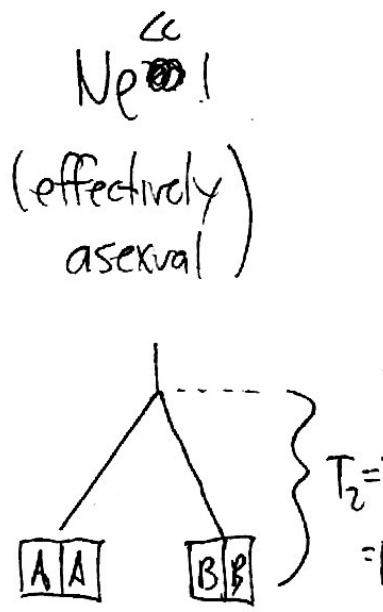
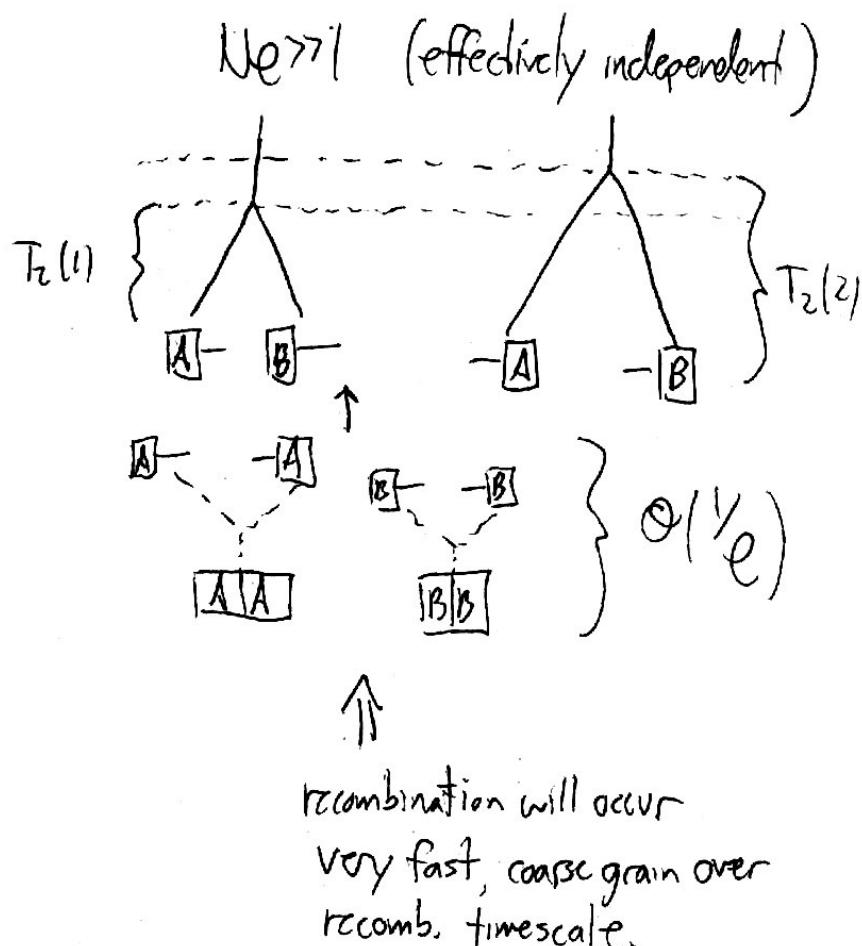


Quasi Linkage Equilibrium (QLE)

Last time, we talked about coalescent models for neutral recombining genomes, and saw that this led to 2 extreme limits:

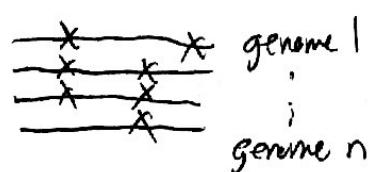


↑
not enough time
for recombination to
occur.



⇒ in between, $N \sim O(1)$ very complicated (ARG)

@ end of day, want mutations in sample, not trees...



⇒ also hard to get selection into this picture...

(2)

So today, want to talk about same concepts from
a forward-time perspective, based on genotype frequencies, $f(\vec{g})$

\Rightarrow to start, let's consider a simple 2-locus model
for genotypes $\vec{g} = (0,0), (1,0), (0,1), (1,1)$, w/o selection

then genotype freqs satisfy the system of SDEs:

$$\frac{df(11)}{dt} = \rho \left[f(10)f(01) - f(11)f(00) \right] + \sqrt{\frac{f(11)}{N}} \eta(11) - f(11) \sum_{\vec{g}' \neq 11} \sqrt{\frac{f(\vec{g}')}{N}} \eta(\vec{g}')$$

$$\frac{df(10)}{dt} = \rho \left[f(11)f(00) - f(10)f(01) \right] + \sqrt{\frac{f(10)}{N}} \eta(10) - f(10) \sum_{\vec{g}' \neq 10} \sqrt{\frac{f(\vec{g}')}{N}} \eta(\vec{g}')$$

$$\frac{df(01)}{dt} = \rho \left[f(11)f(00) - f(10)f(01) \right] + \sqrt{\frac{f(01)}{N}} \eta(01) - f(01) \sum_{\vec{g}' \neq 01} \sqrt{\frac{f(\vec{g}')}{N}} \eta(\vec{g}')$$

$$\frac{df(00)}{dt} = \rho \left[f(10)f(01) - f(11)f(00) \right] + \sqrt{\frac{f(00)}{N}} \eta(00) - f(00) \sum_{\vec{g}' \neq 00} \sqrt{\frac{f(\vec{g}')}{N}} \eta(\vec{g}')$$

and sample comes from:

$$(f(11), f(10), f(01), f(00)) \xrightarrow{\text{Multinomial } (n, \vec{f})} \begin{array}{c} \bullet \\ \bullet \\ \bullet \\ \hline \end{array} \quad n(11), n(10), n(01), n(00)$$

(3)

this system really has only 3 independent eqs,

$$\text{since } f(11) + f(10) + f(01) + f(00) = 1$$

\Rightarrow e.g. can eliminate $f(00) = 1 - f(11) - f(10) - f(01)$
 & work w/ $f(11), f(10), f(01)$.

However, $(f(11), f(10), f(01))$ is not the only basis we could work with.

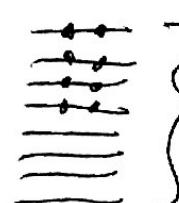
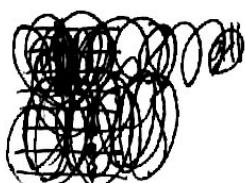
\Rightarrow free to choose any other combination.

\Rightarrow one combination that is often used:

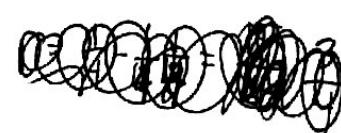
$$\underbrace{f_1 \equiv f(11) + f(10)}_{\substack{\text{marginal allele freq} \\ \text{of mutations @} \\ \text{site 1}}}, \quad \underbrace{f_2 \equiv f(11) + f(01)}_{\substack{\text{allele freq} \\ \text{@ site 2}}}, \quad \underbrace{D \equiv f(11) - f_1 f_2}_{\substack{\text{"Linkage disequilibrium" (LD)}}} = f(11)f(00) - f(10)f(01)$$

\Rightarrow LD is measure of how much double mutant frequency deviates from a model where mutations are totally independent.

e.g. one high-LD scenario
 $(D = \text{large \& positive})$

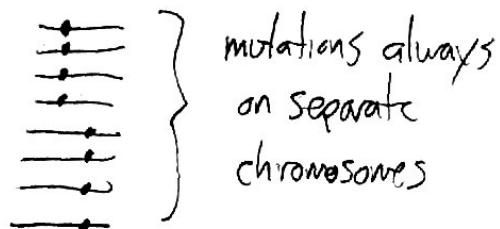


mutations always appear
~~together~~ on same genomes.



$$D = \frac{1}{2} - \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$$

e.g. another high-LD scenario
(LD = large & negative)



$$D = 0 - \frac{1}{2} \cdot \frac{1}{2} = -\frac{1}{4}$$

(4)

\Rightarrow sometimes people measure as correlation coefficient:

$$r = \frac{D}{\sqrt{f_1(1-f_1)f_2(1-f_2)}}$$

e.g. in example 1, $r = +1$
example 2, $r = -1$

\Rightarrow mutation @ site 1 gives you perfect info about site 2, & vice versa.

Why is f_1, f_2, D a good basis? Rewrite SDEs:

$$\frac{\partial f_1}{\partial t} = \frac{\partial f(11)}{\partial t} + \frac{\partial f(10)}{\partial t} = \cancel{\text{noise}} e^{[f(00)f(01) - f(00)f(11)]} + e^{[f(00)f(11) - f(10)f(01)]} + \text{noise (later)}$$

$$\begin{aligned} \frac{\partial f_2}{\partial t} &= 0 + \text{noise}, \quad \frac{\partial D}{\partial t} = \frac{\partial f_{11}}{\partial t} \cancel{- \frac{\partial f_1}{\partial t} f_2} - \frac{\partial f_1}{\partial t} \overset{0}{f_2} - f_1 \frac{\partial f_2}{\partial t} \overset{0}{f_1} \\ &= -pD + \text{noise}. \end{aligned}$$

(5)

In other words, ~~recombination~~ recombination cannot change allele frequencies, can only change linkage disequilibrium. Change in linkage disequilibrium has very simple ~~form~~ form (deterministically)

$$\Rightarrow \frac{dD}{dt} = -rD \Rightarrow D(t) = D(0)e^{-rt} \Rightarrow \text{decays to zero exponentially fast in } r$$

\Rightarrow suggests that if $r \rightarrow \infty$ (compared to what?)

$D \rightarrow 0$ and maybe we can treat 2 locus system as direct product of single locus systems : ~~recombination~~

e.g. pictorially:

we call this limit "Linkage equilibrium", "free recombination", "independent sites", etc.

(6)

we can check this assumption using our now-familiar

Self consistency argument: assume that D is small &

calculate next order correction \Rightarrow this condition is called

"Quasi-linkage equilibrium"
(QLE)

Easiest to see QLE if we focus on rare mutations, $f_1, f_2 \ll 1$.
Then SDEs reduce to (w/ noise now)

$$\frac{df_1}{dt} = \sqrt{\frac{f_{11}}{N}} \eta_{11} + \sqrt{\frac{f_{10}}{N}} \eta_{10} = \sqrt{\frac{f_1 f_2 + D}{N}} \eta_{11} + \sqrt{\frac{f_1 (1-f_2) - D}{N}} \eta_{10}$$

$$\approx \sqrt{\frac{f_1}{N}} \tilde{\eta}_1$$

$$\frac{df_2}{dt} = \sqrt{\frac{f_{11}}{N}} \eta_{11} + \sqrt{\frac{f_{01}}{N}} \eta_{01} = \sqrt{\frac{f_1 f_2 + D}{N}} \eta_{11} + \sqrt{\frac{f_0 (1-f_1) f_2}{N}} \eta_{01} \approx \sqrt{\frac{f_2}{N}} \tilde{\eta}_2$$

$$\frac{df_{11}}{dt} = -\rho D + \sqrt{\frac{f_{11}}{N}} \eta_{11} = \cancel{\rho f_1 f_2} \cancel{\rho f_0 f_1} \rho f_1 f_2 - \rho f_{11} + \sqrt{\frac{f_{11}}{N}} \eta_{11}$$

$$w/ \langle \eta_{11} \tilde{\eta}_1 \rangle = \sqrt{\frac{f_{11}}{f_1}}, \langle \eta_{11} \tilde{\eta}_2 \rangle = \sqrt{\frac{f_{11}}{f_2}}, \langle \tilde{\eta}_1 \tilde{\eta}_2 \rangle = \frac{f_{11}}{\sqrt{f_1 f_2}}$$

(7)

\Rightarrow in QLE, we will assume that dynamics of D (f_{\parallel}) will relax much faster than dynamics of f_1, f_2

\Rightarrow since neutral, know that f_1, f_2 change on timescale

$$T_{\text{drift}} \sim Nf_1, Nf_2$$

\Rightarrow on timescales $\ll T_{\text{drift}}$, f_1 & f_2 are effectively const.

\Rightarrow equation for f_{\parallel} looks like Branching process w/ mutation

w/ effective params: $N_e = \rho f_1 f_2$, $S_e = -\rho$

\Rightarrow solution from Lecture 8, p.4:

Gamma Dist'n w/ shape: ~~2Nρf₁f₂~~ $\alpha = 2N\rho f_1 f_2$

and ~~$f_{\parallel, \text{max}}$~~ $f_{\parallel, \text{max}} = \frac{1 - e^{-\rho t}}{2N\rho} \rightarrow \frac{1}{2N\rho}$

$$\Rightarrow \langle f_{\parallel} \rangle = f_1 f_2, \text{Var}(f_{\parallel}) = \frac{f_1 f_2}{2N\rho}$$

$$\Rightarrow \langle D \rangle = 0 \quad \text{Var}(D) = \frac{f_1 f_2}{2N\rho}$$

(8)

we also know that relaxation timescale is $\sim \frac{1}{e}$

\Rightarrow Now can check our assumptions: $\frac{1}{e} \ll T_{\text{drift}} - Nf_1, Nf_2$

\Rightarrow QLE holds if $Nf_1, Nf_2 \gg 1$.

↳ similar to coalescent picture,
except now explicit f dependence.

\Rightarrow if this is true, then $\langle \hat{\eta}_2 \hat{\eta}_1 \rangle = \sqrt{\frac{f_1^2}{f_1 f_2}} = \sqrt{\frac{1}{(Nf_1)(Nf_2)}} \ll 1$

and $\frac{df_1}{dT}, \frac{df_2}{dT}$ equations really decouple! ✓.

so we showed that $\begin{array}{c} -o-o- \\ | \quad | \\ 1 \quad 2 \end{array} \longrightarrow \begin{array}{c} -o- \otimes -o- \\ | \quad | \\ 1 \quad 2 \end{array}$

but only ~~approximates~~ for sufficiently high freqs where can coarse-grain over recombination time.

\Rightarrow can do same argument for selection, e.g. $X(\vec{g}) = S_1 g_1 + S_2 g_2$

(9)

then can show:

$$\frac{df_1}{dt} = S_1 f_1 + S_2 D + \text{noise}$$

$$\frac{df_2}{dt} = S_2 f_2 + S_1 D + \text{noise}$$

$$\begin{aligned} \frac{dD}{dt} &= \frac{df_1}{dt} - f_2 \frac{df_1}{dt} - f_1 \frac{df_2}{dt} = \text{(cancel terms)} \\ &= (S_1 + S_2 - r) D + \text{noise.} \end{aligned}$$

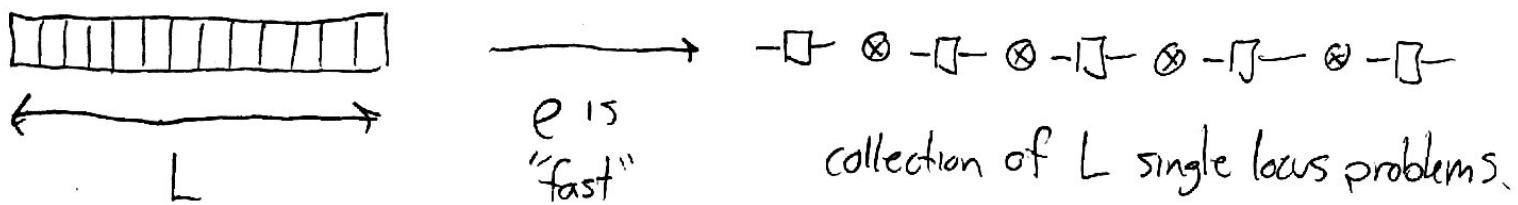
\Rightarrow so if $r > S_1 + S_2 \Rightarrow D(t) \rightarrow 0$ w/ time.

\Rightarrow General conclusion: if r is faster than all other timescales in system then loci evolve independently!

\Rightarrow in practice, people take this argument & run w/ it for entire genome: (though people rarely check it like we just did because QLE is hard...)

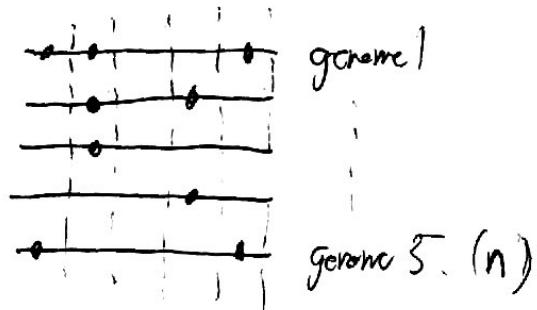
Linkage equilibrium approx ("independent sites")

10



\Rightarrow when it works, one of the most powerful approximations in pop gen.
Since it lets us use single locus results to look @ real data
 \hookrightarrow crazy if you think about it.

now when we draw a sample of individuals,
we can assign mutations independently given
current allele frequencies, $\{f_e(t)\}$



\Rightarrow by definition, $D \approx 0$ & no information in haplotype structure.

\Rightarrow instead, data can be completely summarized by $n_e = \# \text{ of individuals w/ mutation @ site } e$

\Rightarrow 

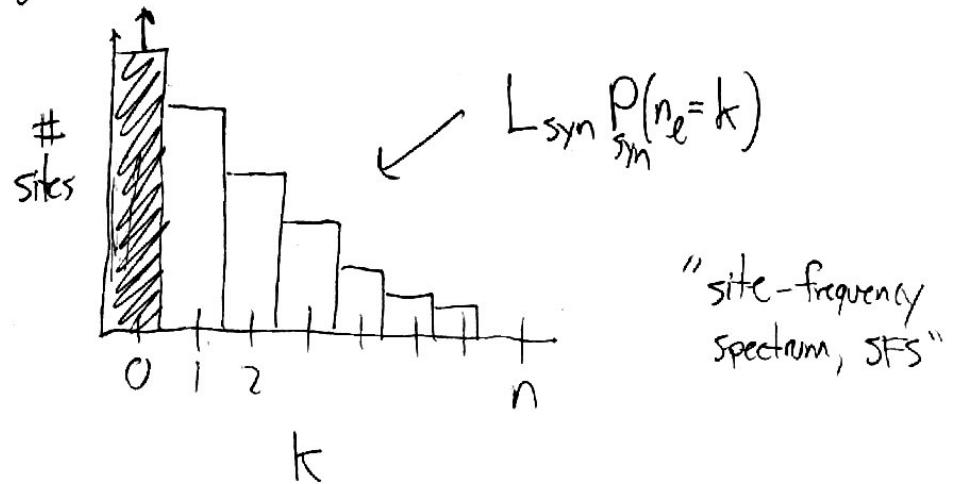
$$\Rightarrow \Pr(n_e = k) = \int \binom{n}{k} f_e^k (1-f_e)^{n-k} p_e(f_e) df_e$$

(11)

@ this point, people often gap "similar" sites together.

e.g. all synonymous sites
(putatively neutral,)

$$\text{so } \pi_e = 0$$



\Rightarrow since there are lots of synonymous sites, & each one is an independent draw from $P_{\text{syn}}(n_e=k)$, then across genome, we get a self-averaged version of $P_{\text{syn}}(n_e=k)$, even from just 1 population! this means we can estimate demography,

$$\text{Since } P_e(f_e) \Leftrightarrow \frac{\partial f_e}{\partial t} = \mu_e + \sqrt{\frac{f_e(1-f_e)}{N(t)}} N(t)$$

(mapping often not possible in closed form, but can do numerically)

$$\Rightarrow \text{if } N(t)=N \Rightarrow \Pr(n_e=k) = \frac{2N\mu}{k} \rightarrow \frac{2N\mu}{f} \text{ when } n \text{ large.}$$

↓
2Nμ when n=2.

\Rightarrow similarly, π_{syn} self averages to $2N\mu/N$, even w/ just 2 samples!

can do same thing for non-synonymous mutations

$$\Pr[n_e = k] = \iiint \binom{n}{k} f_e^k (1-f_e)^{n-k} p(f_e | s_e) \rho(s_e) df_e ds_e$$

✓ ↗ Selection
 random freq @ coeff @
 locus locus

$$\Rightarrow \text{e.g. if } \rho(s) = \cancel{\text{random freq}} \cancel{\text{locus}} (1-\lambda_d) \delta(s) + \lambda_d \delta(s+s_d)$$

✓ ↗
 neutral mutations strongly
 deleterious mut.
 (e.g. LOF mutant)

$$\Rightarrow p(f) = \frac{2N\mu(1-\lambda_d)}{f} + \frac{2N\mu\lambda_d e^{-Nsf}}{f} = \begin{cases} \frac{2N\mu}{f} & \text{for } f \ll \frac{1}{Ns} \\ \cancel{\text{random freq}} \cancel{\text{locus}} & (\text{like synonymous}) \\ \frac{2N\mu(1-\lambda_d)}{f} & \text{for } f \gg \frac{1}{Ns} \end{cases}$$

\Rightarrow similarly for π :

$$\pi_{\text{non}} = (1-\lambda_d) 2N\mu + \lambda_d \frac{2N\mu}{2Ns} \Rightarrow \text{typically we don't know } 2N\mu, \text{ but can estimate it from } \pi_{\text{syn}}$$

$$\Rightarrow \pi_{\text{non}} / \pi_{\text{syn}} = (1-\lambda_d) + \frac{\lambda_d}{Ns} \quad (Ns \gg 1)$$

$\frac{T_{\text{In}}}{T_{\text{S}}}$ provides a measure of "constraint", how much negative selection is going on in population w/in TMRCA.

e.g. in bacteria (E. coli in 2 different people's guts)

often find $T_{\text{Non}}/T_{\text{Syn}} \approx 0.1$.

\Rightarrow this is pretty crazy... suggests that $\lambda_d \gtrsim 0.9$

i.e., 90% of all amino acid changes in bacteria ~~are~~
~~are~~ are sufficiently strongly deleterious

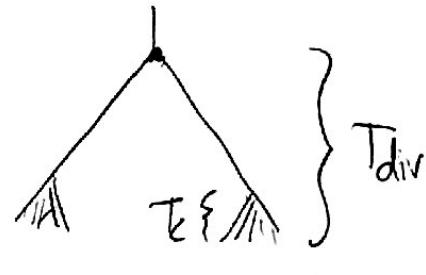
that they are ~~are~~ strongly selected against.

\Rightarrow in practice, people often coarse grain sites & look for constraint on even smaller portions of genome.

\Rightarrow reason is that strongly constrained \approx important for organism
 (interesting biology)

\Rightarrow can also do same thing w/ substitutions between 2 species (dN/dS)

\Rightarrow more time for mutations to occur, better signal



When do we expect QLE to work?

\Rightarrow think about collection of 2 locus mini problems.

$$\rho_{\text{eff}} = r \Delta l \rightarrow \begin{array}{l} \text{distance between sites} \\ \hookrightarrow \text{recombination rate per site.} \end{array}$$

if guesstimate N from π_{syn} : $N = \frac{\pi_{\text{syn}}}{2r}$

$$\Rightarrow \cancel{D_{\max}} \sim \frac{1}{N\rho} \sim \frac{r}{r(\pi \Delta l)} \approx \frac{r}{\Delta l} \quad \text{if } \Delta l \approx \frac{1}{\pi} \quad (\text{neighboring SNPs})$$

~~in most organisms we've measured,~~ $\frac{r}{\Delta l} \sim \mathcal{O}(1)$
 (weird right?) \rightarrow so neutral muts right on boundary of ok.

$$\Rightarrow \text{selected mutations needs: } \frac{s}{r} \ll 1$$

$$\Rightarrow \text{if } r \sim N \sim 10^{-8} - 10^{-10} \Rightarrow \text{need } s \ll 10^{-8} - 10^{-10}$$

\Rightarrow bad approximation most of time. Need some other method to predict evolution in this case.