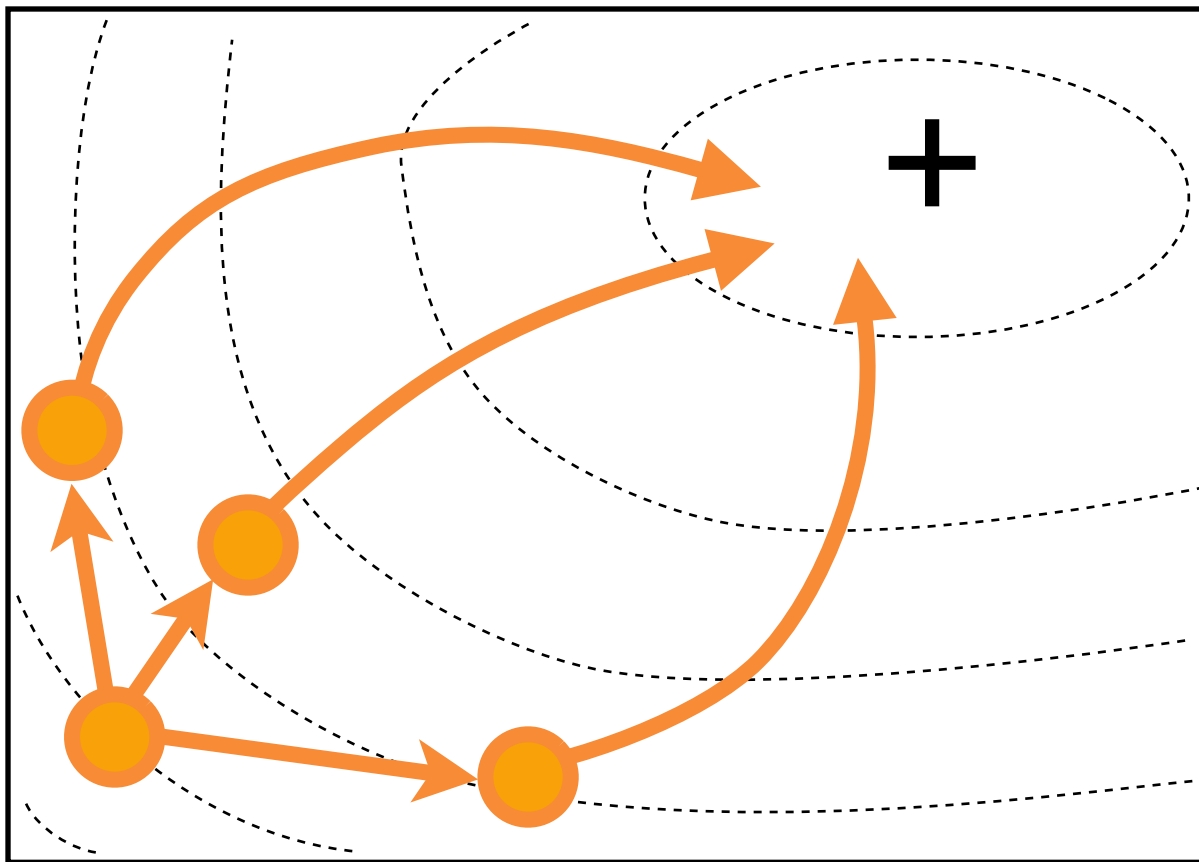


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Find mutations

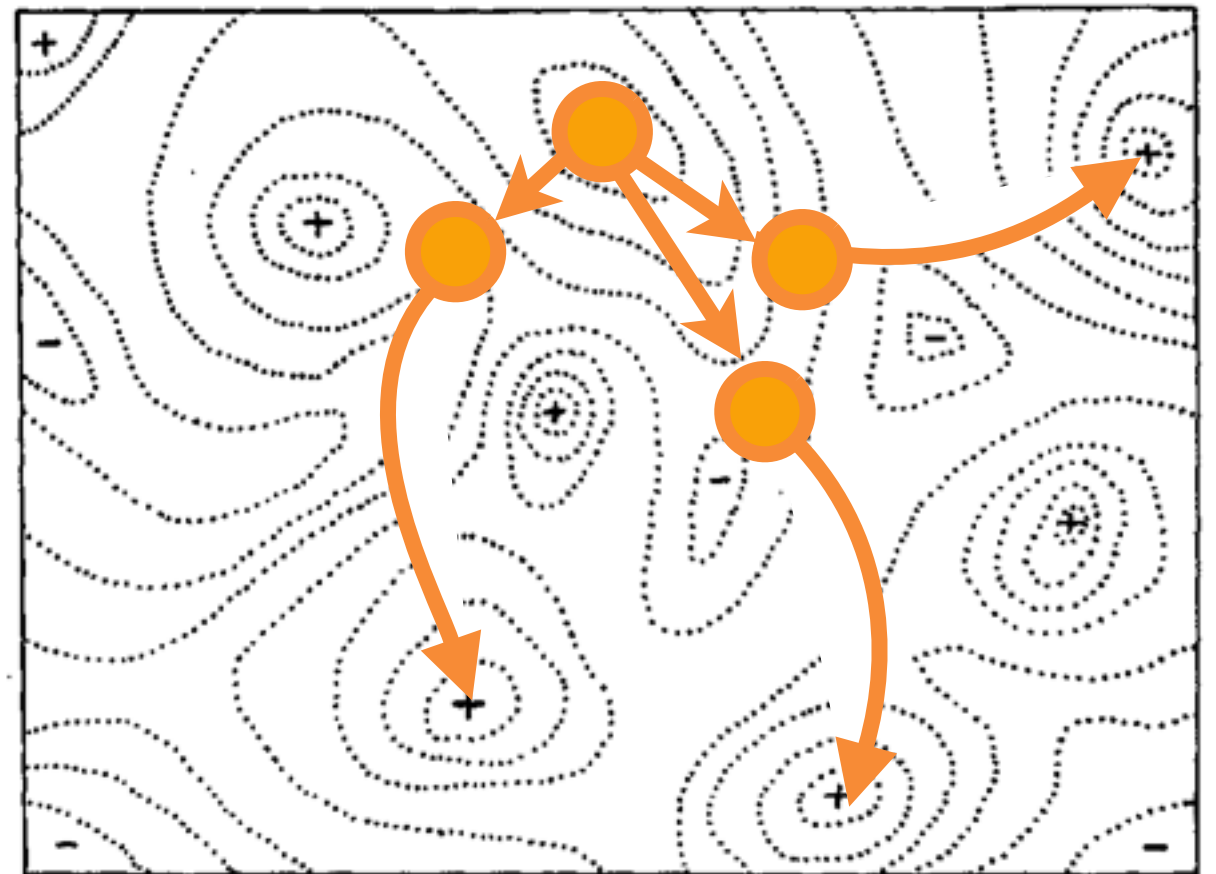
# How does prior evolution affect future evolutionary outcomes?

Mutations are beneficial in all backgrounds



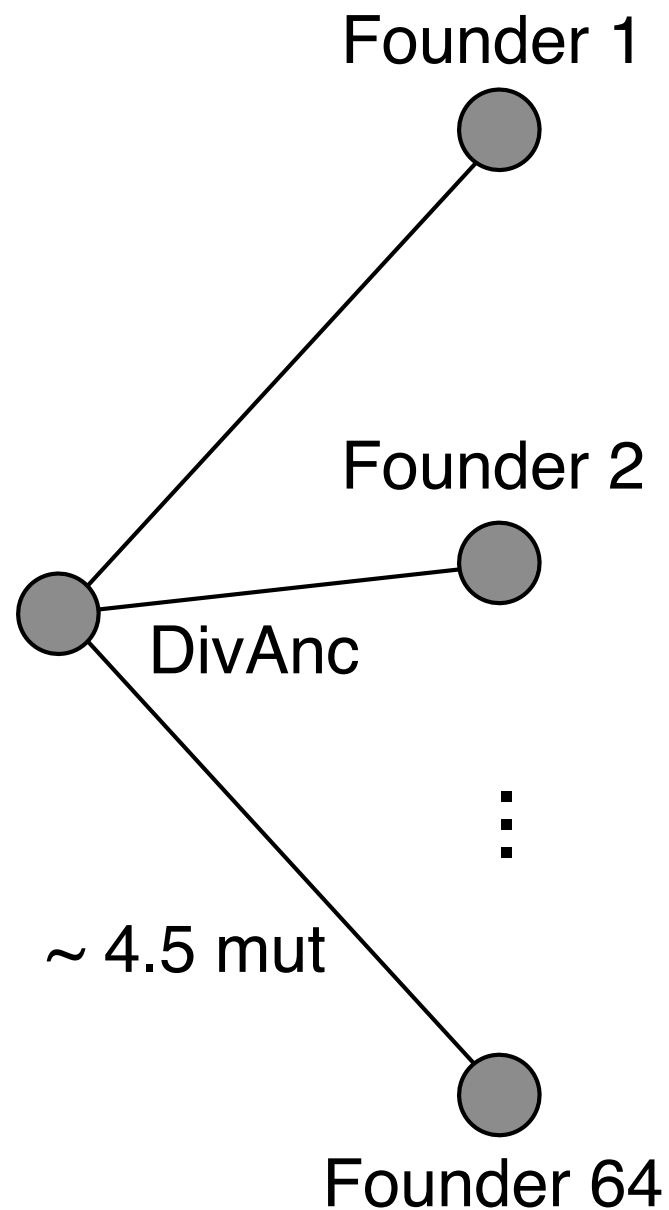
smooth fitness landscape

Mutations are beneficial in specific backgrounds



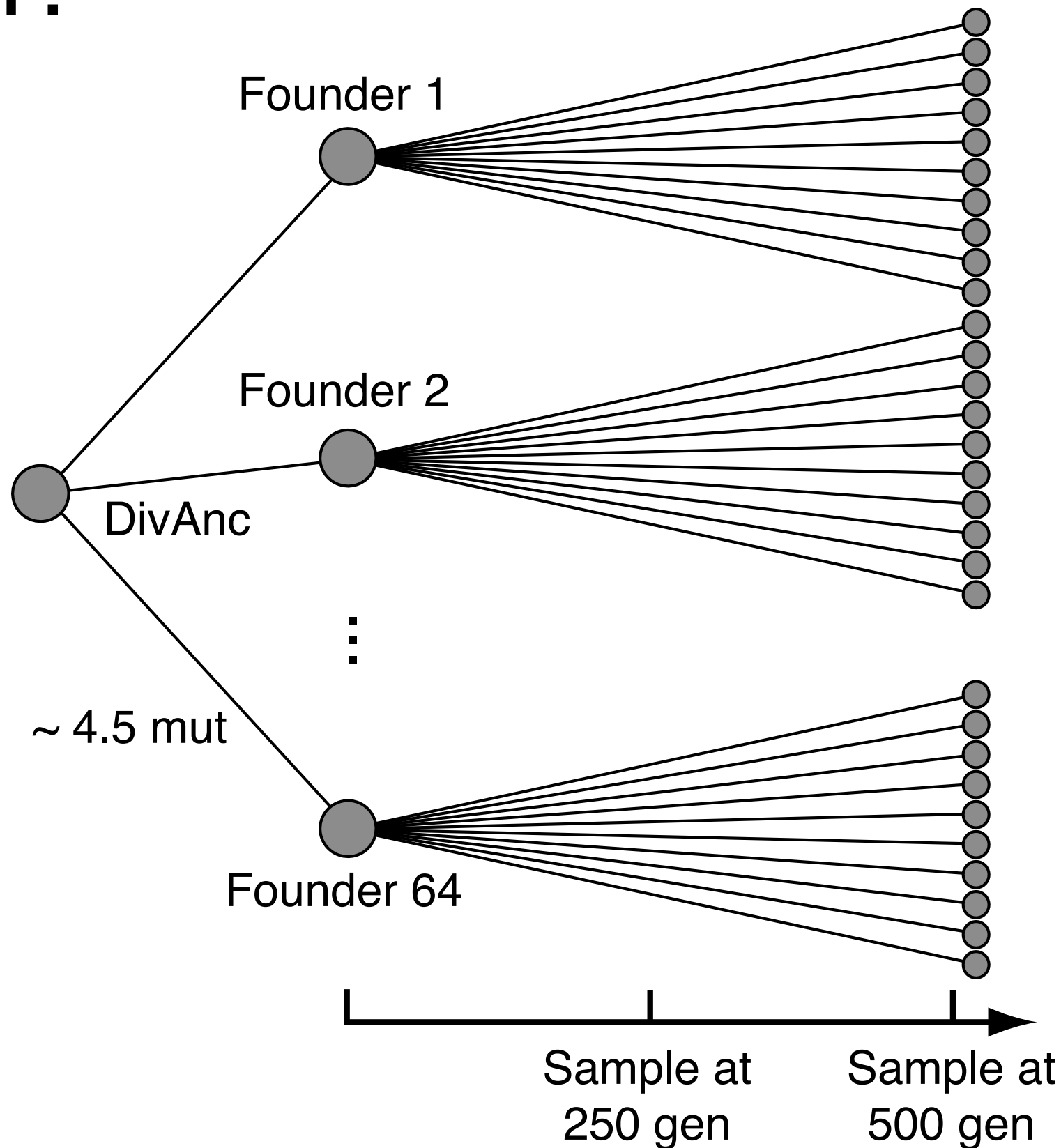
rugged fitness landscape

# Does initial genotype affect further evolution?

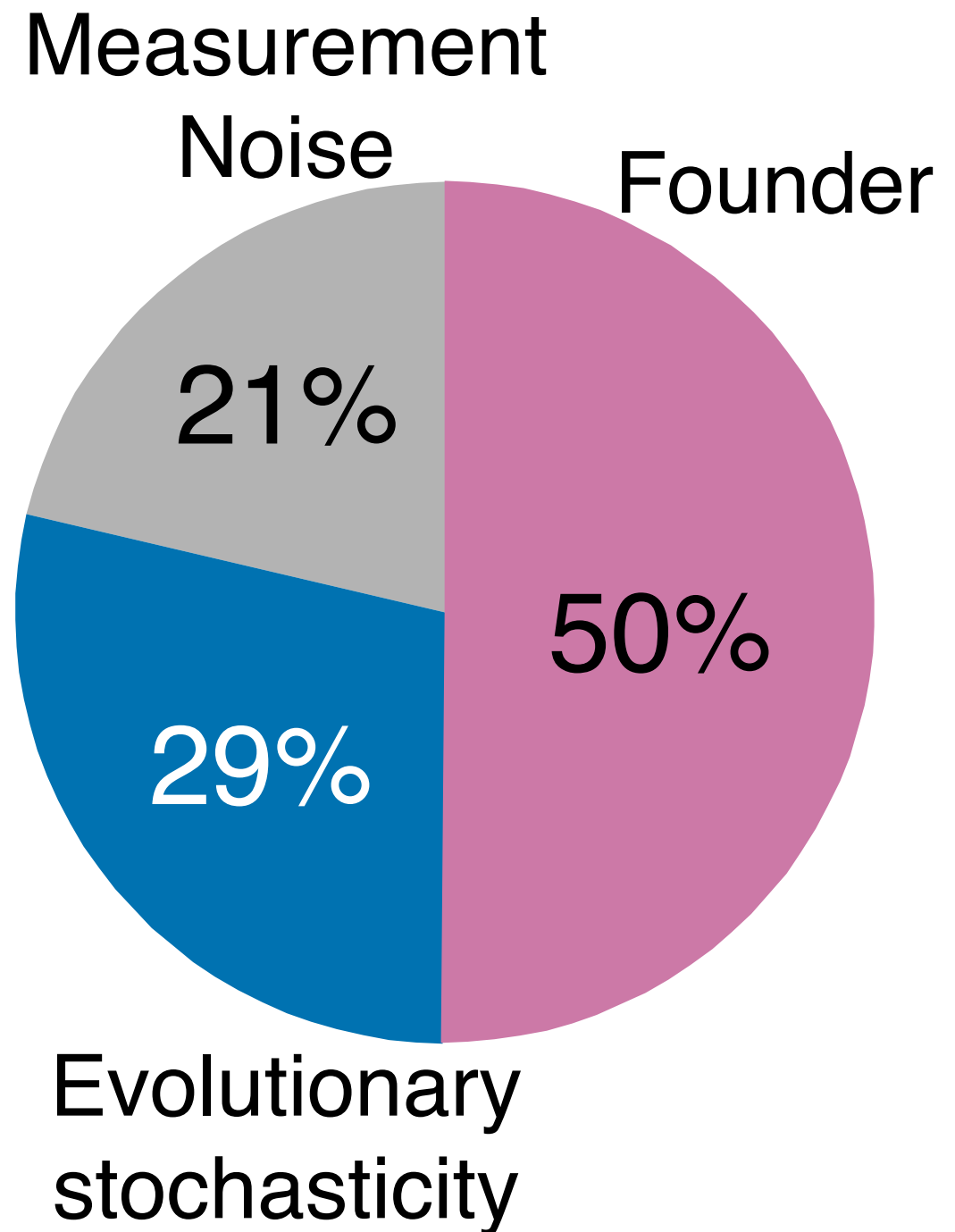
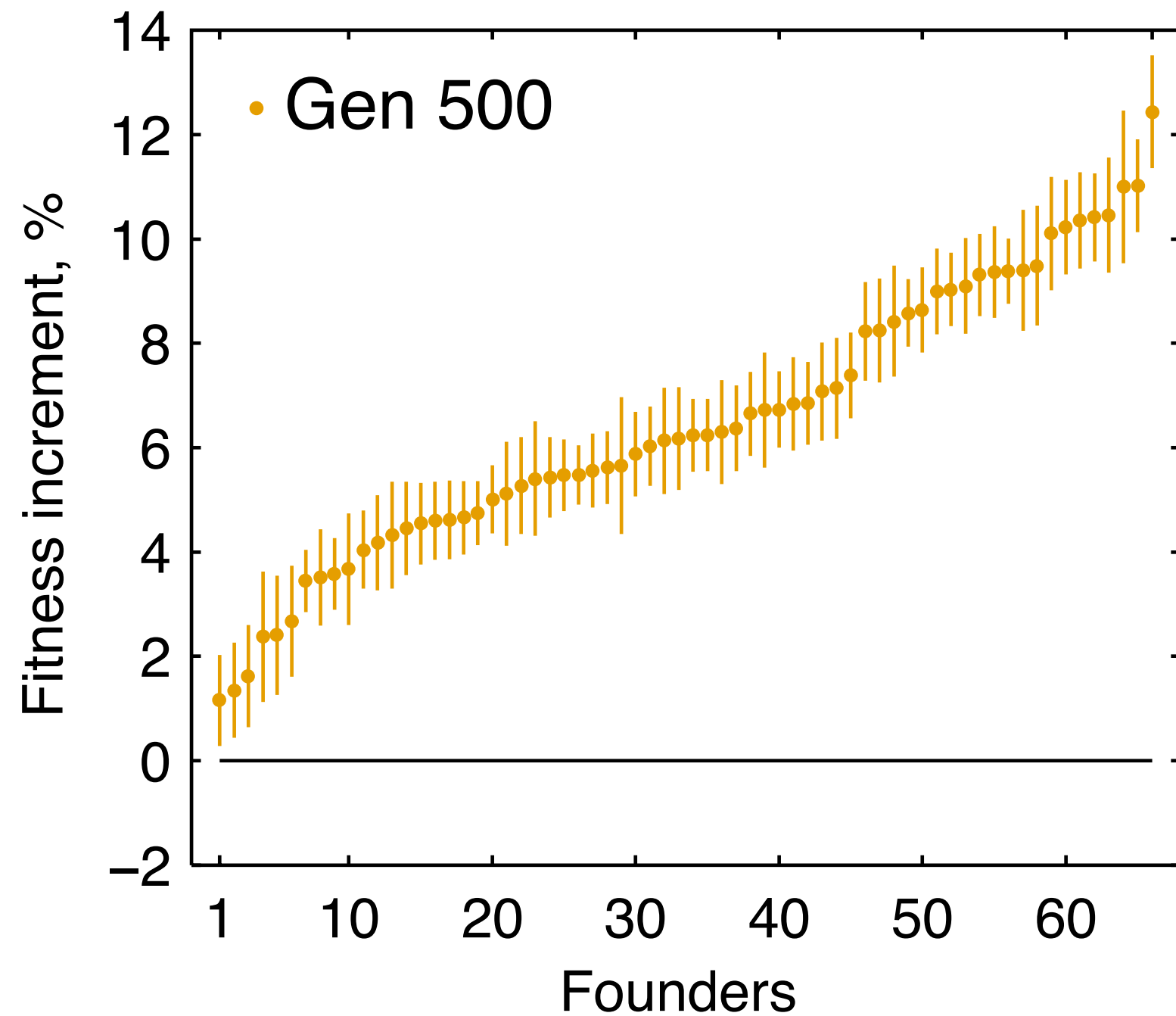




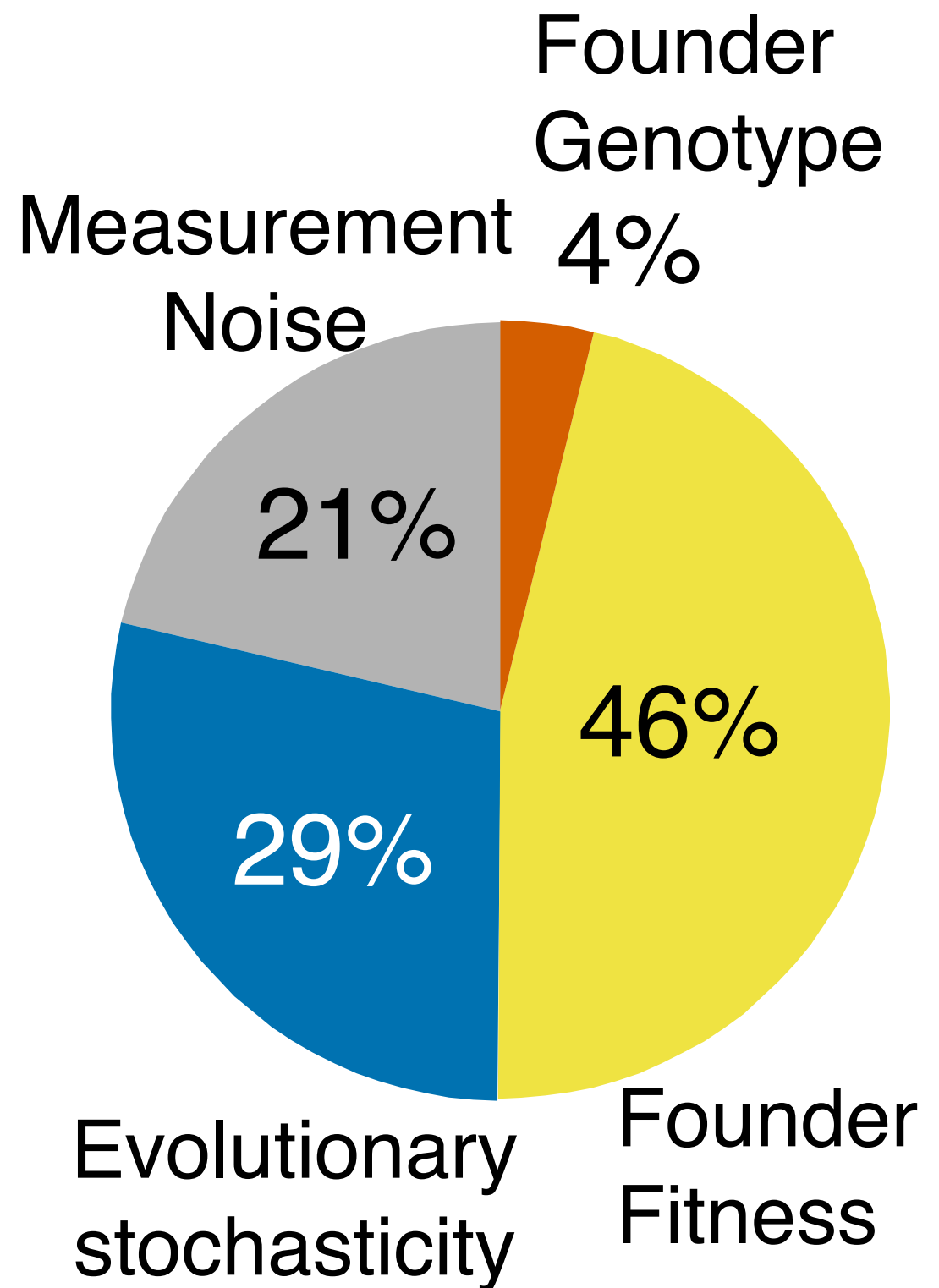
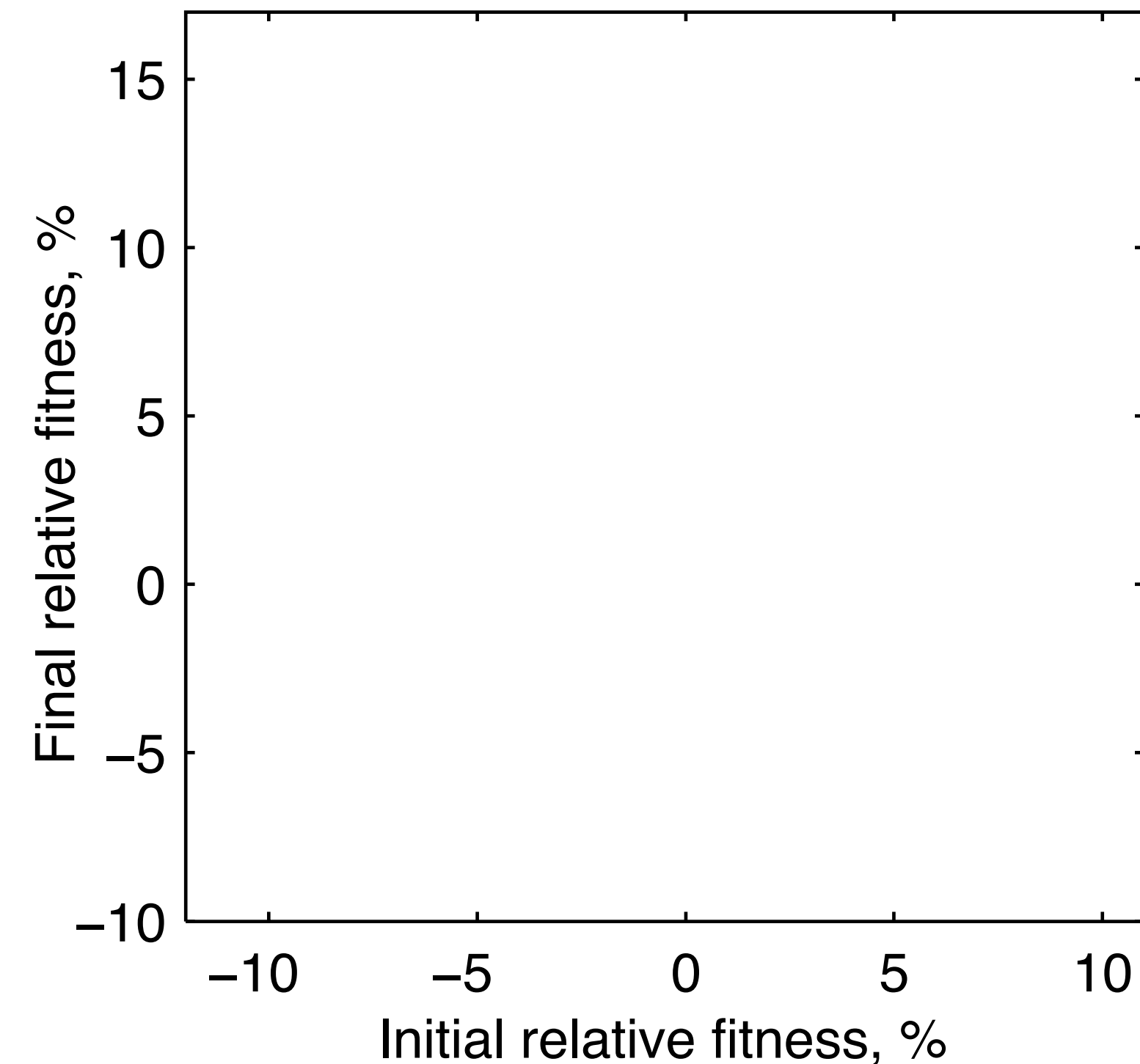
# Does initial genotype affect further evolution?



# Rate of adaptation varies among Founders



# Adaptation rate declines with Founder fitness

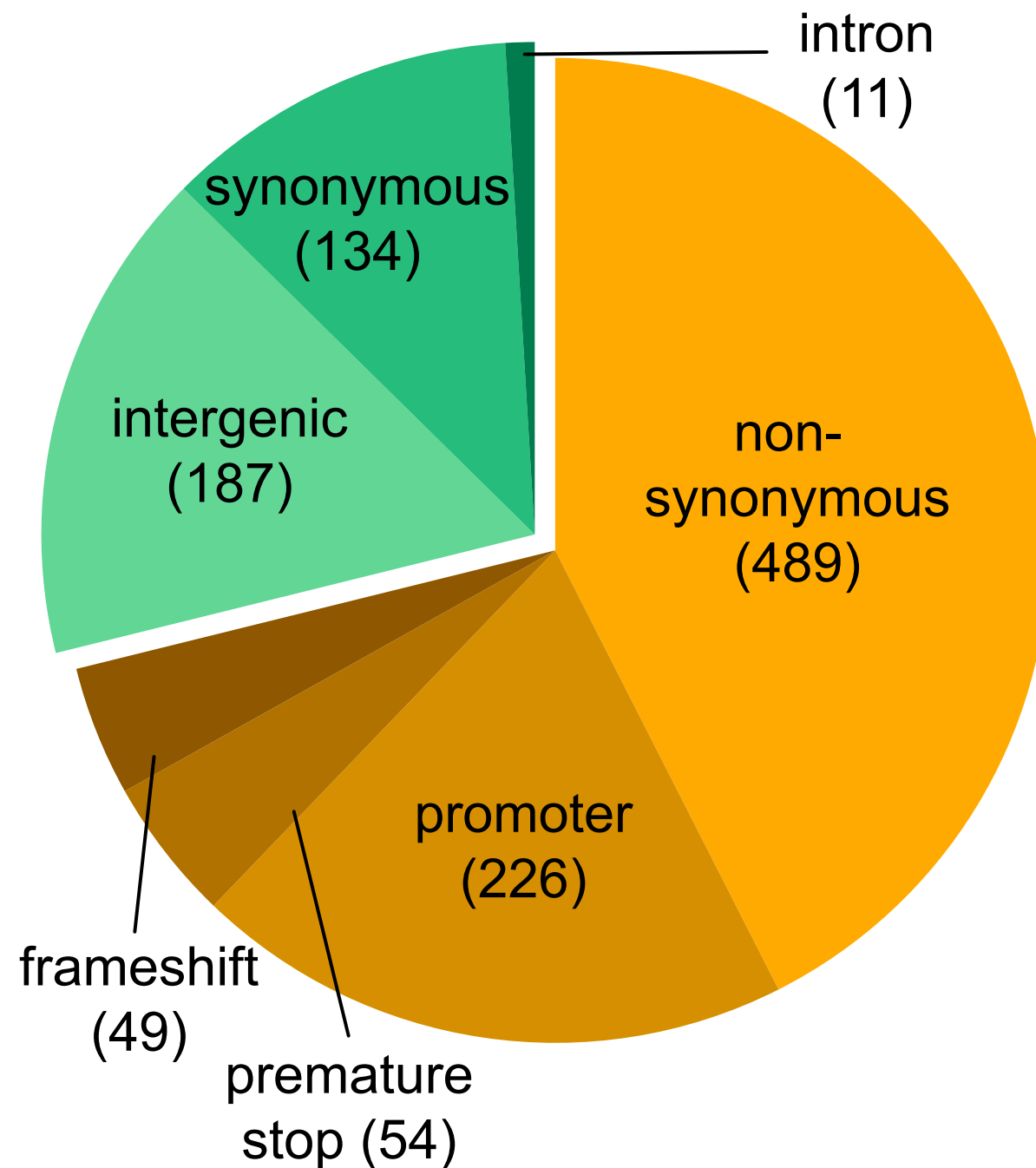




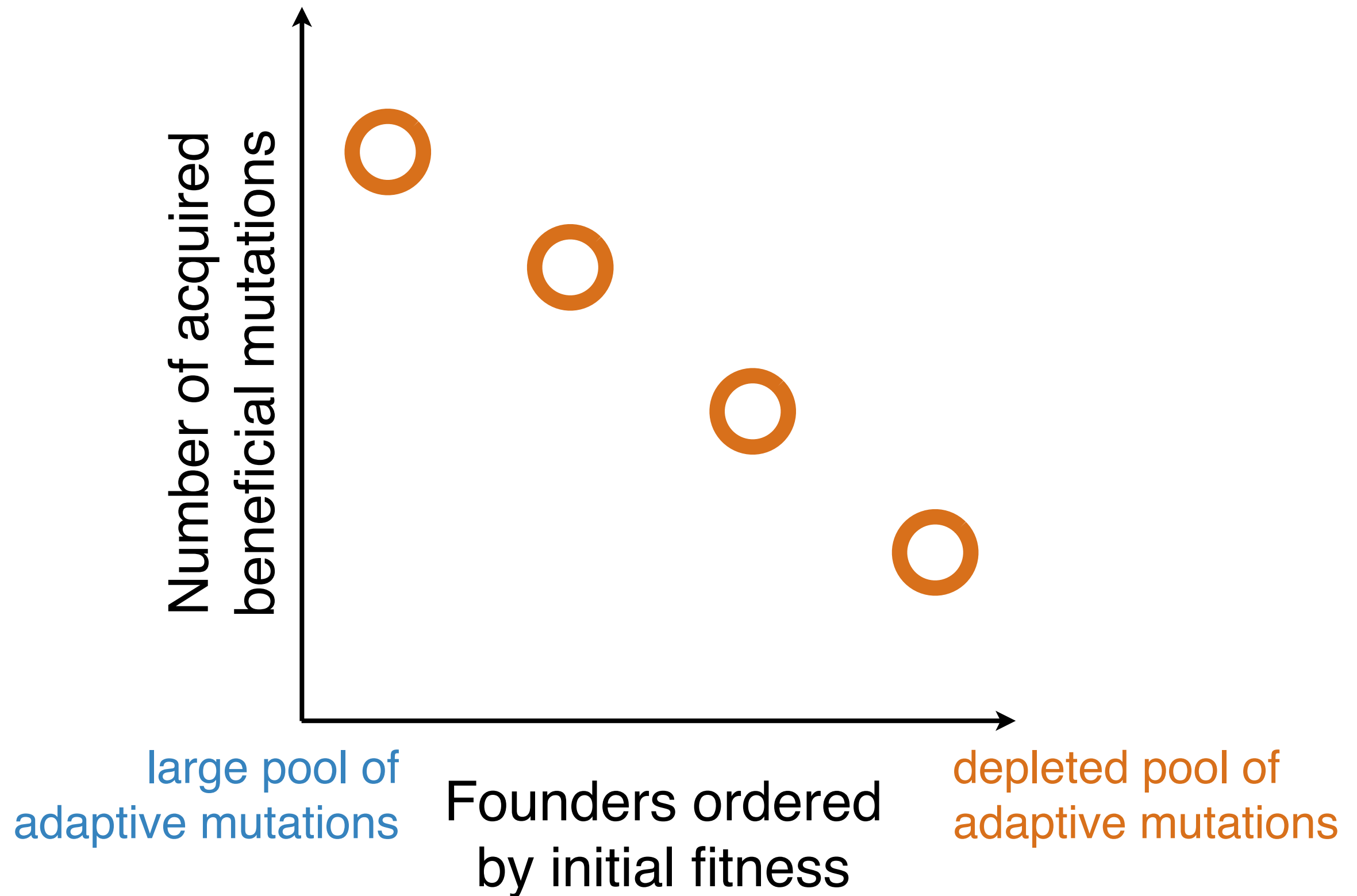
Why does “adaptability”  
decline with fitness?

# Sequenced full genomes of 104 adapted clones

Found 1150 mutations

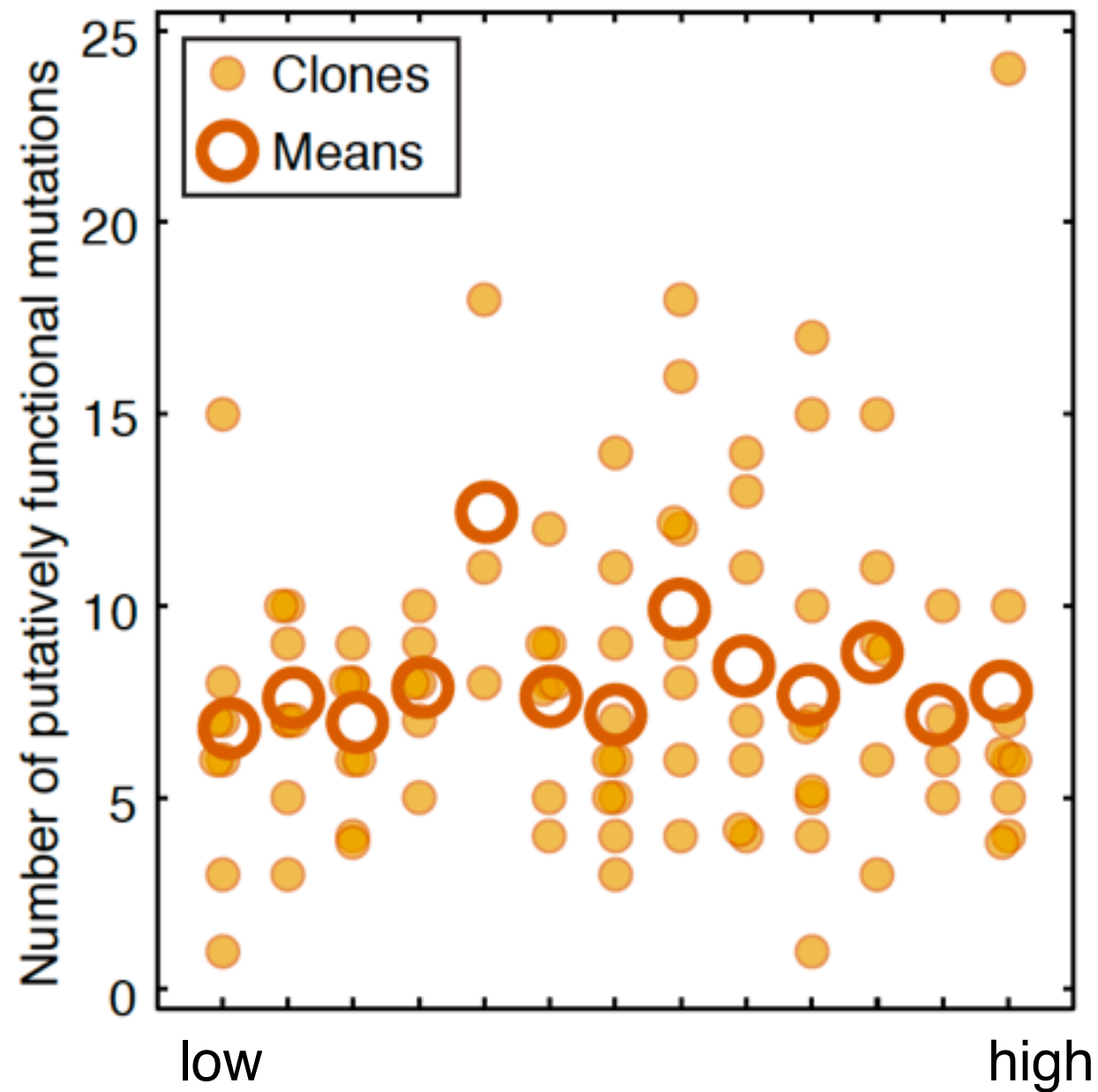


# Hypothesis: “Running out” of beneficial mutations



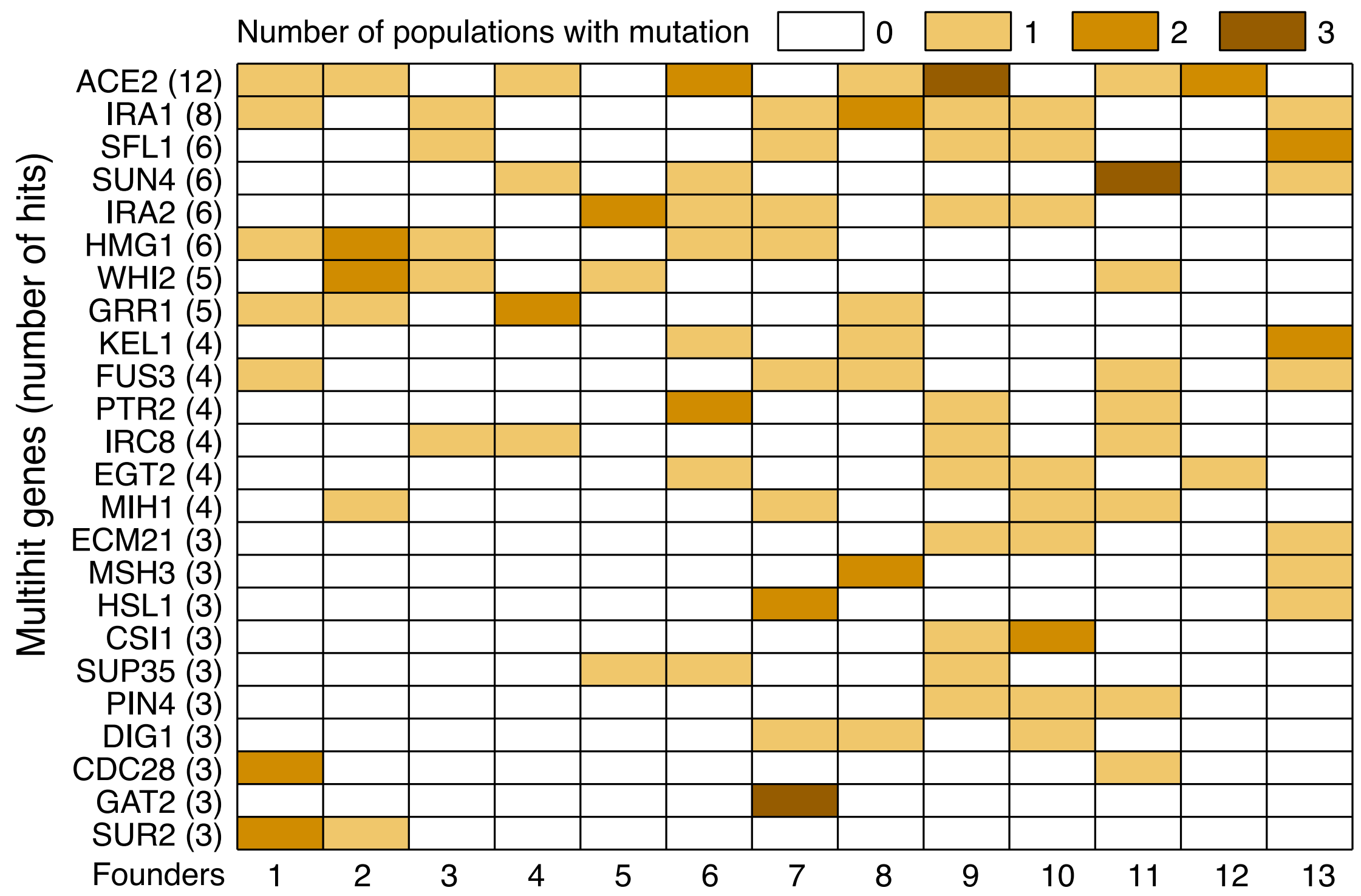


# No support for the running out of beneficial mutations



Founders ordered  
by initial fitness

# Hypothesis: Different Founders acquire different mutations

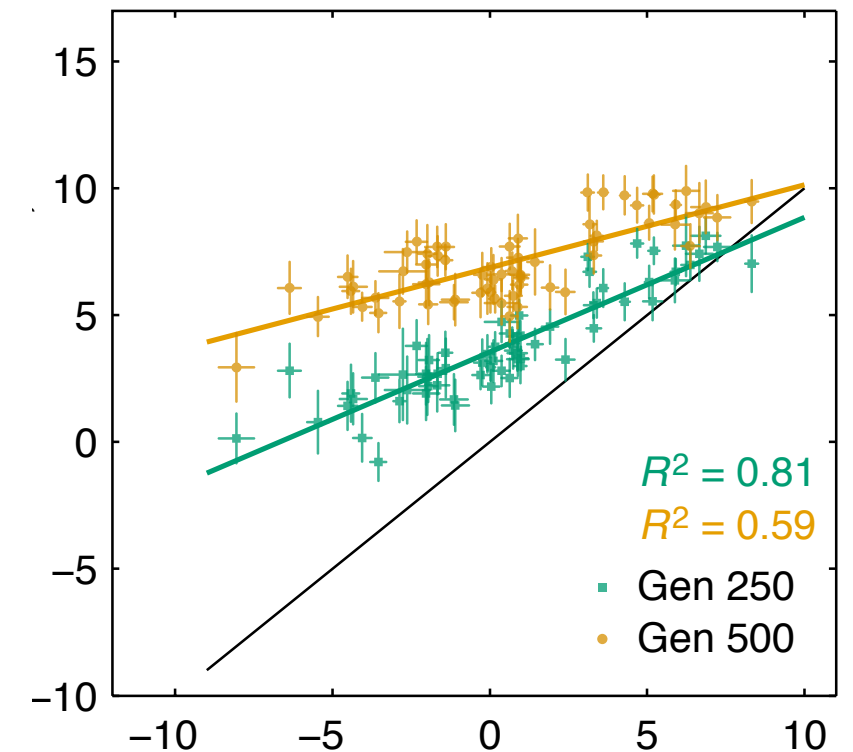
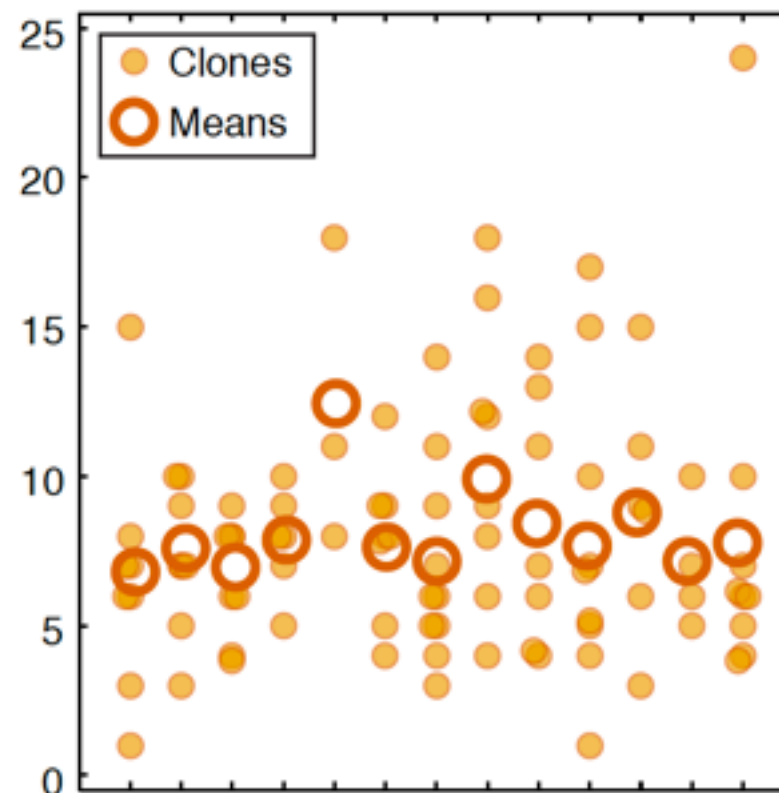
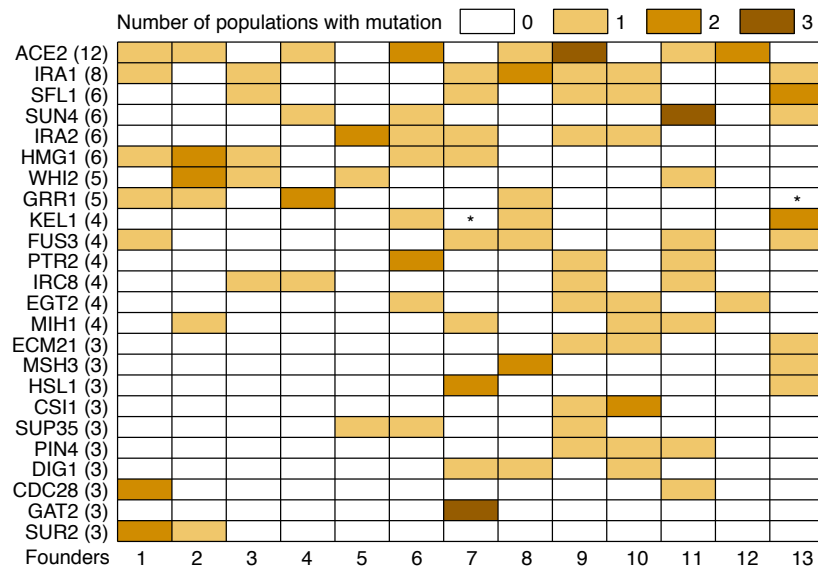


# Consider all data together

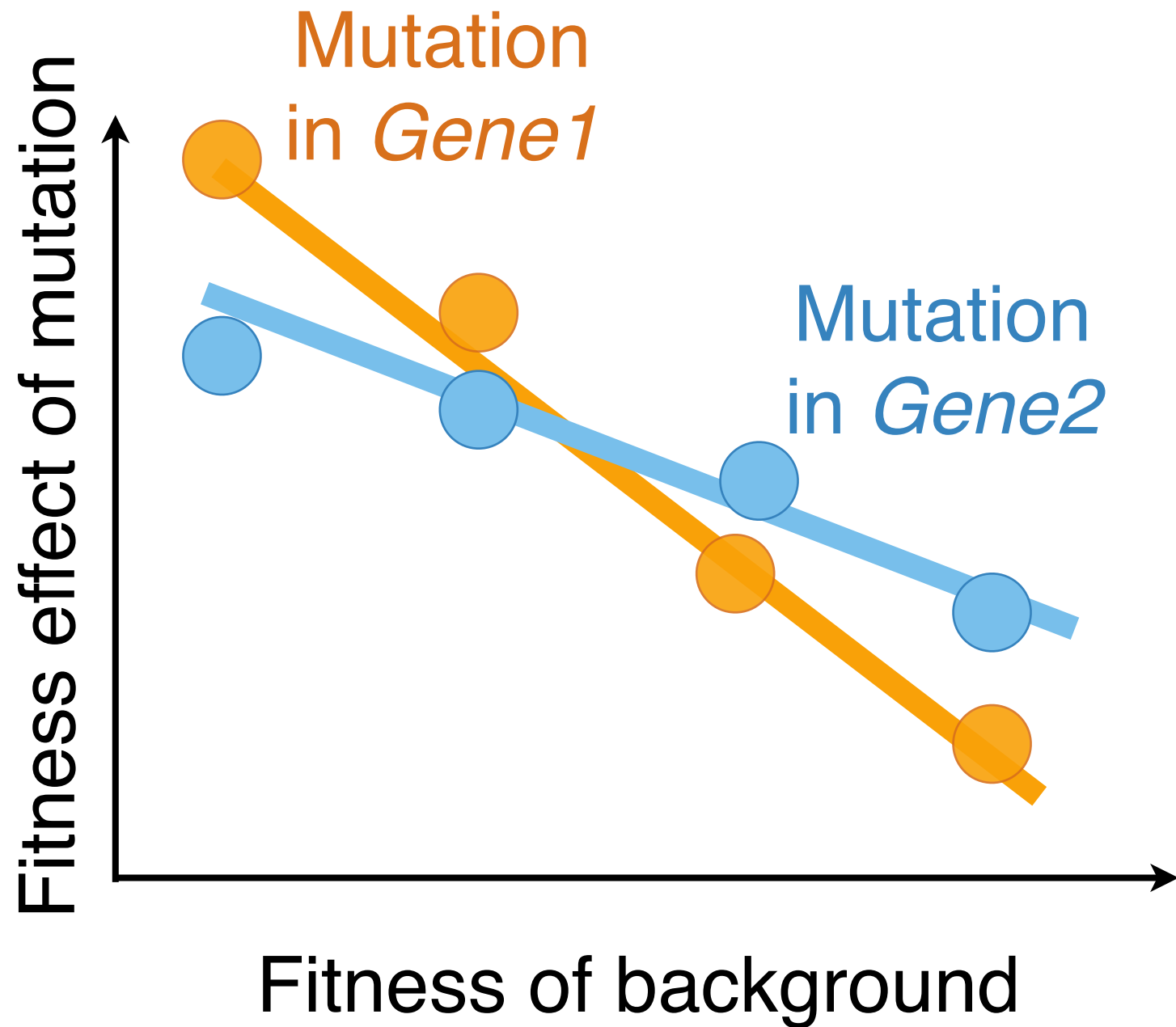
All Founders acquire mutations from the same pool

All Founders acquire same number of mutations

Fitter Founders adapt slower

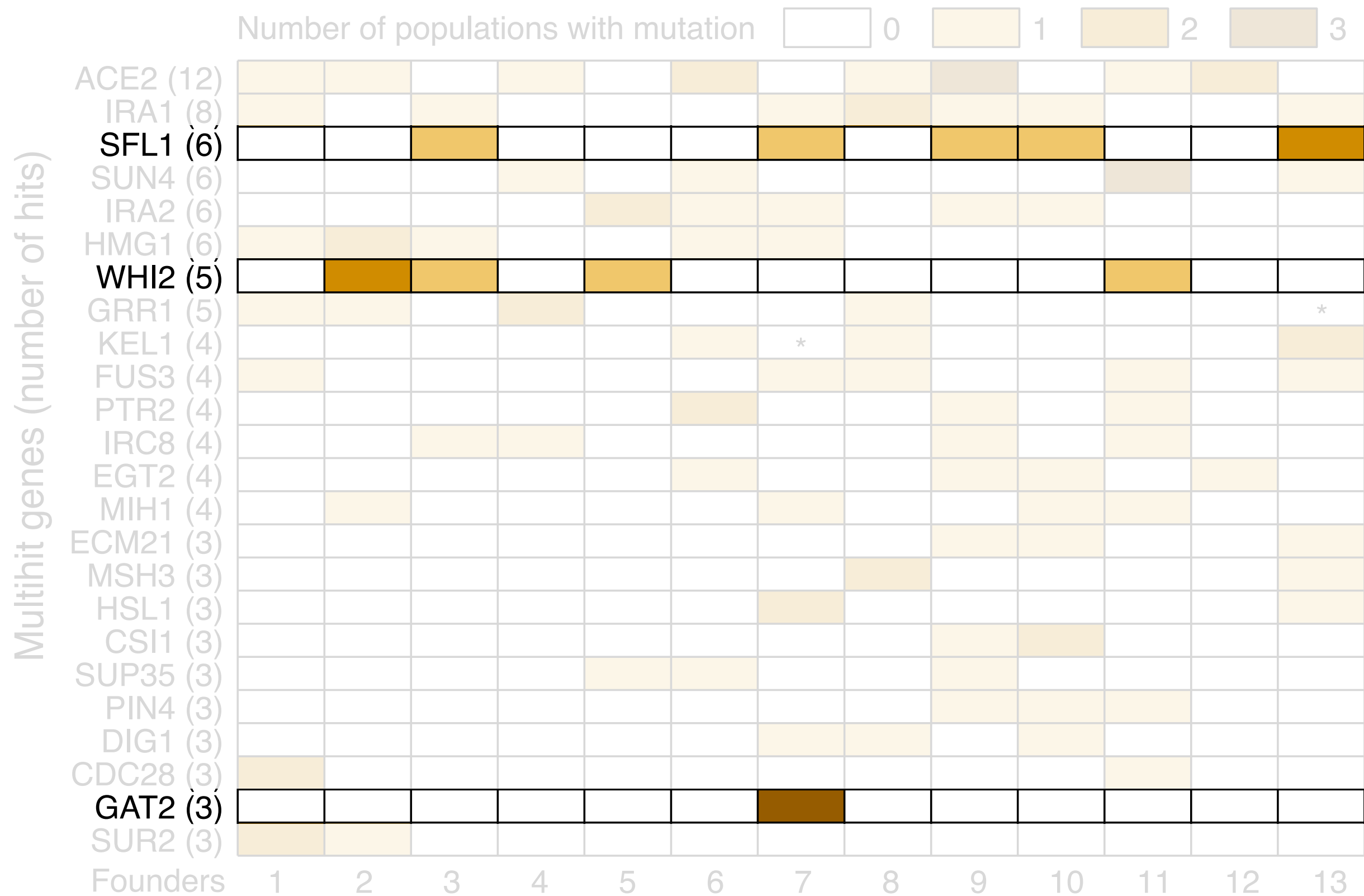


# “Diminishing returns” epistasis

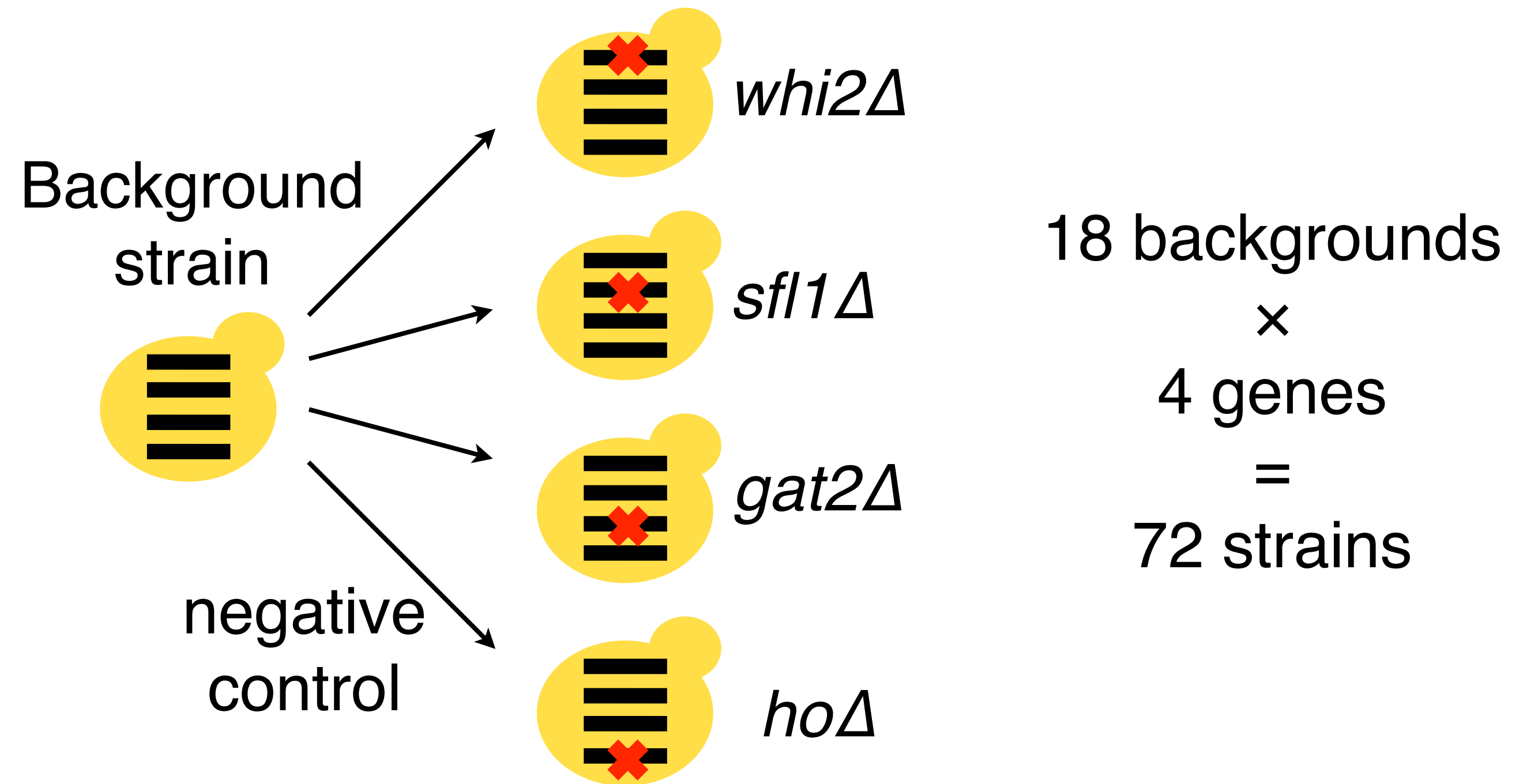




# Test diminishing returns epistasis by allele replacement

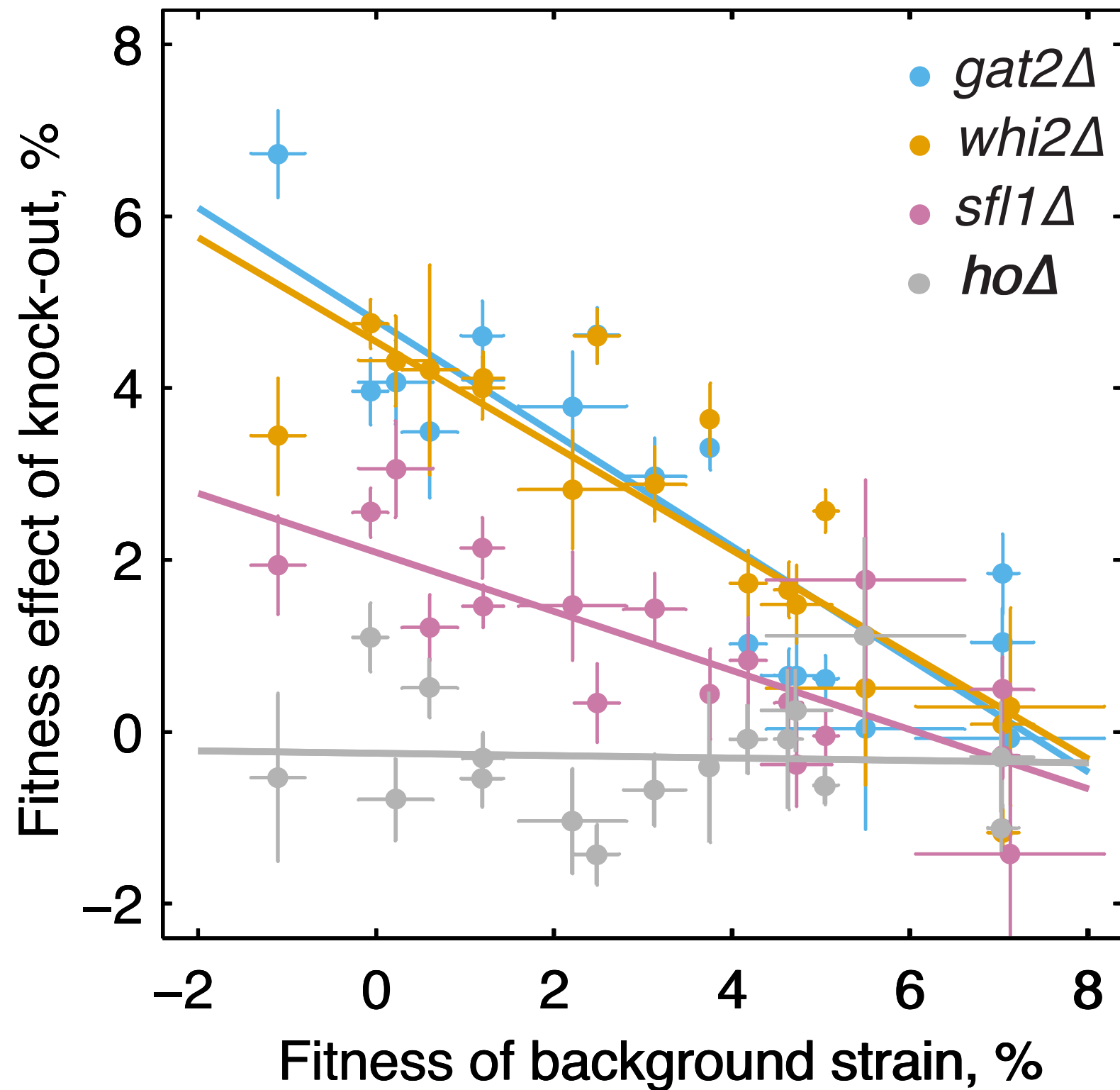


# Knock-out genes in different genetic backgrounds



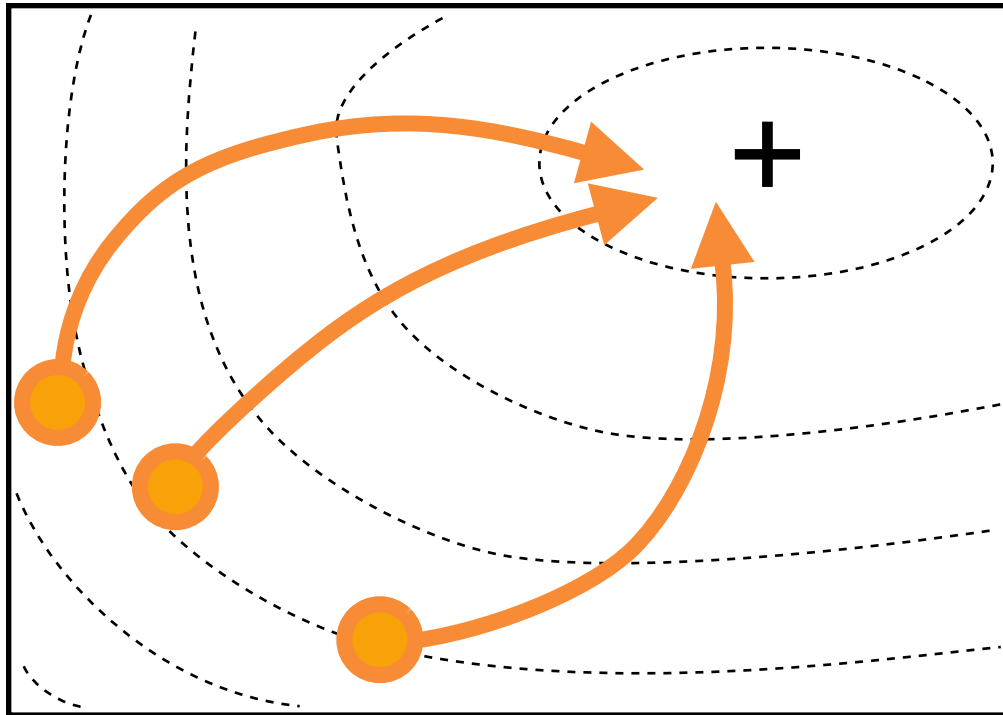
Measure fitness of knock-out strains

# Diminishing returns epistasis supported

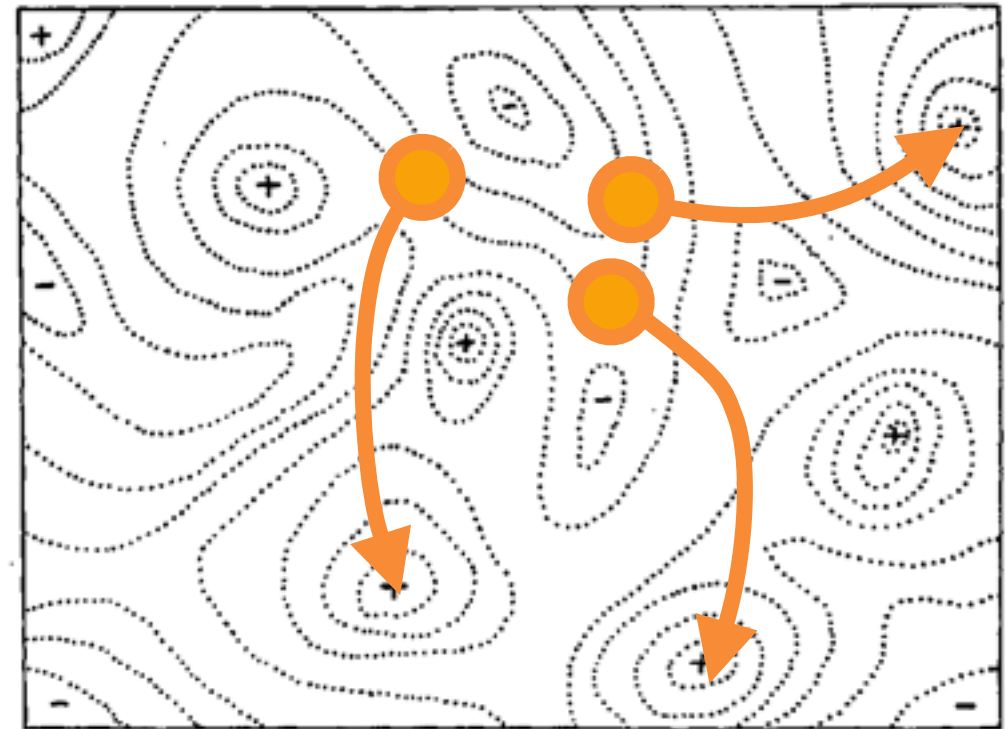


# Fitness landscape structure

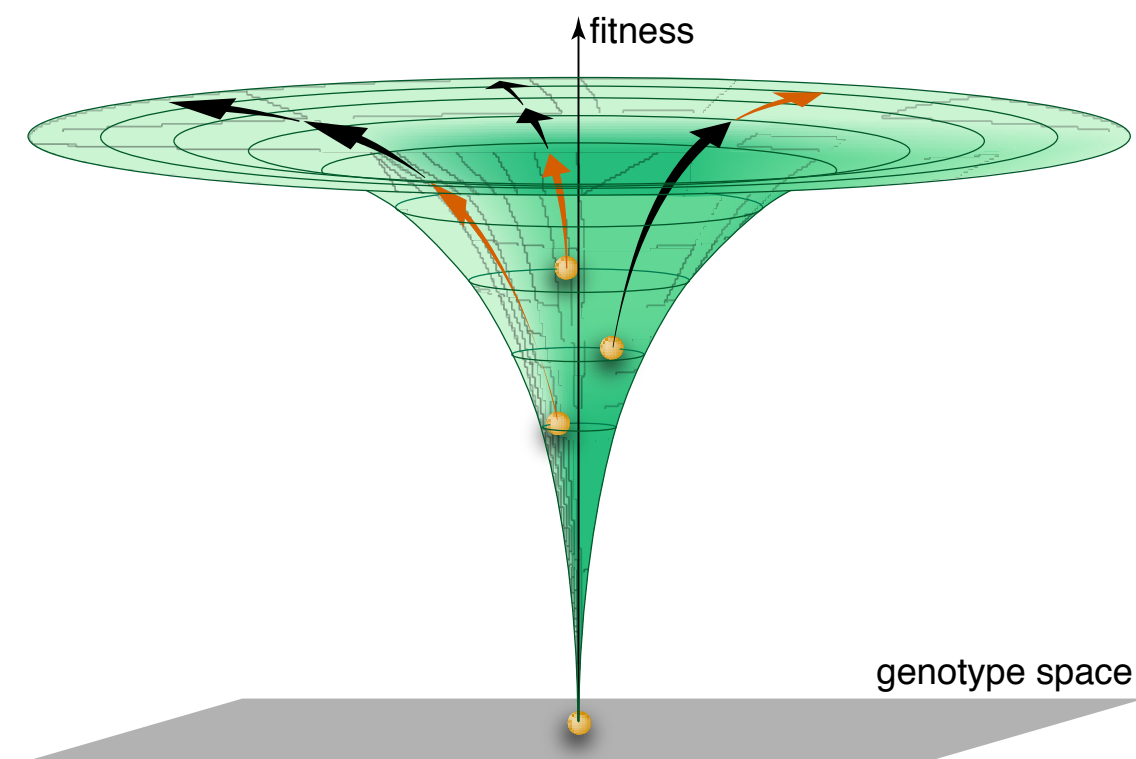
Classic smooth



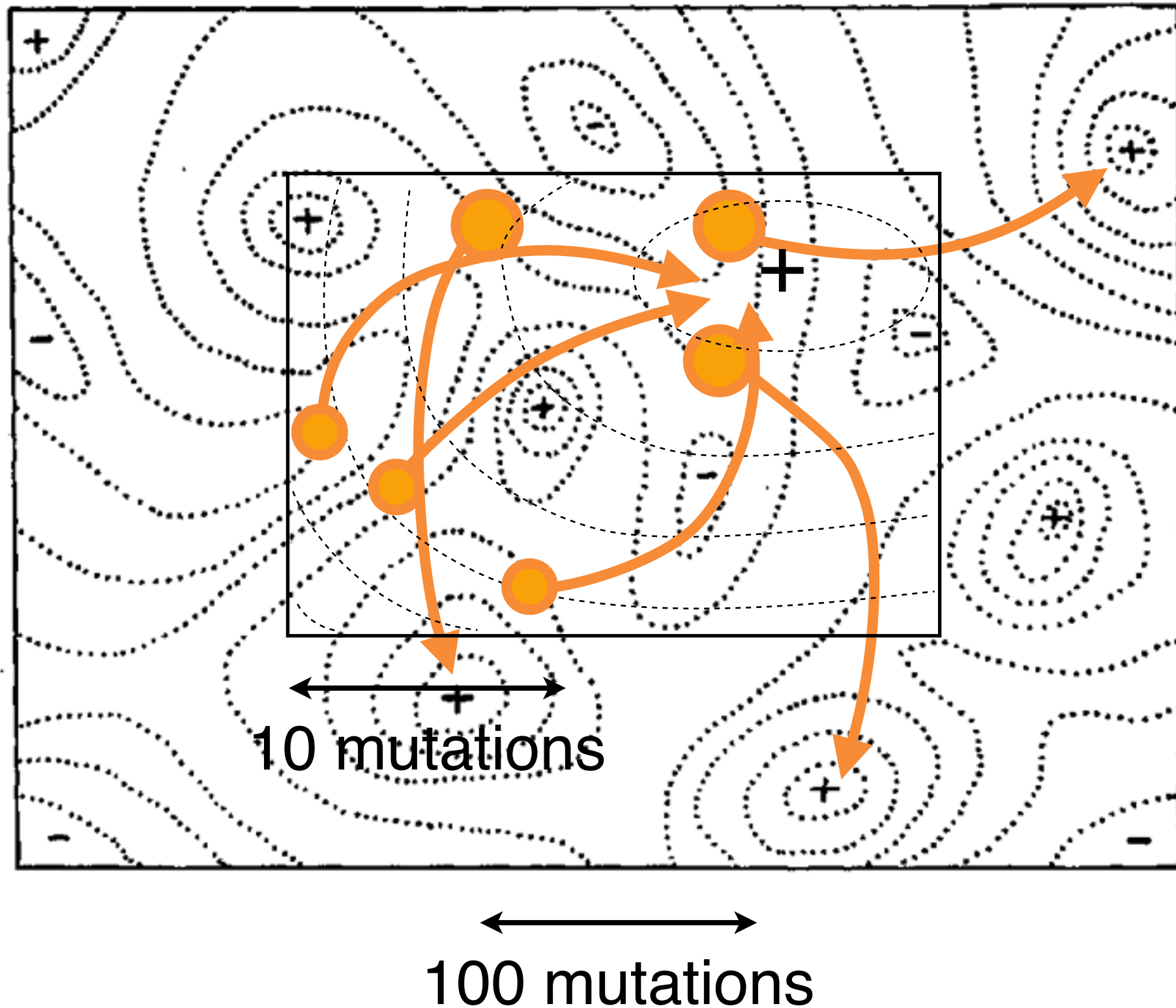
Classic rugged



Diminishing returns

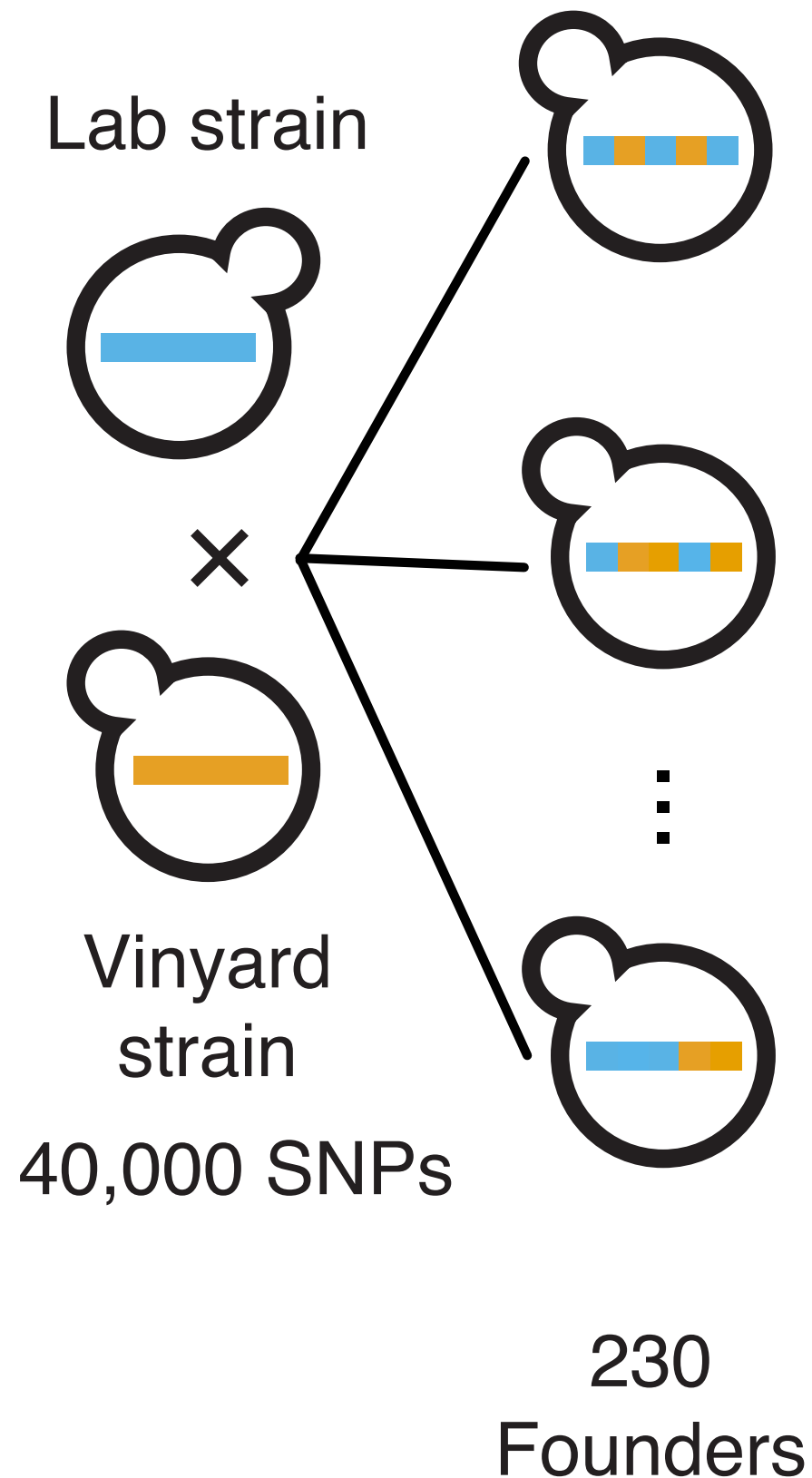


# Ruggedness likely depends on scale of genetic divergence



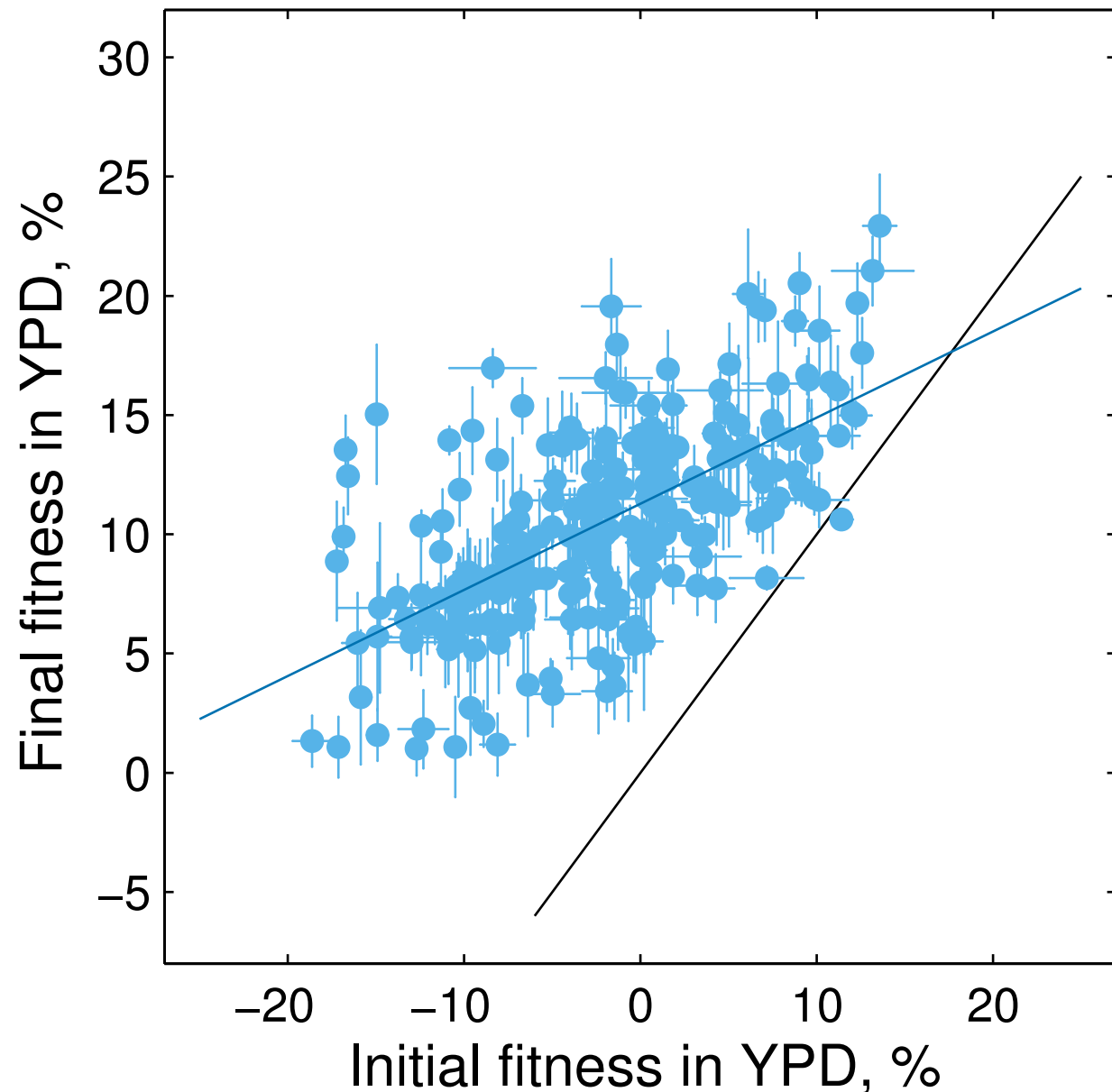


# Adaptation from divergent strains

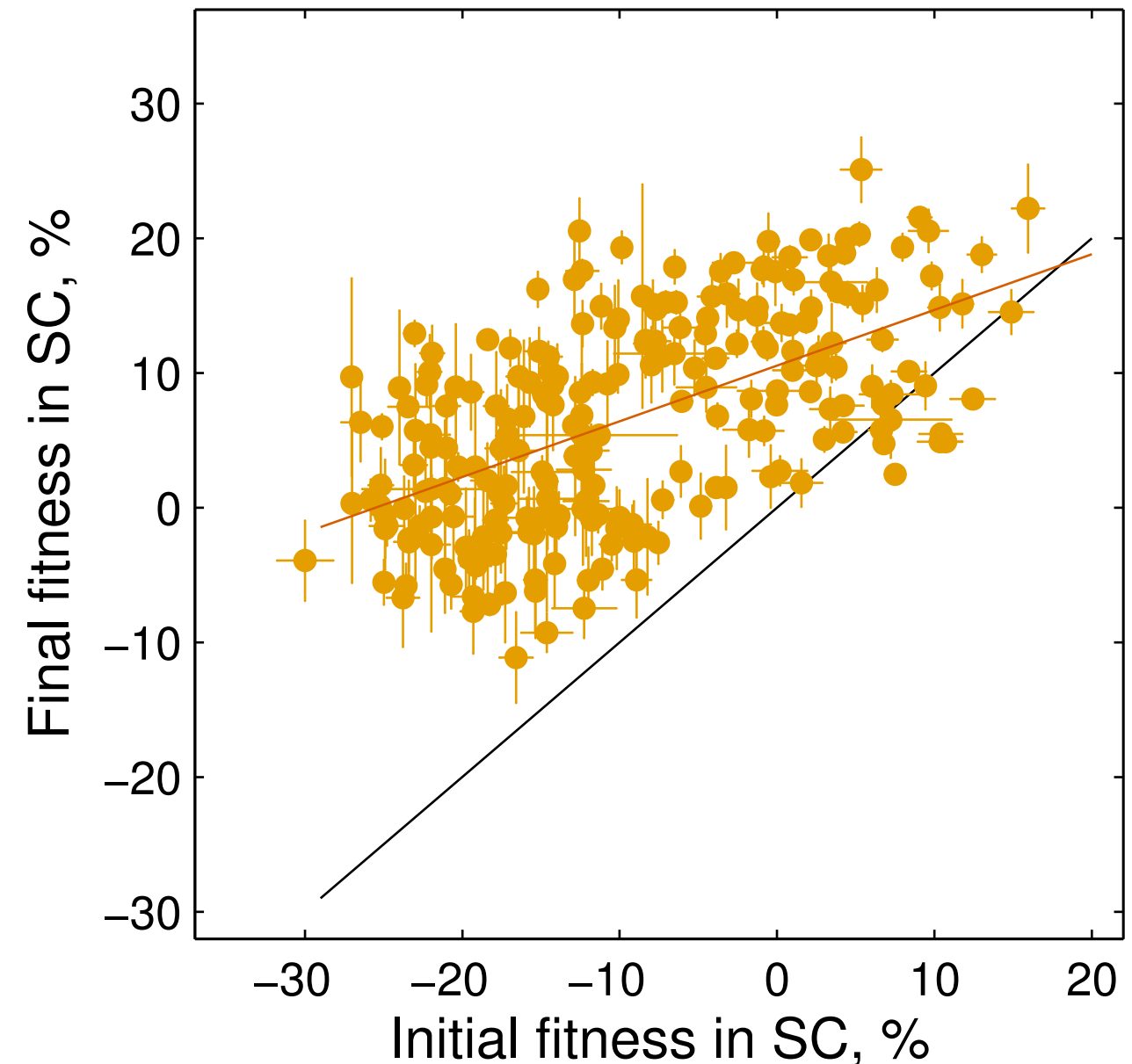


# Observe similar decline in adaptability with fitness among divergent strains

Evolved in YPD + 30°C

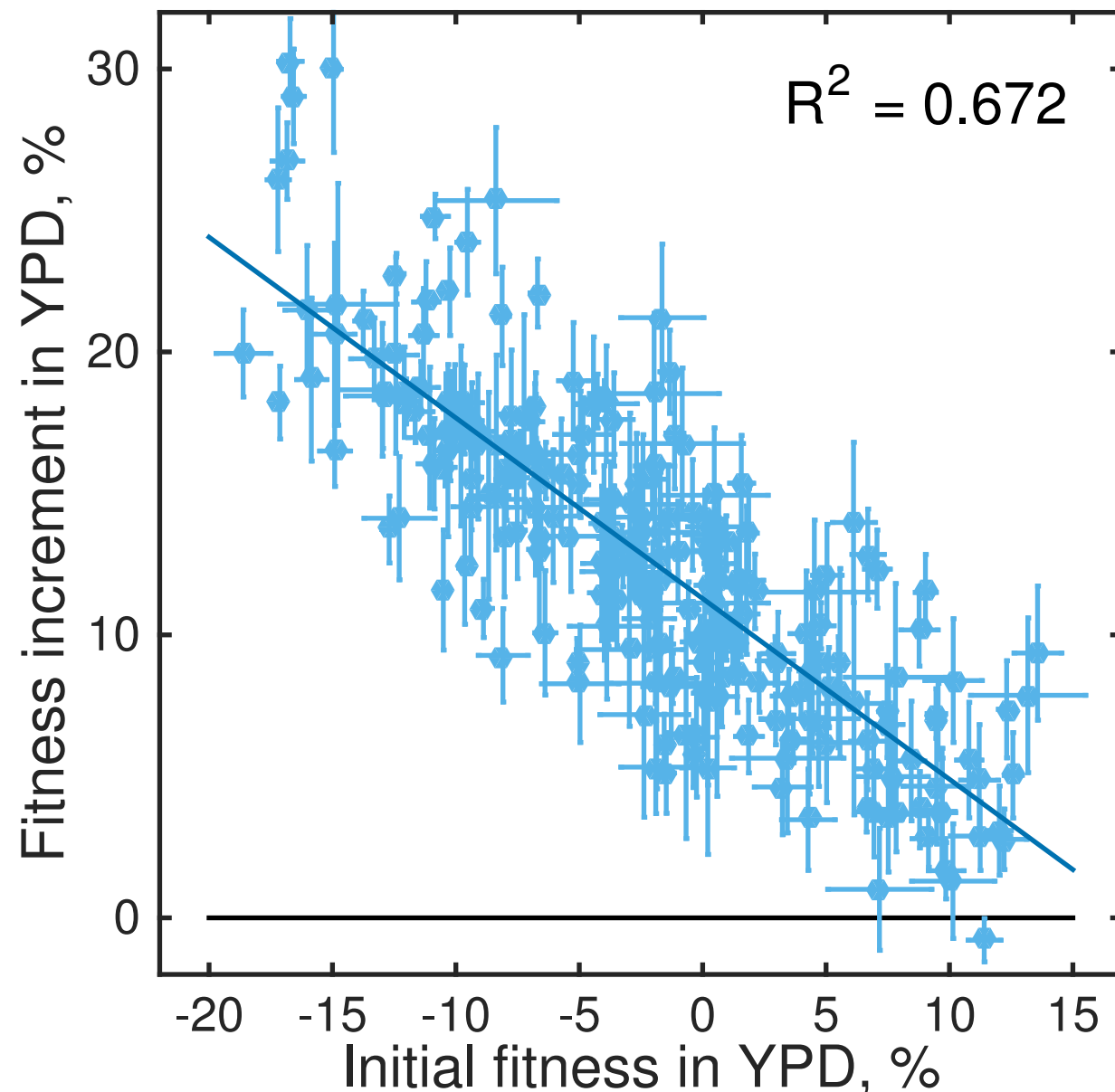


Evolved in SC + 37°C

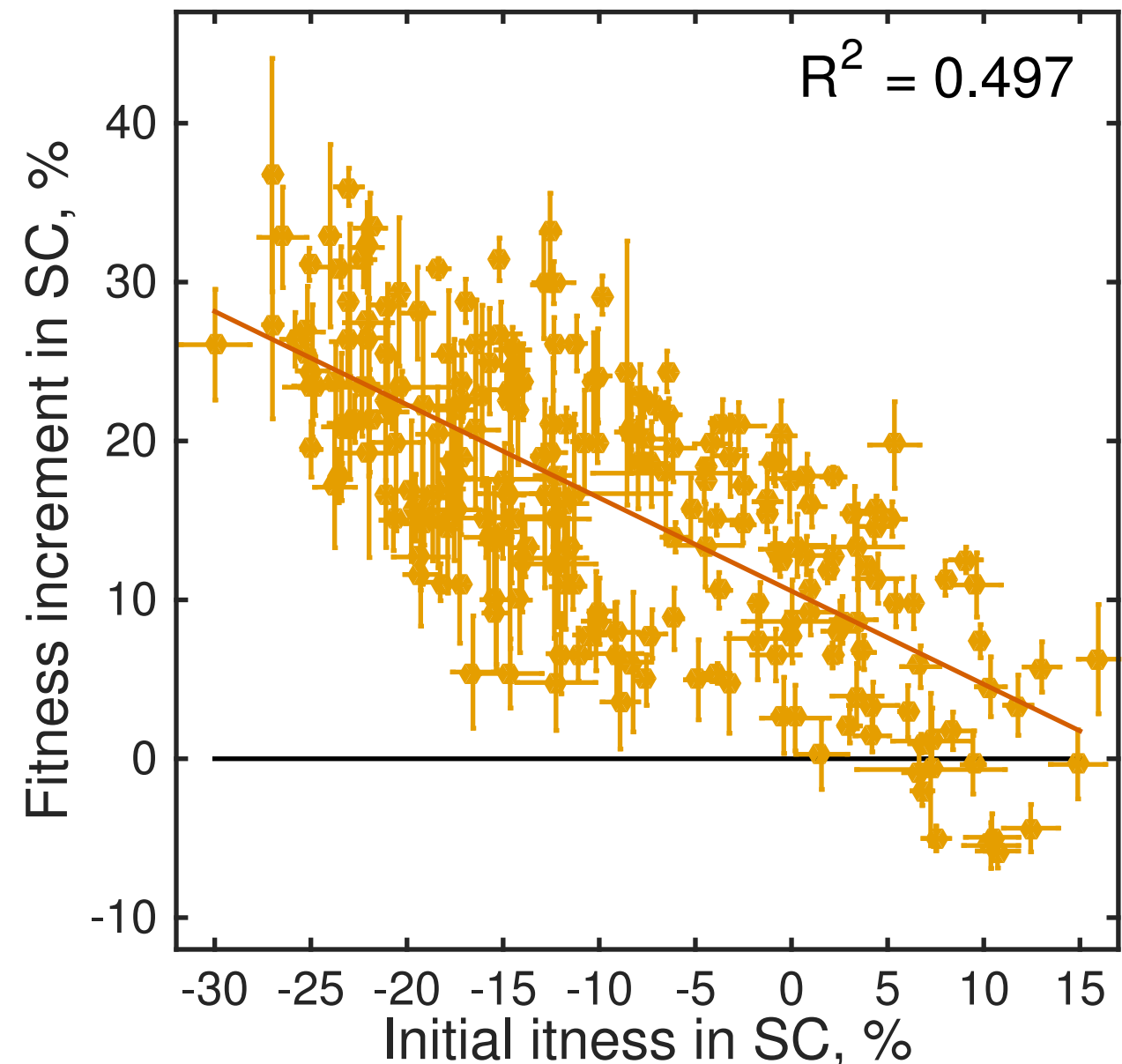


# Observe similar decline in adaptability with fitness among divergent strains

## Evolved in YPD + 30°C

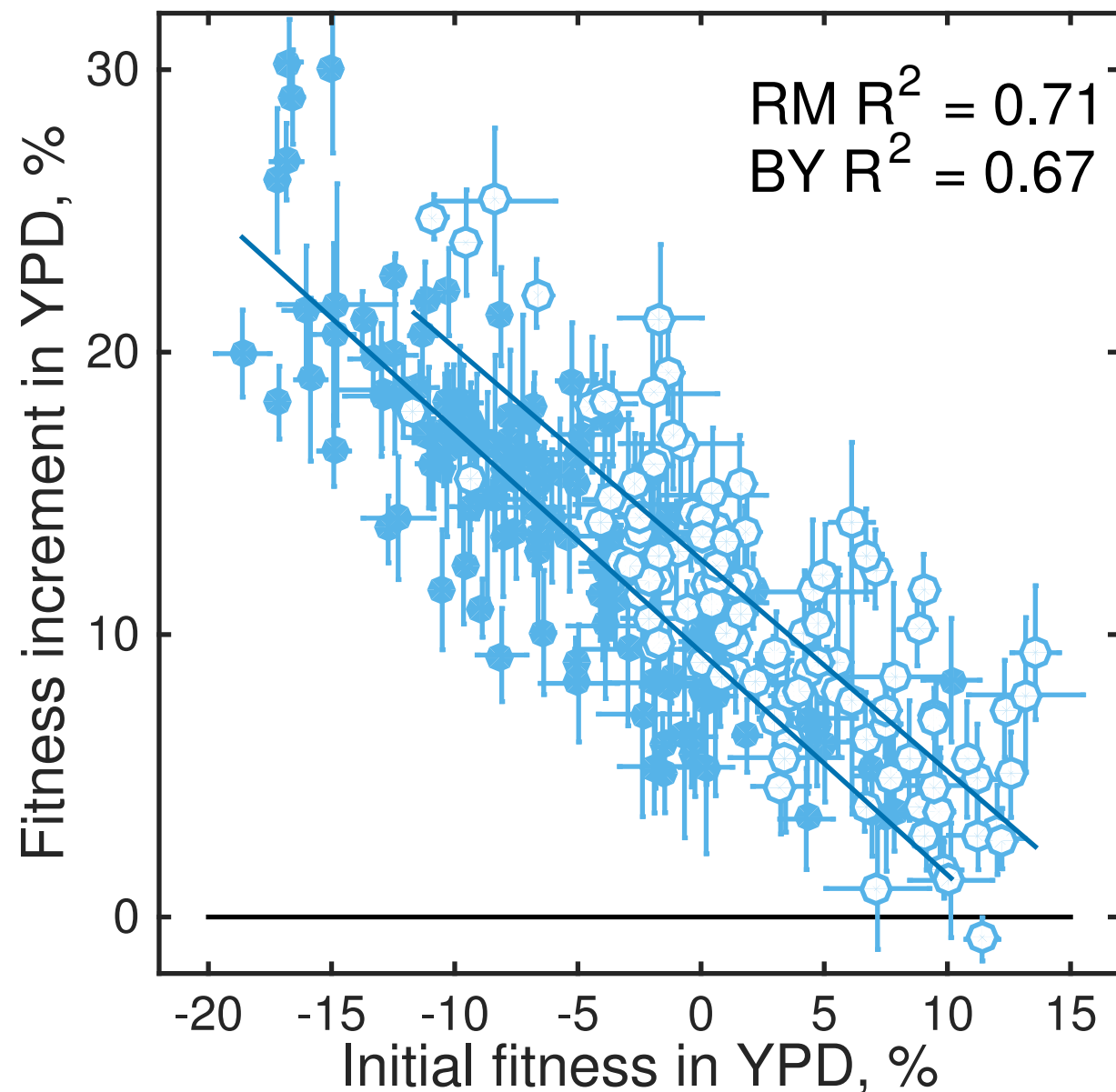


## Evolved in SC + 37°C

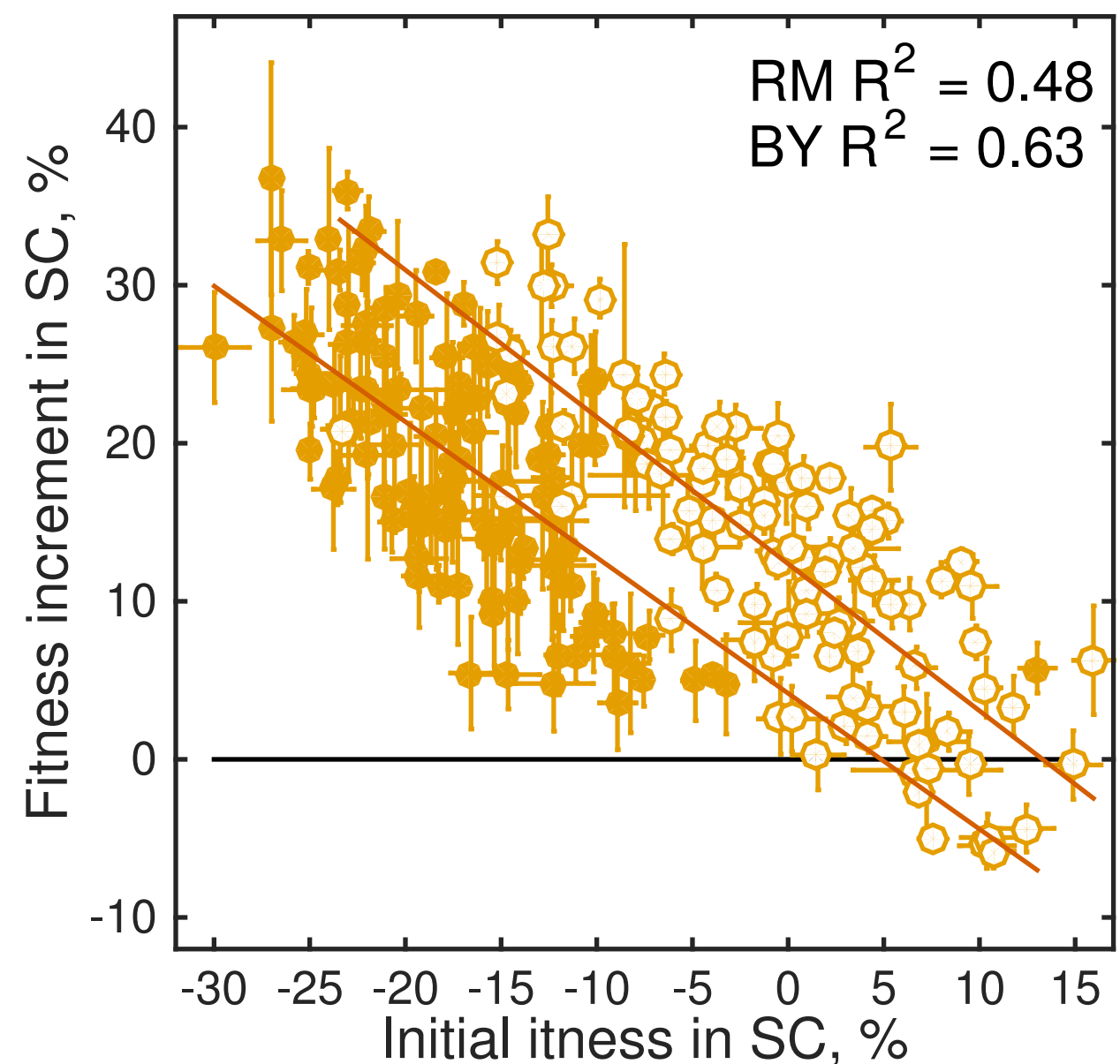


# SNP at KRE33 locus affects adaptability

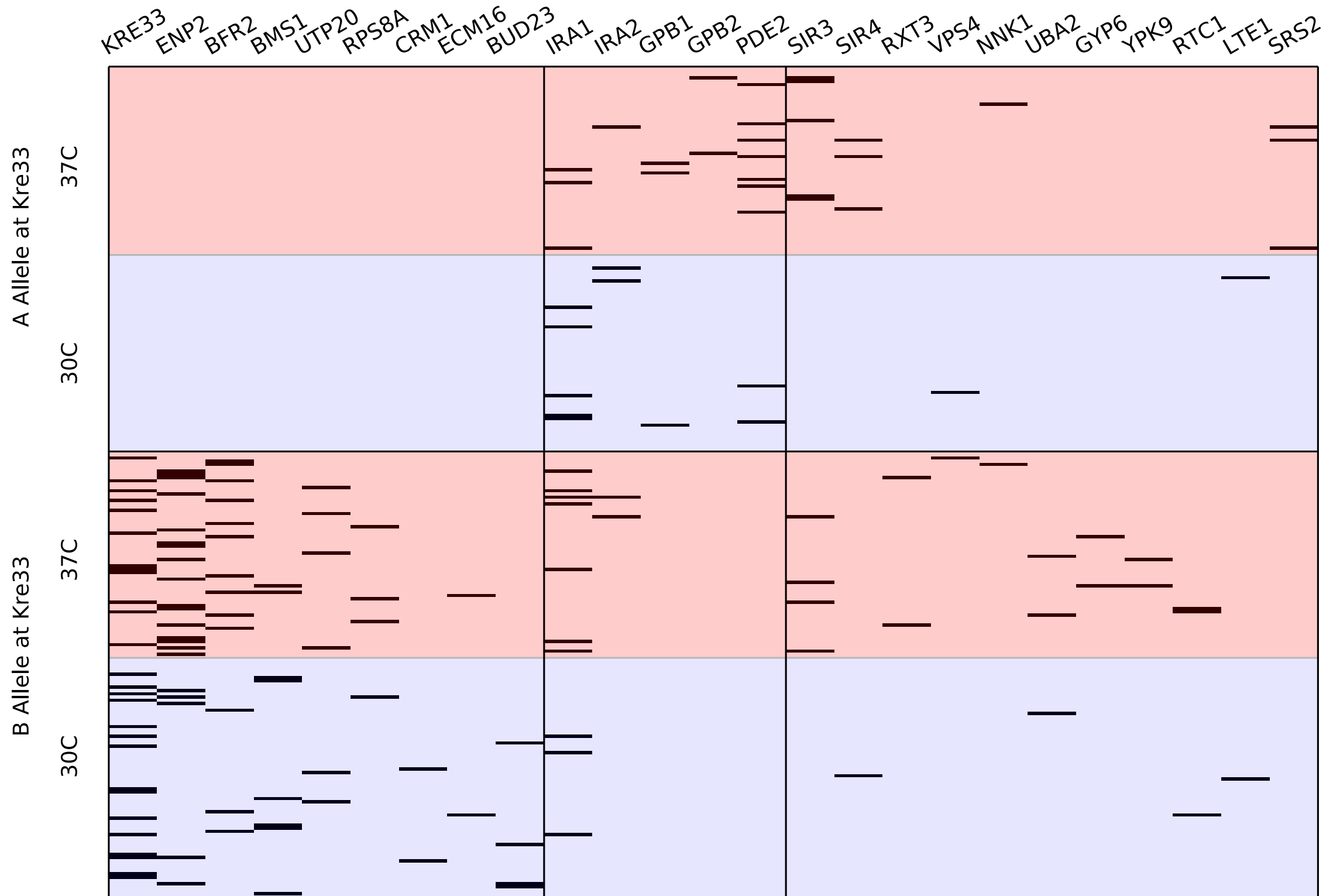
Evolved in YPD + 30°C



Evolved in SC + 37°C



# SNP at KRE33 locus affects the pool of adaptive mutations





# Conclusions

“Rule of declining adaptability” = rate of adaptation declines with initial fitness

Negative “diminishing returns” epistasis is at least partially responsible

Pool of adaptive mutations is common to most closely related genotypes

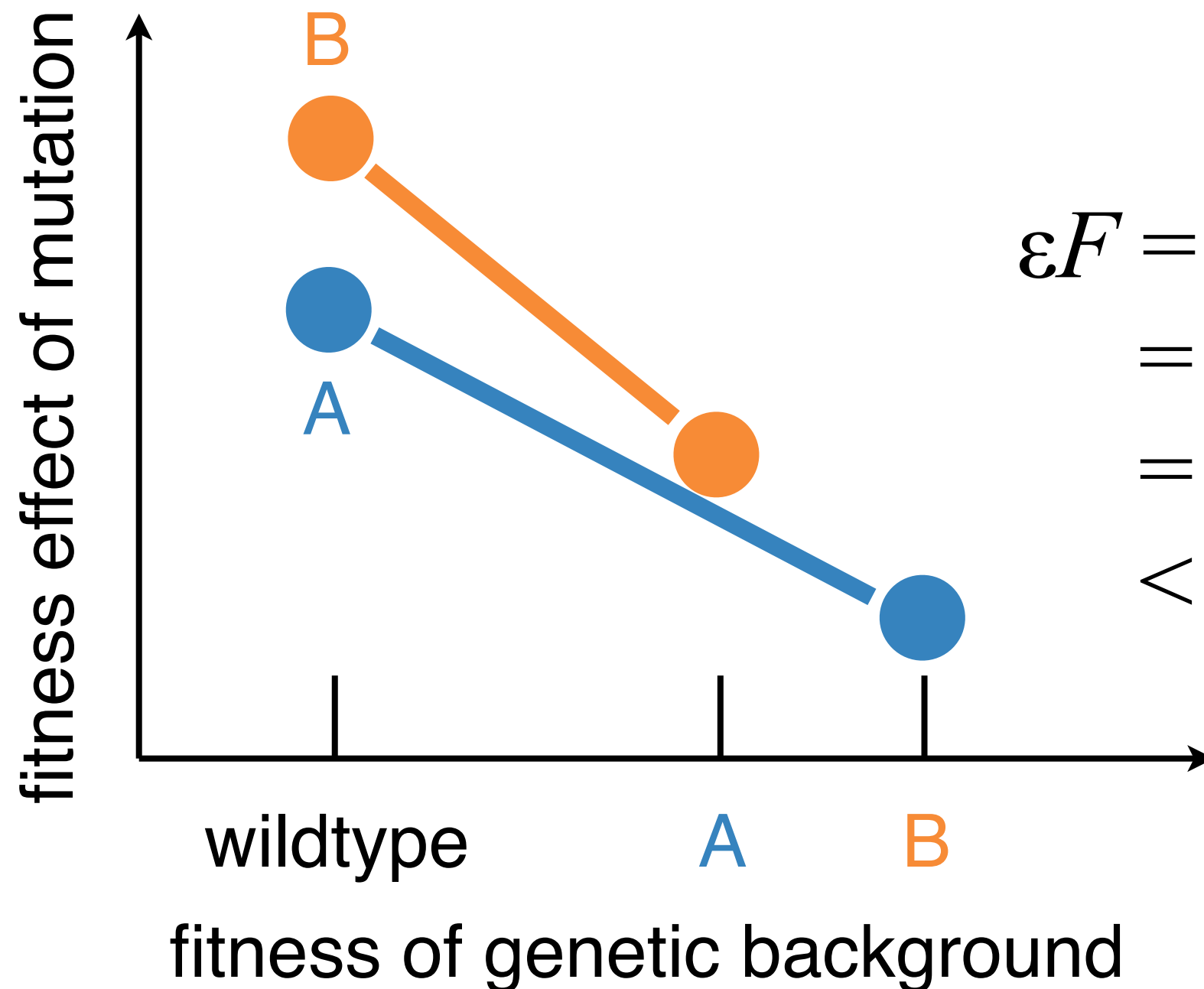
Some (rare) mutations dramatically change the pool of adaptive mutations

Part 2.

Where does epistasis come from?

# Epistasis between two mutations

“Diminishing returns” epistasis between two beneficial mutations



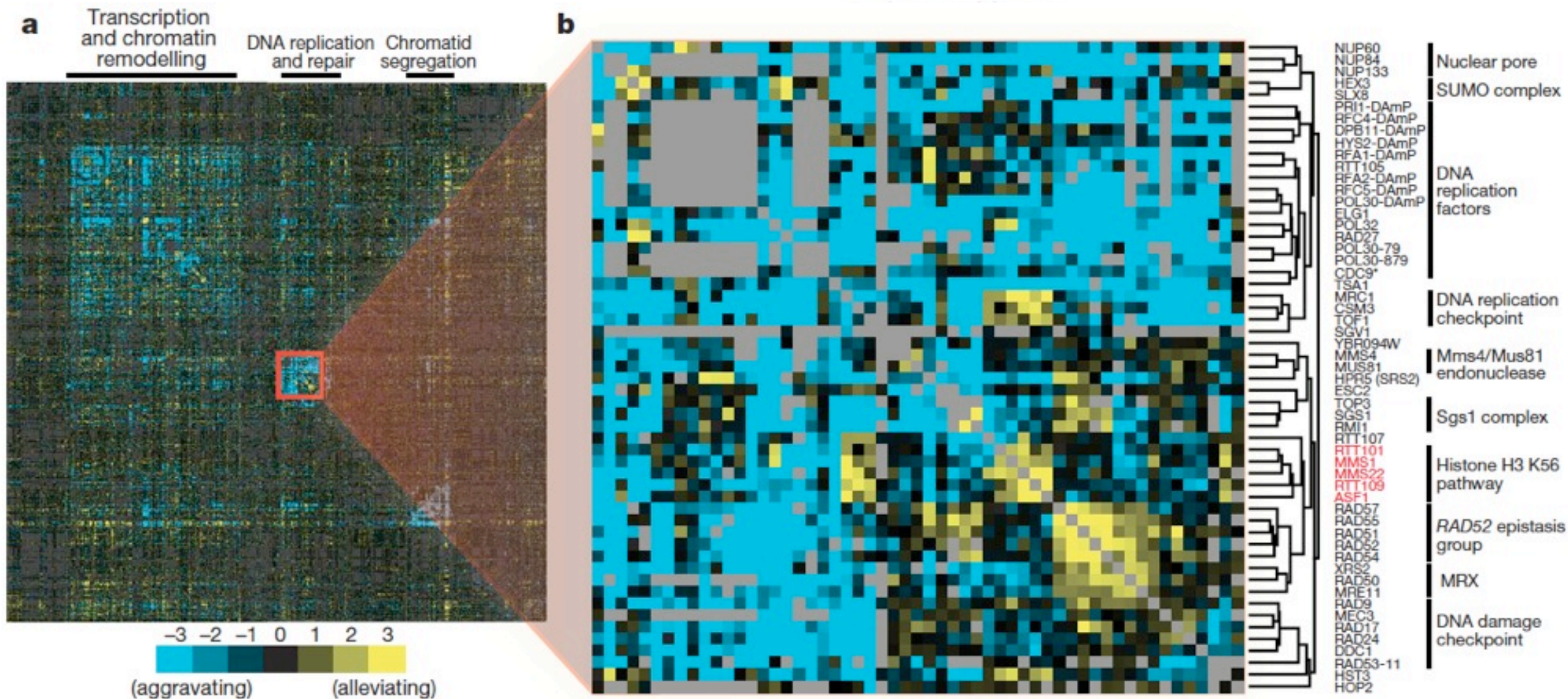
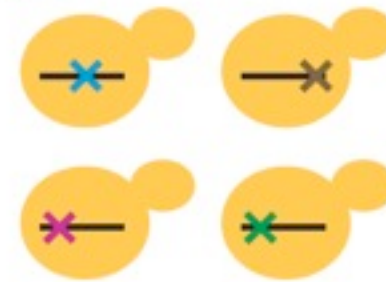
$$\Delta F_{B|A} < \Delta F_{B|WT}$$

$$\Delta F_{A|B} < \Delta F_{A|WT}$$

$$\begin{aligned}\varepsilon F &= \Delta F_{B|A} - \Delta F_{B|WT} \\ &= \Delta F_{A|B} - \Delta F_{A|WT} \\ &= \Delta F_{AB} - \Delta F_A - \Delta F_B \\ &< 0\end{aligned}$$

# Systematic measurements of epistasis between gene knock-out mutations

*Saccharomyces cerevisiae*



We measured epistasis. So what?



# Do we have a null expectation for epistasis?

## Defining genetic interaction

Ramamurthy Mani\*, Robert P. St.Onge<sup>†</sup>, John L. Hartman IV<sup>‡</sup>, Guri Giaever<sup>§</sup>, and Frederick P. Roth\*<sup>¶||</sup>

**Properties Expected of an Ideal Definition for Identifying Functional Relationships.** Gene function can be defined at multiple levels of specificity (27). By definition, there are few genes that hold any given specific function, and gene pairs sharing a specific function should then also be rare. Therefore, if interaction (either synthetic or alleviating) is to be an ideal indicator of specific functional relationships, the vast majority of gene pairs should be noninteracting. An ideal definition for interaction should then yield a distribution of observed double-mutant fitness values that closely approximates the expected distribution over most gene pairs.

# Do we have a null expectation for epistasis?

## Quantitative analysis of fitness and genetic interactions in yeast on a genome scale

Anastasia Baryshnikova<sup>1,2,10</sup>, Michael Costanzo<sup>1,10</sup>, Yungil Kim<sup>3,4</sup>, Huiming Ding<sup>1</sup>, Judice Koh<sup>1</sup>, Kiana Toufighi<sup>1</sup>, Ji-Young Youn<sup>1,2</sup>, Jiongwen Ou<sup>5</sup>, Bryan-Joseph San Luis<sup>1</sup>, Sunayan Bandyopadhyay<sup>3</sup>, Matthew Hibbs<sup>6</sup>, David Hess<sup>7</sup>, Anne-Claude Gingras<sup>8</sup>, Gary D Bader<sup>1,2</sup>, Olga G Troyanskaya<sup>9</sup>, Grant W Brown<sup>5</sup>, Brenda Andrews<sup>1,2</sup>, Charles Boone<sup>1,2</sup> & Chad L Myers<sup>3</sup>

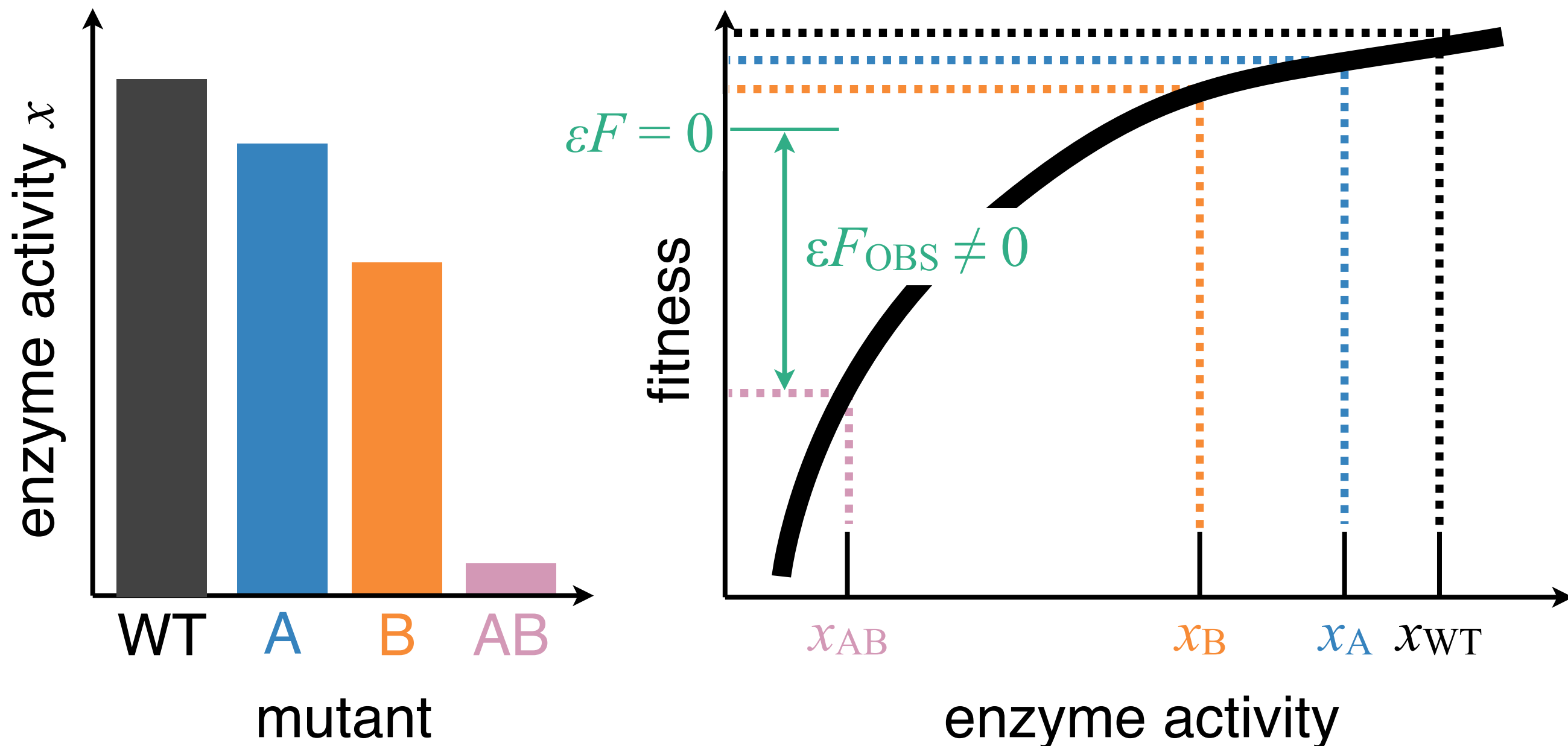
size is related to the corresponding query mutation. From **Eq. 1** above, we assume that observed colony area is a function of the single mutations fitness defects as well as time:

$$C_{ij} = \alpha \cdot i_j \cdot f_j \cdot t \cdot s_{ij} \cdot e \quad \text{Eq. 2}$$

as in most cases  $\varepsilon_{ij} \approx 0$  because genetic interactions are rare. Due to the

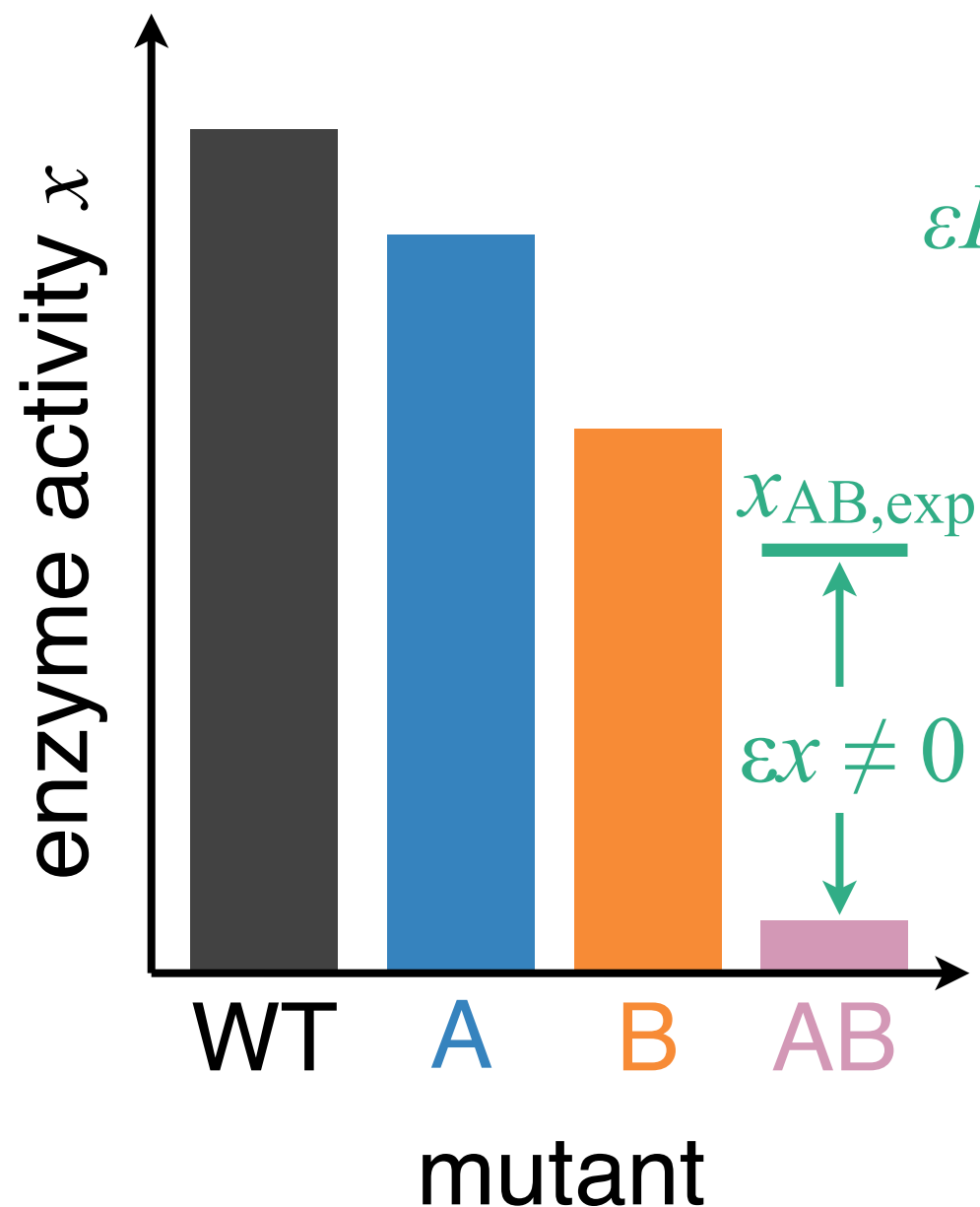
# Is “no epistasis” a biologically meaningful expectation?

Simple example: epistasis between mutations in the same enzyme

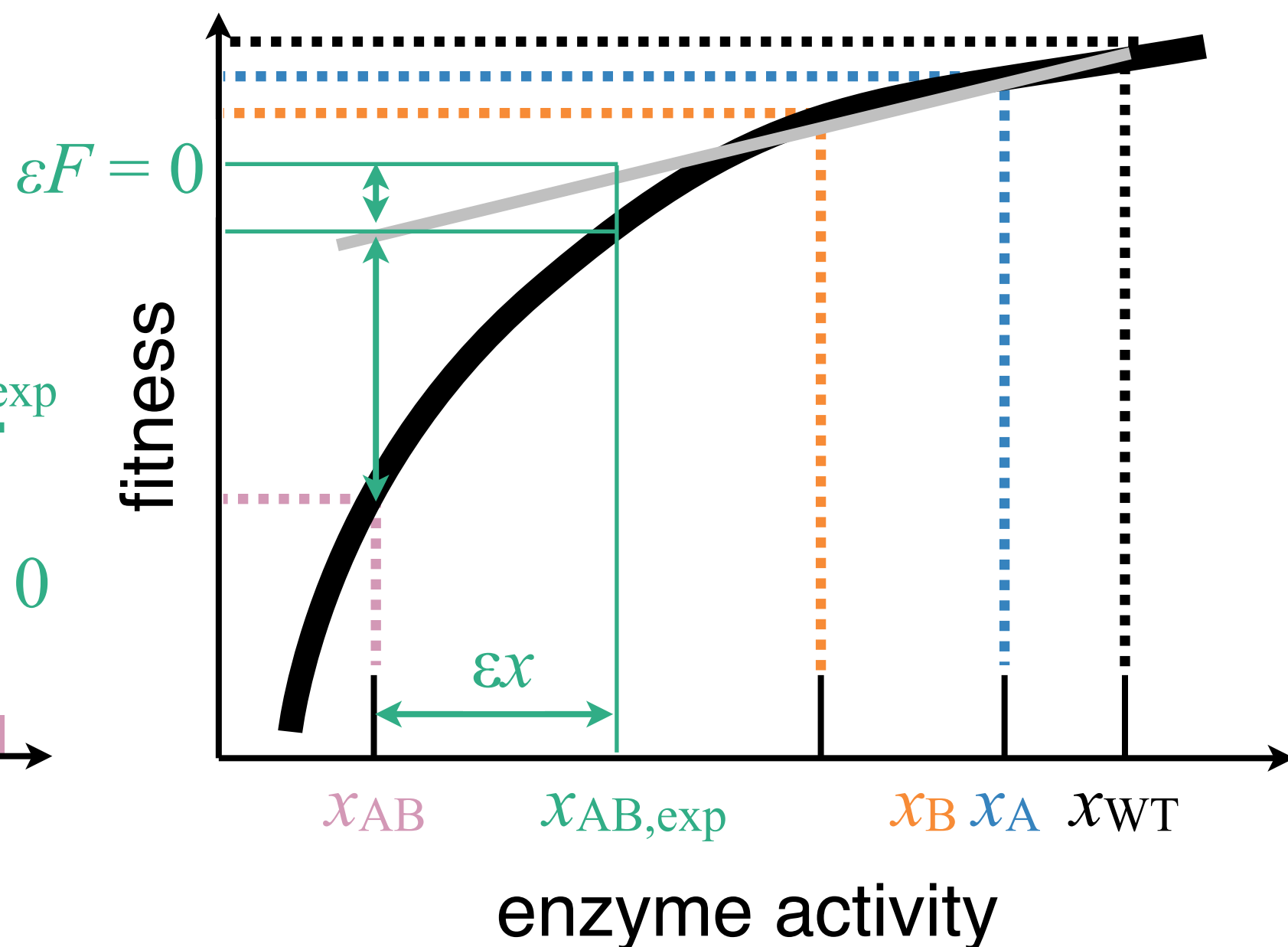


# Within-protein epistasis arises from two contributions

1. Activity is not additive



2. Fitness is a non-linear function of activity





# Within-protein epistasis arises from two contributions

$F = F(x)$  Fitness is function of activity

$\delta^A x, \delta^B x, \delta^{AB} x$  Effects of mutations on activity

$\varepsilon^{AB} x = \delta^{AB} x - \delta^A x - \delta^B x$  Epistasis for activity

$\delta^A F, \delta^B F, \delta^{AB} F$  Effects of mutations on fitness

$\varepsilon^{AB} F = \delta^{AB} F - \delta^A F - \delta^B F$  Epistasis for fitness

$$\varepsilon^{AB} F = C \cdot \varepsilon^{AB} x + H \cdot \delta^A x \cdot \delta^B x$$

“Propagation of epistasis”

# Epistasis between enzymes also arises from two contributions

$F = F(x_1, x_2)$       Fitness is function of activities  $x_1, x_2$

$\delta^A x_i, \delta^B x_i, \delta^{AB} x_i$       Effects of mutations on activity  $i$

$\varepsilon^{AB} x_i = \delta^{AB} x_i - \delta^A x_i - \delta^B x_i$       Epistasis for activity  $i$

$\delta^A F, \delta^B F, \delta^{AB} F$       Effects of mutations on fitness

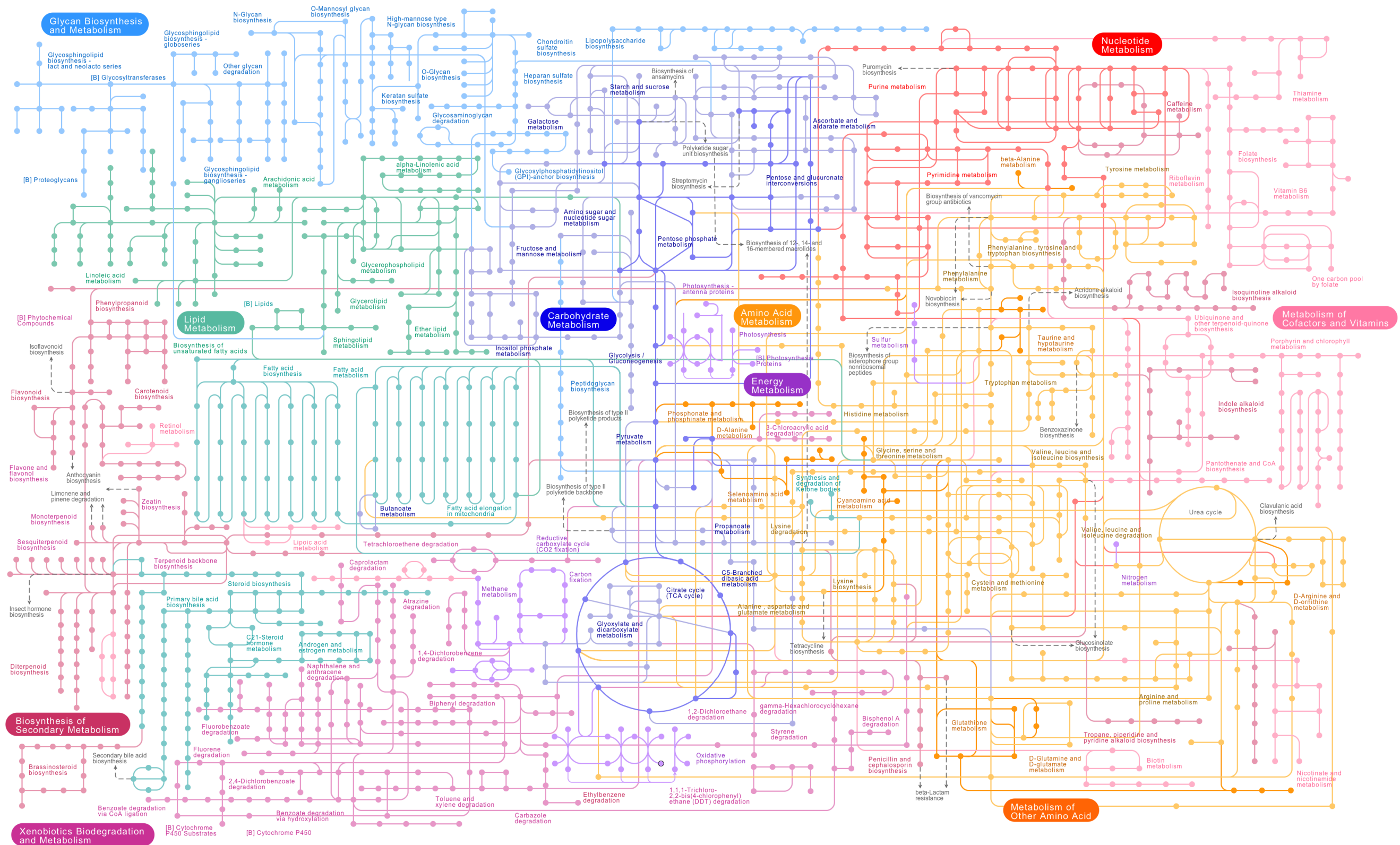
$\varepsilon^{AB} F = \delta^{AB} F - \delta^A F - \delta^B F$       Epistasis for fitness

$$\varepsilon^{AB} F = \mathbf{C} \cdot \varepsilon^{AB} \mathbf{x} + (\delta^A \mathbf{x})^T \cdot \mathbf{H} \cdot \delta^B \mathbf{x}$$

“Propagation of epistasis”



# What is function $F$ ?



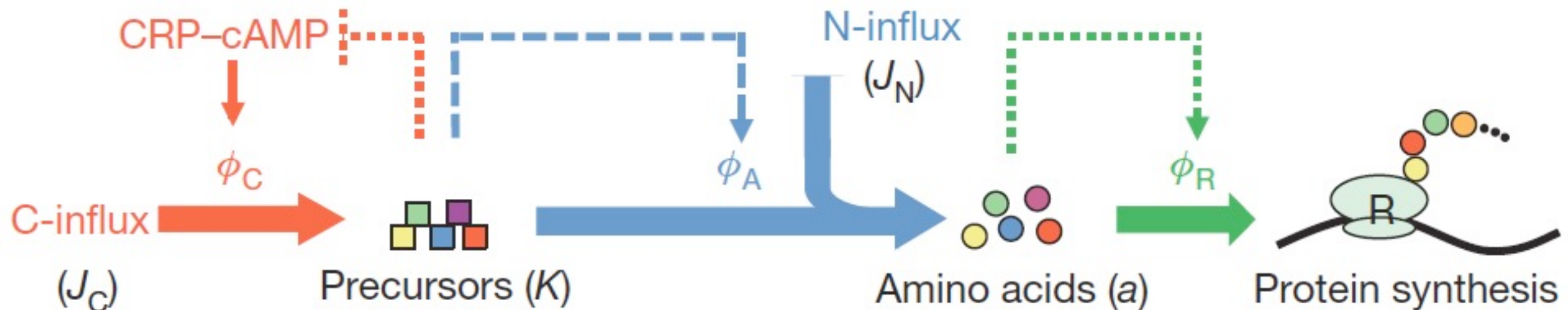
$F(x_1, x_2, \dots, x_n)$  must be impossibly complex?

# Bacterial growth laws offer a simple “coarse” description of metabolism



Terry Hwa

$$F = \frac{C}{1/x_C + 1/x_N + 1/x_T}$$



T. Hwa's model predicts epistasis  
between mutations that affect different  
cellular processes

$$F = \frac{C}{1/x_C + 1/x_N + 1/x_T}$$

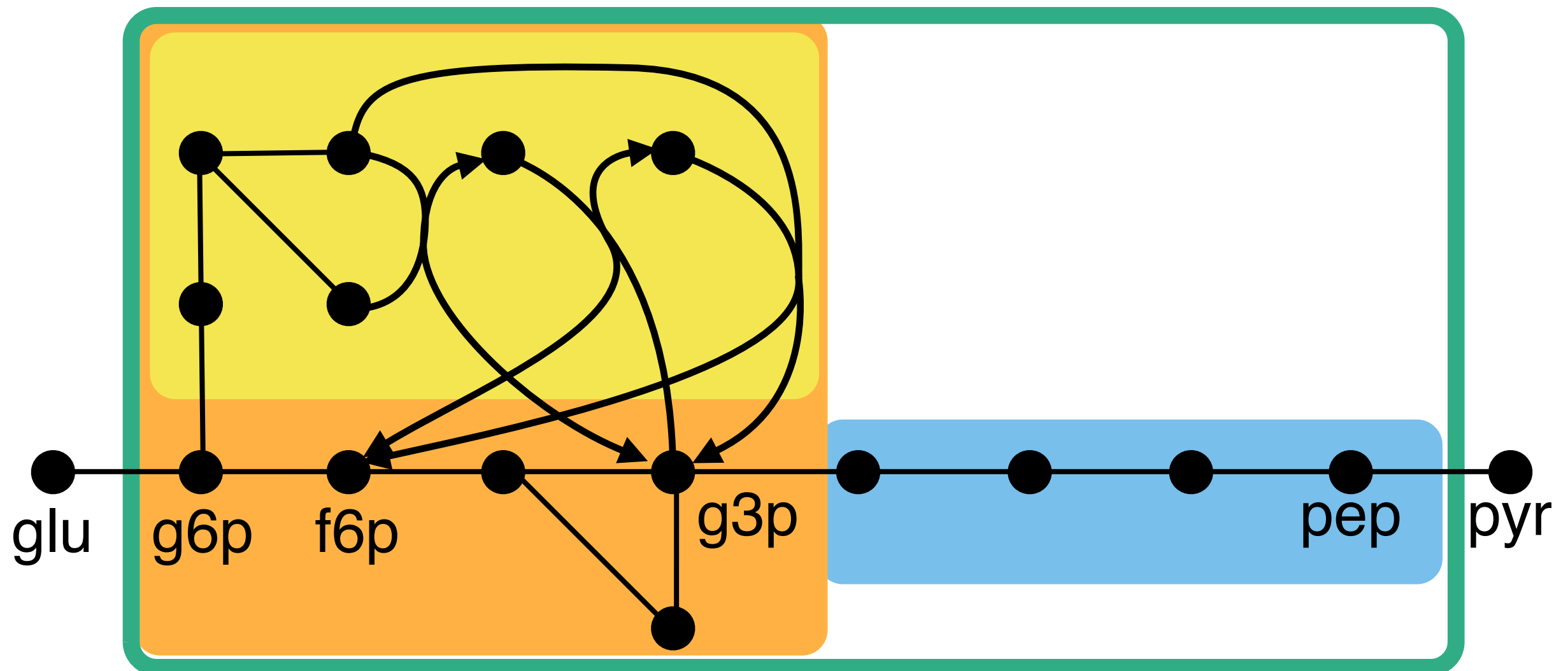
Mutation A reduces only  $x_C$

Mutation B reduces only  $x_N$

$$\varepsilon^{AB}F > 0$$

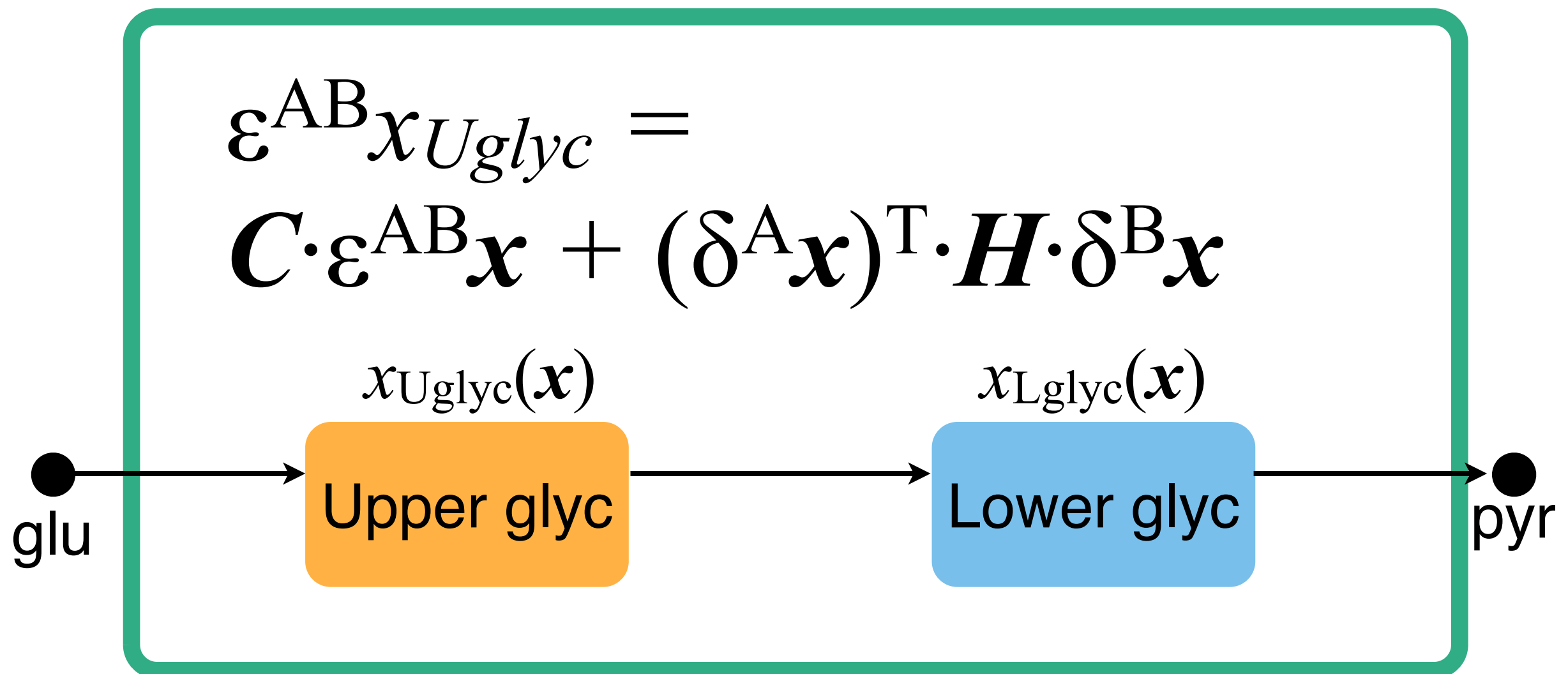
# What about epistasis between mutations that affect the same process?

1. Break down metabolic networks into pathways
2. Characterize pathways by effective parameters
3. Propagate epistasis



# What about epistasis between mutations that affect the same process?

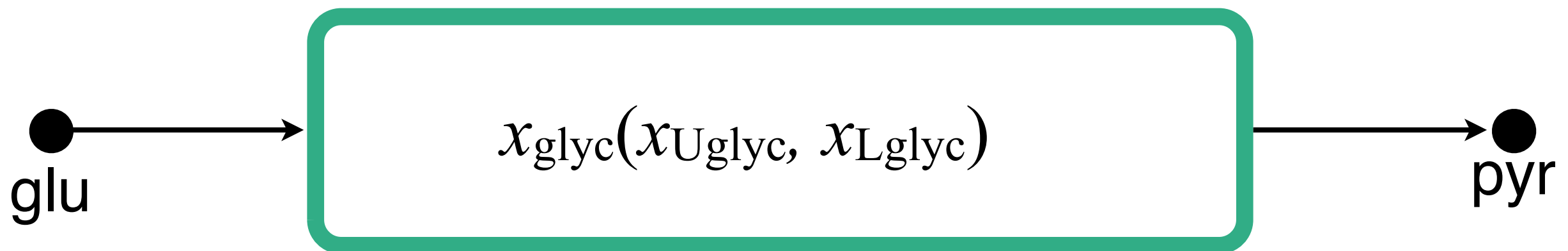
1. Break down metabolic networks into pathways
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# What about epistasis between mutations that affect the same process?

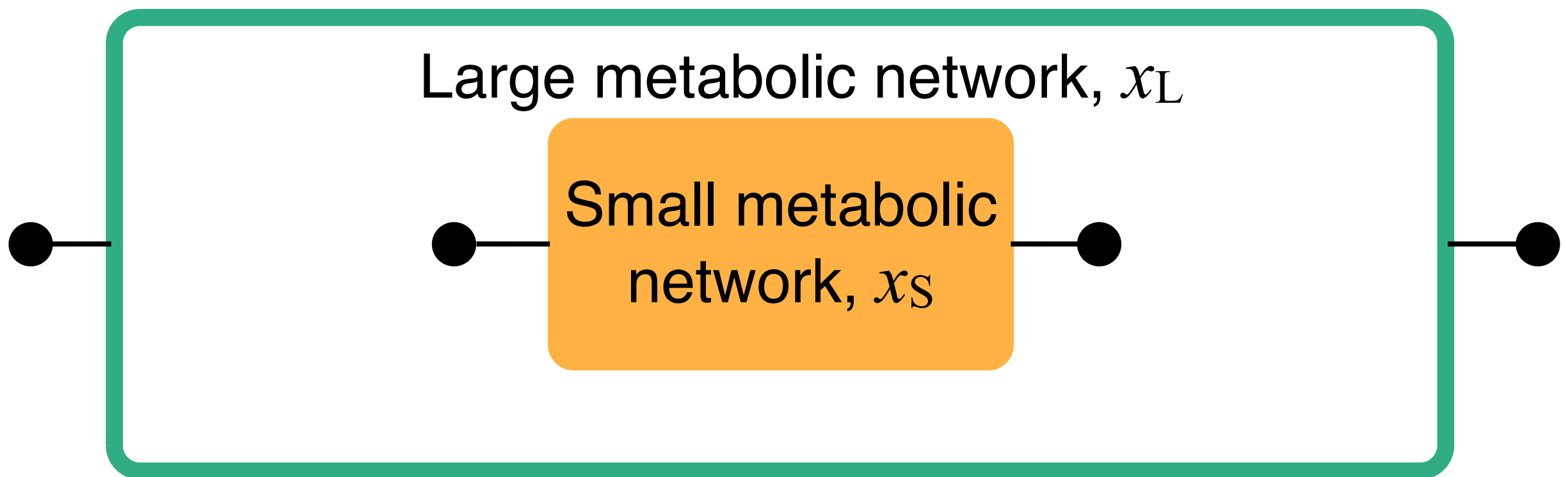
1. Break down metabolic networks into pathways
2. Characterize pathways by effective parameters
3. Propagate epistasis

$$\varepsilon^{\text{AB}}x_{\text{glyc}} = \mathbf{C} \cdot \varepsilon^{\text{AB}}\mathbf{x} + (\delta^{\text{A}}\mathbf{x})^{\text{T}} \cdot \mathbf{H} \cdot \delta^{\text{B}}\mathbf{x}$$





# Some properties of propagation of epistasis can be derived for metabolic networks with unsaturated reactions



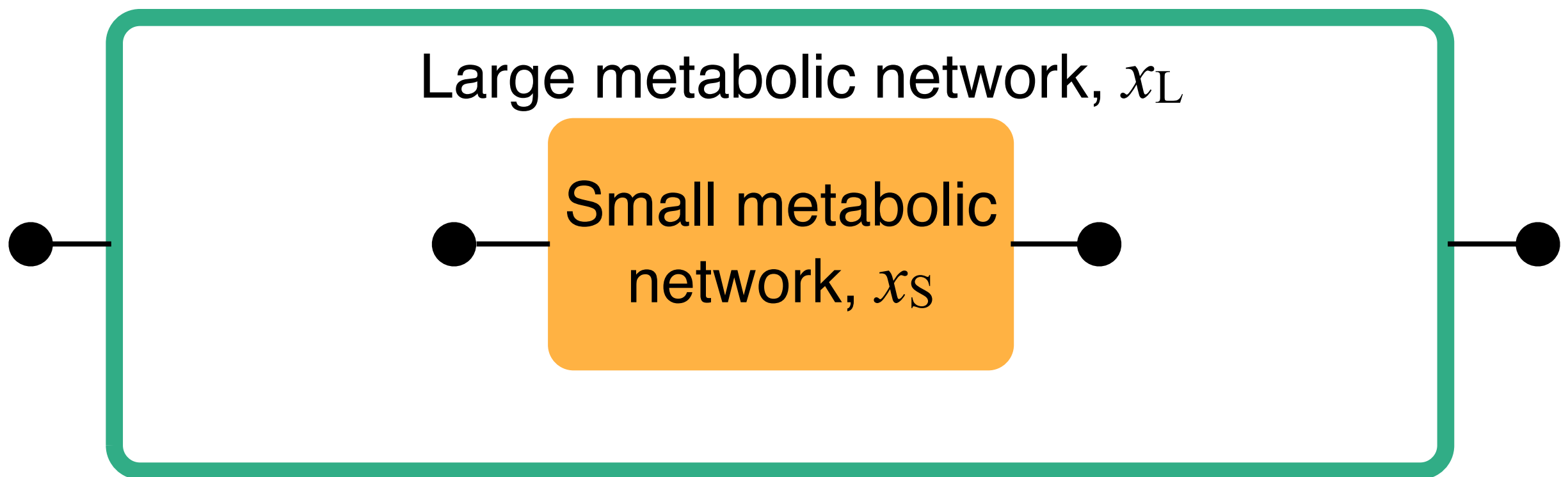
Mutations A and B reduce only  $x_S$  with  $\varepsilon^{AB}x_S \neq 0$

$$\varepsilon^{AB}x_L = C \cdot \varepsilon^{AB}x_S + \delta^A x_S \cdot H \cdot \delta^B x_S$$

$$C > 0$$

$$H < 0$$

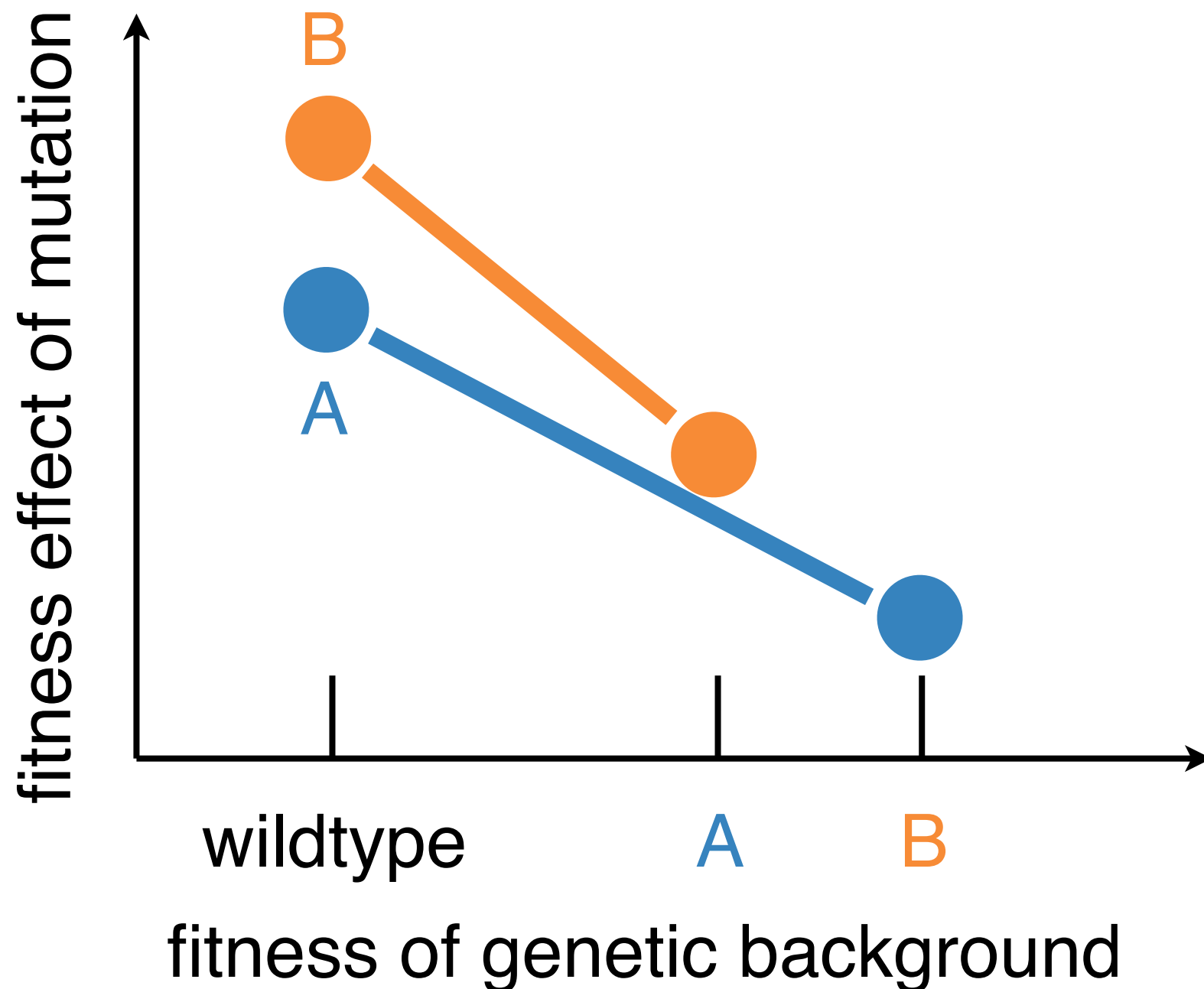
# Some properties of propagation of epistasis can be derived for metabolic networks with unsaturated reactions



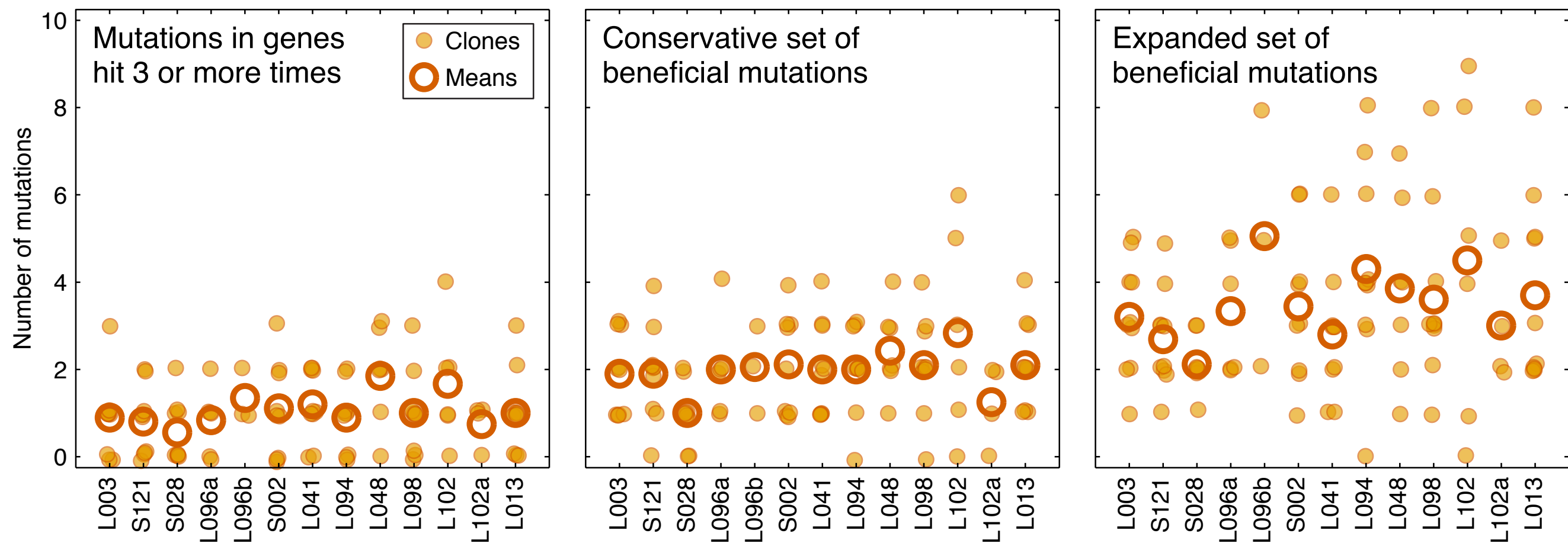
Mutations A and B reduce only  $x_S$  with  $\varepsilon^{AB}x_S \neq 0$

Negative epistasis tends to accumulate  
but positive epistasis does not

# Can propagation of epistasis explain prevalence of negative epistasis for fitness in evolution experiments?



# Number of mutations indistinguishable among Founders



Founders ordered by initial fitness