

# Experimental evolution on microbial fitness landscapes

Arjan de Visser

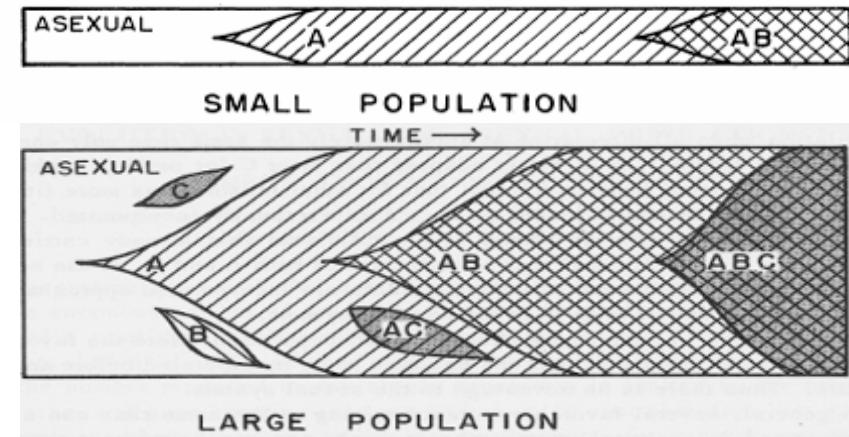
Laboratory of Genetics



# Factors determining course of evolution

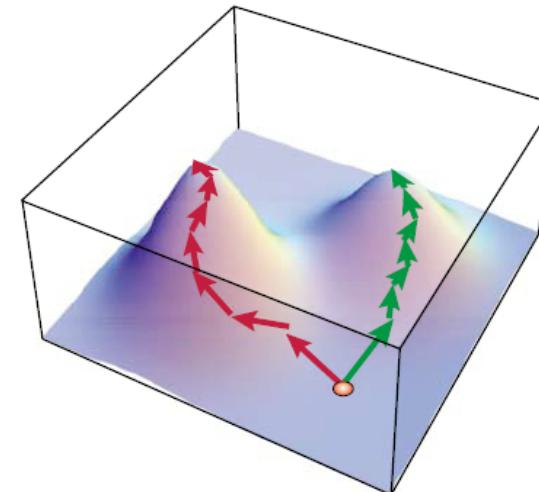
## 1. Population dynamics

- genetic variation
- population size



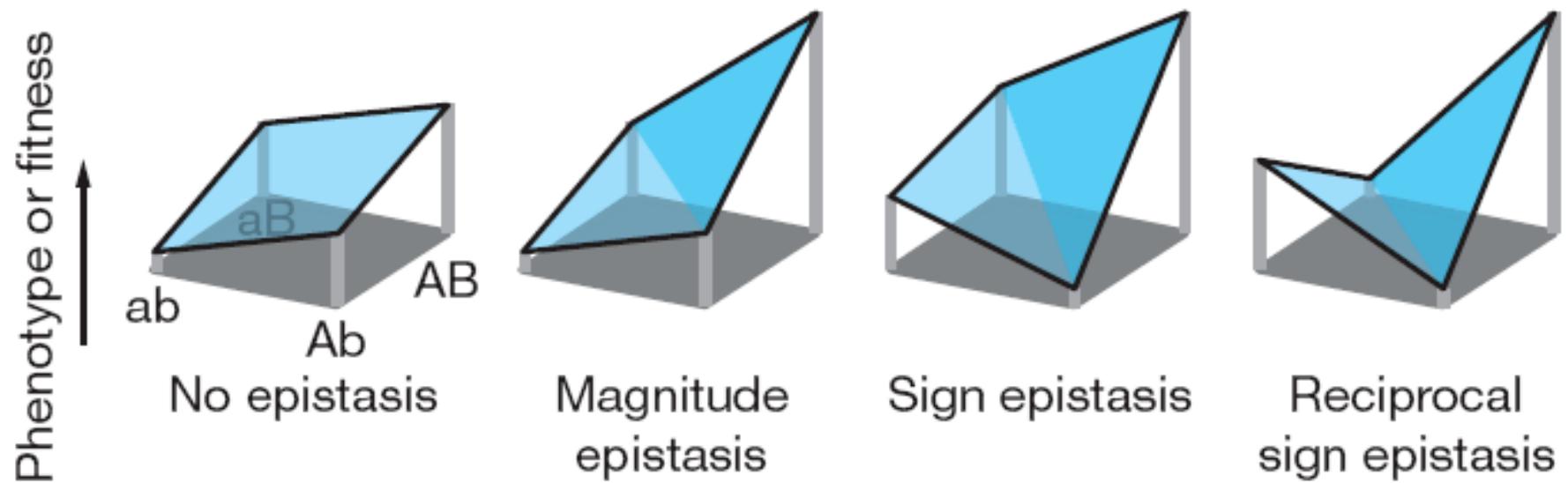
## 2. Fitness landscape

- environment
- genetic architecture



# Fitness landscape topography

- Ultimate cause: largely unknown
- Proximate cause: epistasis



(Poelwijk *et al.* 2007 *Nature*)

Little known about topography of real fitness landscapes

# Microbial experimental evolution



- Microbial populations evolve in laboratory for many generations in relatively short time
- Direct observation of evolution under controlled conditions
- Access to intermediates using ‘frozen fossil bed’
- Analyze fitness landscape via construction of mutants

# 3 studies of empirical fitness landscapes

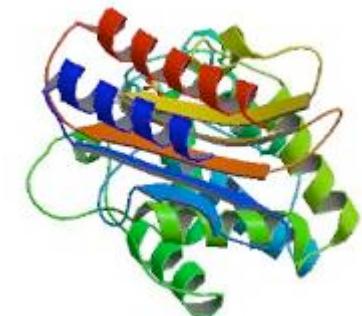
I. Landscape topography inferred from dynamics of adaptation in **bacteria**



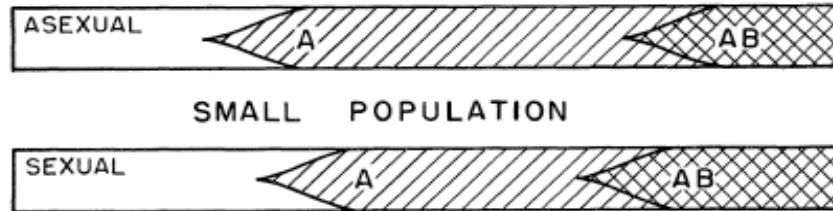
II. Systematic study of landscape involving marker mutations in a **fungus**



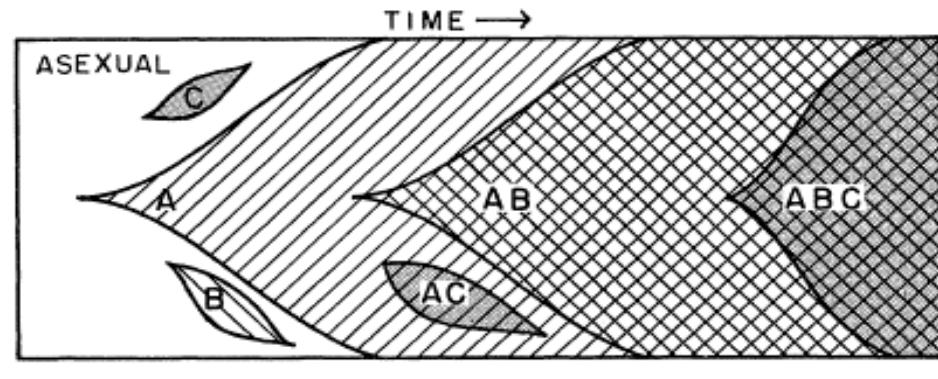
III. Molecular exploration of the landscape of an **enzyme**



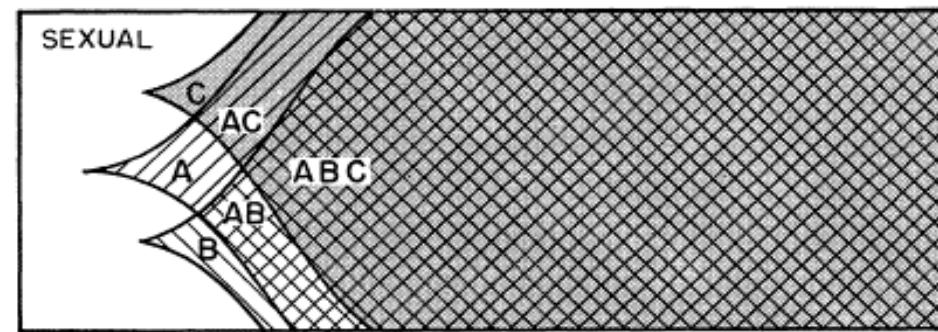
# I. Prelude: Adaptation in asexual populations



(after H.J. Muller 1932)



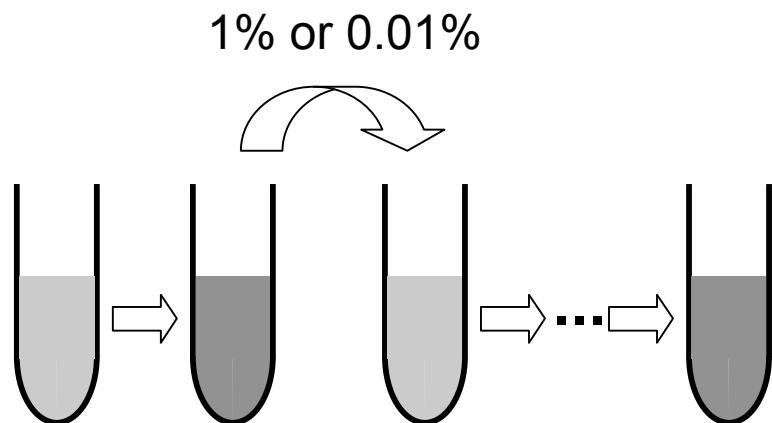
LARGE POPULATION



Consequences **clonal interference**:

- **Short term:** *slower substitution of larger-benefit mutations*
- **Long term:** diminishing returns from mutation supply on rate of adaptation

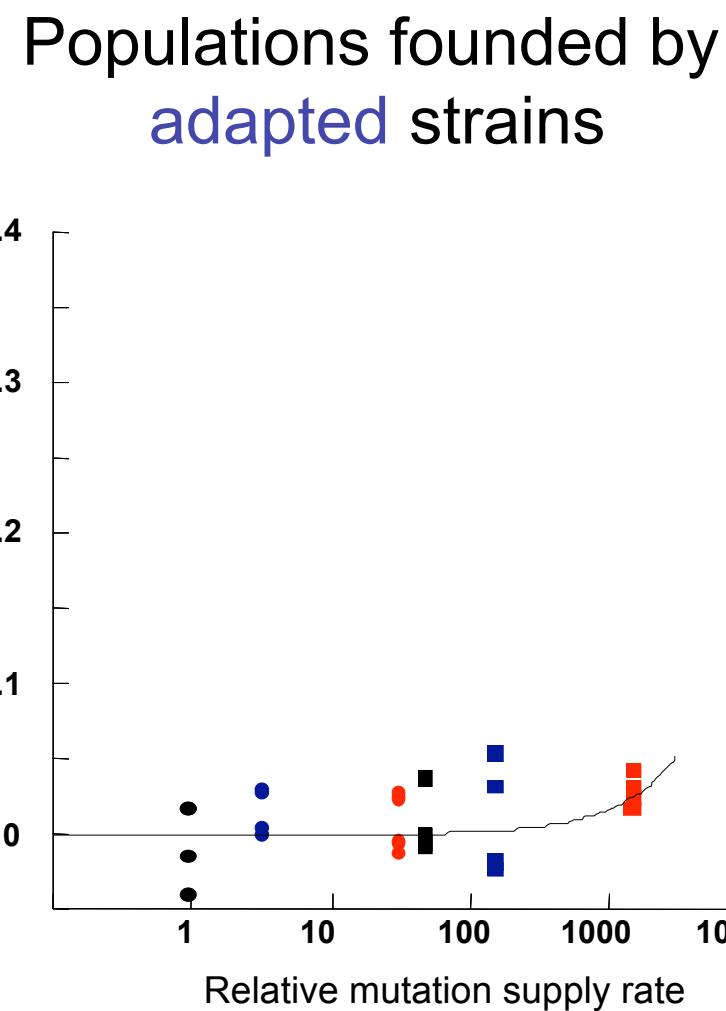
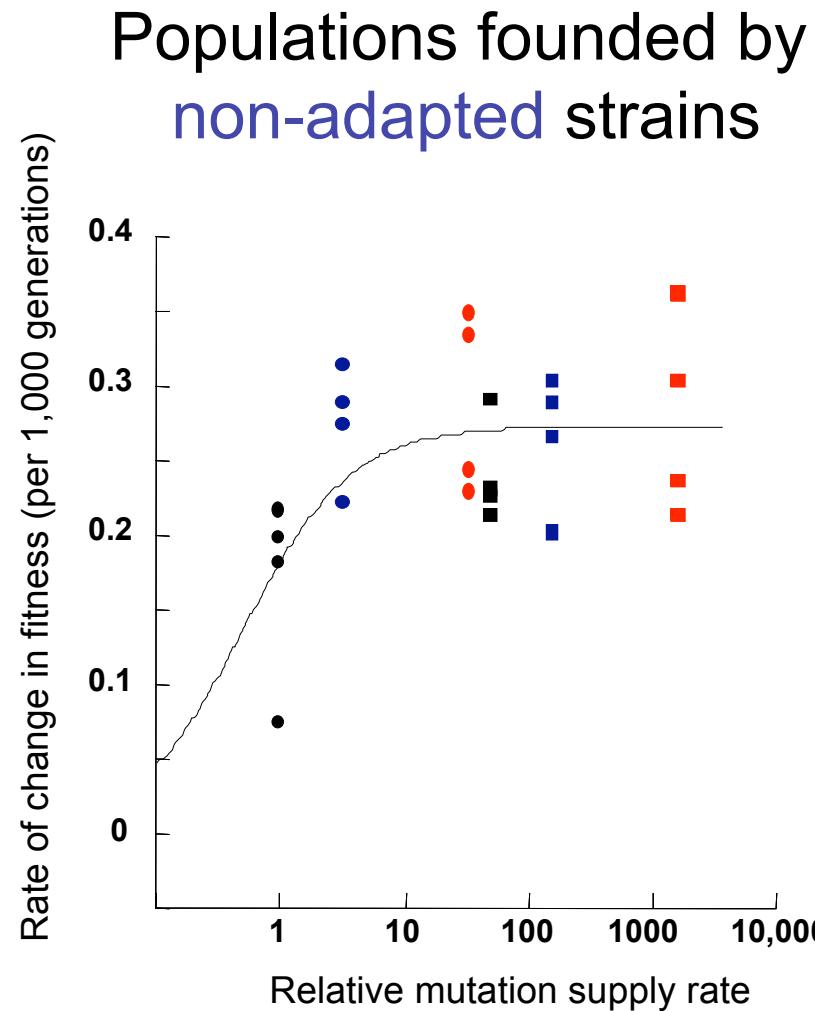
# I. Prelude: Test long-term effects CI in *E. coli*



## Manipulated variables

- Mutation rate:  
wt, *mutY*<sup>-</sup>, *mutS*<sup>-</sup>
- Population size:  
transfer cell number
- Fraction beneficial mutations:  
strain with low and strain with high level of adaptation

# I. Prelude: Adaptive benefit from increased mutation supply rate



(de Visser, Zeyl, Gerrish, Blanchard & Lenski 1999 *Science*)

# I. Fitness landscapes and population size

- So far, large populations adapt faster than small populations -- although this advantage may be small due to clonal interference
- However, relationship between rate of adaptation and population size may also depend on **fitness landscape**

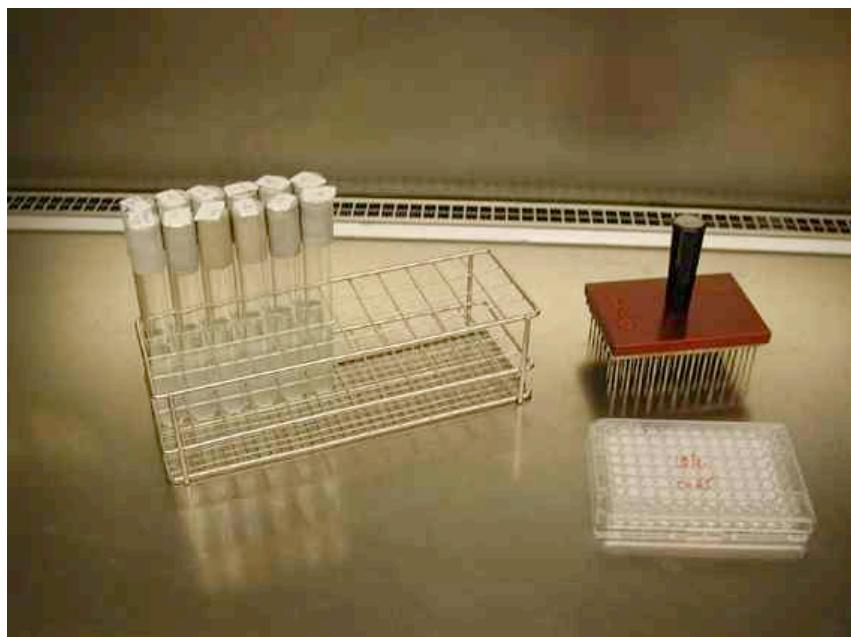


• **Smooth landscape:** *single peak* reached fastest by large populations taking big steps on single-best trajectory



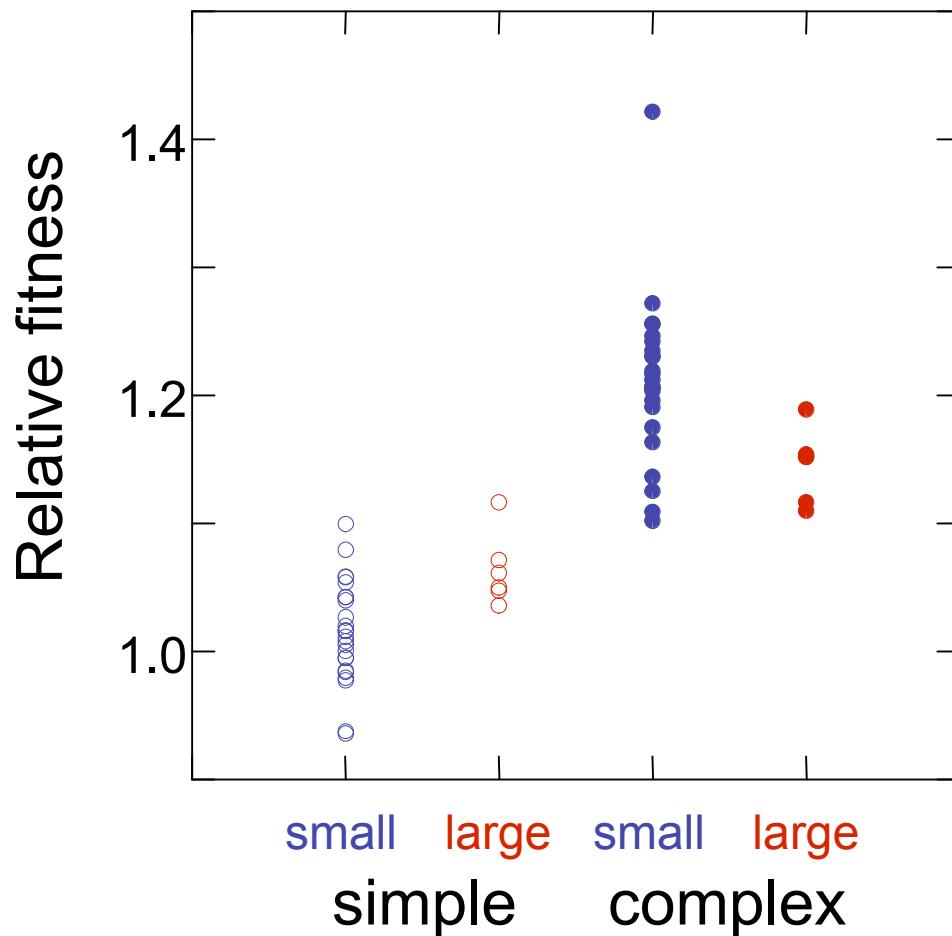
• **Rugged landscape:** *global peak* may be reached faster by smaller populations exploring different trajectories due to greater role drift

# I. Test of interaction between population size and environmental complexity



- 24 small ( $N_e \sim 5 \times 10^5$ ), 6 large *E. coli* populations ( $N_e \sim 2.5 \times 10^7$ )
- Simple (single C-source) and complex (multiple C-sources) nutrient environment (same  $N_e$ )
- 500 generations

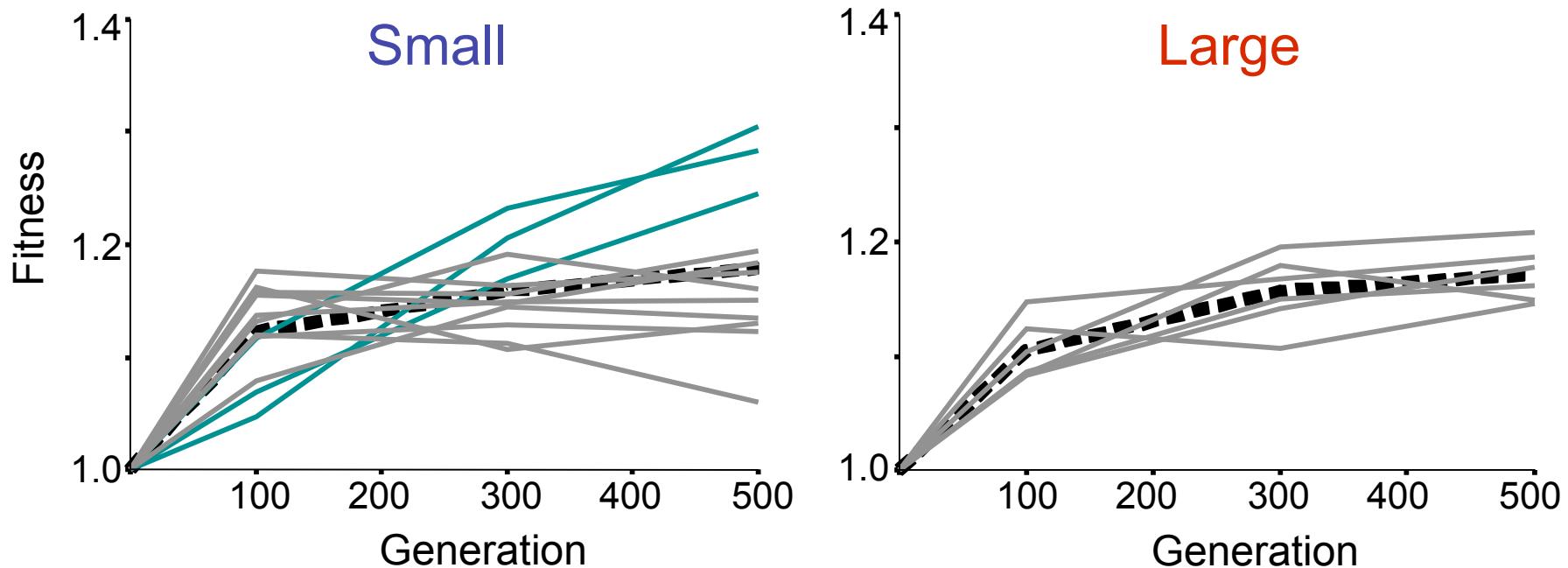
# I. Fitness after 500 generations



- Among-population variance only for small  $N$ : higher adaptive variability small  $N$
- $N \times$  environment interaction: adaptive advantage small  $N$  in complex environment

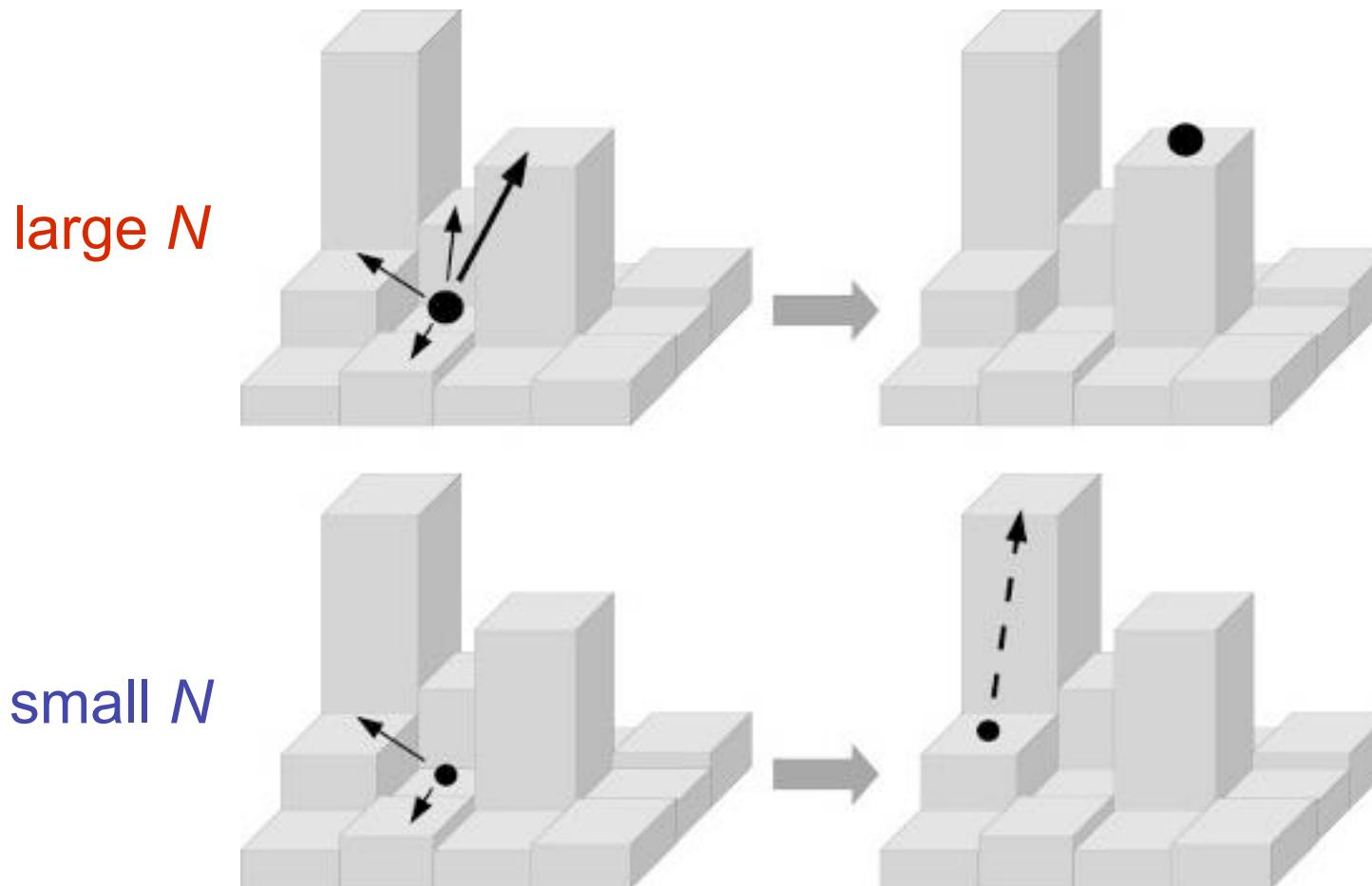
(Rozen, Habets, Handel & de Visser 2008 *PLoS One*)

## I. Fitness trajectories complex environment



- Greater slope variation in trajectories of small populations
- Highest final fitness reached by initially slowest small populations

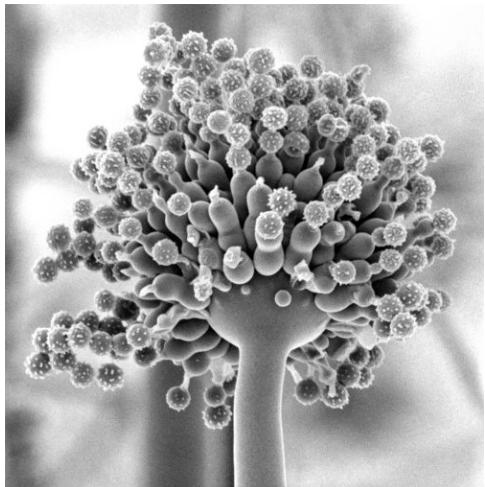
# I. The perils of greed



(Handel & Rozen 2009 *BMC Evol. Biol.*)

## II. Adaptation on empirical fungal fitness landscapes

with **Su-Chan Park** and **Joachim Krug** (University of Cologne)



*Aspergillus niger* (courtesy of Mycology Online & Maren Klich)

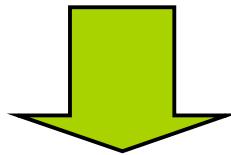
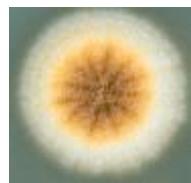
- Construct all possible  $2^8$  combinations of mutations at 8 loci, and measure fitness
- Simulate adaptation with and without recombination on this landscape

## II. *Aspergillus niger* strain construction



Heterozygous diploid strain

<i>fnw</i>	<i>arg</i> <sup>-</sup>	<i>pyr</i>	<i>leu</i> <sup>-</sup>	<i>phe</i> <sup>-</sup>	<i>lys</i> <sup>-</sup>	<i>oli</i> <sup>R</sup>	<i>crn</i> <sup>R</sup>
+	+	+	+	+	+	+	+
<i>olv</i>	+	+	+	+	+	+	+

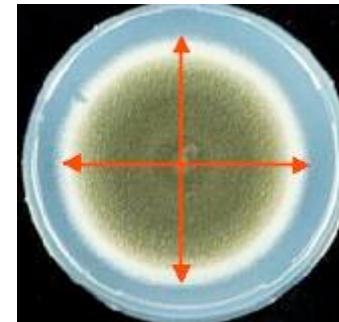
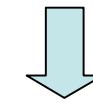


- $2^8 = 256$  possible haploid segregants carrying different combinations of mutations
- Among ~2,500 segregants tested, 186 found

Among 186 strains isolated,

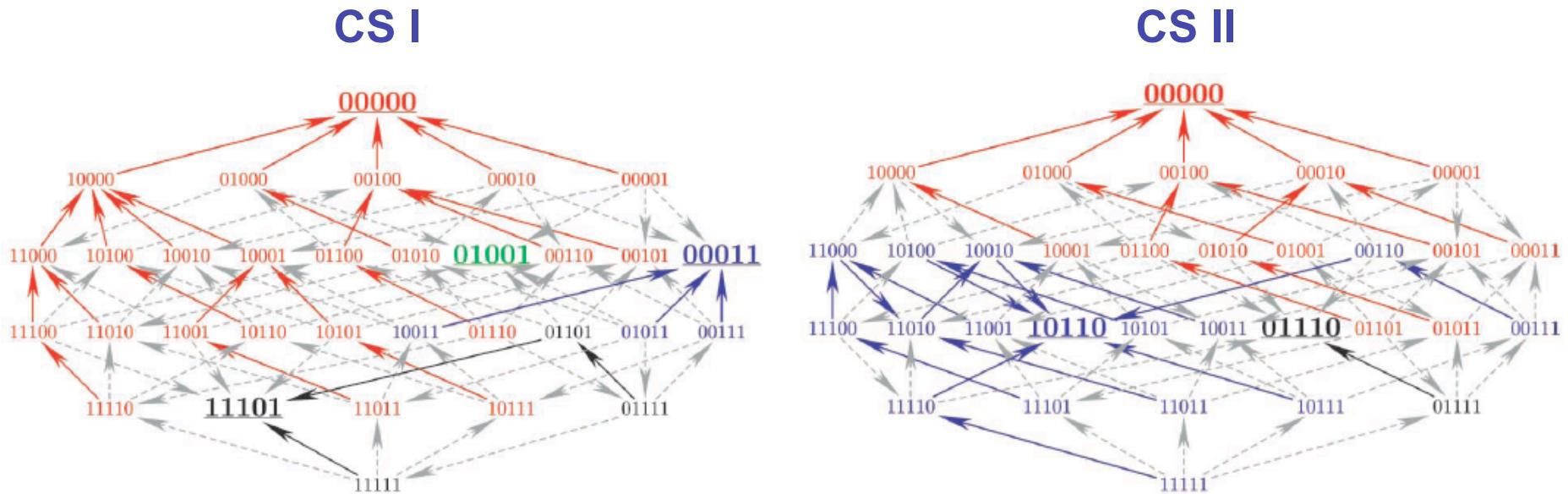
2 Complete Subsets (CS) of strains involving all  $2^5 = 32$  combinations of 5 individually deleterious mutations:

- CS I: *arg*<sup>-</sup>, *pyr*, *leu*<sup>-</sup>, *oli*<sup>R</sup>, *crn*<sup>R</sup>
- CS II: *arg*<sup>-</sup>, *pyr*, *leu*<sup>-</sup>, *phe*<sup>-</sup>, *oli*<sup>R</sup>



**Fitness:**  
growth rate on supplemented medium

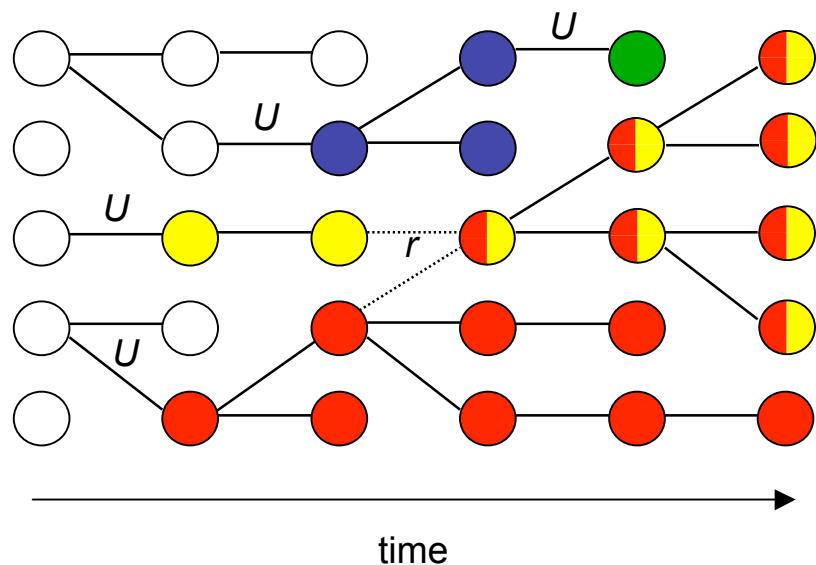
## II. Arrow plots of *A. niger* landscapes



(de Visser, Park & Krug 2009 *Am. Nat.*)

- **Ruggedness:** several local maxima (underlined)
- Of 120 possible pathways from **11111** to **00000**, only ~20% are accessible by natural selection
- Basins of attraction under ‘greedy walk’ dynamics shown in color

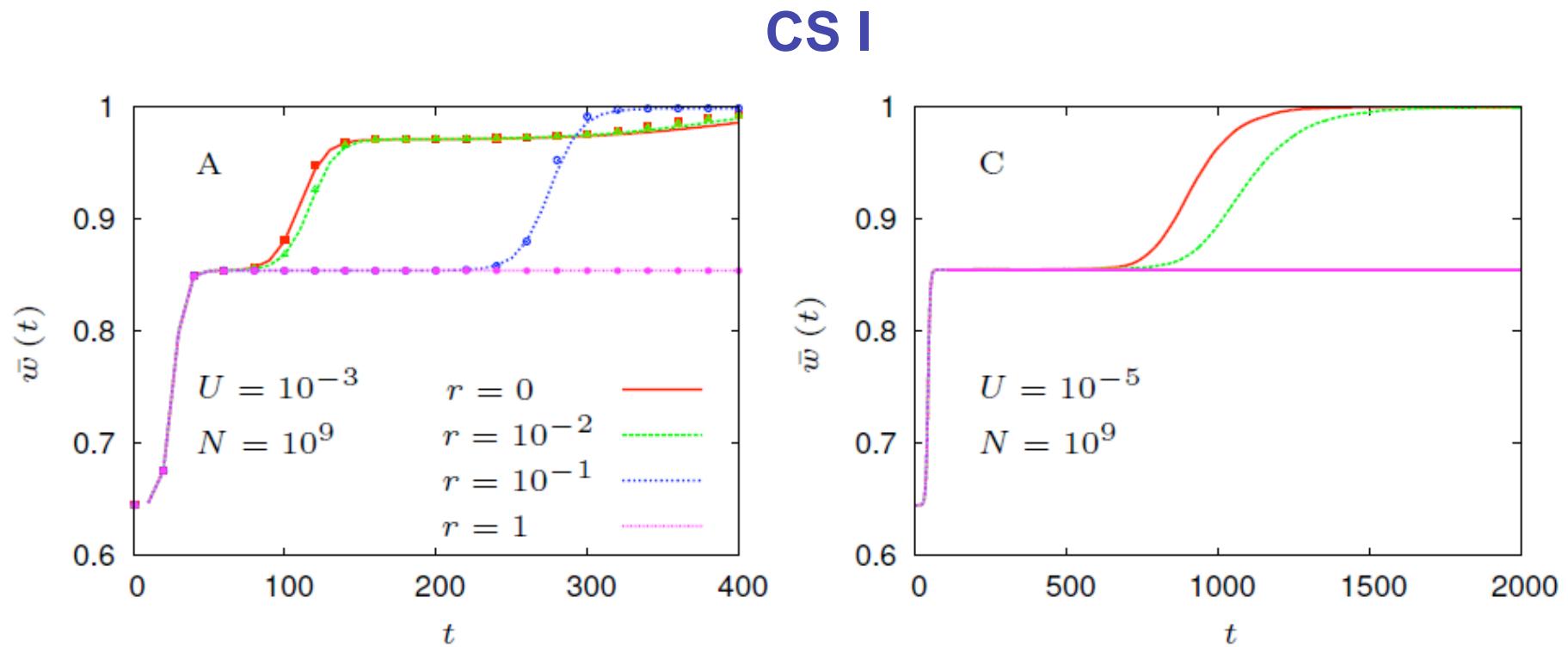
## II. Simulations using Wright-Fisher model



- Constant haploid population size  $N$
- Reproduction with probability proportional to genotype's fitness  $w$
- Mutation with probability  $U$  per genome ( $L = 5$ ) per generation, changes into neighboring genotype
- Free recombination with probability  $r$  per pair of individuals per generation

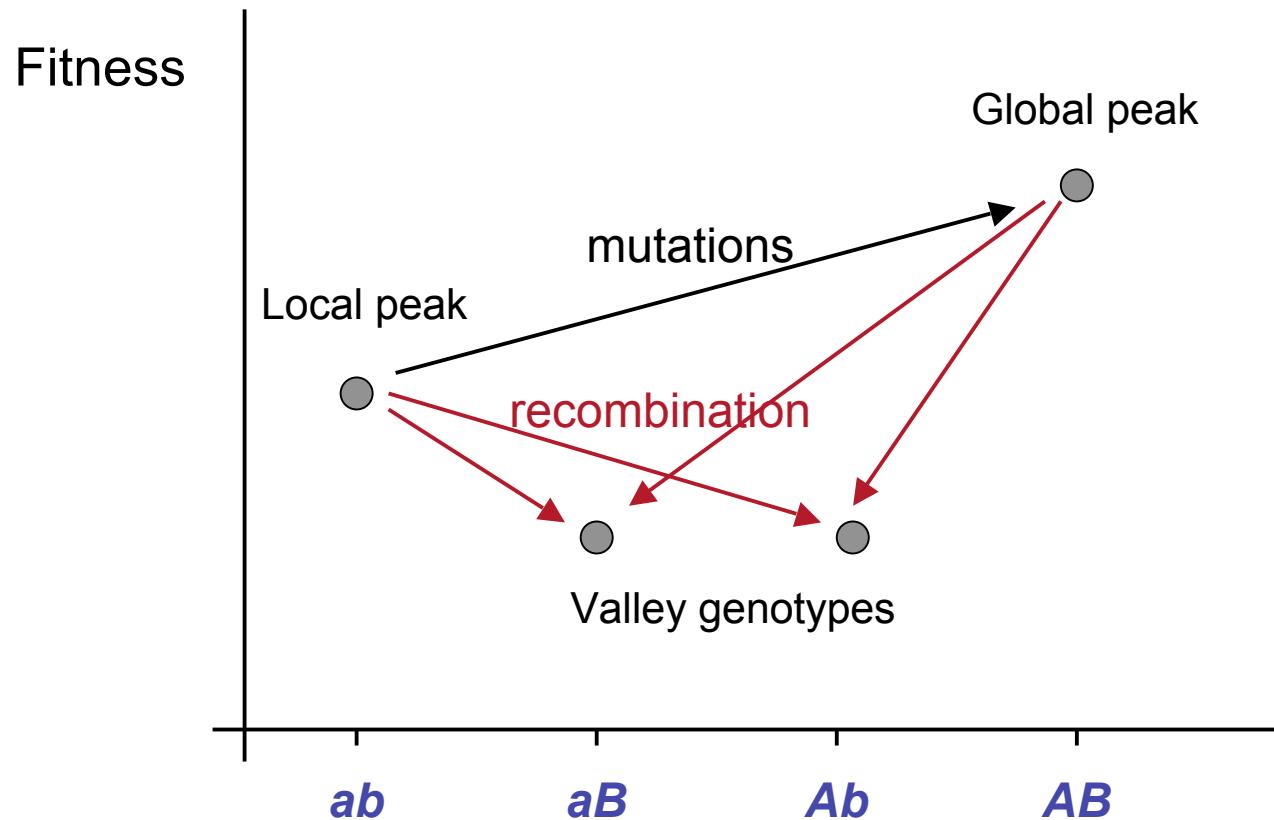
Does sex facilitate adaptation on the *A. niger* landscapes?

## II. Adaptation on *A. niger* landscapes



- Sex generally slows down adaptation
- Only for intermediate  $U$  and  $r$  and on one landscape (CS I), sex accelerates escape from local maximum

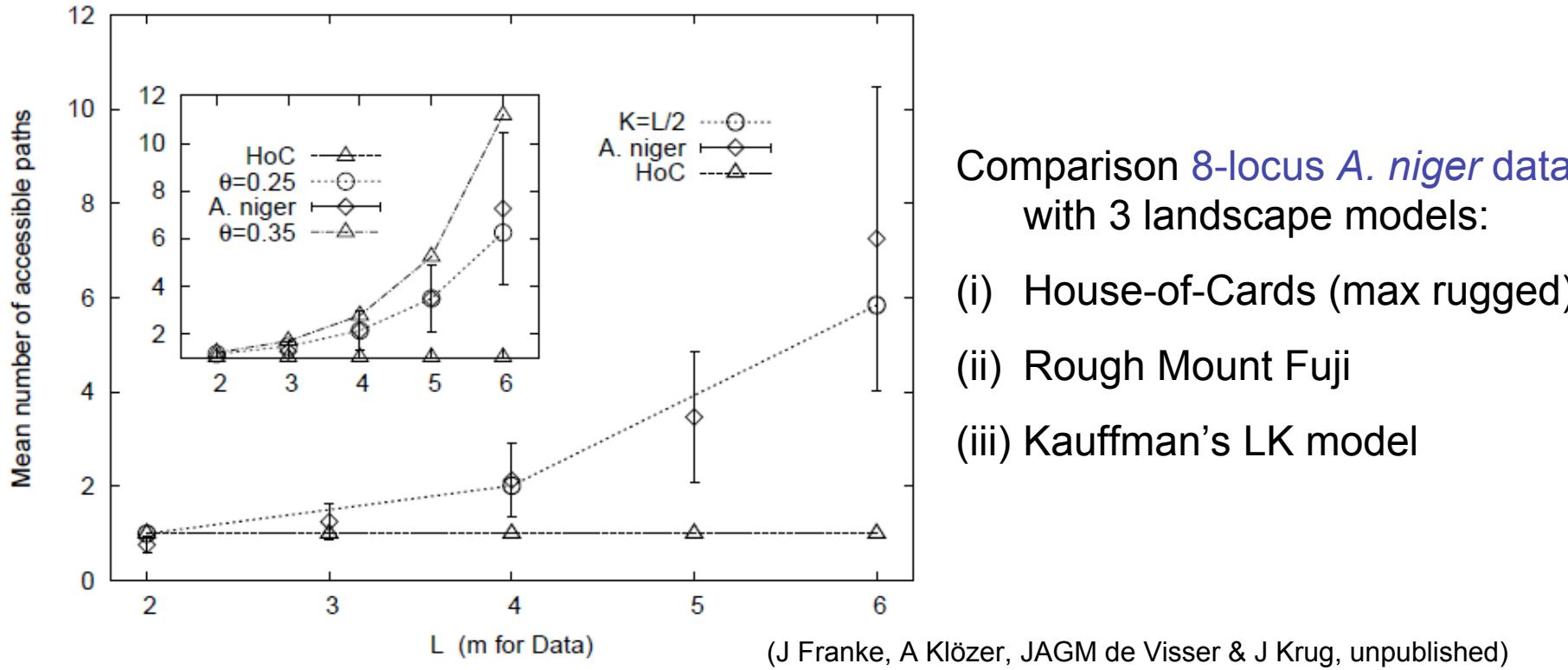
## II. Sex breaks down escape genotypes



How reconcile these results with wide-spread sex?

- *A. niger* data not representative
- accessibility global peak may increase with number of loci

## II. Accessibility increases with $L$

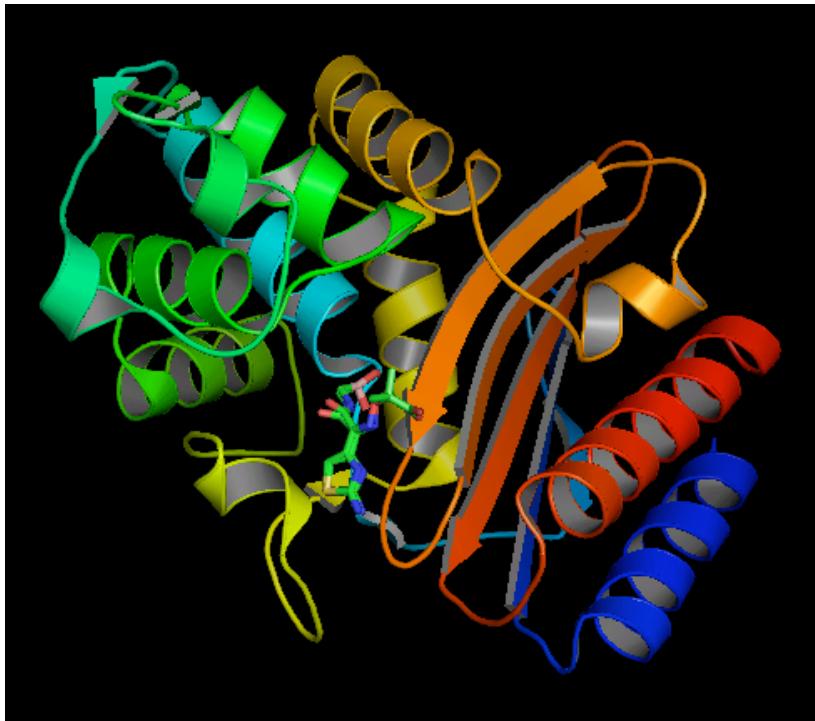


- Except for HoC model, accessibility increases with  $L$ : decreased accessibility per pathway overwhelmed by (factorial) increase in path number
- *A. niger* landscape fits models with intermediate ruggedness

### III. Exploring the fitness landscape of an antibiotic resistance enzyme

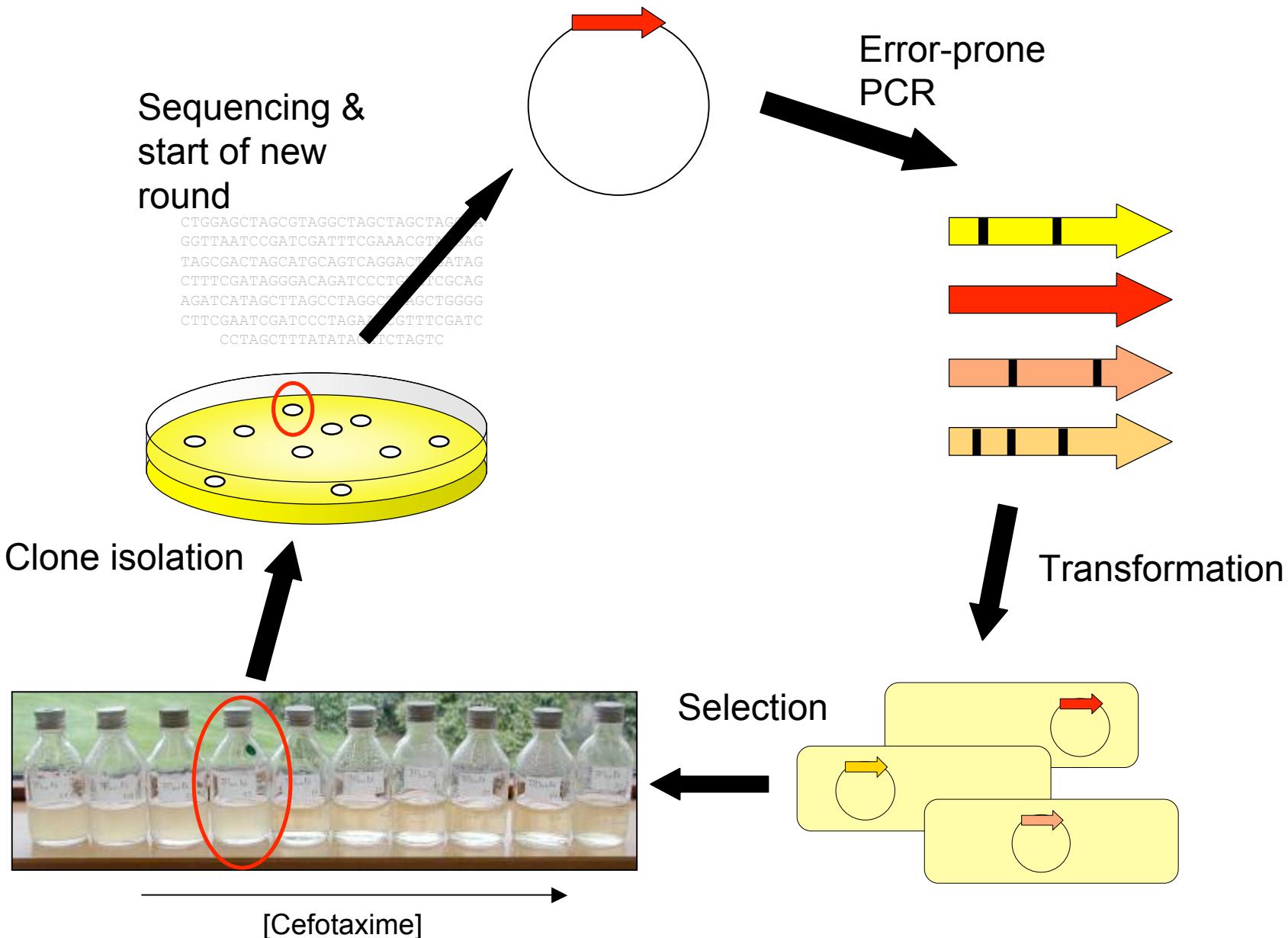
work by **Merijn Salverda**

#### TEM-1 $\beta$ -lactamase

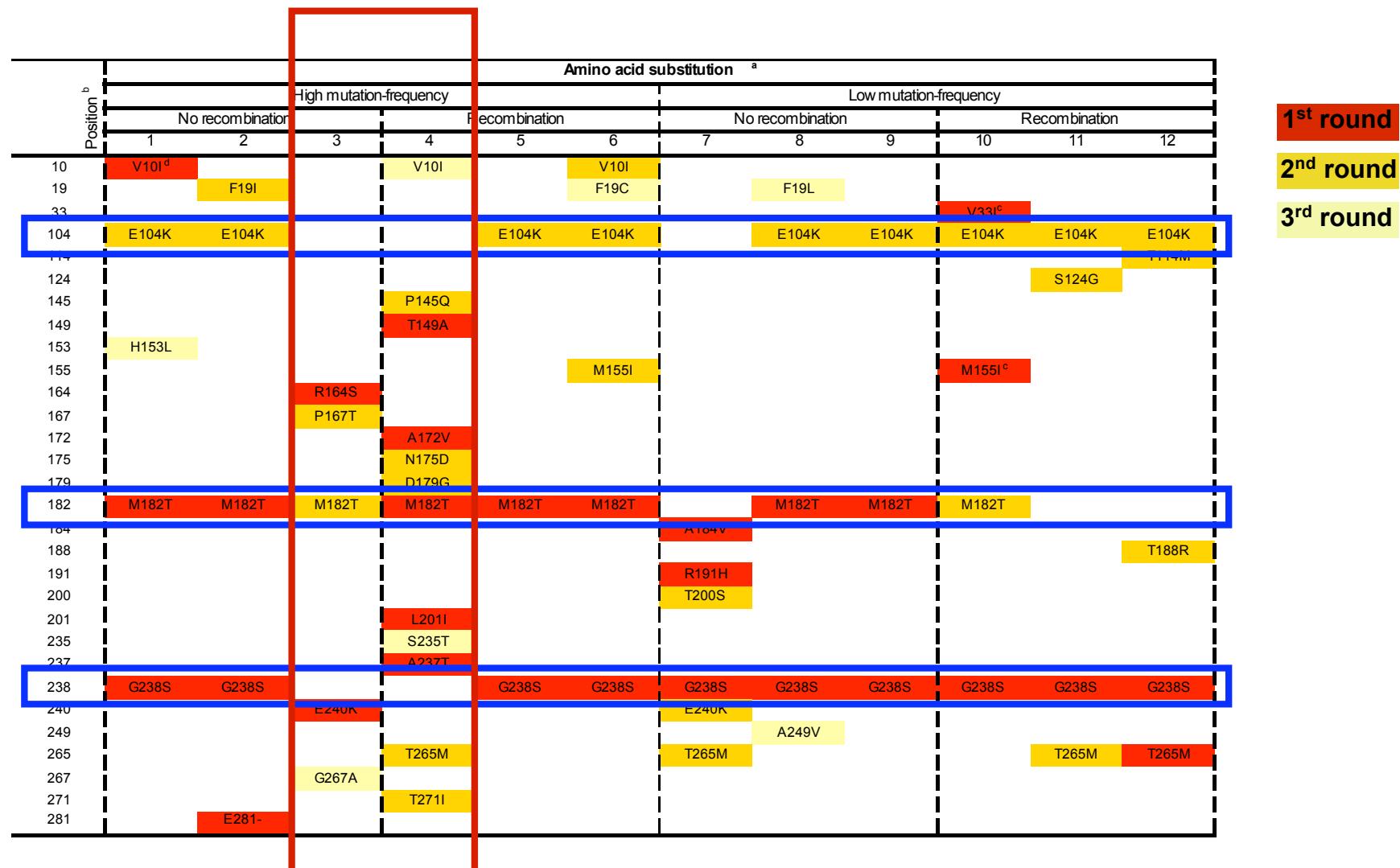


- Breaks down  $\beta$ -lactam antibiotics (penicillins, cephalosporins)
- Low resistance to cefotaxime (Ctx)
- Five point mutations can jointly increase Ctx-resistance ~100,000-fold:  
g4205a / A42G / E104K / M182T / G238S  
(Stemmer 1994, Hall 2002, Weinreich *et al.* 2006)

### III. *In vitro* evolution



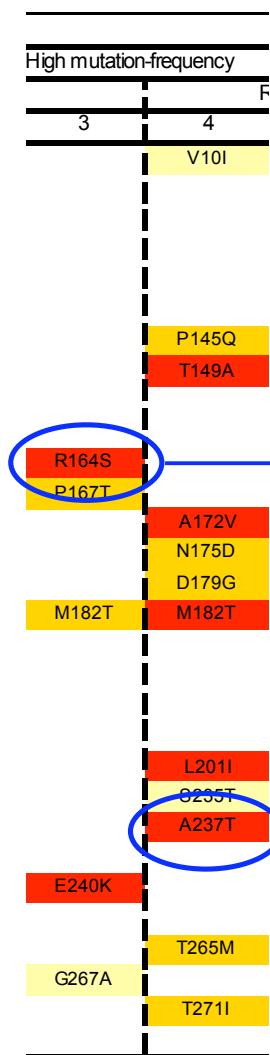
### III. Amino acid substitutions 1<sup>st</sup> experiment



**A42G/ E104K/ M182T/ G238S**

(Salverda, Dellus, Gorter, Debets, van der Oost, Hoekstra, Tawfik & de Visser 2011 PLoS Genet)

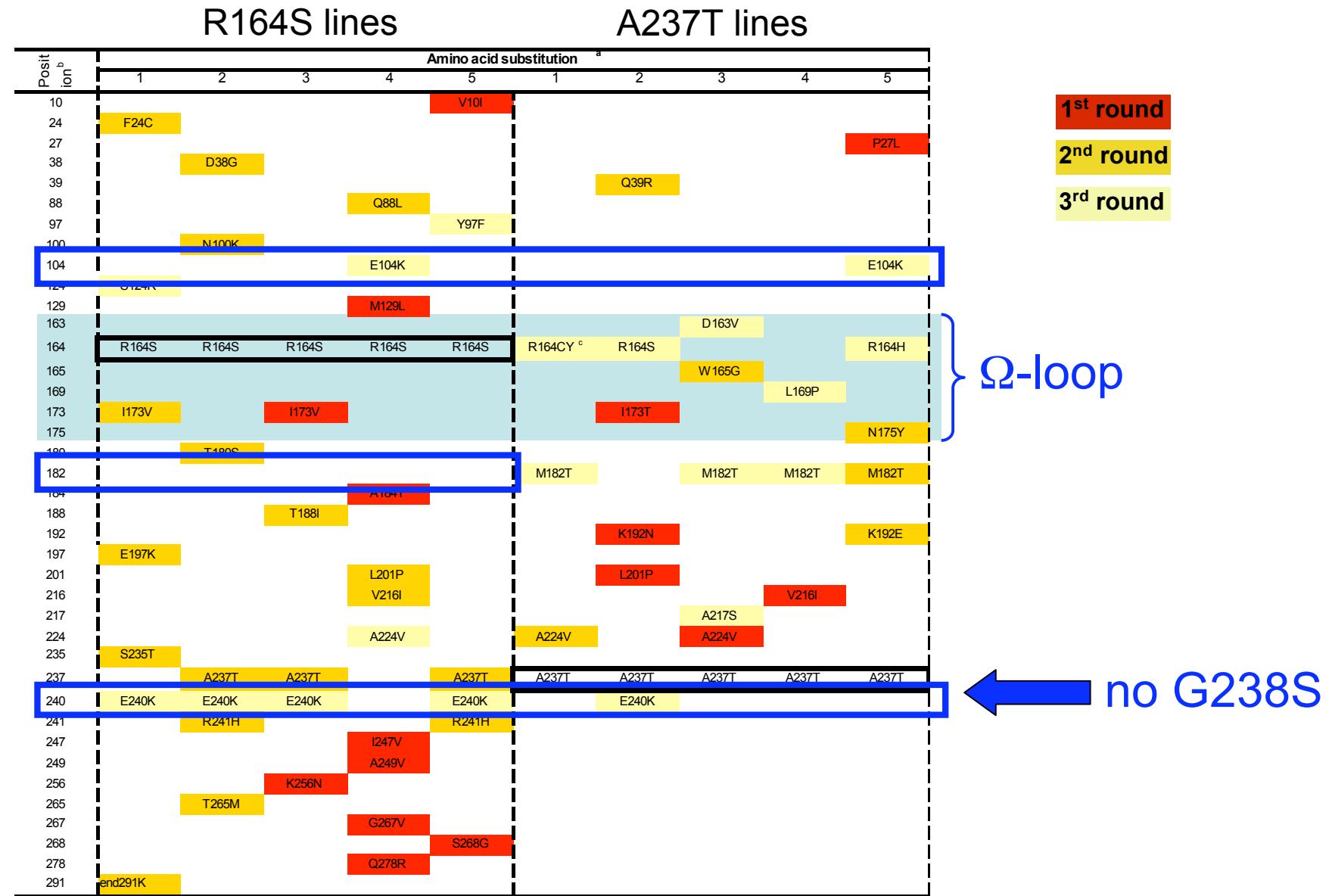
### III. Candidate early epistatic mutations in two lines lacking large-effect mutation G238S



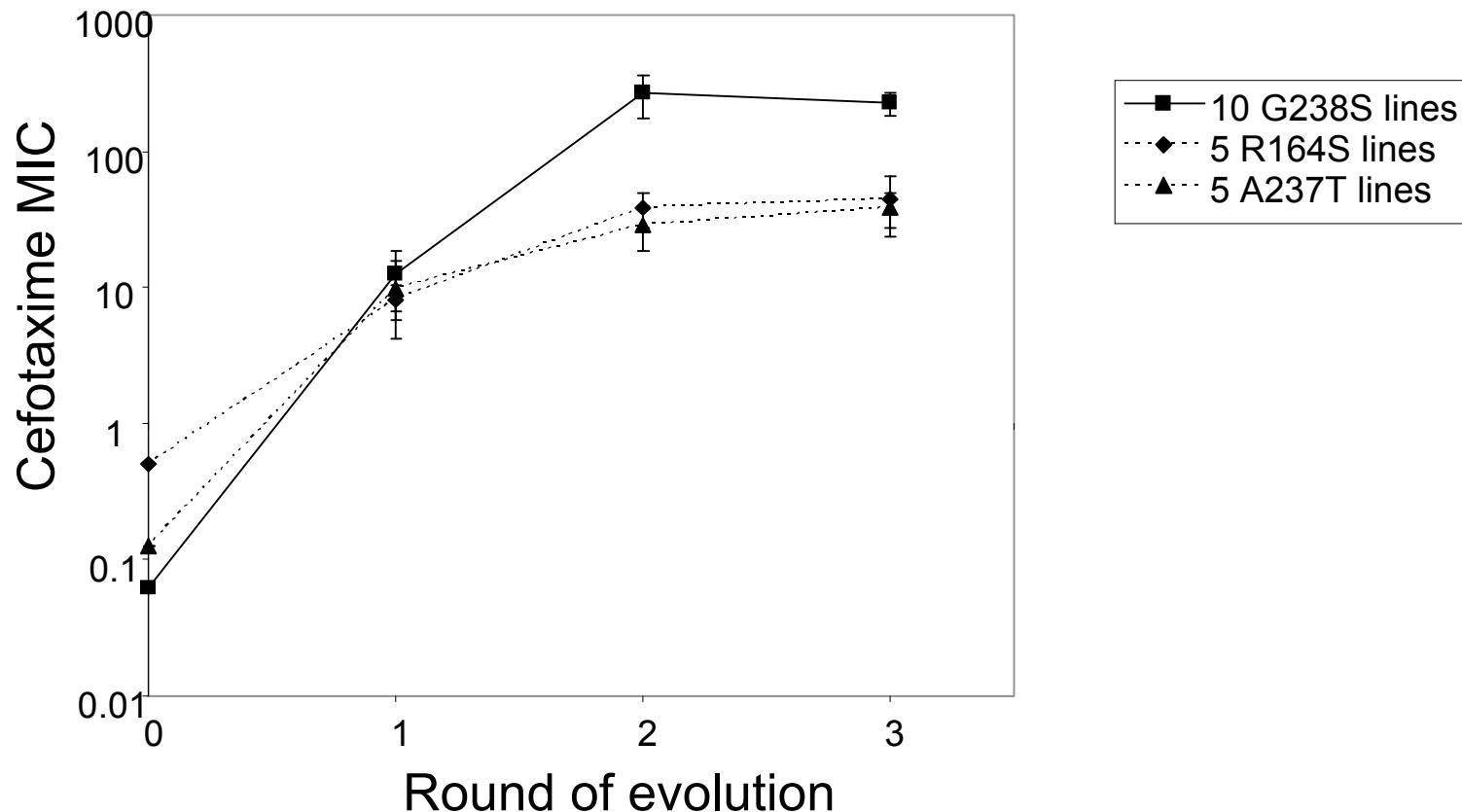
Increases Ctx-resistance and has negative epistatic interaction with G238S (Giakkoupi *et al.* 2000)

Frequently found in clinical isolates, never in combination with G238S

### III. Amino acid substitutions 2<sup>nd</sup> experiment



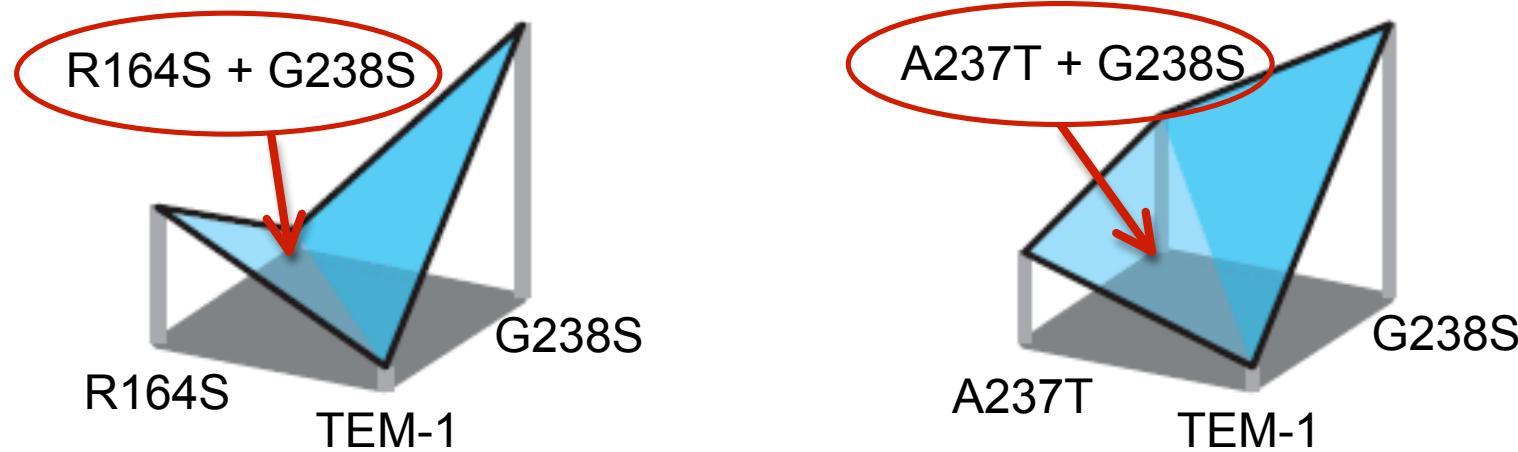
### III. Alternative pathways approach lower adaptive peaks



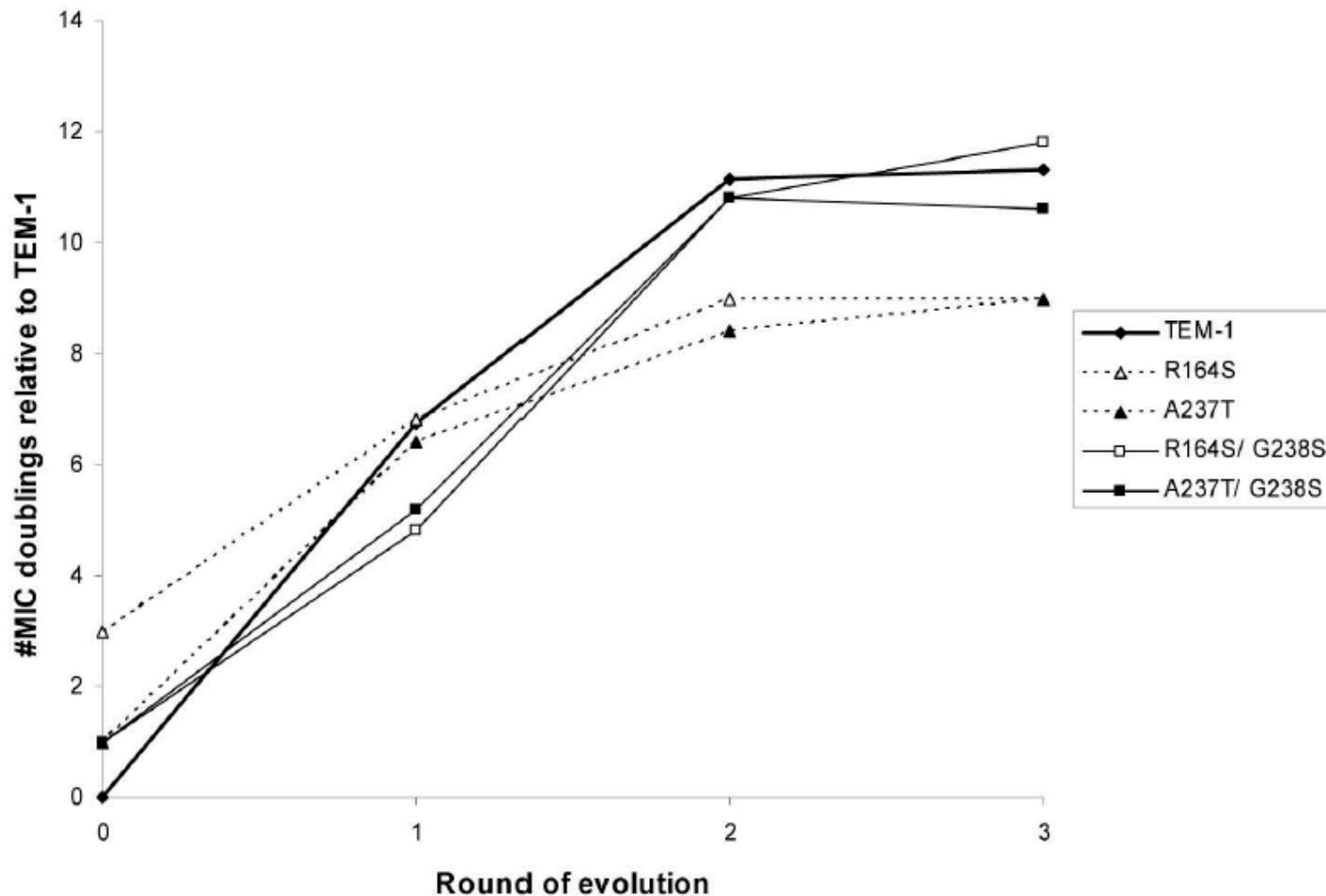
- Fitness landscape TEM-1 for  $\text{ctx}^R$  contains multiple adaptive peaks
- Initial mutations can determine which peak is approached

### III. Constraints from epistatic pairs

- Do pairs of negatively interacting mutations cause adaptive constraints: is only solution **reversion**, or can **novel mutations** neutralize negative interaction?



### III. MIC trajectories 3<sup>rd</sup> experiment



### III. Amino acid substitutions 3<sup>rd</sup> experiment

G238S/R164S lines      G238S/A237T lines

position	L1	L2	L3	L4	L5	IL1	L2	L3	L4	L5
4						S4N				
7			H7Q							
16	F16L									
17		A17S								
21	L21H									
24							P27S	F24S		
27										
38			D38V							
47	I47V									
56		I56T								
60		F60Y								
92		G92D					G92D			
104	E104V	E104K	E104Q	E104K						
115						D115E				
120										R120K
124										S124N
153			H153R					H153R		
164	S164R	S164R	S164R	S164R	S164R		R164G			
174							P174R			
182	M182T		M182T			M182T	M182T			
184	A184V									
188		T188K								
192	K192N									
197		E197K								
198			L198V				L198I			
208	I208K									
224							A224V	A224V		
235							S235T			
237						T237A	A237T	A237T	A237T	A237T
238	G238S									
240	E240K									
268							S268G			
275		R275Q						R275Q	R275L	
289			H289Y							
MIC1	2	4	1	2	1	4	4	4	1	2
MIC2	64	64	128	128	256	512	32	128	64	64
MIC3	512	64	128	512	256	512	32	128	64	64

1<sup>st</sup> round

2<sup>nd</sup> round

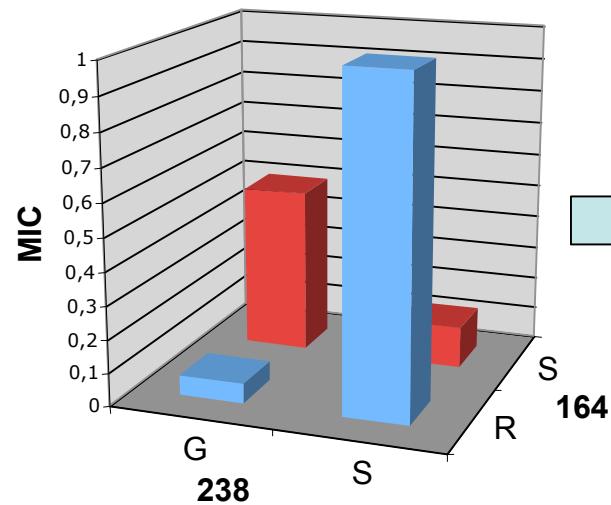
3<sup>rd</sup> round

- Adaptive constraints from sign epistasis can be alleviated both by reversion and novel mutations

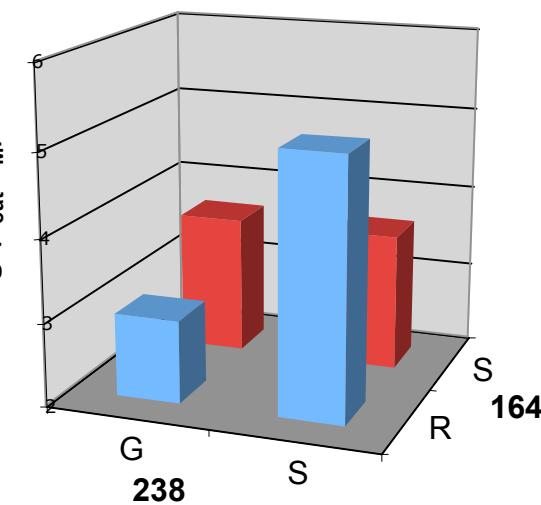
- Strength of constraint varies and correlates with strength epistasis

### III. Biochemical basis epistatic interaction

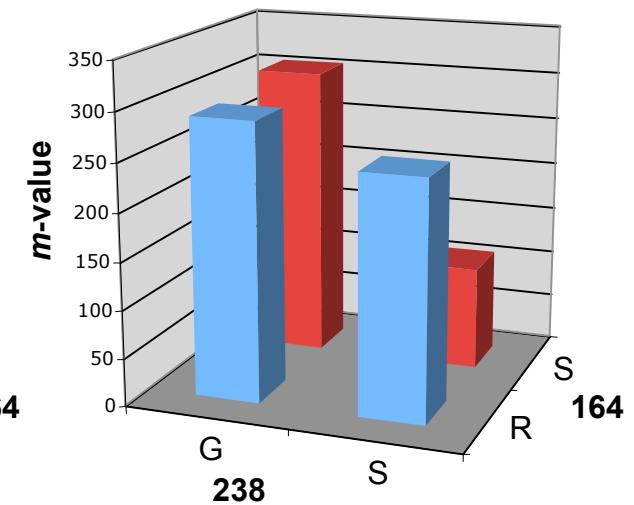
Cefotaxime resistance



Catalytic efficiency



Enzyme dosage



- Sign epistasis between beneficial mutations G238S and R164S results from combined epistatic effects at level of **enzyme activity** and **dosage**

# Summary and outlook

- Fitness landscapes can be studied empirically, either by studying dynamics and repeatability of phenotypic and molecular changes *or* by constructing mutants carrying all possible mutations of sets of mutations
- These studies increasingly show that real fitness landscapes are substantially rugged
- Future questions:
  - How does topography affect evolutionary predictability?
  - What are the evolutionary causes of different topographies?

# Acknowledgments

## **1<sup>st</sup> study**

- Danny Rozen
- Michelle Habets
- Cliff Zeyl
- Phil Gerrish
- Jeff Blanchard
- Rich Lenski
- Andreas Handel

## **2<sup>nd</sup> study**

- Joachim Krug
- Su-Chan Park
- Jasper Franke
- Andreas Klözer

## **3<sup>rd</sup> study**

- Merijn Salverda
- Eynat Dellus
- Florien Gorter
- John van der Oost
- Dan Tawfik
- Fons Debets
- Rolf Hoekstra