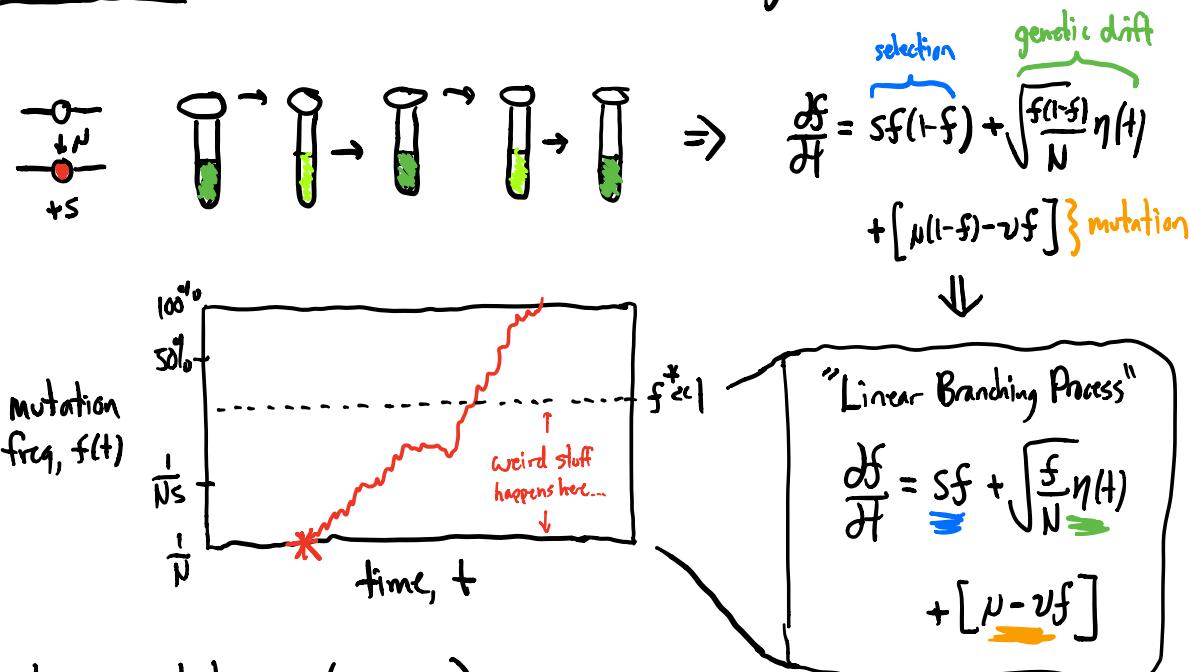


Announcements: PSET 2 DUE TUES [Anita problem session - see Slack]

Last time: Quick review - how did we get here?



w/ no mutations ($\mu=v=0$)

⇒ Generating function

$$H(z, t) \equiv \langle e^{-z f(t)} \rangle = \exp \left[- \frac{z \langle f(t) \rangle}{1 + \frac{z}{2} \langle f^2(t) \rangle G_v^2(t)} \right]$$

$\langle f(t) \rangle = f_0 e^{st}$
 $\langle f^2(t) \rangle = \frac{1 - e^{-st}}{2 N s f_0}$

Time-dependent extinction probability:

$$P_{ext}(t) \equiv \lim_{z \rightarrow \infty} H(z, t) = \exp \left[- \frac{2}{G_v^2(t)} \right] \equiv 1 - p_s(t) \leftarrow \begin{matrix} \text{"survival} \\ \text{probability"} \end{matrix}$$

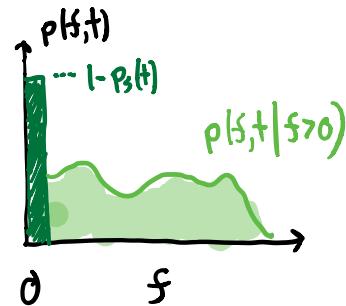
- Today:
- ① what can we learn from these formal results?
 - ② can we do the same thing by cheating ("heuristics")

$$\text{Basic behavior follows from } C_V(t) = \frac{1 - e^{-st}}{Ns_0} \approx \begin{cases} \ll 1 & \text{when } t \ll t^* \\ \gg 1 & \text{when } t \gg t^* \end{cases}$$

$$w/ \quad C_V^2(t^*) \sim 1 \quad \Rightarrow \quad t^*(N, s, f_0) \approx \begin{cases} \infty & \text{if } s > 0; f_0 \gg \frac{1}{N|s|} \quad [C_V^2(t) \text{ always } \leq 1] \\ Nf_0 & \text{if } f_0 \ll \frac{1}{N|s|} \\ \frac{1}{s} \log(N|s|f_0) & \text{if } s < 0; f_0 \gg \frac{1}{N|s|} \end{cases}$$

\Rightarrow for survival probability,

\Rightarrow can anticipate that distribution, $p(s,t)$ will be ~case 2 dist'n that is a mixture of 2 different kinds of paths:



$$p(s,t) \approx \underbrace{[1-p_s(t)]}_{\text{extinct paths}} \delta(s) + \underbrace{p_s(t) \cdot p(s,t|s>0)}_{\text{non-extinct paths}} \rightarrow \begin{matrix} \text{distribution of } s(t) \\ * \text{conditioned on survival} * \end{matrix}$$

\Rightarrow what can we say about $p(f,+|f>0)$?

easy to calculate mean by decomposing:

$$\langle f(t) \rangle = 0 \cdot (1 - p_s(t)) + p_s(t) \cdot \langle f(t) | f > 0 \rangle$$

$$\hookrightarrow \langle f(t) | f > 0 \rangle = \frac{\langle f(t) \rangle}{p_s(t)} = \begin{cases} f_0 e^{st} & \text{if } t \ll t^* \\ \frac{e^{st} - 1}{2Ns} & \text{if } t \gg t^* \end{cases}$$

↖ independent of f_0 !

Depending on selection coefficient, $t \gg t^*$ limit looks like:

$$\langle f(t) | f > 0 \rangle \xrightarrow{t \gg t^*} \begin{cases} \frac{e^{st}}{2Ns} & \text{if } s > 0; t \gg \frac{1}{s} \\ \frac{1}{2N} & \text{if } t \ll \frac{1}{|s|} \\ \frac{1}{2N|s|} & \text{if } s < 0; t \gg \frac{1}{|s|} \end{cases}$$

← looks like deterministic case, w/ different $f_0 \approx \frac{1}{2Ns}$

← grows linearly in time.

← saturates @ const value.

corresponding survival probabilities are:

$$p_s(t) \xrightarrow{t \gg t^*} \begin{cases} 2Ns f_0 & \text{if } s > 0; t \gg \frac{1}{s} \\ 2Ns_0 / t & \text{if } t \ll \frac{1}{|s|} \\ 2Ns |s| f_0 e^{-\frac{1}{2Ns} t} & \text{if } s < 0; t \gg \frac{1}{|s|} \end{cases}$$

← saturates @ constant value.

perfectly set up so that

$$\langle f(t) | f > 0 \rangle \cdot p_s(t) = \langle f(t) \rangle$$

\Rightarrow can use same argument for full distribution, $p(f, t | f > 0)$
 via the generating function:

$$H(z, t) = \langle e^{-z \cdot f(t)} \rangle = [1 - p_s(t)] \cdot e^{-z \cdot 0} + p_s(t) H(z, t | f > 0) \quad ?$$

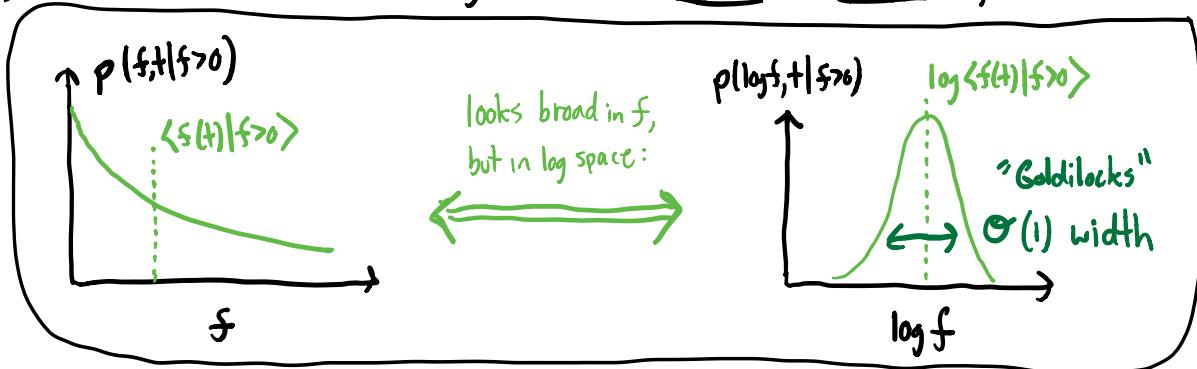
\Downarrow rearrange to obtain:

$$H(z, t | f > 0) = \frac{H(z, t) - [1 - p_s(t)]}{p_s(t)} = \frac{e^{-\frac{z \langle f(t) \rangle}{1 + \frac{z \langle f(t) \rangle}{2} C_V^2(t)}} - e^{-\frac{z}{C_V^2(t)}}}{1 - e^{-\frac{z}{C_V^2(t)}}}$$

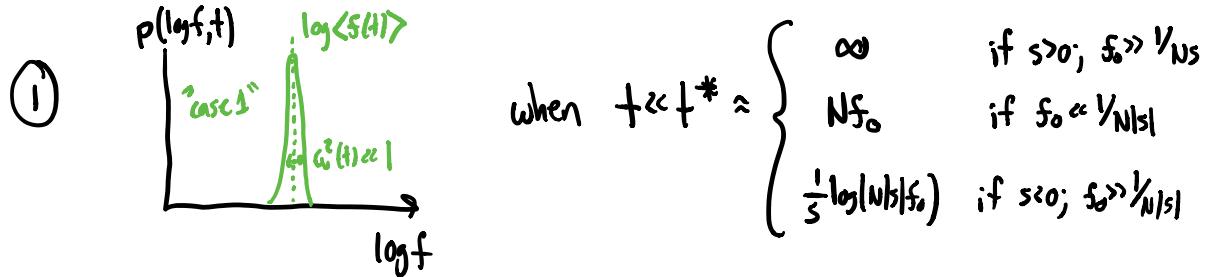
\Rightarrow when $t \gg T^*$ ($C_V^2(t) \gg 1$), can Taylor expand exponentials

$$\stackrel{(p \parallel, in notes)}{\Rightarrow} H(z, t | f > 0) \xrightarrow{t \gg T^*} \left(1 + z \cdot \underbrace{\frac{\langle f(t) \rangle C_V^2(t)}{2}}_{{\langle f(t) | f > 0 \rangle \text{ from before}}} \right)^{-1} \sim \frac{1}{1 + az}$$

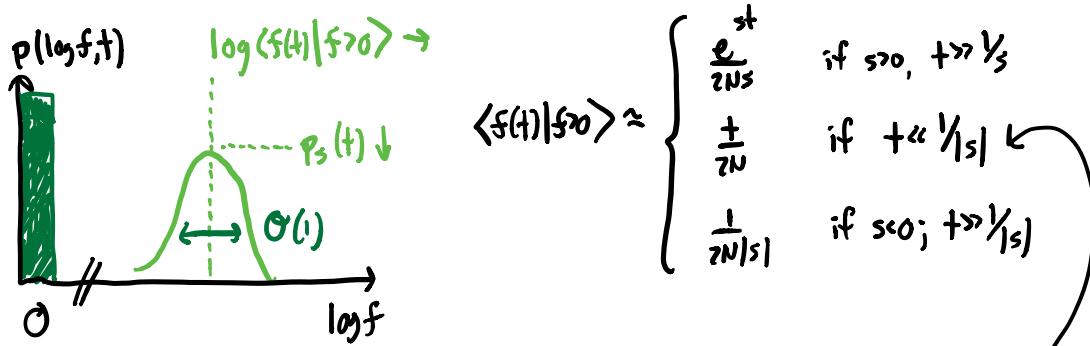
By "method of wikipedian", recognize as exponential distribution, $p(u) \propto e^{-u/\langle u \rangle}$



Putting everything together, we have:



② when $t \gg t^*$



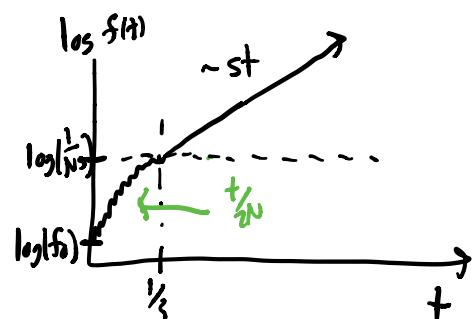
\Rightarrow if $t^* \ll t \ll 1/s$, then

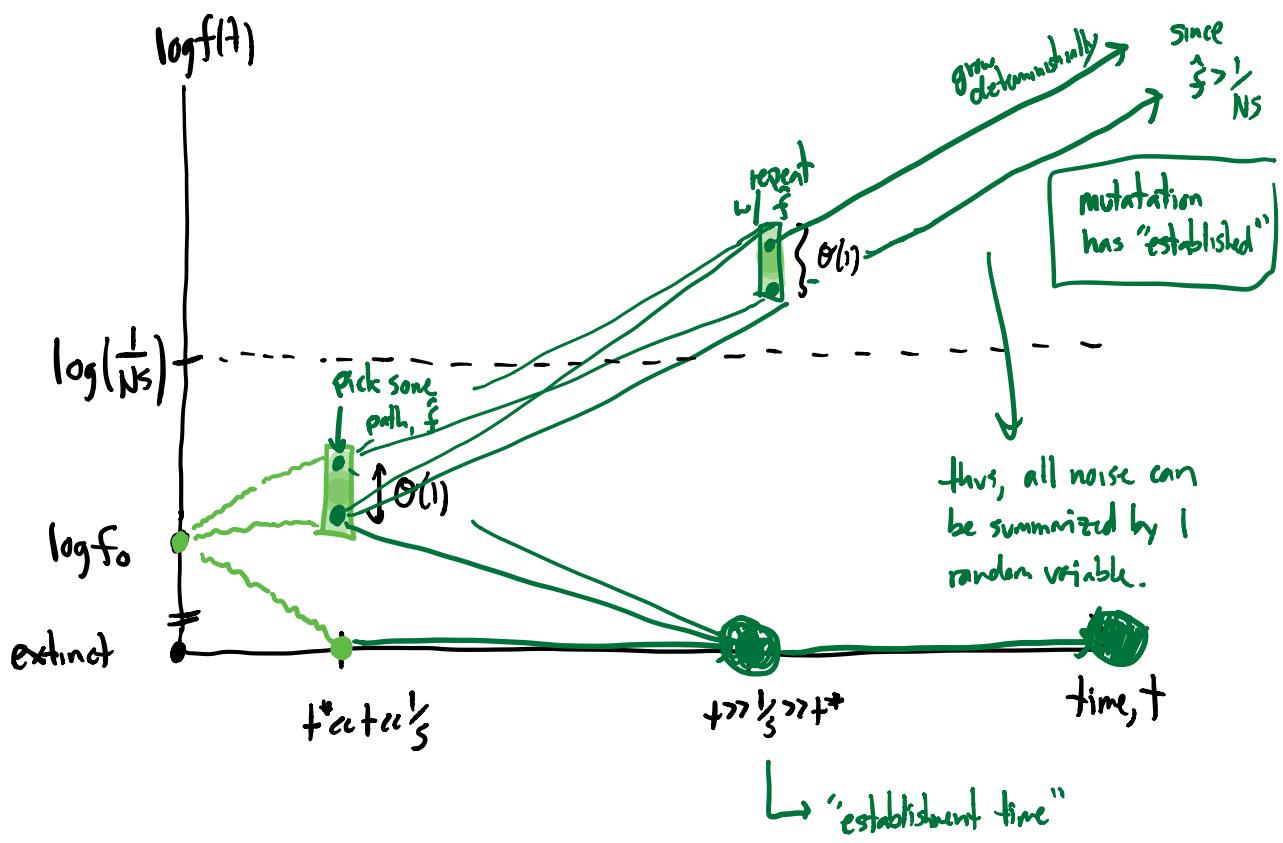
\Rightarrow mutations look neutral ($s=0$) even when $Ns \gg 1$!

\Rightarrow need $\sim Nf$ generations to go from $f_0 \rightarrow f$

For beneficial mutation ($s > 0$):

faster than deterministic @
early times ($t \ll 1/s$, $f(t) \ll 1/Ns$)





$$\Rightarrow \text{for } s > 0, +\gg \frac{1}{\zeta}, \quad f(t) = v e^{st}$$

↳ "random variable"
(but const in time)

$$\begin{aligned} \Rightarrow H_v(z) &= \langle e^{-z \cdot v} \rangle = \langle e^{-z \cdot e^{-st}} \cdot f(t) \rangle \\ &= H_f(z e^{-st}, +) \xrightarrow{+\gg \frac{1}{\zeta}} \left(1 + \frac{z}{2Ns}\right)^{-1} \end{aligned}$$

$$\Rightarrow v \sim \text{Exponential}\left(\frac{1}{2Ns}\right) \Rightarrow v \sim \frac{1}{2Ns} \cdot \underbrace{\text{Exponential}(1)}_{\text{"c"}}$$

Get full picture w/ "asymptotic matching"

Step 1: pick some time t_i s.t. $t_i \gg \frac{1}{s}$
but $f(t_i) \ll \frac{1}{2}$

$$\Rightarrow \text{need } s: \left[\frac{1}{s} \ll t_i \ll \frac{1}{s} \log(Ns) \right]$$

$$\Rightarrow @ \text{time } t_i, f(t_i) = v e^{st_i}$$

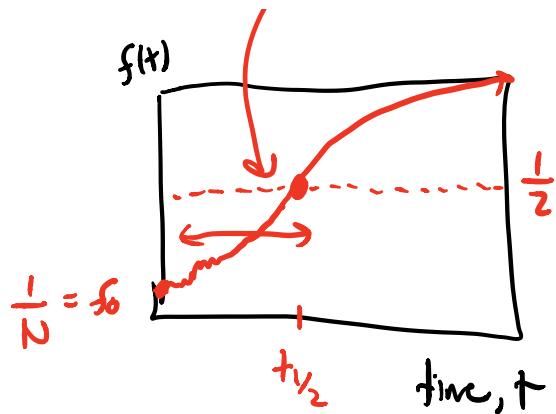
Step 2: use t_i as starting point for deterministic model:

$$\left(\partial_t f = sf(1-f) \right)$$
$$\Rightarrow f(t) = \frac{f(t_i) e^{s(t-t_i)}}{f(t_i) e^{s(t-t_i)} + 1 - f(t_i)} \approx \frac{f(t_i) e^{s(t-t_i)}}{f(t_i) e^{s(t-t_i)} + 1}$$

Step 3: plug in for $f(t_i) = v e^{st_i}$

$$\Rightarrow \boxed{f(t) = \frac{v e^{st}}{v e^{st} + 1}} \Rightarrow \boxed{\text{independent of time } t_i! \checkmark}$$

A red curved arrow points from the left side of the first box to the right side of the second box.



How long do go from $f_0 = \frac{1}{N}$
(new mutation) to $f = \frac{1}{2}$?

$$\Rightarrow f(t_{1/2}) = \frac{ve^{st_{1/2}}}{ve^{st_{1/2}} + 1} = \frac{1}{2}$$

$$\Rightarrow t_{1/2} = \frac{1}{s} \log\left(\frac{1}{v}\right) = \underbrace{\frac{1}{s} \log(Ns)}_{\text{deterministic part.}} + \underbrace{\log\left(\frac{2}{c}\right)}_{\text{stochastic part.}}$$

$$\approx \frac{1}{s} \log(Ns) \pm O\left(\frac{1}{s}\right)$$

$$\Rightarrow \text{fixation time, } T_{\text{fix}} = 2t_{1/2} = \frac{2}{s} \log(Ns) \pm O\left(\frac{1}{s}\right)$$

