

Some ticks of the time

1) Variable returns (works equally as well as variable costs)

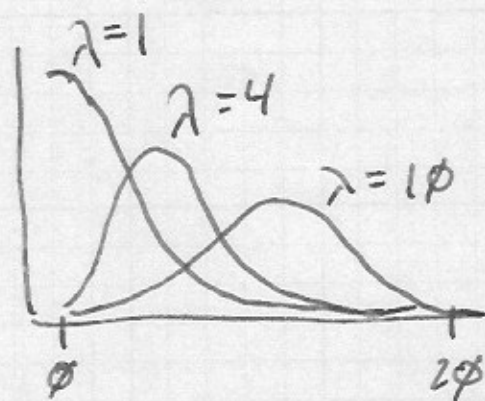
Before, we had!

	Patel 1	2	3
$p =$	<del>1</del>	3	5
	1		

Let's assume that these are means of ~~probabi~~ a gain that follows a probability distribution

- We are exploring a landscape, and  $p$  scales with the number of encounters that we have with a particular food.
- The number of encounters within a particular period of time  $M_i = \frac{\lambda_i}{m}$  follows a Poisson process where the probability is:

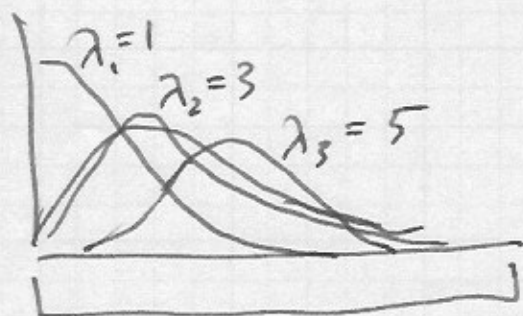
$$f_{M_i}(m_i | \lambda_i) = e^{-\frac{\lambda_i}{m} t} \frac{(\frac{\lambda_i}{m} t)^{m_i}}{m_i!}$$



Say  $p_i = (1, 3, 5)$  is the mean gain,  
and each encounter results in +1 gain.

So gain  $\sim$  encounter, and  $p_i \sim \lambda_i$

So:



$M_{\max} = 15$

Now we have to account for the probability  
of finding different gains during a foraging bout

patch 1  $M_i$

0	0.36
1	0.36
2	0.18
3	0.06
4	0.01
5	...

$dpois(0, \lambda_1)$

$dpois(1, \lambda_1)$

$dpois(2, \lambda_1)$

$\leftarrow$  goes to zero

$M_{\max}$

Because we cut off  $\text{prob}(M_i = m_i)$  @  $M_{\max} = 15$ , we  
have to ensure ~~the total~~ our revised probabilities  
sum to 1

$$p_i(M_i = m_i) \quad \text{Pr}(M_i = m_i) = \frac{p_i}{\sum_{m_i=1}^{M_{\max}} p_i}$$

14.3

Returning ~~as~~ to our canonical equation,  
we have

$$S(x, t) = \max_i \left\{ (1-d_i) f_i S(x - c_i + p_i, t+1) + (1-d_i)(1-f_i) S(x - c_i, t+1) \right\}$$

Survival probability

@  $x$  from  $t=t \rightarrow T$  if 1) Don't die, AND find food

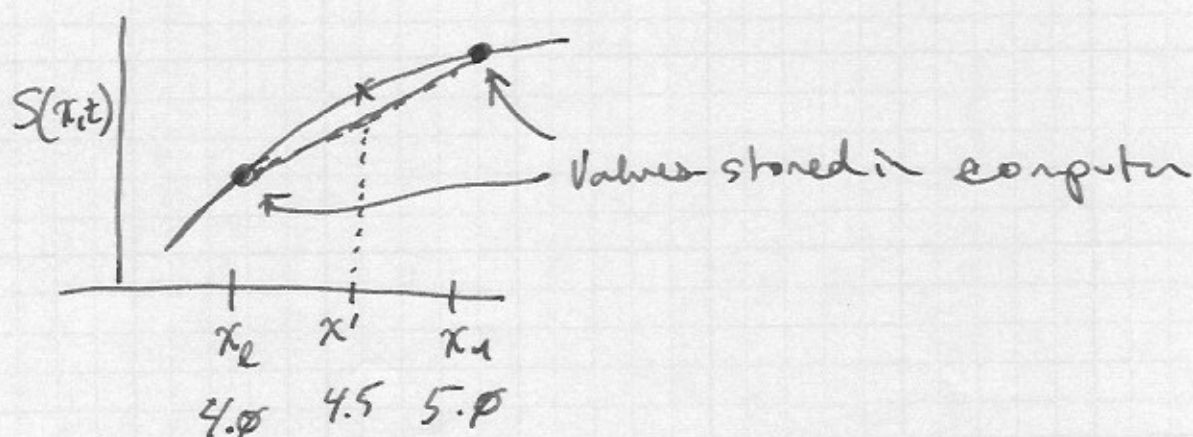
~~AND~~ OR 2) Don't die, AND find  $\phi$  food

we want: 1) Don't die, AND find food w/  $n_1$  value  
OR 2) Don't die, AND find food w/  $n_2$  value  
:  
etc.

$$S(x, t) = \max_i \left\{ (1-d_i) \sum_{n=\phi}^{M_{\max}} \text{pr}(n_i = n) S(x - c_i + n_i, t+1) \right\}$$



14.4 Now consider Noninteger states  
 -  $p_i$  is not an integer...



$x_l$  = integer part of  $x'$   
 - make sure  $x_l > x_c$

$\text{floor}(x')$

$x_u = x_l + 1$

$\text{floor}(x') + 1$

If  $\text{diff}(x_u, x_l)$  is small, we can assume  $x'$   
 $(x_u - x_l)^2$  falls on the line in between

= Linear interpolation

$q_x = x' - x_l$       ⊗ weight  $x_u, x_l$  based on  $q_x$

$$S(x', t) = q_x S(x_u, t) + (1 - q_x) S(x_l, t)$$

A new interpretation: Per-period fitness

- When patch  $i$  is visited, state changes as before, but in addition, there is a contribution to total ~~per~~ reproduction  $\psi_i(x)$

$W(x, t) = \text{maximum expected accumulated fitness b/w } t \text{ and } T \text{ given that } X(t) = x$

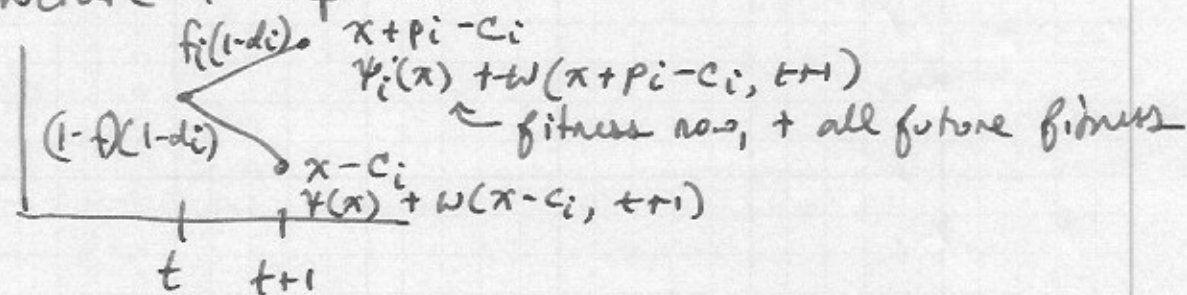
Stochastic events of survival, finding food

Sum of fitness gains by visiting patches

if  $T = \text{last chance to accumulate fitness}$   
 then  $W(x, T) = \max_i \psi_i(x)$

if  $T-1 = \text{last chance}$ , then  $W(x, T) = \emptyset$

Patch structure + reproduction



$$W(x, T) = \max_i \psi_i(x)$$

$$W(x, T) = \emptyset \text{ if } x \leq x_c \text{ for all } t$$

$$W(x, t) = \max_i \left[ d_i \psi_i(x) + (1-d_i) f_i \left\{ \psi(x) + W(x + p_i - c_i, t+1) \right\} + (1-d_i)(1-f_i) \left\{ \psi(x) + W(x - c_i, t+1) \right\} \right]$$

14.6

After Algebra:

$$= \max_i \left[ \psi(\pi) + (1-d_i) f_i \{w(x+p_i-c_i, t+1)\} + \right. \\ \left. (1-d_i)(1-f_i)w(\pi-c_i, t+1) \right]$$



15.1

15.1

$$W(x, T) = \max_i \psi_i(x)$$

$$\hookrightarrow i^*(x, T)$$

$Q(x, t)$  = expected accumulated reproduction between  $t$  &  $T$  for an organism always going to patch  $i^*(x, T)$  ~ denote  $i^*$

What is  $Q(x, T)$ ?

$$Q(x, T) = \psi_{i^*}(x)$$

$$Q(x, t) = \psi_{i^*}(x, t) + d_{i^*}(x, t) \cdot 0 + (1 - d_{i^*}(x, t)) f_{i^*}(x, t) \times Q(x + p_{i^*}(x, t) - c_{i^*}(x, t), t+1) + (1 - d_{i^*}(x, t))(1 - f_{i^*}(x, t)) Q(x - c_{i^*}(x, t), t+1)$$

$$\frac{W(x, 1)}{Q(x, 1)} \geq 1 \quad \text{if close to 1, not much selection pressure for optimal decision-making}$$

Maybe we should consider some cost of cognitive load.

Compare  $Q(x, 1)$  with  $W_S(x, t)$

$$W_S(x, t) = \max_i \left[ \psi_i(x) + (1 - d_i)(1 - f_i) W_S(x - c_i - s + p_i, t+1) + (1 - d_i)(1 - f_i) W_S(x - c_i - s, t+1) \right]$$

$$\text{where } W_S(x, 1) \geq Q(x, 1)$$

depending on  $s$

where  $\min(s)$  where  $W_S(x, 1) > Q(x, 1)$

is the max. cognitive load that would support evolution of optimal decision-making

## 15.2 Random patch Selection

$R(x, t)$  = accumulated reproduction from  $t \rightarrow T$   
 given  $X(t) = x$  and Random patch choice

~ equal prob of choosing any patch

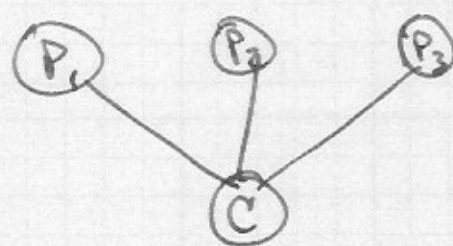
$$R(x, T) = \frac{1}{3} \sum_i \psi(x) + \text{B.C.}$$

$$R(x, t) = \frac{1}{3} \left\{ \sum_i \left[ \psi(x) + (1-d_i) f_i R(x + p_i - c_i, t+1) + (1-d_i)(1-f_i) R(x - c_i, t+1) \right] \right\}$$

Now compare  $W(x, 1)$   $Q(x, 1)$   $W_S(x, 1)$ ,  $R(x, 1)$

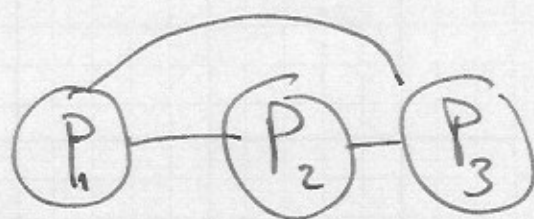
Now suppose we want to model movement rules into patches

Before



state: energy  $x$

Now



state: energy  $x$

where you are  
 & where you are going  $s$

$(x, s)$

$d_{ij}$  = cost of mortality while traveling from patch  $i$  to  $j$

$c_{ij}$  = cost of moving from patch  $i$  to patch  $j$



15.3

$$W(x, i, T) = \max_j \gamma_j (x - c_{ij}) (1 - d_{ij})$$

$$W(x, i, T) = \max_j \left[ \gamma_j (x - c_{ij}) + (1 - d_{ij})(1 - f_j) \gamma_j \right.$$

$$+ (1 - d_{ij})(1 - d_j) f_j W(x - c_{ij} - c_j + p_j, j, t+1)$$

$$\left. + (1 - d_{ij})(1 - d_j)(1 - f_j) W(x - c_{ij} - c_j, j, t+1) \right]$$

Now we will have a  $(x_{\max} - x_c) \times (\# \text{ patch}) \times T$  array

for  $W$   
 $\alpha D$