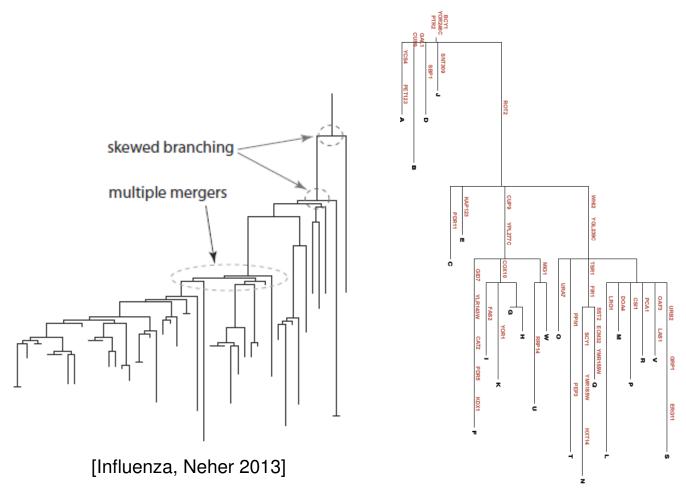
Genetic Diversity When Selection is Pervasive



Michael Desai, Harvard University

Genetic Diversity in the Interference Selection Limit



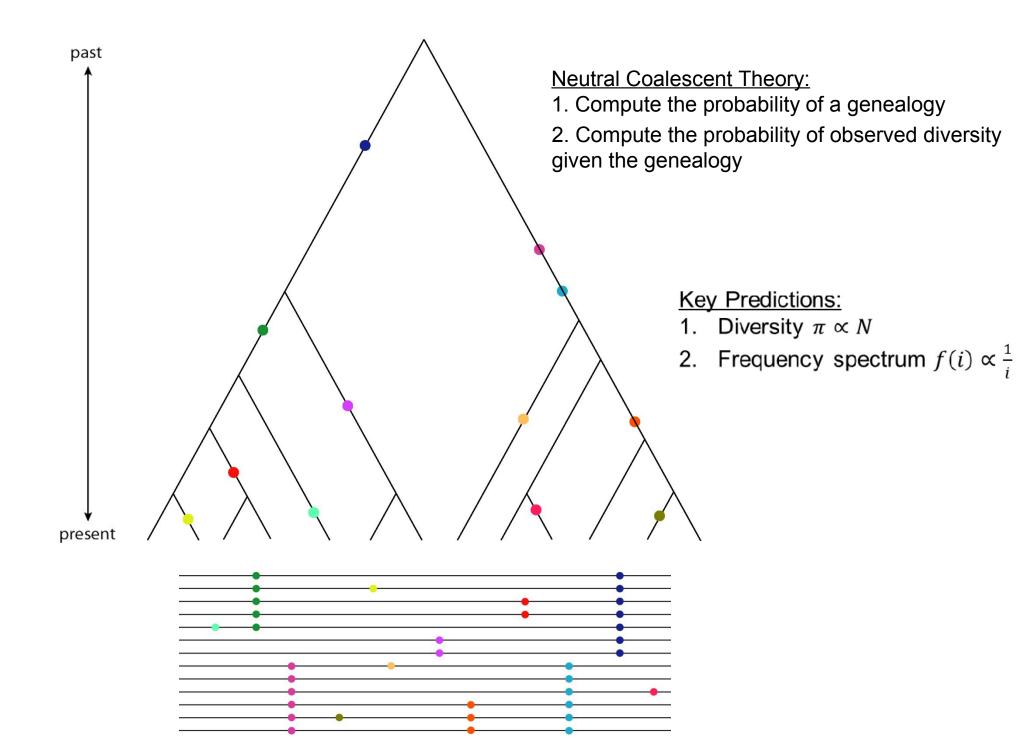


[Experimental Yeast Populations]

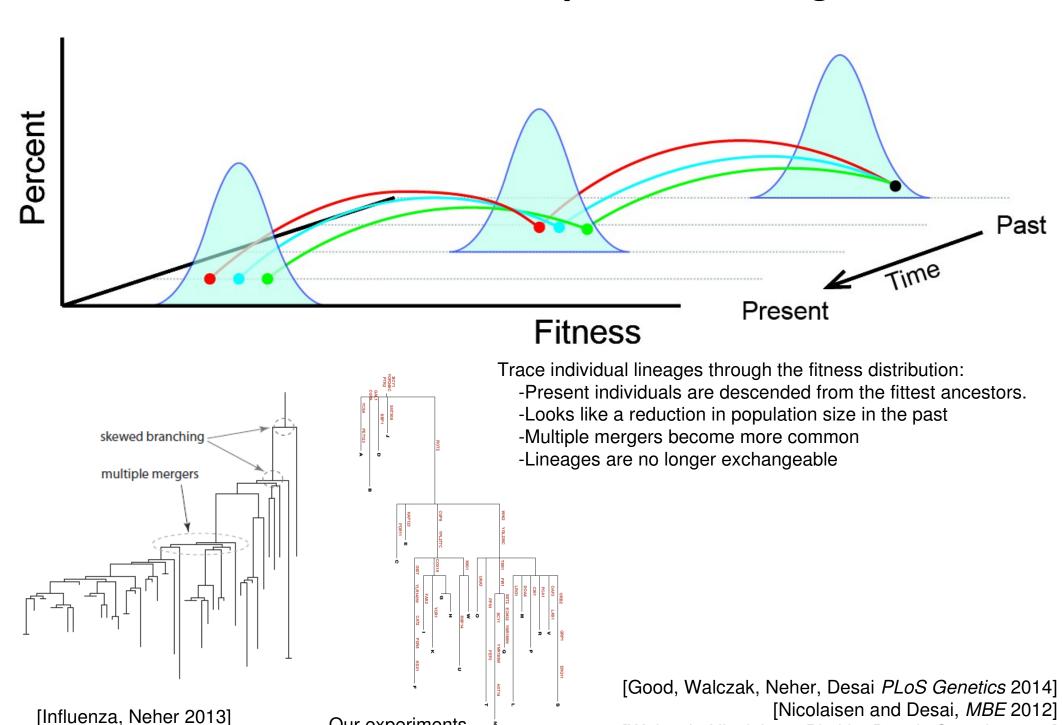
Ben Good

How does pervasive natural selection alter patterns of genetic diversity?

Standard methods describe neutral evolution



Selection and the Shape of Genealogies



[Walczak, Nicolaisen, Plotkin, Desai, *Genetics* 2012]

Our experiments

How does purifying selection shape diversity?

A simple model:

Population size: N

Mutation rate: *U*

Fitness effects: $\rho(s)$

Recombination rate: R

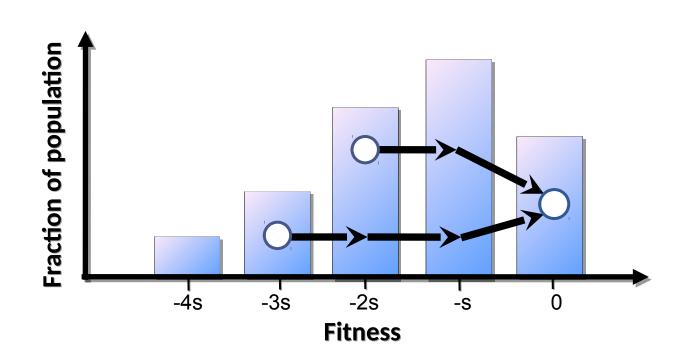
An even simpler model:

Population size: N

Neutral mutation rate: U_n

Deleterious mutation rate: U_d

Fitness effect: s



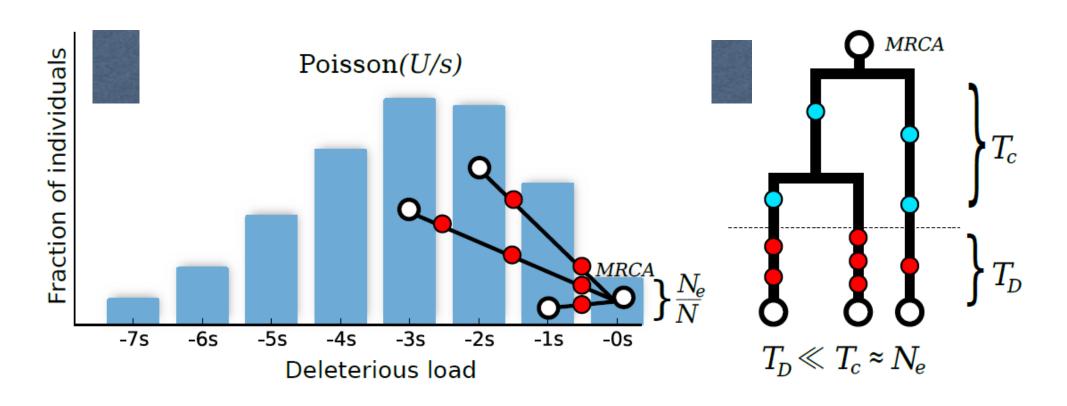
Structured Coalescent:

Steady state distribution of fitness within the population.

"Migrate" between fitness classes by mutations.

Exchangeability within each fitness class.

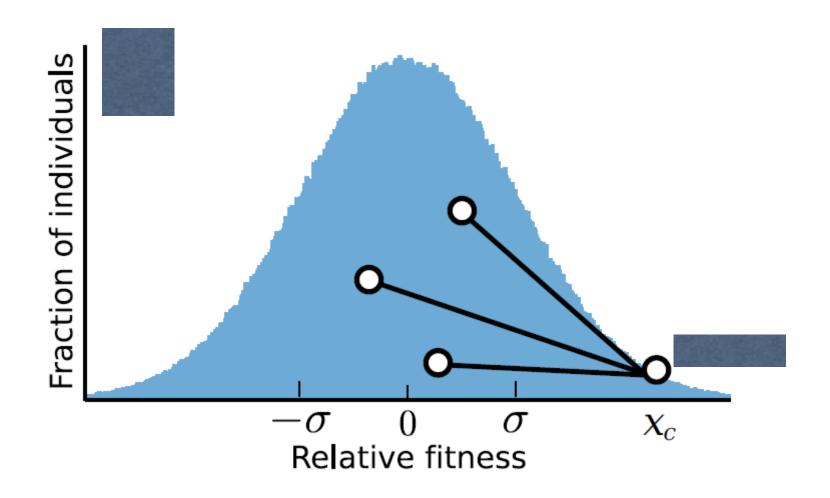
Strong purifying selection: "Background Selection"



Strong purifying selection reduces effective population size. Exact in the limit $Ns = \infty$ while holding NU/Ns constant.

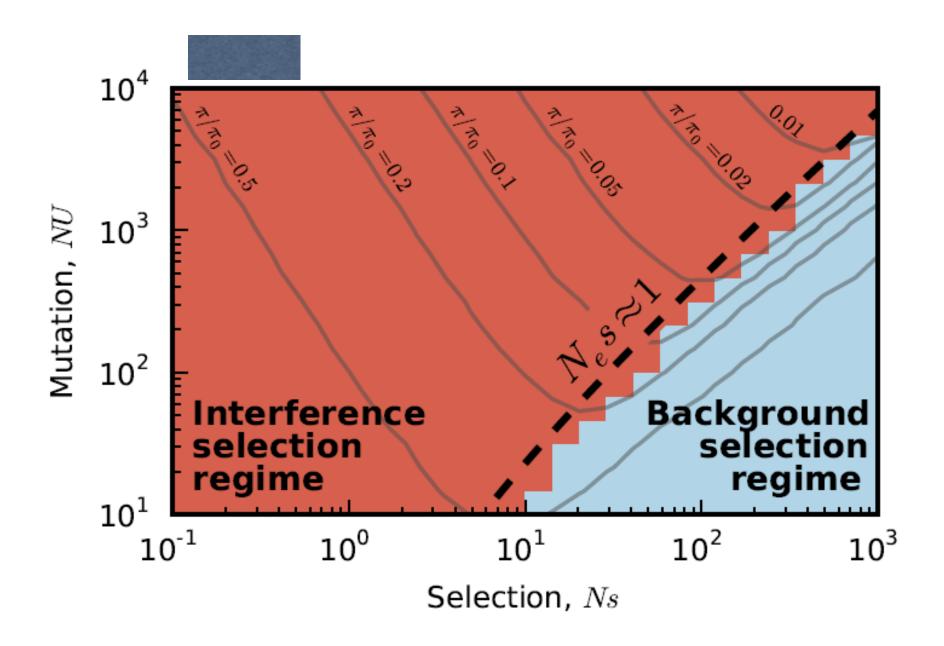
Corrections for large but finite *Ns* from the Structured Coalescent

What about weak or pervasive selection?

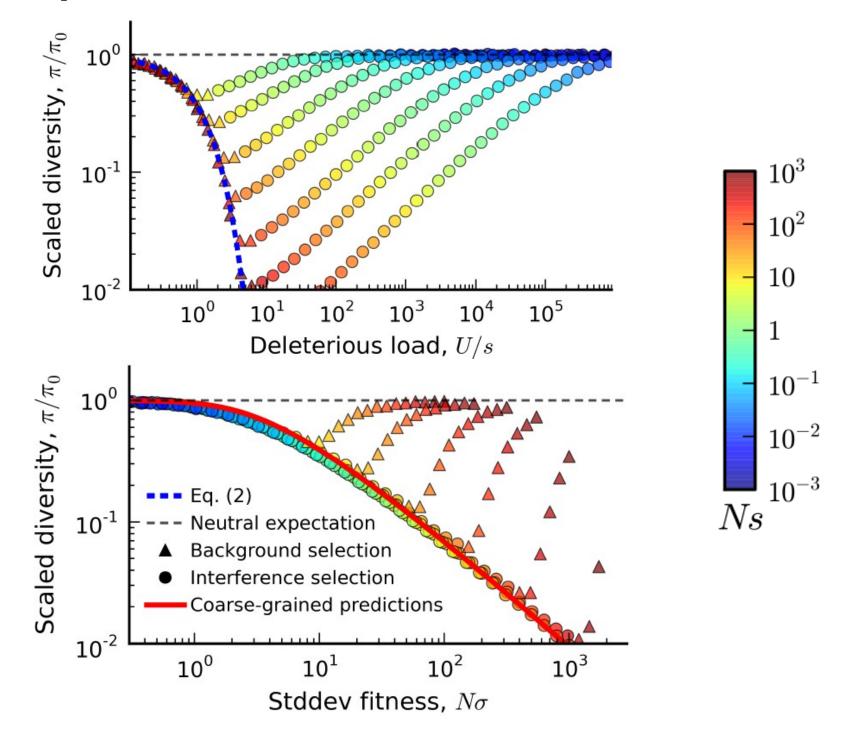


When $Nse^{-U/s} \sim 1$ or less, the distribution fluctuates too much underneath, so the structured coalescent does not make sense.

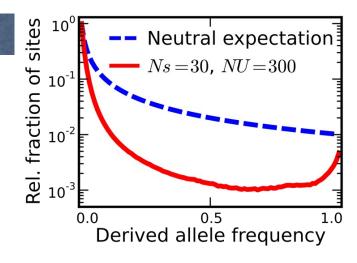
BGS/Structured Coalescent Break Down for Weak Selection

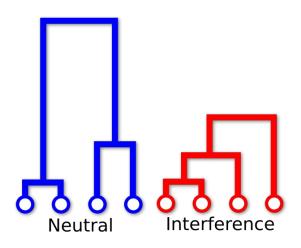


Collapse with U/s (BGS) or with N σ (IS)



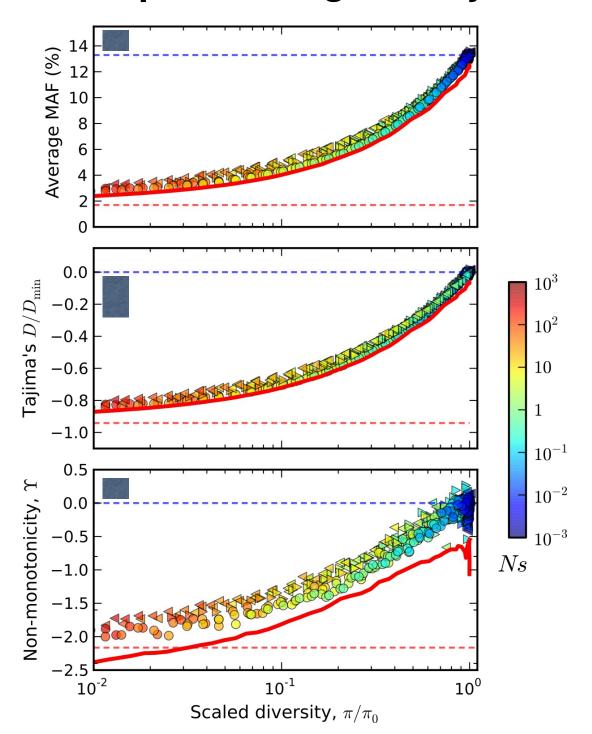
Interference Selection collapse holds generally



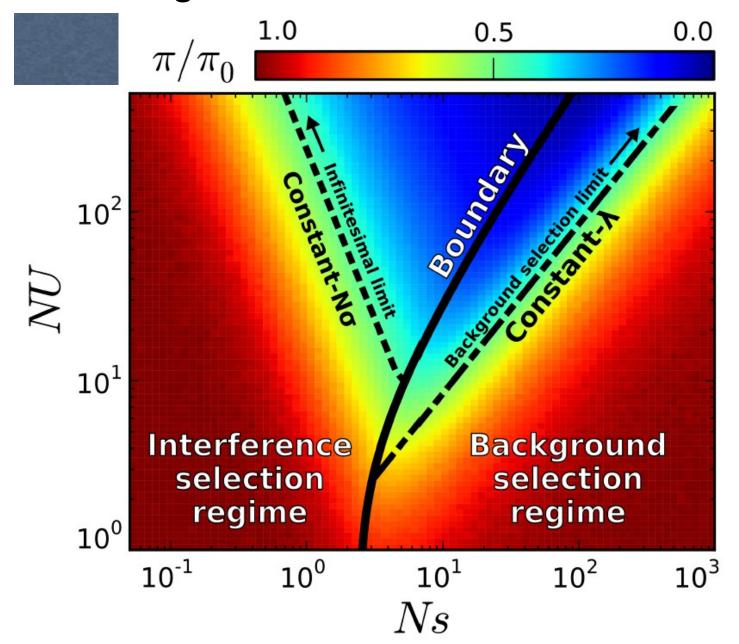


Legend:

- Asexual
- --- Neutral expectation
- NR = 10
- Coarse-grained predictions
- \blacktriangleleft NR = 100
- --- Large $N\sigma$ limit

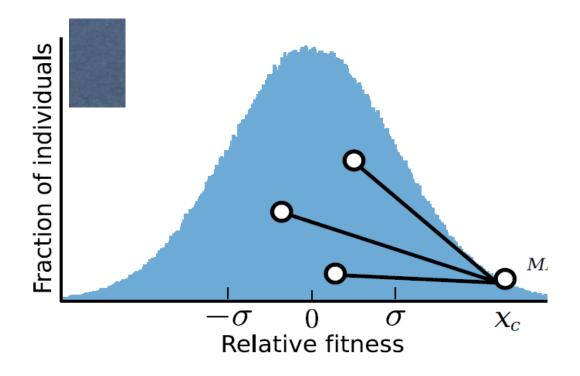


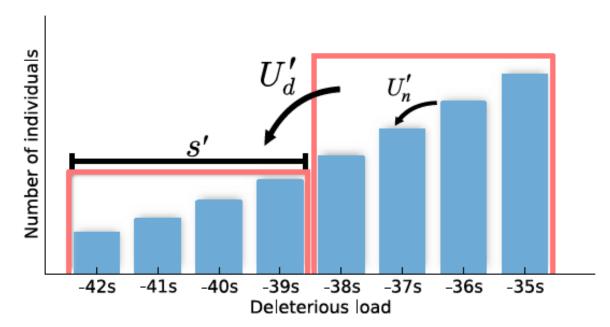
Two limits: background selection and interference selection



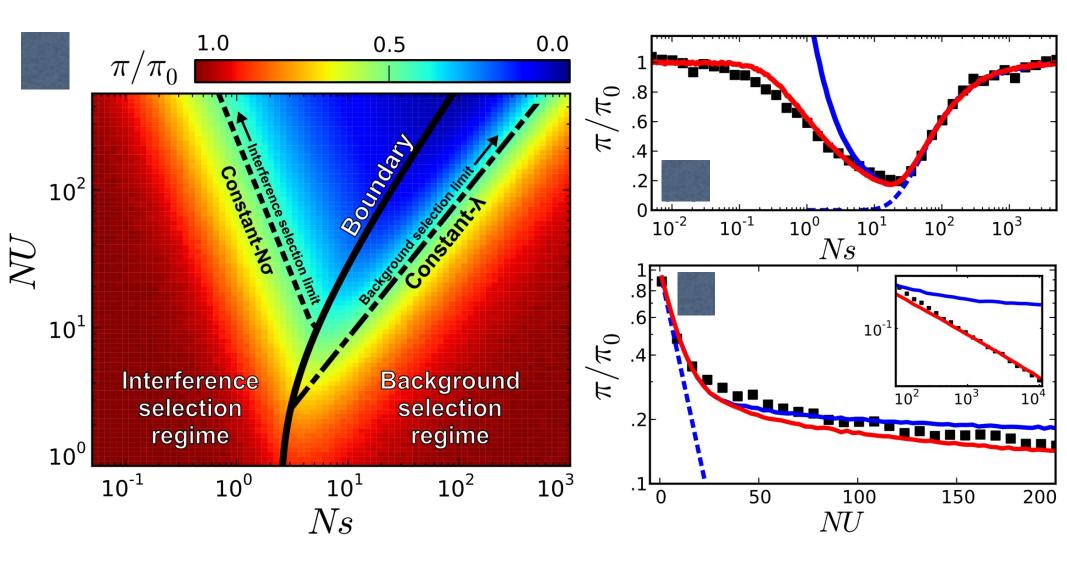
Background selection: $Ns = \infty$ while holding NU/Ns constant Inteference selection: Ns = 0 while holding $N\sigma$ constant

Intuition: "coarse-graining" the fitness distribution





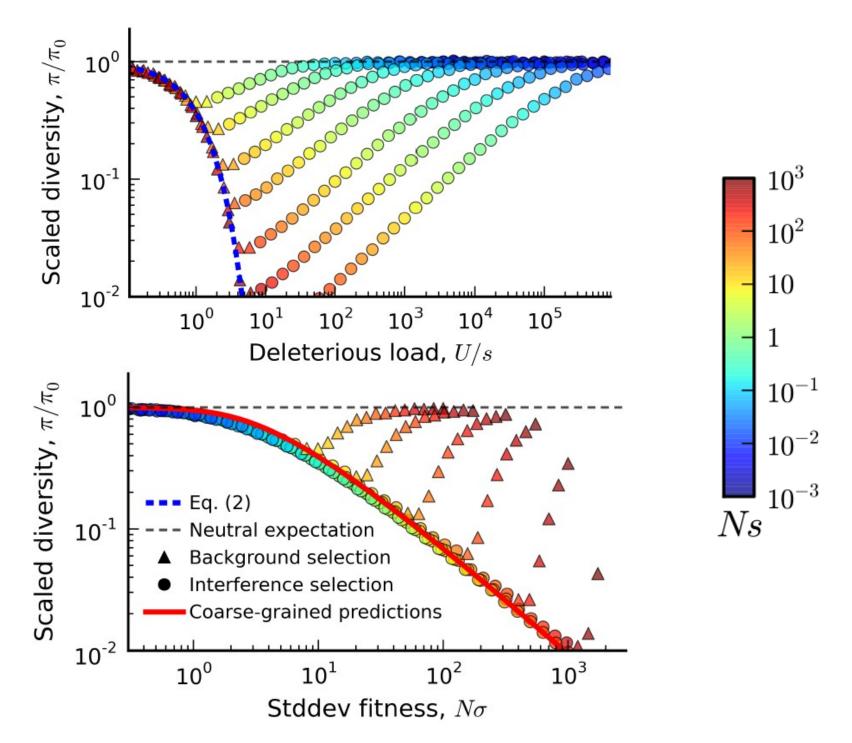
This allows us to predict diversity



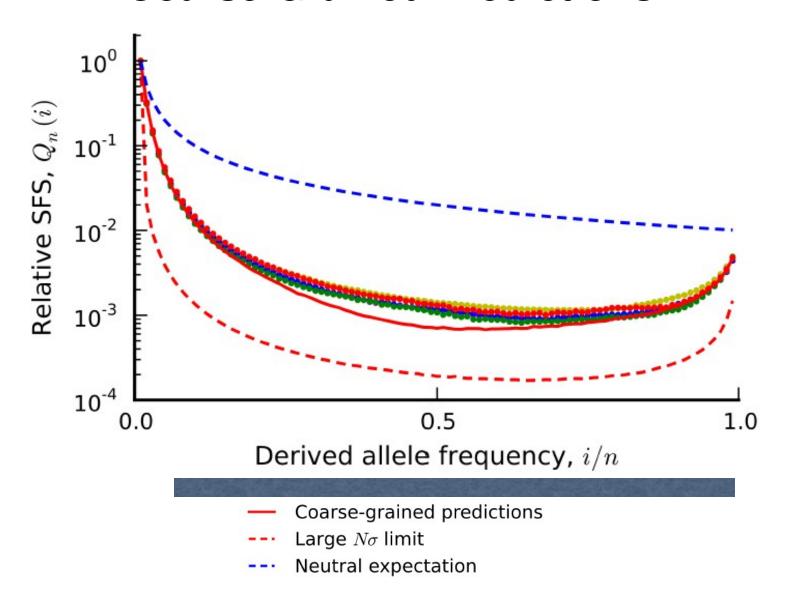
There is a fundamental problem of identifiability:

Many different parameter values lead to *identical* patterns of diversity.

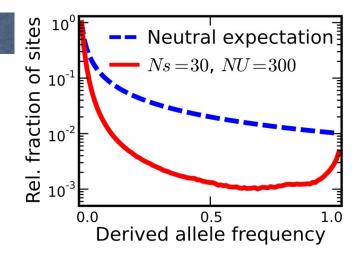
Coarse-Grained Predictions

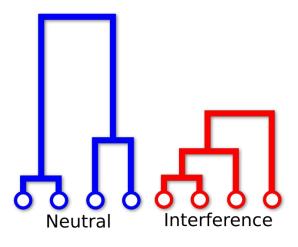


Coarse-Grained Predictions



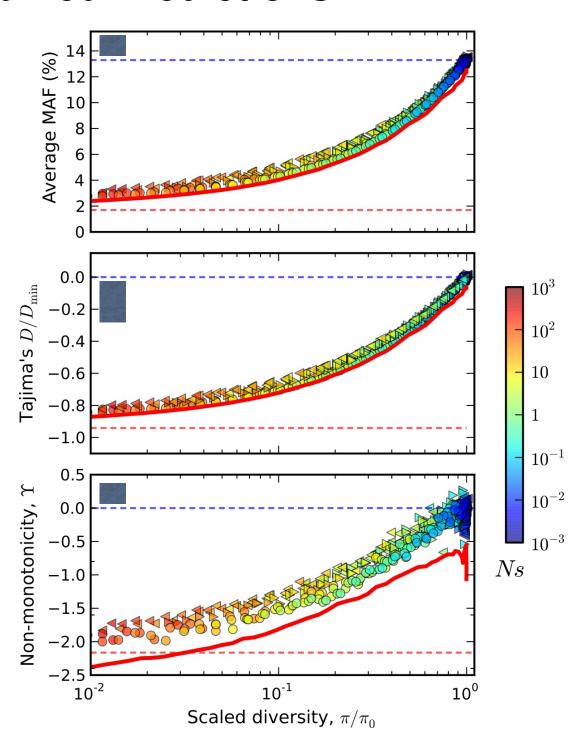
Coarse-Grained Predictions



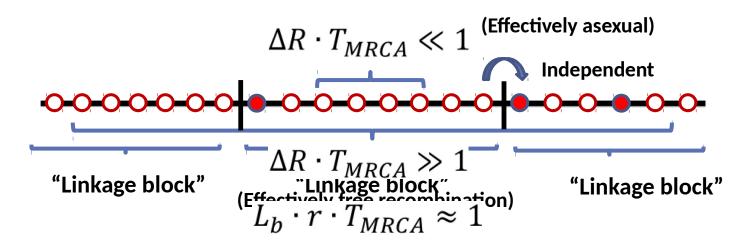


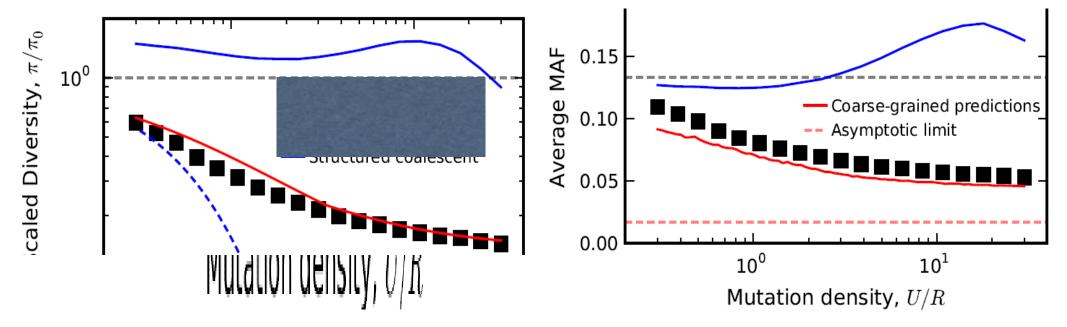
Legend:

- Asexual
- --- Neutral expectation — Coarse-grained predictions
- NR = 10 \blacktriangleleft NR = 100
- --- Large $N\sigma$ limit

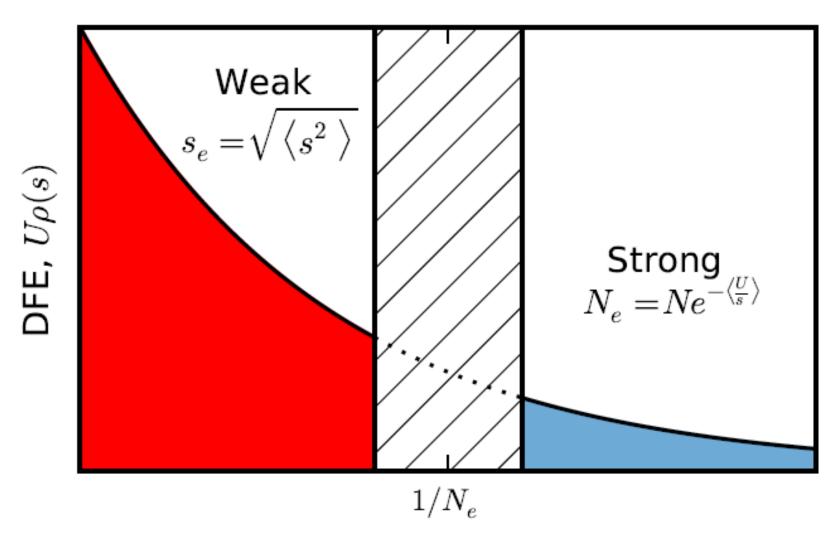


A Linkage-Block Approximation for Recombining Genomes



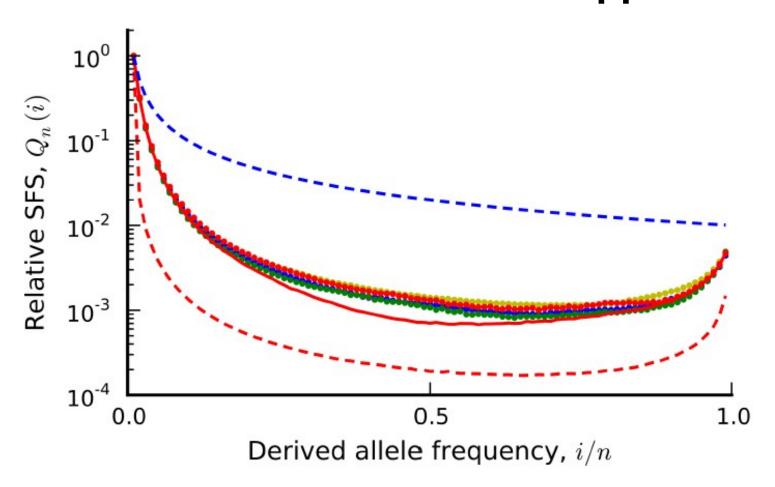


Distributions of Fitness Effects



Deleterious fitness effect, s

Interference Selection Still Applies



- • Sexual (NR = 10), Ns = 30, NU = 354
- • TruncatedExp($s_{\mathrm{max}}/\overline{s}=3$), $N\overline{s}=10$, NU=2230
- • Uniform(0, $s_{
 m max}$), $Ns_{
 m max}$ = 28.5, NU = 1000
- Finite sites ($L\!=\!10^5$), $Ns\!=\!21.4$, $NU\!=\!600$
- • Single-s, Ns = 30, NU = 300
- Coarse-grained predictions
- --- Large $N\sigma$ limit
- -- Neutral expectation

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