

Variables

- x population size, later vector of population sizes of different types
- x' population sizes at next timestep
- t time
- x_t population sizes at time t
- w growth rate of population per timestep
- W matrix of growth rates. Here W is usually diagonal, and $W_{i,i}$ is the growth rate of type i
- s fitness. $w = (1 + s)$. s is simply a more convenient way to represent fitness when it is close to 1
- μ mutation rate. Rate of switching away from a type per timestep
- M mutation matrix. Entry $M_{i,j}$ row i column j is the rate of switching from j to i per timestep
- D diagonal matrix
- V Matrix of eigenvectors of $W \times M$. We just find V and D so that $W \times M = V \times D \times V^{-1}$

Eigen example with long genome

L length of genome. Could also be length of a part of the genome that we are considering.

i index for type with i mutations vs. the optimal type.

x_i population size for type i

ν mutation rate per site. Chance of a single site to change in a single timestep. Assuming independence, global rate would be $1 - (1 - \nu)^L$, but I assume it is simply νL .

$p_{i \rightarrow i+1}$ Chance to mutate from i mutations to $i + 1$, to add one additional mutation

$p_{i \rightarrow i-1}$ Chance to mutate from i mutations to $i - 1$, to correct one mutation

M mutation matrix. $M_{i,i+1} = p_{i \rightarrow i+1}$, $M_{i,i-1} = p_{i \rightarrow i-1}$

W fitness matrix for genome. I assume that only the optimal type has fitness > 1 , so $W_{1,1} = (1 + s)$

```
> library(expm); options(warn=-1)
```

Growth of single population

$$x' = w \cdot x$$

Therefore:

$$x_t = w^t x_0$$

Example

```
> w = 0.9 ;  
x0 = 1
```

```
w=0.9;x0=1
```

```
> T = 1...20 ; x = rep(0, length(T));  
for(t in T) x_t = w^t x0
```

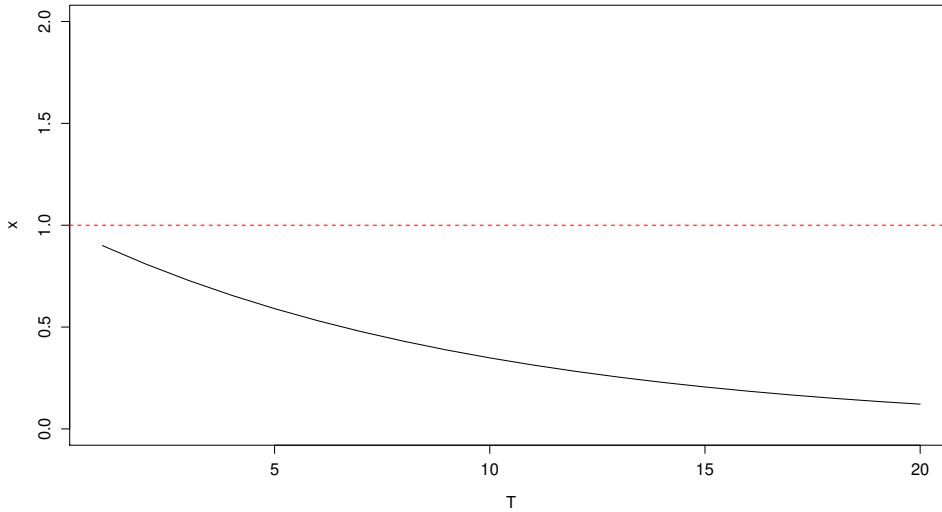
```
T=1:20;x=rep(0,length(T));for(t in T) x[t]=w^t*x0
```

```
> y = x
```

$y=x$

```
> plot_xt()
```

```
plot_xt()
```



```
>
```

We can represent two types as a vector

```
> x0 =  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ 
```

```
x0=(t(matrix(c(1, 1), 1, 2 )))
```

```
>
```

The growth rates can then be represented by a diagonal matrix

```
> W =  $\begin{pmatrix} 1.1 & 0 \\ 0 & 0.95 \end{pmatrix}$ 
```

```
W=(t(matrix(c(1.1, 0, 0, 0.95), 2, 2 )))
```

```
>
```

$$x' = Wx$$

```
> W %*% x0
```

```
> W %*% x0
```

```
W %*% x0
```

[, 1]	
[1,]	1.1
[2,]	0.95

```
>W %*% W %*% x0
```

```
W %*% W %*% x0
```

[, 1]	
[1,]	1.21
[2,]	0.9025

```
>
```

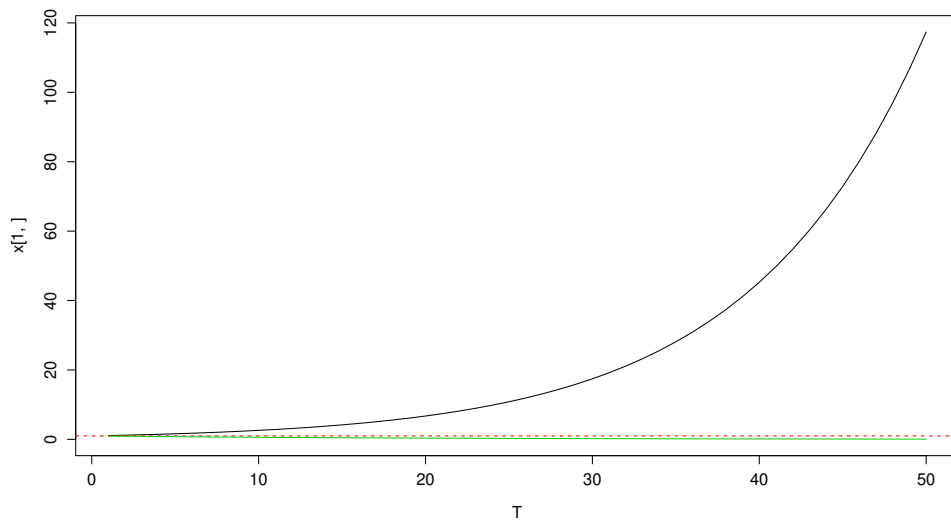
$$x_t = W^t x_0$$

```
>T = 1...50 ; x = matrix(0, 2, length(T));  
  for(t in T) x[,t] = (W%^t) %*% x0
```

```
T=1:50;x=matrix(0,2,length(T));for(t in T) x[,t]=(W%^t) %*% x0
```

```
> plot_xt2()
```

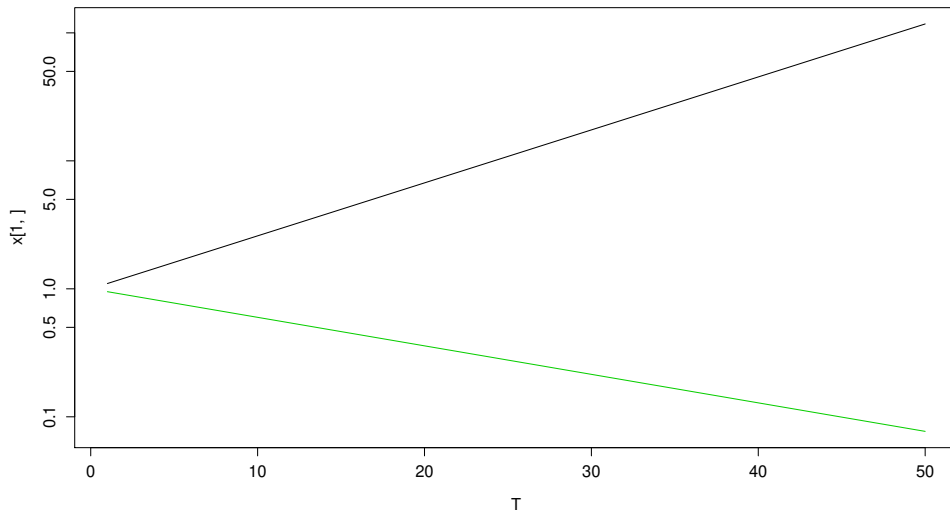
```
plot_xt2()
```



```
> plot_log_xt2()
```

```
>
```

plot_log_xt2()



>

Mutations are represented by a stochastic matrix

```
> mu = 0.1;
```

$$M = \begin{pmatrix} 1 - \mu & 0 \\ \mu & 1 \end{pmatrix}$$

```
mu=0.1;M=(t(matrix(c(1-mu, 0, mu, 1), 2, 2 )))
```

```
> x0 =  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ ; x0
```

```
x0=(t(matrix(c(1, 1), 1, 2 )));x0
```

	[,1]
[1,]	1
[2,]	1

```
>M %*% x0
```

```
M %*% x0
```

	[,1]
[1,]	0.9
[2,]	1.1

```
>
```

Now

$$x' = W \times M \times x$$

or

$$x_t = (W \times M)^t x_0$$

If we diagonalize $W \times M$:

$$W \times M = V \times D \times V^{-1}$$

with D diagonal, and V the eigenvectors of $W \times M$

$$(W \times M)^t = V D^t V^{-1}$$

D^t is dominated by the largest eigenvalue.

```
> x0 = ( 1 ); mu = 0.2; M = ( 1 - mu  0 ); W = ( 1.1  0 )
      ( 1 );                               ( 0  0.95 )
```

```
<, 2, 2 )); W=(t(matrix(c(1.1, 0, 0, 0.95), 2, 2 )))
```

```
> T=1:150;x=matrix(0,2,length(T))
  for(t in T) x[,t]=((W**M) ^t) ** x0
```

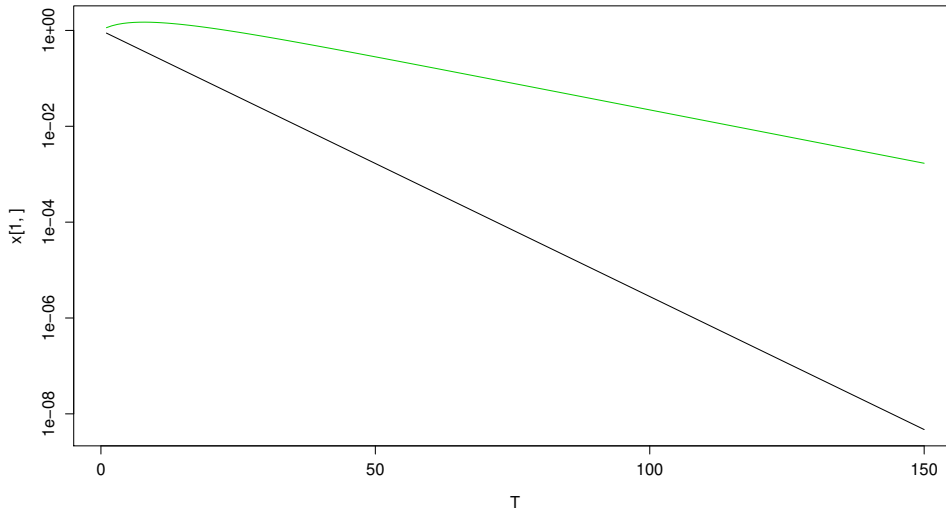
```
T=1:150;x=matrix(0,2,length(T))
```

```
> for(t in T) x[,t]=((W**M) ^t) ** x0
```

```
T=1:150;x=matrix(0,2,length(T))
> for(t in T) x[,t]=((W%*%M) %^%t) %*% x0
```

```
> plot_log_xt2()
```

```
plot_log_xt2()
```



```
>
```

values of μ to try: 0.01, 0.08, 0.2

```
> eigen((W%*%M))
```

```
eigen((W%*%M))
```

```
eigen() decomposition
```

```
$values
```

```
[1] 0.95 0.88
```

```
$vectors
```

	[,1]	[,2]
[1,]	0	0.345705358827356
[2,]	1	-0.93834311681711

```
>
```

```
> z = sapply(seq(0, 0.2, length = 100), function(mu) { M =  $\begin{pmatrix} 1 - \mu & 0 \\ \mu & 1 \end{pmatrix}$ ; W =  $\begin{pmatrix} 1.1 & 0 \\ 0 & 0.95 \end{pmatrix}$ ; eigen((W*%M))$values }
```

```
<(1.1, 0, 0, 0.95), 2, 2 ))); eigen((W*%M))$values }
```

```
> plot(seq(0, 0.2, length = 100), z[1,], type = "l",
       ylim = c(0.8, 1.1), xlab = "mu", ylab = "Eigenvalues")
lines(seq(0, 0.2, length = 100), z[2,]); v()
```

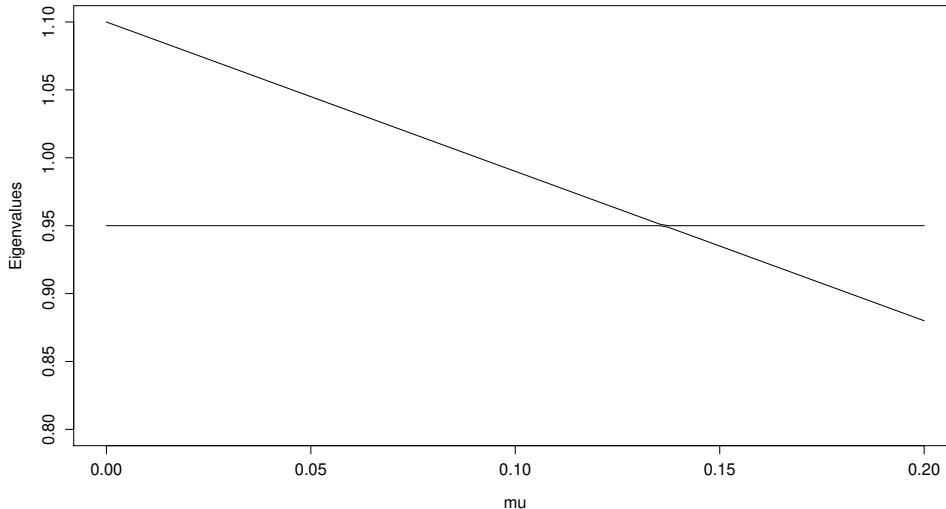
```
>
```

```
> z = sapply(seq(0, 0.2, length = 100), function(mu) { M =  $\begin{pmatrix} 1 - \mu & \mu/100 \\ \mu & 1 - \mu/100 \end{pmatrix}$ ; W =  $\begin{pmatrix} 1.1 & 0 \\ 0 & 0.95 \end{pmatrix}$ ; eigen((W*%M))$values }
```

```
> plot(seq(0, 0.2, length = 100), z[1,], type = "l",
       ylim = c(0.8, 1.1), xlab = "mu", ylab = "Eigenvalues")
lines(seq(0, 0.2, length = 100), z[2,]); v()
```

```
> plot(seq(0,0.2,length=100),z[1,], type="l",  
       ylim=c(0.8,1.1),xlab="mu",ylab="Eigenvalues")  
lines(seq(0,0.2,length=100),z[2,]);v()
```

```
plot(seq(0,0.2,length=100),z[1,], type="l",  
+     ylim=c(0.8,1.1),xlab="mu",ylab="Eigenvalues")  
> lines(seq(0,0.2,length=100),z[2,]);v()
```



```
>
```

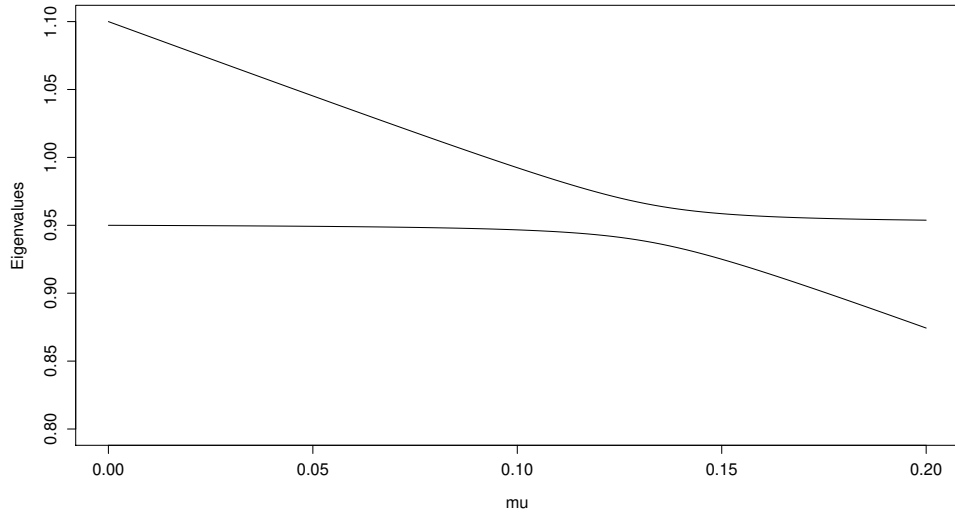
```
> z = sapply(seq(0, 0.2, length = 100), function( $\mu$ ) { $M$  =  
  ( $\begin{matrix} 1 - \mu & \mu/100 \\ \mu & 1 - \mu/100 \end{matrix}$ );  $W = \begin{pmatrix} 1.1 & 0 \\ 0 & 0.95 \end{pmatrix}$ ; eigen(( $W \%* \% M$ ))$values}
```

```
<(1.1, 0, 0, 0.95), 2, 2)); eigen(( $W \%* \% M$ ))$values}
```

```
> plot(seq(0,0.2,length=100),z[1,], type="l",  
       ylim=c(0.8,1.1),xlab="mu",ylab="Eigenvalues")  
lines(seq(0,0.2,length=100),z[2,]);v()
```



```
plot(seq(0,0.2,length=100),z[1,], type="l",  
+     ylim=c(0.8,1.1),xlab="mu",ylab="Eigenvalues")  
> lines(seq(0,0.2,length=100),z[2,]);v()
```



>

$$M = \begin{pmatrix} 1 - \mu & 0 \\ \mu & 1 \end{pmatrix}; W = \begin{pmatrix} w & 0 \\ 0 & 1 \end{pmatrix}$$

$$W \times M = \begin{pmatrix} w & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 - \mu & 0 \\ \mu & 1 \end{pmatrix} = \begin{pmatrix} (1 - \mu)w & 0 \\ \mu & 1 \end{pmatrix}$$

$$W \times M \times \begin{pmatrix} 1 \\ \alpha \end{pmatrix} = \begin{pmatrix} (1 - \mu)w & 0 \\ \mu & 1 \end{pmatrix} \begin{pmatrix} 1 \\ \alpha \end{pmatrix} = \begin{pmatrix} (1 - \mu)w \\ \mu + \alpha \end{pmatrix}$$

so to keep ratio

$$\frac{\mu + \alpha}{(1 - \mu)w} = \frac{\alpha}{1}$$

$$\mu = \alpha(w(1 - \mu) - 1)$$

$$\alpha = \frac{\mu}{w(1 - \mu) - 1}$$

only positive if $w(1 - \mu) > 1$.

If we write $w \equiv (1 + s)$ then $(1 + s)(1 - \mu) > 1$

or $s \gtrsim \mu$

Genome of length L

Assume optimal sequence is

AAGCGGCTACTGCAAACGTC...



Manfred Eigen



Peter Schuster

Genome of length L

Assume optimal sequence is

AAGCGGCTACTGCAAACGTC...

We can work with a binary genome where

- 1 identical to optimum
- 0 different from optimum

ν is mutation rate per site, optimum to non-, or back.

1111111111111000000...

Assume only one mutation per timetep, rate $L\nu$.

i number of 0s.

0s i

1s $L - i$

$$p_{i \rightarrow i+1} = \frac{L-i}{L} L\nu = (L-i)\nu$$

$$p_{i \rightarrow i-1} = \frac{i}{L} L\nu = i\nu$$

```
> nu = 0.01; L = 10
```

```
nu=0.01;L=10
```

```
>
```

$$p_{i \rightarrow i+1} = (L - i)\nu$$

$$p_{i \rightarrow i-1} = i\nu$$

```
> M=matrix(0,L+1,L+1);for(i in 1:(L+1))M[i,i]=1;
```

```
M=matrix(0,L+1,L+1);for(i in 1:(L+1))M[i,i]=1;
```

```
> for(i in 0...(L-1)) {M[i+2,i+1] = (L-i)nu; M[i+1,i+1] = M[i+1,i+1] - (L-i)nu}
```

```
for(i in 0:(L-1)) {M[i+2,i+1]=(L-i)*nu;M[i+1,i+1]=M[i+1,i+1]-(L-i)*nu}
```

```
> for(i in 1...(L)) {M[i,i+1] = i nu; M[i+1,i+1] = M[i+1,i+1] - i nu}
```

```
for(i in 1:(L)) {M[i,i+1]=i*nu;M[i+1,i+1]=M[i+1,i+1]-i*nu}
```

```
> M
```

M

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]	[,11]
[1,]	0.9	0.01	0	0	0	0	0	0	0	0	0
[2,]	0.1	0.9	0.02	0	0	0	0	0	0	0	0
[3,]	0	0.09	0.9	0.03	0	0	0	0	0	0	0
[4,]	0	0	0.08	0.9	0.04	0	0	0	0	0	0
[5,]	0	0	0	0.07	0.9	0.05	0	0	0	0	0
[6,]	0	0	0	0	0.06	0.9	0.06	0	0	0	0
[7,]	0	0	0	0	0	0.05	0.9	0.07	0	0	0
[8,]	0	0	0	0	0	0	0.04	0.9	0.08	0	0
[9,]	0	0	0	0	0	0	0	0.03	0.9	0.09	0
[10,]	0	0	0	0	0	0	0	0	0.02	0.9	0.1
[11,]	0	0	0	0	0	0	0	0	0	0.01	0.9

```
>s = 0.08; W = matrix(0, L + 1, L + 1); for(i in 0...L) Wi+1,i+1 = 1
```

```
s=0.08;W=matrix(0,L+1,L+1);for(i in 0:L) W[i+1,i+1]=1
```

```
> W1,1 = 1 + s; W
```



```
W[1,1]=1+s;W
```

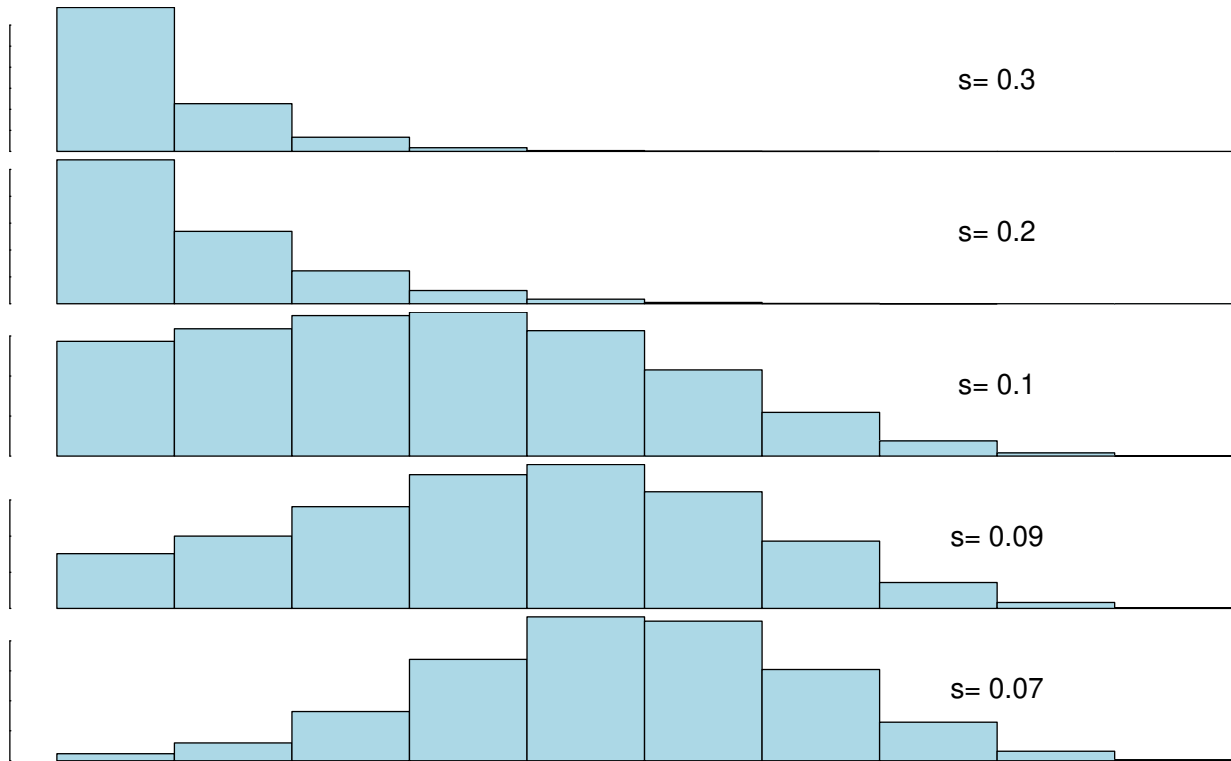
	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]	[,11]
[1,]	1.08	0	0	0	0	0	0	0	0	0	0
[2,]	0	1	0	0	0	0	0	0	0	0	0
[3,]	0	0	1	0	0	0	0	0	0	0	0
[4,]	0	0	0	1	0	0	0	0	0	0	0
[5,]	0	0	0	0	1	0	0	0	0	0	0
[6,]	0	0	0	0	0	1	0	0	0	0	0
[7,]	0	0	0	0	0	0	1	0	0	0	0
[8,]	0	0	0	0	0	0	0	1	0	0	0
[9,]	0	0	0	0	0	0	0	0	1	0	0
[10,]	0	0	0	0	0	0	0	0	0	1	0
[11,]	0	0	0	0	0	0	0	0	0	0	1

```
>round( max.eigen(W %*% M)$vector,2 )
```

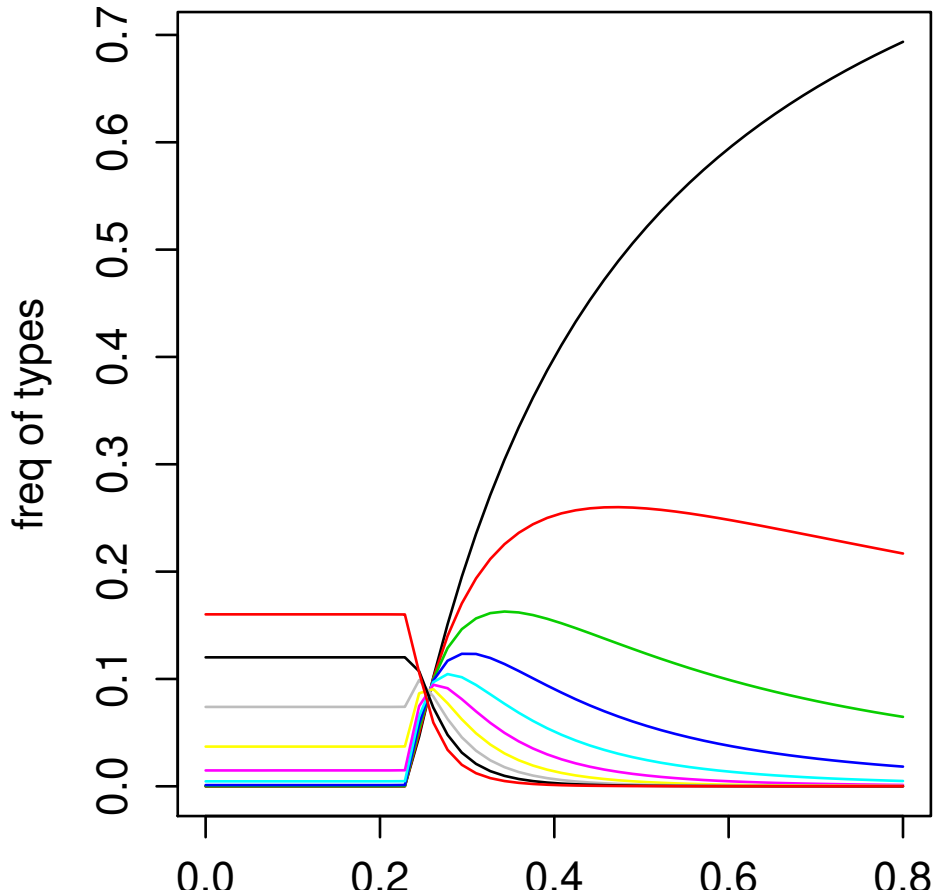
```
round( max.eigen(W %*% M)$vector,2 )
```

	freq
i=0	0.01
i=1	0.02
i=2	0.05
i=3	0.12
i=4	0.21
i=5	0.24
i=6	0.2
i=7	0.11
i=8	0.04
i=9	0.01
i=10	0

```
>
```



L= 20 nu= 0.01



$$p_{i \rightarrow i+1} = (L - i)\nu$$

$$p_{i \rightarrow i-1} = i\nu$$

so:

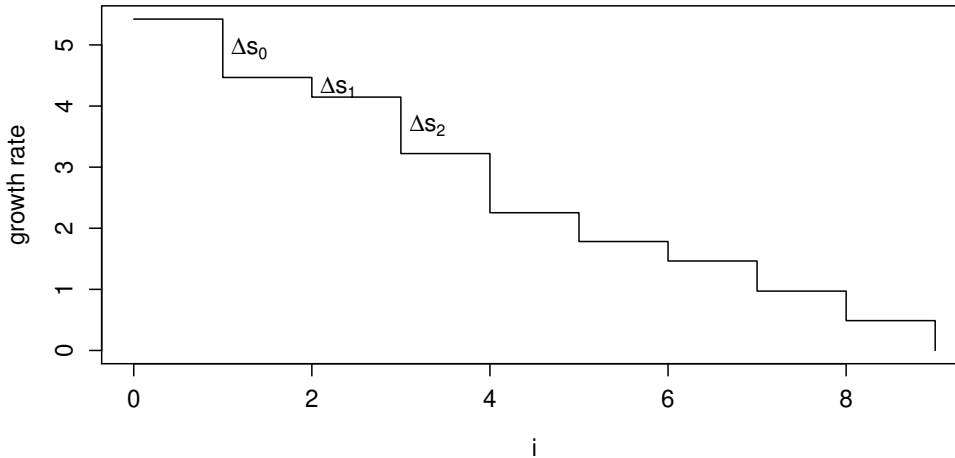
$$p_{0 \rightarrow 1} = L\nu$$

Where is the error catastrophe?

since $\mu = L\nu$

$$s \gtrsim L\nu$$

$$p_{i \rightarrow i+1} = (L - i)\nu$$



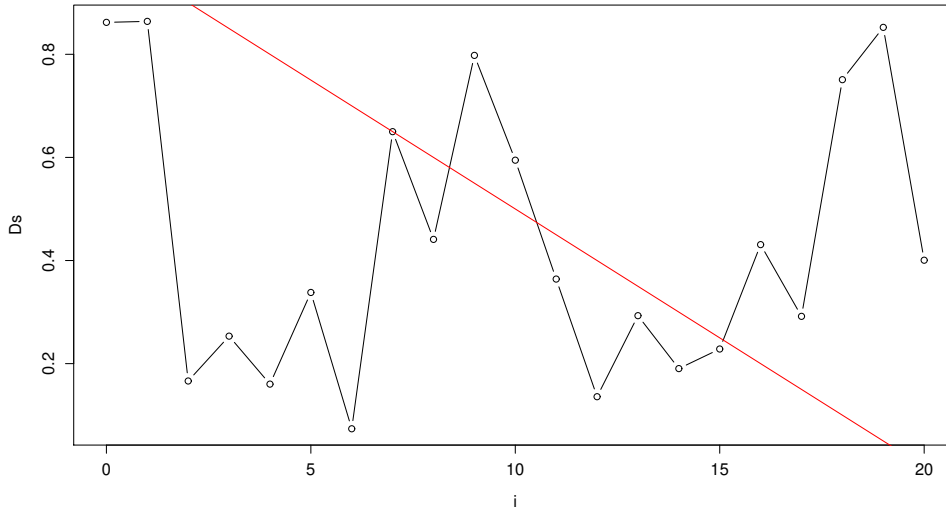
If change in growth from i to $i + 1$ is Δs_i

Population will be where $\Delta s_i > (L - i)\nu$

```
> L=20; Ds=runif(L+1); nu=0.05; i=0:L  
plot( i, Ds,type="b"); lines(i, (L-i)*nu,col=2);v()
```

```
L=20; Ds=runif(L+1); nu=0.05; i=0:L
```

```
> plot( i, Ds,type="b"); lines(i, (L-i)*nu,col=2);v()
```



```
>
```