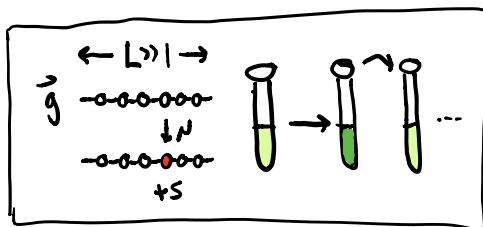


- Announcements: ① PSET 3 DUE 3/9/21 (only 3 problems this time, \Rightarrow but #1 is long!)
 ② weekly comments

Last time: Multi-locus models of evolution ($L \gg 1$)



$$\frac{d\vec{f}(\vec{g})}{dt} = \underbrace{[X(\vec{g}) - \bar{X}(t)] f(\vec{g})}_{\text{selection (nonlinear)}} + \underbrace{\sum_{\vec{g}'} M(\vec{g}' \rightarrow \vec{g}) f(\vec{g}') - M(\vec{g} \rightarrow \vec{g}') f(\vec{g})}_{\text{mutation (linear, "local")}}$$

$$+ e \sum_{\vec{g}_1, \vec{g}_2} T(\vec{g}_1, \vec{g}_2 \rightarrow \vec{g}) f(\vec{g}_1) f(\vec{g}_2) - e f(\vec{g}) \quad \text{recombination (nonlinear, non-local)}$$

$$+ \sqrt{\frac{f(\vec{g})}{N}} \eta(\vec{g}) - f(\vec{g}) \sum_{\vec{g}'} \sqrt{\frac{f(\vec{g}')}{N}} \eta(\vec{g}') \quad \text{genetic drift (stochastic)}$$

No closed-form sol'n in the general case...

\Rightarrow Need asymptotic approximations (i.e., limits of N, L, e, M , etc...)

① Large N / "mean field" (genetic drift is negligible)

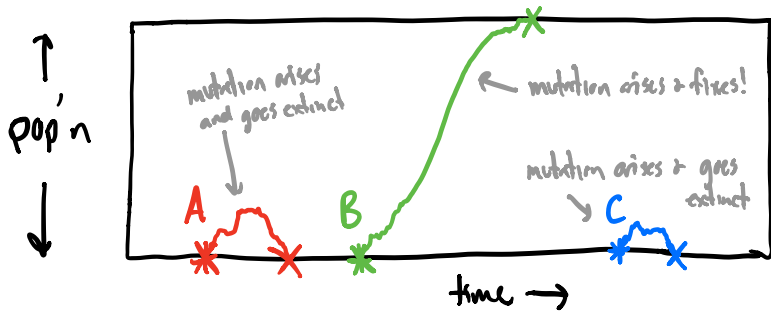
\Rightarrow \times inconsistent when $L \gg 1$! (e.g. $\frac{\# \text{ individuals}}{\# \text{ genotypes}} \ll 1$)

Today:

② Successive mutations regime
 (i.e., mutation is small correction)

$$\frac{d\vec{f}(\vec{g})}{dt} = \sim (x - \bar{x}) + \cancel{\sim \frac{1}{L} \mu} + \sim e + \sim \frac{\sum \eta}{\sqrt{N}}$$

⇒ i.e. new mutations fix or go extinct before next one occurs...



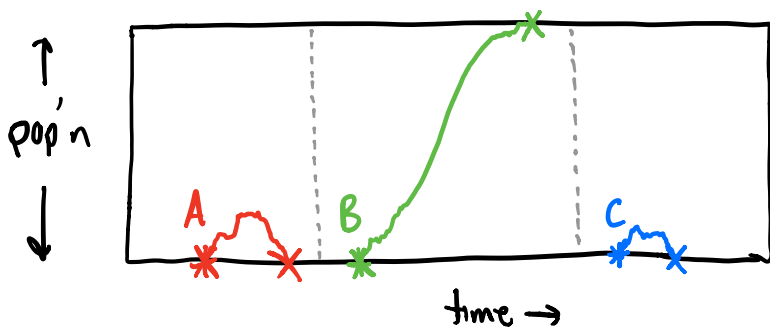
⇒ @ any given time, only 2 genotypes present:

"current wildtype" $\vec{g}_0 = (1, 0, 1, 1, 0, 0, 0)$

"single mutant" $\vec{g}_m = (1, 0, 1, 1, 0, 1, 0)$
 ↓ mutation @ site l

⇒ what can recombination do? Nothing! (on average)

⇒ then each mutation looks like $L=1$ model w/



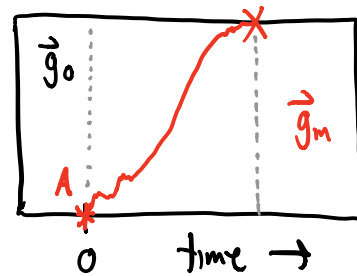
$$S_{\text{eff}} \equiv X(\vec{g}_m) - X(\vec{g}_0)$$

⇒ in this case, know exactly what happens:

(i) w/ prob $P_{\text{fix}} = \frac{2s}{1 - e^{-2Nes}}$,

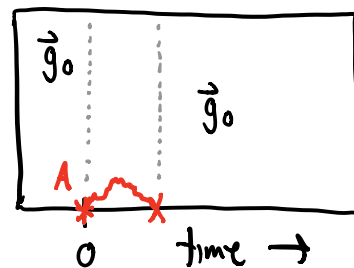
⇒ mutation fixes ("sweeps")

⇒ $\vec{g}_0 \rightarrow \vec{g}_m$; repeat!

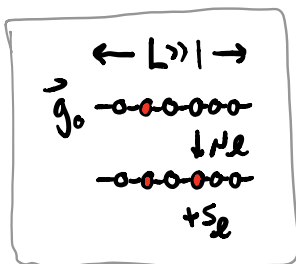


(ii) otherwise, mutation goes extinct

⇒ \vec{g}_0 stays put.



⇒ New: when $L > 1$, multiple different mutations are possible...



⇒ useful to define **fitness effects**

$$s_e \equiv X(\vec{g}_0 + \text{mut'n @ site } e) - X(\vec{g}_0)$$

along w/ **distribution of fitness effects ("DFE")**



$$N(s | \vec{g}_0) \equiv \sum_{e=1}^L N_e \delta(s - s_e)$$

↑
technically
depends on \vec{s}_0

↓
prob. of drawing
a mutation w/ effect s .

⇒ mutations w/ fitness effect s are produced as
Poisson process w/ rate $N\mu(s)$.

⇒ if each successful w/ prob $P_{\text{fix}}(s)$,

then successful mutations also Poisson Process

w/ total rate $R \equiv \int_0^{\infty} N\mu(s) \cdot P_{\text{fix}}(s) ds = \sum_{\ell} N_{\ell} \mu_{\ell} P_{\text{fix}}(s_{\ell})$

⇒ ① time until next successful mutation is born is:

$T_{\text{test}} \sim \text{Exponential}\left(\frac{1}{R}\right)$

[similar to $N\mu \rightarrow 0$
case in lecture 10]

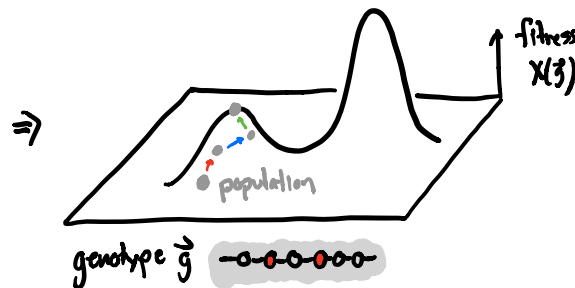
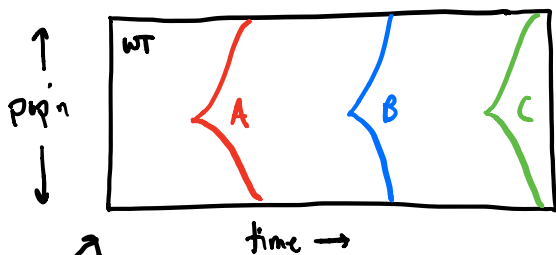
⇒ ② probability that it was site ℓ :

$$P_{\ell} = \frac{N_{\ell} \mu_{\ell} P_{\text{fix}}(s_{\ell})}{R} = \frac{N_{\ell} \mu_{\ell} P_{\text{fix}}(s_{\ell})}{\sum_{\ell} N_{\ell} \mu_{\ell} P_{\text{fix}}(s_{\ell})}$$

⇒ ③ $\vec{g}_0 \rightarrow \vec{g}_m \Rightarrow$ recalculate $\mu(s|\vec{g}_0) \Rightarrow$ repeat! ✓

When approx is valid: will check carefully below...

⇒ simple algorithm for modeling evolution (not just pop gen)



- ① time until next successful mutation is born
 $T_{est} \sim \text{Exponential}(\frac{1}{R})$
- ② probability that it was site l :
 $p_l = \frac{N_{l_0} \mu_{lx}(z_0)}{R} = \frac{\mu_{lx}(z_0)}{\sum_l \mu_{lx}(z_0)}$
- ③ Recalculate local $\mu(s)$; repeat!

"evolution as Hill-Climbing"

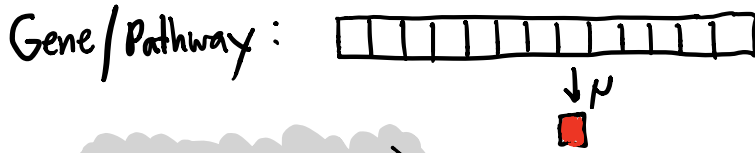
- or -

"evolution as optimization"

Note: even in these simplified settings,

⇒ fundamental limits to optimization picture...

Example: maintaining a useful function

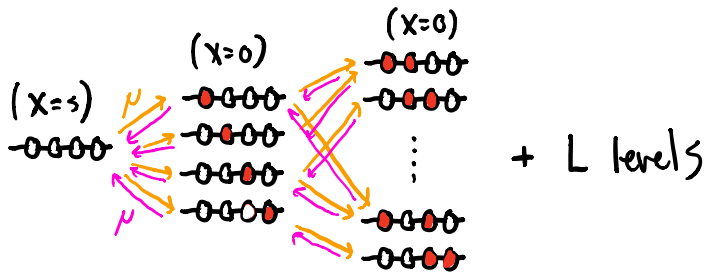


(e.g. for eating some low-level nutrient)

L ways to "break" (each w/ rate μ)

$$X(\vec{0}) \equiv s; X(\text{else}) \equiv 0$$

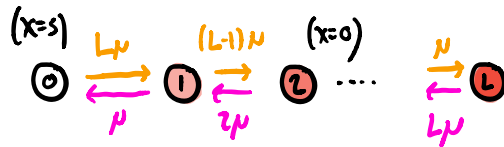
Genotype network:



Key simplification:

"equivalence class"

$$f_k \equiv \sum_{|j|=k} f(j)$$



↳ can track:

$$P_k(t) \equiv \Pr[f_k(t)=1] = \langle f_k(t) \rangle$$

How can $P_k(t)$ change?

$$\underline{k=0}: \quad d_t P_0 = \overbrace{N\mu P_{fix}(s) P_1}^{\text{incoming}} - \overbrace{NL\mu \cdot P_{fix}(-s) \cdot P_0}^{\text{outgoing}} \xrightarrow{t \rightarrow 0} 0$$

$$\Rightarrow \frac{P_0}{P_1} = \frac{N\mu P_{fix}(s)}{NL\mu P_{fix}(-s)} = \frac{1}{L} e^{2Ns}$$

$$\underline{k=1}: \quad d_t P_1 = NL\mu P_{fix}(-s) P_0 - N\mu P_{fix} P_1 \quad (\text{from } 0 \text{ class}) \\ + N \cdot 2\mu \cdot \left(\frac{1}{N}\right) P_2 - N(L-1)\mu \left(\frac{1}{N}\right) P_1 \xrightarrow{t \rightarrow \infty} 0$$

$$\Rightarrow P_2 = \frac{(L-1)}{2} P_1$$

$$\underline{k=2}: \quad \Rightarrow P_3 = \frac{L-2}{3} \cdot P_2 = \frac{(L-1)(L-2)}{3 \cdot 2} \cdot P_1$$

$$\Rightarrow P_k = \frac{1}{L} \frac{L!}{k!(L-k)!} P_1$$

$$\Rightarrow 1 - P_0 = \sum_{k=1}^L P_k = \frac{1}{L} (2^L - 1) P_1$$

$$\Rightarrow \boxed{f_1 = (1-f_0) \frac{L}{(2^L-1)}}$$

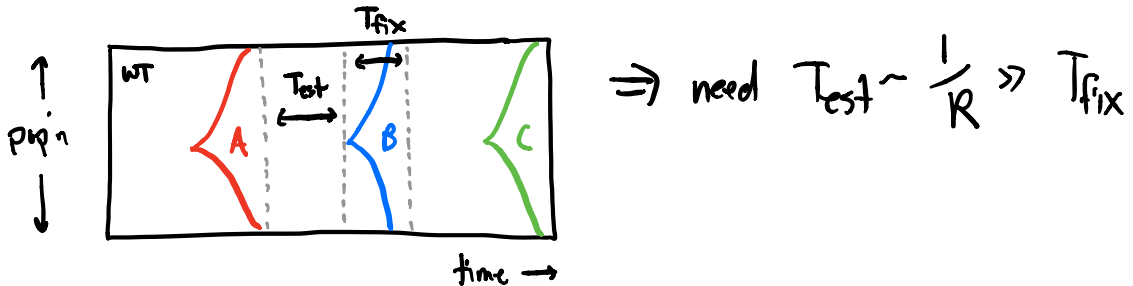
$$\Rightarrow \boxed{\frac{P_0}{1-P_0} = \exp \left[\underbrace{2Ns}_{\text{"drift barrier"}} - \underbrace{\log(2^L-1)}_{\text{"entropy of genotype space"}} \right]} = \frac{\text{prob having function}}{\text{prob broken}}$$

$$\Rightarrow \frac{0.9}{0.1} = \exp[2] \leq \exp \left(2Ns - \overbrace{\log(2^L-1)}^{\propto L} \right)$$

$\hookrightarrow \boxed{s \geq \frac{1}{N}}$ "drift barrier"

\hookrightarrow e.g. compare to deterministic case: $\bar{f}_0 = 1 - \frac{L\mu}{s}$

When is successive mutations regime a good approx?



E.g. Neutral mutations ($\mu(s) = U_n \delta(s)$)

$$P_{\text{fix}}(0) = \frac{1}{N} \Rightarrow R = NU_n \left(\frac{1}{N}\right) = U_n ; T_{\text{fix}} \sim \mathcal{O}(N)$$

$$\Rightarrow \text{need } \frac{1}{U_n} \gg N \Rightarrow \boxed{NU_n \ll 1} \text{ "weak mutation"} \\ \text{"weak selection"}$$

E.g. Strongly beneficial mutations ($\mu(s) = U_b \delta(s - s_b)$; $Ns_b \gg 1$)

$$\Rightarrow P_{\text{fix}}(s) \approx 2s \Rightarrow R = 2NU_b s ; T_{\text{fix}} = \frac{2}{s} \log(Ns)$$

$$\Rightarrow \text{need } \frac{1}{2NU_b s} \gg \frac{2}{s} \log(Ns) \Rightarrow \boxed{NU_b \ll \frac{1}{\log(Ns_b)}} \ll 1$$

\Rightarrow what does this look like
for some "real" parameter values?

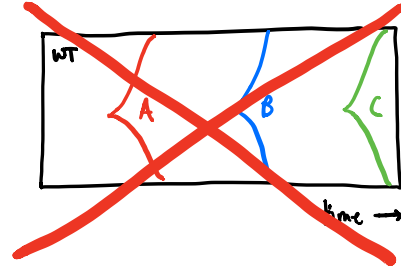
\Rightarrow e.g. HW2 problem 4

$$U_b \sim 5 \times 10^{-6}, s_b \sim 0.02$$

just for L.O.F. muts.

\Rightarrow e.g. if $N \sim 10^5 \Rightarrow 2N \log_2(N) = 3$

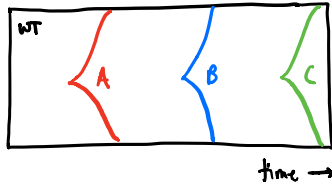
\Rightarrow successive mutations picture does not apply!



\Rightarrow what do things look like instead?

Neutral theory & the Coalescent

Successive mutations:



$$\frac{ds(\bar{g})}{dt} = \underbrace{\sim (x-\bar{x})}_{\text{blue}} + \underbrace{\sim L\mu}_{\text{orange}} \xrightarrow{\epsilon} + \underbrace{\sim e}_{\text{purple}} + \underbrace{\sim \frac{\pi}{2\mu}}_{\text{green}}$$

\Rightarrow ~ 1 variant present @ high freqs \Rightarrow solved by reducing to $L=1$ model

* But genomes in data separated by multiple mut'n's
(e.g. humans, 2 individuals differ by ~ 1 mut / 1000 bp)

\Rightarrow need to understand what's going on in these cases...

$$\frac{ds(\bar{g})}{dt} = \underbrace{\sim (x-\bar{x})}_{\text{blue}} + \underbrace{\sim L\mu}_{\text{orange}} + \underbrace{\sim e}_{\text{purple}} + \underbrace{\sim \frac{\pi}{2\mu}}_{\text{green}}$$

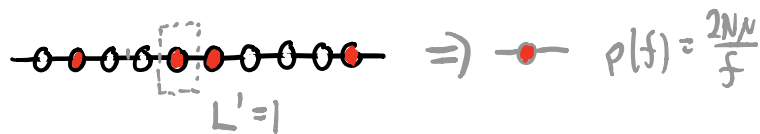
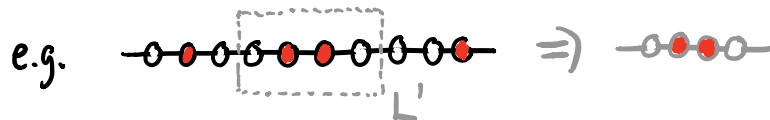
\Rightarrow one other limit that's well understood:
neutral evolution in nonrecombining genome

when $X(\vec{g})=0 \rightarrow e=0$, left with: $(\mu_e = \nu_e)$

$$\frac{df(\vec{g})}{dt} = \underbrace{\sum_{|\vec{g}'-\vec{g}|=1} \sum_e \mu_e f(\vec{g}') \left[g_e(1-g_e') + (1-g_e)g_e' \right]}_{\text{incoming mutations}} - \underbrace{\sum_e \nu_e f(\vec{g})}_{\text{outgoing mutations}}$$

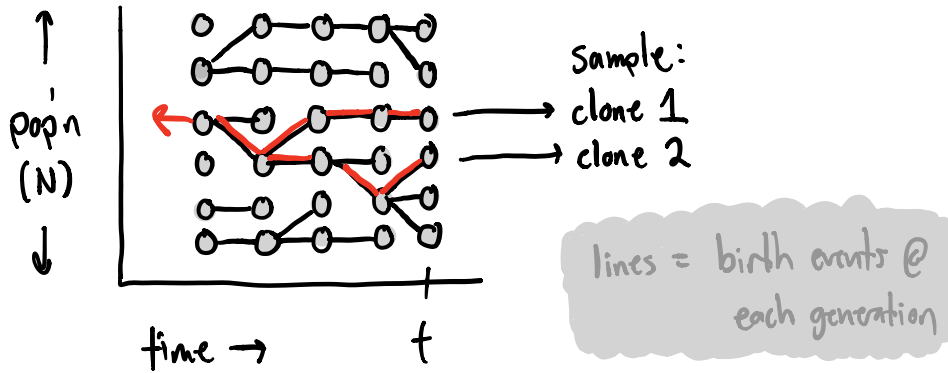
$$+ \sqrt{\frac{f(\vec{g})}{N}} \eta(\vec{g}) - f(\vec{g}) \sum_{\vec{g}'} \sqrt{\frac{f(\vec{g}')}{N}} \eta(\vec{g}') \quad \text{genetic drift}$$

Key insight: sites don't actually influence each other (because neutral)

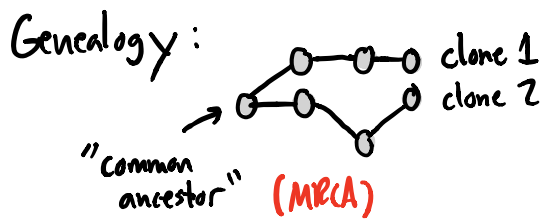


\Rightarrow 2nd key insight: can take $L'=0$ —

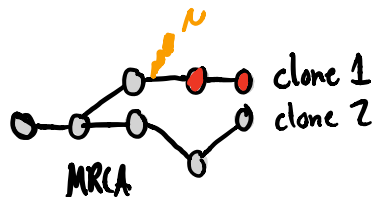
E.g. simulation of neutral pop'n in Wright-Fisher model:



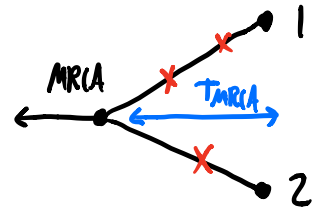
⇒ key insight: lines also = **genealogical relationships** backward in time!



⇓ differences between sampled individuals = mutations on genealogy



⇒ Mutations occur as Poisson Process
w/ rate μ_e on each branch



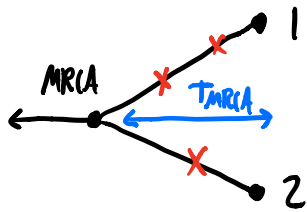
↳ total # of muts
 $\text{Poisson}(2\mu_e T_{MRCA})$

⇒ 2 extreme limits:

(1) $\mu_e T_{MRCA} \ll 1 \Rightarrow 0$ or 1 mutations on whole tree

$$\Rightarrow \Pr[\text{genetic diff} @ \text{site } e] =$$

(2) $\mu_e T_{MRCA} \gg 1 \Rightarrow$ lots of forward & backward mutations along each branch.



$$\Rightarrow \Pr[\text{genetic diff} @ \text{site } e] =$$

⇒ Key Question: what sets T_{MRCA} ?