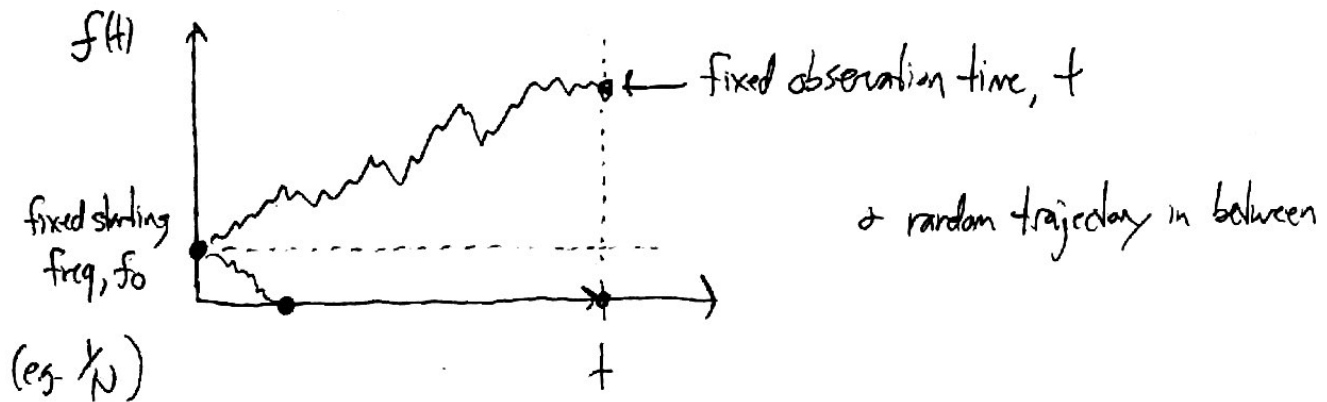


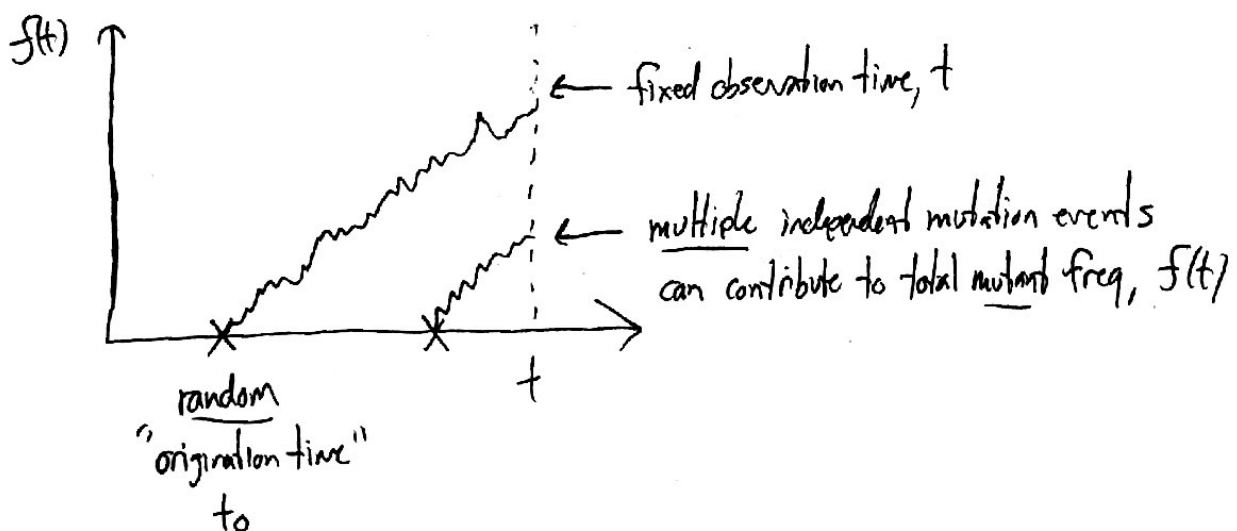
# Dynamics of Linear Branching Processes II (mutations)

①

so far, we have focused on scenarios of the form:



in practice, often interested in frequencies of mutant types that arise spontaneously due to random mutation events, whose origination time is not known:



$\Rightarrow$  to understand this case, need to understand B.P. w/ mutations:

$$\frac{df}{dt} = \mu + sf + \sqrt{\frac{f}{N}} \eta(t)$$

Due to linearity, can also write this process as sum over independent mutation events,

$$f(t) = \int_0^t dt_0 \sum_{i=1}^{\Theta(t_0)} f_i(t | f(t_0) = \frac{1}{N}) = \int_0^t dt_0 \Theta(t_0) \times \left[ \frac{1}{\Theta(t_0)} \sum_{i=1}^{\Theta(t_0)} f_i(t | f(t_0) = \frac{1}{N}) \right]$$

where  $\Theta(t_0)$  is the (random) # of mutations produced in generation  $t_0$  [ $\Theta(t_0) \sim \text{Poisson}(N\mu)$ ]

and  $f_i(t | f(t_0) = \frac{1}{N})$  is random trajectory of mutation event that occurred @ generation  $t_0$  (can predict w/ previous  $\mu=0$  results)

e.g. on average,

$$\langle f(t) \rangle = \int_0^t dt_0 \underbrace{\langle \Theta(t_0) \rangle}_{N\mu} \underbrace{\langle f(t | f(t_0) = \frac{1}{N}) \rangle}_{\frac{1}{N} e^{s(t-t_0)}} = \frac{\mu}{s} (e^{st} - 1)$$

$\downarrow$   
 same as solving  $d_t \langle f \rangle = \mu + s \langle f \rangle$

$\Rightarrow$  what about distribution of  $f(t)$ ?

$\Rightarrow$  can again return to generating function,  $H(z,t) = \langle e^{-z f(t)} \rangle$

Repeating our derivation for the  $\mu=0$  case, we find that  $H(z,t)$  satisfies the partial differential equation,

$$\frac{\partial H}{\partial t} = \left[ s z - \frac{z^2}{2N} \right] \frac{\partial H}{\partial z} - \underbrace{\mu z H}_{\rightarrow \text{new term}}$$

this time, it will be most interesting to consider an initial condition with no mutant individuals:  $H(z,0) = e^{-z \cdot 0} = 1$

this PDE can be solved using the method of characteristics (actually, a slight variant of what we did before)

can skip to solution on p. 4 below

In particular, note that if we define the function,

$$\psi(t) \equiv \log H(\underbrace{z^*(t_f - t)}_{\varphi(t)}, t_f - t), \quad \text{where } z^* \text{ is characteristic curve from } \mu=0 \text{ case}$$

then  $\psi(t)$  satisfies the ODE

$$\frac{d\psi(t)}{dt} = \mu \varphi(t) \quad w/$$

$$\begin{aligned} \psi(0) &= \log H(\varphi(0), t_f) \\ \psi(t) &= \log H(\varphi(t), 0) = 0 \end{aligned}$$

thus, we have

$$\underbrace{\psi(t)}_0 = \underbrace{\psi(0)}_{\log H(z,t)} + \int_0^t \mu \varphi(t') dt' \Rightarrow H(z,t) = e^{-\mu \int_0^t \varphi(t') dt'}$$

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plugging in for the characteristic curve, we have:

$$H(z,t) = e^{-\mu \int_0^t \frac{ze^{st'}}{1 + \frac{z}{2Ns}(e^{st'} - 1)} dt'} = e^{-2N\mu \log\left(1 + \frac{z}{2Ns}(e^{st} - 1)\right)} = e^{-2N\mu \log\left(1 + \frac{z}{2Ns}(e^{st} - 1)\right)}$$

or

$$H(z,t) = \left(1 + z \cdot \frac{e^{st} - 1}{2Ns}\right)^{-2N\mu}$$

⇒ can recognize as generating function for Gamma distribution w/  
shape  $\alpha = 2N\mu$  and scale  $(e^{st} - 1)/2Ns \equiv s_{\max}$

$$p(f) df = \frac{1}{\Gamma(2N\mu)} \left(\frac{f}{s_{\max}}\right)^{2N\mu-1} e^{-f/s_{\max}} \left(\frac{df}{s_{\max}}\right)$$

↳ dynamic version of mutation-selection-drift balance

What does this distribution look like?

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From properties of Gamma dist'n, know that

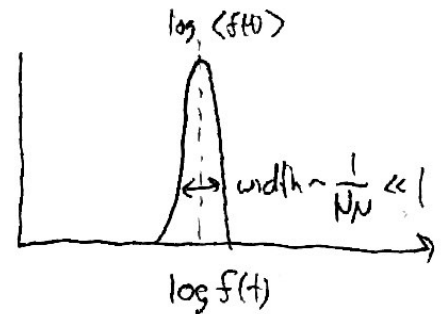
$$\langle f(t) \rangle = \alpha f_{\max} = 2N\mu \cdot \left( \frac{e^{st} - 1}{2Ns} \right) = \frac{\mu}{s} (e^{st} - 1) \quad \left[ \begin{array}{l} \text{same as from} \\ \text{SOE} \end{array} \right]$$

$$\text{Var}(f(t)) = \alpha f_{\max}^2 = \frac{1}{2N\mu} \langle f(t) \rangle^2 \Rightarrow \boxed{C_v(t) = \frac{1}{2N\mu}}$$

$\Rightarrow$  unlike single trajectory, no longer a transition w/ t.

Instead, like equilibrium mut-sel-drift balance, 2 very diff behaviors depend on  $N\mu$

case 1: when  $N\mu \gg 1$ , dist'n strongly peaked around characteristic value,  $\langle f(t) \rangle$ :



e.g. for deleterious mutations ( $s < 0$ )

$$\langle f(t) \rangle = \frac{\mu}{|s|} (1 - e^{-|s|t}) \rightarrow \bar{f} = \frac{\mu}{s}$$

"deterministic mut-sel balance"

can understand from integral eq'n:

$$f(t) = \int_0^t dt_0 \underbrace{\Theta(t_0)}_{\text{law of large #s}} \times \underbrace{\left[ \frac{1}{\Theta(t_0)} \sum_{i=1}^{G(t_0)} f_i(t | f_i(t_0) = \frac{1}{N}) \right]}_{\text{deterministic value}}$$

law of large #s:  $\langle \Theta(t_0) \rangle = N\mu$   $\times$   $\langle f_i(t | f_i(t_0) = \frac{1}{N}) \rangle = \frac{1}{N} e^{s(t-t_0)}$  = deterministic value

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can calculate spread w/ perturbative approach:

let  $f(t) = \bar{f} + \delta f(t)$  w/  $\delta f(t) \ll \bar{f}$ . then plugging into SDE, obtain:

$$\frac{d(\delta f)}{dt} = -|s| \delta f(t) + \sqrt{\frac{\bar{f}}{N}} \eta(t) \Rightarrow \text{classic physics problem}$$

(Brownian particle in quadratic potential)  
w/  $r = |s|$ ,  $D = \bar{f}/N$ .

$$\Rightarrow \text{Var}(\delta f) \sim \sqrt{\frac{D}{r}} \sim \frac{\bar{f}}{N} \sqrt{\dots}$$

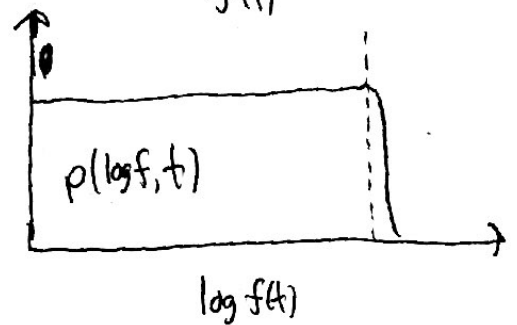
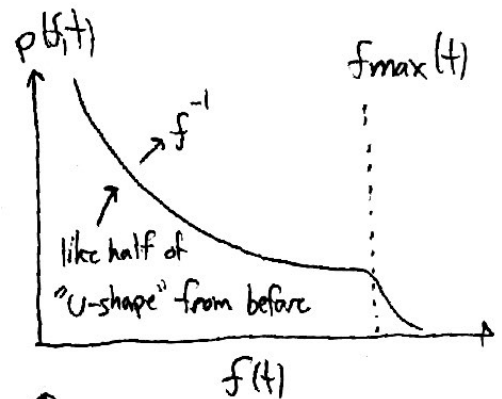
$\Rightarrow$  Note: this general approach (almost) always works when  $f$  is close to avg value.  
(might recognize as mean-field theory from stat phys.)

case 2: In opposite extreme ( $N\mu \ll 1$ )  
dist'n of  $f$  becomes very broad:

$$p(f, t) \approx 2N\mu f^{-1} e^{-f/f_{\max}}$$

$\Downarrow$  or, in log space

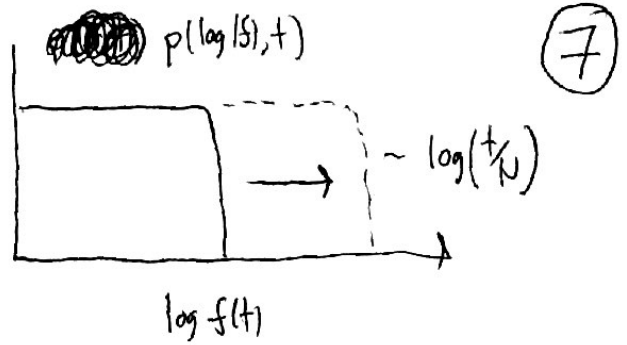
$$p(\log f, t) \approx 2N\mu e^{-\log(f/f_{\max})} \approx 2N\mu \Theta(f_{\max} - f)$$



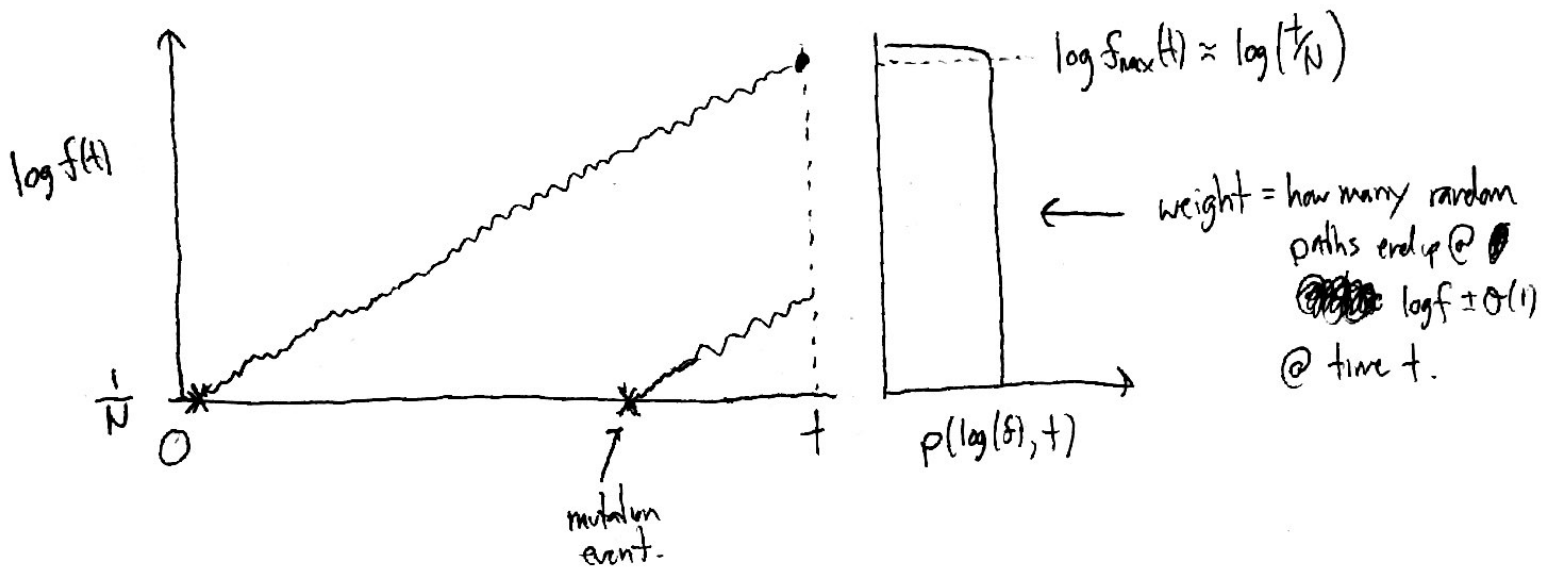
Interpretation: most weight near  $f=0$ , but w/ prob  $\sim N\mu$  considerable chance of observing any order of magnitude between 0 &  $f_{\max}$   
(little chance of  $f \gg f_{\max}$ )

e.g. for neutral mutation,  $f_{\max}(t) = t/2N$

$$\Rightarrow p(f, t) = \frac{2N\mu}{f} e^{-\frac{2N\mu}{f}}$$



\* can understand dist'n as contribution from @ most 1 random mutation event between 0 and t. Helps to visualize as:



$\Rightarrow$  can see that  $f_{\max} \sim t/2N$  is largest size mutation could have reached if it occurred @ earliest possible time ( $t=0$ ).

$\Rightarrow$  hence, little probability of seeing  $f \gg f_{\max}$ .

$\Rightarrow$  for  $f \ll f_{\max}$ , we see that to contribute to  $\log f \pm O(1)$ , must have arisen at least  $\sim Nf$  generations earlier, so that there is time to drift from  $1/N \rightarrow f$ .

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What range of times contribute?

$\Rightarrow$  since  $f \sim \frac{t}{N} \Rightarrow \log f \pm \Theta(1) \Leftrightarrow \Delta t \sim Nf$

probability of surviving until  $t$ :  $\sim \frac{1}{Nf}$

$\Rightarrow$  mutations w/  $t_0 \ll t - Nf$  are much less likely to be alive to contribute to  $\log f \pm \Theta(1)$  @ time  $t$ .

Putting everything together, have:

$$p(\log f, t) \Delta \log f \approx N\mu \times Nf \times \left(\frac{1}{Nf}\right) = N\mu \checkmark$$

✓  
prob that mutation occurs per gen.

↓  
prob that mutation occurs in right-time window to contribute to  $\log(f) \pm \Theta(1)$

↘  
prob that mutation survives long enough to reach  $\log f \pm \Theta(1)$  at time  $t$ .

$\Rightarrow$  thus, "U"-shape in  $p(f, t)$  arises because probability of surviving until  $\log f(t) \approx \log f \pm \Theta(1)$  is balanced by larger # of times that can contribute.   
↳ origin.

What about selected mutations?

$\Rightarrow$  just like single trajectory, selected mutations indistinguishable from neutral mutations when  $t \ll \frac{1}{s}$  (since  $f_{\text{max}}(t) \sim \frac{t}{N}$ )



For deleterious mutations ( $s < 0$ ),  $f_{max}(t) \rightarrow \frac{1}{2N|s|}$

$\Rightarrow p(f,t) \rightarrow p(f) = \frac{2N\mu}{f} e^{-2Ns f}$

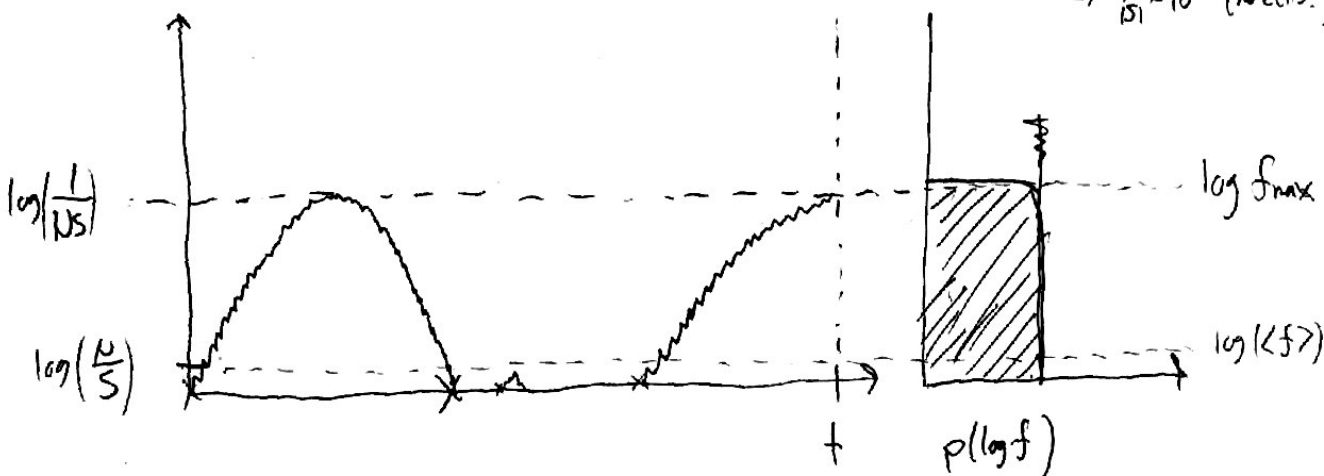
$\Rightarrow$  i.e. mostly zero, but w/ prob  $\sim N\mu$  chance to grow as large as  $\sim \frac{1}{|s|}$

$\Rightarrow$  very different than avg!

e.g. ABX resistance mut w/  $|s| \sim 10^{-2}$ ,  $N \sim 10^6$ ,  $\mu \sim 10^{-10}$

$\Rightarrow \frac{\mu}{|s|} \sim 10^{-8}$  (no cells!)  $\frac{1}{Ns} \sim \frac{1}{10^4}$  (100 cells!)

can again visualize w/ paths:

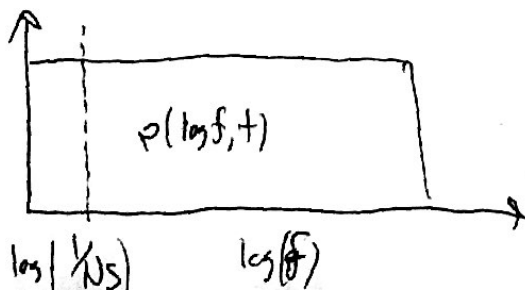


Now mutations can't get much larger than  $\frac{1}{|s|}$ , can't survive much longer than  $\frac{1}{|s|}$

$\Rightarrow$  mutations can only contribute if they arose in last  $\sim \frac{1}{|s|}$  gens.

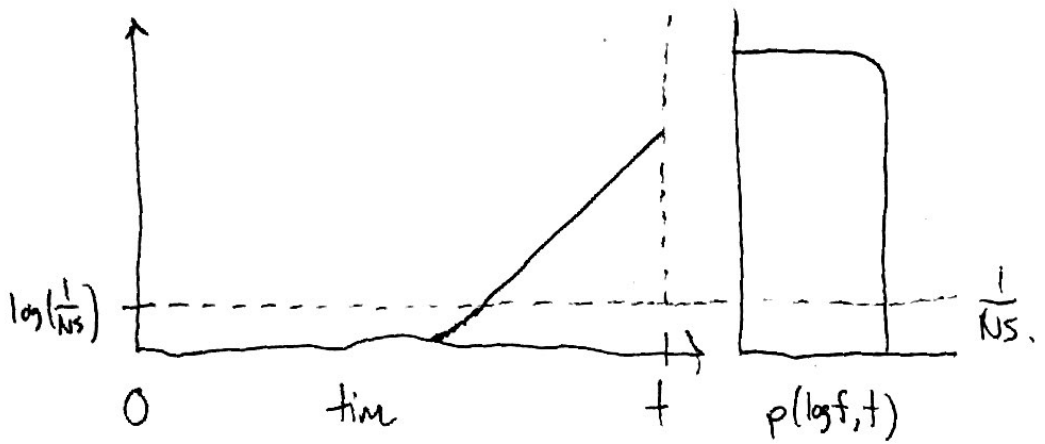
$\Rightarrow$  hence like neutral case w/  $t_{eff} \sim \frac{1}{|s|}$

Finally, for beneficial mutations,  $f_{max}(t) \rightarrow \frac{e^{st}}{2Ns} \gg \frac{1}{Ns}$



$\Rightarrow$  weird... knew that  $f(t)$  not drifting when  $f \gg \frac{1}{Ns}$ .

can again understand from picture:



mutations grow as  $f(t) \sim \frac{1}{Ns} e^{s(t-t_0)} \Rightarrow \log f \pm O(1) \Leftrightarrow \Delta t = \frac{1}{s}$

Putting it together:  $p(\log f, t) \Delta \log f = N\mu \times (\frac{1}{s}) \times (s) = N\mu \checkmark$

prob that mutation occurs in right time window to contribute to  $\log f \pm O(1)$

prob that mutation survives drift to reach  $\log f \pm O(1)$

$\Rightarrow$  same as neutral dist'n, but very different reason underneath.

$\Rightarrow$  differences become important when considering full path,  $f(t)$ .

$\Rightarrow$  path of beneficial mutation  $\approx$  deterministic once  $f(t) \gg \frac{1}{Ns}$ .

i.e. for large  $t$ , expect to capture all randomness in single #,

$f(t) = v e^{st}$  (just like  $\mu=0$  case)

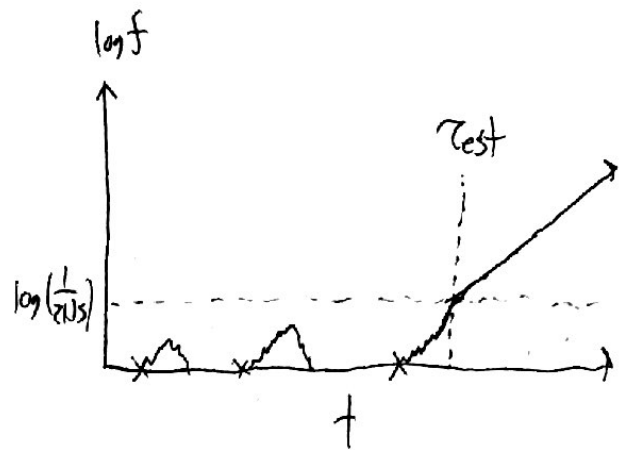
$\Rightarrow$  Find  $v \sim \text{Gamma}(2N\mu, \frac{1}{2Ns})$  for  $t \gg \frac{1}{s}$ . (independent of time  $\checkmark$ )

Often helpful to rewrite  $v$  as a time,  $f(t) = \frac{1}{2Ns} e^{s(t-\tau_{\text{est}})}$

or  $\tau_{\text{est}} \equiv \frac{1}{s} \log\left(\frac{1}{2Ns v}\right)$

\* this is known as establishment time.

$\Rightarrow$  intuitively, it's the time that  $f(t)$  would have reached  $\frac{1}{2Ns}$  if it grew deterministically for the whole time.



(i.e. roughly time that mutation arose & survived drift)

$\Rightarrow$  can show that when  $N\mu \ll 1 \Rightarrow \tau_{\text{est}} \sim \text{Exponential}\left(\frac{1}{2N\mu s}\right) \pm \mathcal{O}\left(\frac{1}{s}\right)$

Has simple interpretation: mutations occur @ rate  $N\mu$  per gen & survive drift w/ prob  $\sim s$ .

$\Rightarrow$  successful mutations occur as Poisson process w/ rate  $\sim N\mu s$ .

$\Rightarrow$  "limited by supply of new mutations" (i.e. increasing  $N$  or  $\mu$  by const factor decreases  $\tau_{\text{est}}$  by same amount)

$\Rightarrow$  same picture also helps us understand behavior when  $N\mu \gg 1$ .   
 $\Rightarrow$  in this case,  $\sim N\mu$  establishment events contribute to  $v$  (rate  $\uparrow$  time window to contribute  $\downarrow$ )  $\left(2Ns \times \frac{1}{s} = 2N\mu\right)$    
 each w/ typical size  $\sim \frac{1}{Ns} e^{st}$ .

in this case, have  $\tau_{\text{est}} = -\frac{1}{s} \log(N\mu) \pm \mathcal{O}\left(\frac{1}{s\sqrt{N\mu}}\right)$

(12)

(deterministic and negative)

$\Rightarrow$  negative because multiple mutation events contribute. initially grows much faster than  $e^{st}$

$\Rightarrow$  time it takes mutation to reach  $f=1/2$  is  $t_{1/2} \sim \frac{1}{s} \log\left(\frac{s}{\mu}\right)$

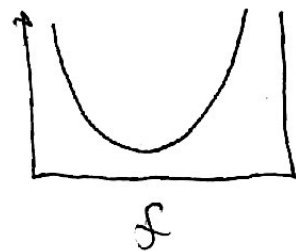
(independent of  $N$ , weakly dependent on  $\mu \Rightarrow$  limited mainly by strength of selection)

compare to  $N\mu \ll 1$  case:

$$t_{1/2} \sim \text{Exponential}\left(\frac{1}{N\mu s}\right) + \frac{1}{s} \log(NS) \sim \frac{1}{N\mu s} \quad (N\mu \ll 1/\log(NS))$$

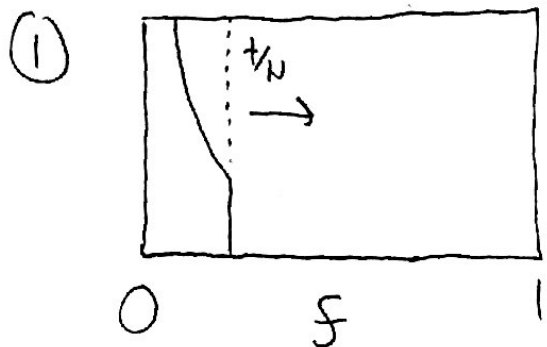
Finally, can use our new found knowledge to understand what was going on w/ "U"-shaped stationary dist'n from full single-locus model:

$$p(f) \propto f^{2N\mu-1} (1-f)^{2N\mu-1} e^{2Ns f} \xrightarrow{s=0} f^{N\mu-1} (1-f)^{N\mu-1}$$

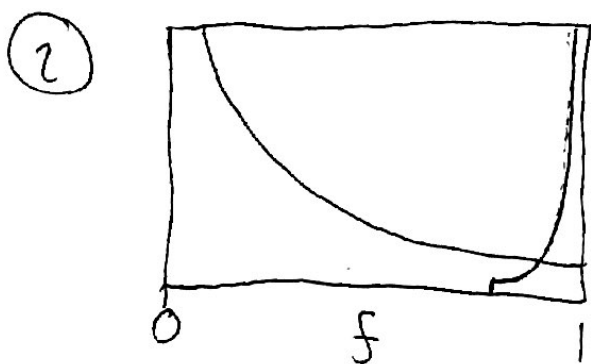


So for neutral mutations ( $s=0$ ), movie is following:

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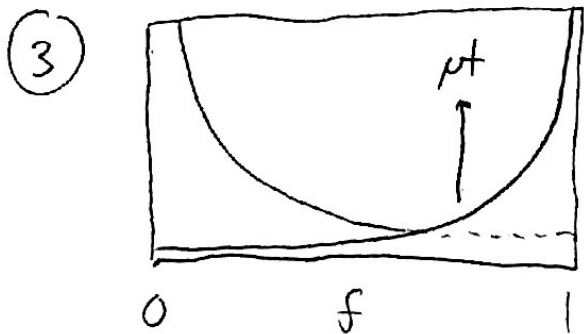


Need  $t \sim N$  generations for left ~~side~~ half ~~to form~~ of "U-shape" to form from forward mutations from  $f=0$ .



Now chance for ~~right~~ right half of "U-shape" to form from back-mutations from  $f=1$

↳ but initially, ~~the~~ height of right half is small (since low probability to reach  $f=1$ )



Rate that mutations reach  $f=1$  from  $f=0$  is

$N\mu \times (\frac{1}{N}) = \mu \Rightarrow$  need  $t_{eq} \sim \frac{1}{\mu}$  generations before  $f=0$  &  $f=1$  are equally likely.

\* this time scale is super long! e.g. humans  $\sim \mu = 10^{-8}$

so  $t_{eq} \sim 10^8$  generations  $\Rightarrow$  ~~the~~  $> 1$  billion years (way longer than time since human-chimp split)

~~most recent common ancestor~~ most recent common ancestor of humans lived  $< 10^5$  years ago so not enough time for human pop to reach stationary dist'n.

Later we will see that this is true more generally:

when  $N\mu \ll 1$ , never enough time for neutral stationary dist'n to equilibrate in time since common ancestor of population.

$\Rightarrow$  instead, more relevant dist'n is quasi-stationary dist'n:

$$p(f) \approx \frac{2N\mu}{f} \quad (\text{valid for } s=0, t \gg N, \text{ but } t \ll 1/\mu)$$

compare to strongly deleterious case,

$$p(f) \approx \frac{2N\mu}{f} e^{-2N|s|f} \quad \text{valid for } N|s| \gg 1, t \gg 1/|s|$$