MATH 142: Mathematical Modeling, Homework 2

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library(ggplot2)

Part A

From $N_{t+1}=(1+R_0)N_t=N_t+R_0N_t$, you notice that the population increases by R_0N_t . Thus, you can easily derive $\frac{dN}{dt}=R_0N$.

Part B

$$\frac{dN}{dt} = R_0 N$$

$$\rightarrow \int \frac{dN}{N} = \int R_0 dt$$

$$\rightarrow \ln(N) = R_0 t + c$$

$$\rightarrow N = ce^{R_0 t}$$

$$\rightarrow N(t_0) = c_1 e^{R_0 t_0} = N_0$$

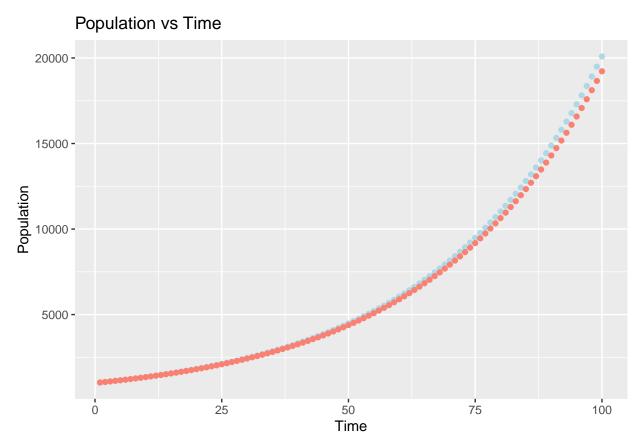
$$\rightarrow c_1 = \frac{N_0}{e^{R_0 t_0}} = \frac{1000}{e^{.03*0}} = 1000$$

Thus, $N(t) = 1000e^{.03t}$

Part C

```
x <- c(1:100)
ode <- 1000*exp(.03*x)
reccurence <- (1.03)^x*1000

ggplot(aes(x = x, y = ode), data = NULL) +
   geom_point(color = "lightblue") +
   geom_point(aes(x = x, y = reccurence), data = NULL, color = "salmon") +
   labs(title="Population vs Time", x = "Time", y = "Population")</pre>
```



From the graph, I notice that the two solutions are very similar, but eventually, it becomes clear that the solution to the reccurence equation grows slower as time increases.

Part A

To find the eigenvalues, $det(L - \lambda I) = det(\begin{bmatrix} 1 - \lambda & 1 \\ \frac{1}{2} & \frac{3}{2} - \lambda \end{bmatrix}) = (1 - \lambda)(\frac{3}{2} - \lambda) - \frac{1}{2} = \lambda^2 - \frac{5}{2}\lambda + 1 = \frac{1}{2}(2x - 1)(x - 2)$. Thus, $\lambda_1 = \frac{1}{2}, \lambda_2 = 2$.

The largest eigenvalue in magnitude is $\lambda_2 = 2 > 1$, which means the population will grow without bounds.

Part B

To find the eigenvector for $\lambda_1 = \frac{1}{2}$, $L - \lambda_1 I = \begin{bmatrix} 1 - \frac{1}{2} & 1 \\ \frac{1}{2} & \frac{3}{2} - \frac{1}{2} \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & 1 \\ \frac{1}{2} & 1 \end{bmatrix}$. Thus, the eigenvector is $\begin{bmatrix} -2 \\ 1 \end{bmatrix}$.

To find the eigenvector for $\lambda_2=2, L-\lambda_1I=\begin{bmatrix}1-2&1\\\frac{1}{2}&\frac{3}{2}-2\end{bmatrix}=\begin{bmatrix}-1&1\\\frac{1}{2}&-\frac{1}{2}\end{bmatrix}$. Thus, the eigenvector is $\begin{bmatrix}1\\1\end{bmatrix}$.

$$\begin{bmatrix} 1 \\ 4 \end{bmatrix} = a \begin{bmatrix} -2 \\ 1 \end{bmatrix} + b \begin{bmatrix} 1 \\ 1 \end{bmatrix} \rightarrow \begin{cases} 1 & = & -2a + b \\ 4 & = & a + b \end{cases} \rightarrow \begin{cases} a & = & 1 \\ b & = & 3 \end{cases}$$

$$N_k = \begin{bmatrix} 1 & 1 \\ \frac{1}{2} & \frac{3}{2} \end{bmatrix}^k (1 \begin{bmatrix} -2 \\ 1 \end{bmatrix} + 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix}) = (\frac{1}{2})^k \begin{bmatrix} -2 \\ 1 \end{bmatrix} + 3(2)^k \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$N_4 = (\frac{1}{2})^4 \begin{bmatrix} -2 \\ 1 \end{bmatrix} + 3(2^4) \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -.125 \\ .0625 \end{bmatrix} + \begin{bmatrix} 48 \\ 48 \end{bmatrix} = \begin{bmatrix} 47.875 \\ 48.0625 \end{bmatrix}$$

```
L <- matrix(data = c(1, 1, 1/2, 3/2), nrow = 2, byrow = T)
(L %*% L %*% L) %*% (1*matrix(c(-2, 1)) + 3*matrix(c(1, 1)))
```

```
## [,1]
## [1,] 47.8750
## [2,] 48.0625
```

Part C

The eigenvector associated with the largest eigenvalue in magnitude is $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$. This means that the stable age distribution is $\begin{bmatrix} \frac{1}{1+1} \\ \frac{1}{1+1} \end{bmatrix} = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}$.

Part A

$$L = \begin{bmatrix} 0 & b \\ s & 1 - m \end{bmatrix}$$

Part B

$$det(L-\lambda I) = det\begin{pmatrix} \begin{bmatrix} 0-\lambda & .9 \\ .4 & .9-\lambda \end{bmatrix} \end{pmatrix} = (-\lambda)(.9-\lambda) - .36 = \lambda^2 - .9\lambda - .36 = (\lambda+.3)(\lambda-1.2) \doteq 0 \rightarrow \lambda_1 = -.3, \lambda_2 = 1.2$$
 Thus, the long-term growth factor is 1.2.

$$L - 1.2I = \begin{bmatrix} -1.2 & .9 \\ .4 & -.3 \end{bmatrix} \rightarrow u_2 = \begin{bmatrix} 3 \\ 4 \end{bmatrix}. \text{ Thus, the stable age distribution is } \begin{bmatrix} \frac{3}{7} \\ \frac{4}{3} \end{bmatrix} = \begin{bmatrix} \frac{3}{7} \\ \frac{4}{7} \end{bmatrix}$$

Part A

 $L_{1,1}$: m of all cells die in each day and t of juvinile become tethered, (1-m)(1-t)

 $L_{1,2}$: it is impossible to go from juvinile to freely swimming, 0

 $L_{1,3}$: m of all cells die in each day and u of the tethered become freely swimming, (1-m)u

 $L_{2,1}$: freely swimming does not give birth to juvinile, 0

 $L_{2,2}$: m of all cells die in each day and c of the juvinile become tethered, (1-m)(1-c)

 $L_{2,3}$: m of all cells die in each day and b of the tethered will divide and create a juvinile, (1-m)b

 $L_{3,1}$: m of all cells die in each day, (1-m)

 $L_{3,2}$: m of all cells die in each day and c of juvinile become tethered, (1-m)c

 $L_{3,3}$: m of all cells die in each day and u tethered become freely swimming, (1-m)(1-u)

Thus,
$$L = \begin{bmatrix} (1-m)(1-t) & 0 & (1-m)u \\ 0 & (1-m)(1-c) & (1-m)b \\ (1-m)t & (1-m)c & (1-m)(1-u) \end{bmatrix}$$

Part B

$$L = \begin{bmatrix} .6 & 0 & .7 \\ 0 & .4 & .5 \\ .3 & .5 & .2 \end{bmatrix}$$

(i)

```
L \leftarrow matrix(c(.6, 0, .7, 0, .4, .5, .3, .5, .2), nrow = 3, byrow = T)
L
```

```
## [,1] [,2] [,3]
## [1,] 0.6 0.0 0.7
## [2,] 0.0 0.4 0.5
## [3,] 0.3 0.5 0.2
```

eigen(L)

```
## eigen() decomposition
## $values
## [1] 1.0504880 0.5021306 -0.3526186
##
## $vectors
## [,1] [,2] [,3]
## [1,] -0.7764170 0.8197641 -0.5220384
## [2,] -0.3840705 -0.5611153 -0.4719745
## [3,] -0.4996665 -0.1146141 0.7104336
```

From above,
$$\lambda_1 = 1.05$$
, $\overrightarrow{u_1} = \begin{bmatrix} .78 \\ .38 \\ .5 \end{bmatrix}$,

$$\lambda_2 = .5, \ \overrightarrow{u_2} = \begin{bmatrix} .82\\ -.56\\ -.11 \end{bmatrix},$$

$$\lambda_3 = -.35, \overrightarrow{u_3} = \begin{bmatrix} -.52\\ -.47\\ .71 \end{bmatrix}$$

The largest eigenvalue in magnitude is $\lambda_1 = 1.05 > 1$, which means the population will grow without bounds.

The stable fraction is
$$\begin{bmatrix} \frac{.78}{.78+.38+.5} \\ \frac{.38}{.78+.38+.5} \\ \frac{.5}{.78+.38+.5} \end{bmatrix} = \begin{bmatrix} .47\\ .23\\ .3 \end{bmatrix}$$

(ii)

```
results <- t(as.matrix(c(0, 100, 0)))
results <- rbind(results, t(L %*% as.matrix(c(0, 100, 0))))

L_new <- L
for(i in 1:9){
    L_new <- L_new %*% L
    results <- rbind(results, t(L_new %*% as.matrix(c(0, 100, 0))))
}

colnames(results) <- c("F", "J", "T")
results <- cbind("k" = 0:10, results)
round(results, 4)</pre>
```

```
F.Growth <- c()
J.Growth <- c()
T.Growth <- c()

F.Fraction <- c()
J.Fraction <- c()
T.Fraction <- c()</pre>
```

```
F.Growth <- cbind(F.Growth, results[i, "F"] / results[i - 1, "F"])
  J.Growth <- cbind(J.Growth, results[i, "J"] / results[i - 1, "J"])</pre>
  T.Growth <- cbind(T.Growth, results[i, "T"] / results[i - 1, "T"])</pre>
  F.Fraction <- cbind(F.Fraction, results[i, "F"] / sum(results[i, c("F", "J", "T")]))
  J.Fraction <- cbind(J.Fraction, results[i, "J"] / sum(results[i, c("F", "J", "T")]))</pre>
  T.Fraction <- cbind(T.Fraction, results[i, "T"] / sum(results[i, c("F", "J", "T")]))
}
growth <- cbind(8:10,
                 t(F.Growth),
                 t(J.Growth),
                t(T.Growth),
                t(F.Fraction),
                 t(J.Fraction),
                 t(T.Fraction))
colnames(growth) <- c("k",</pre>
                       "F_Growth",
                       "J_Growth",
                       "T_Growth",
                       "F_Fraction",
                       "J_Fraction",
                       "T Fraction")
round(growth, 4)
```

```
k F_Growth J_Growth T_Growth F_Fraction J_Fraction T_Fraction
## [1,] 8
            1.0561
                      1.0443
                               1.0485
                                          0.4658
                                                     0.2329
                                                                0.3013
## [2,] 9
             1.0528
                      1.0469
                               1.0503
                                          0.4667
                                                     0.2321
                                                                0.3012
## [3,] 10
             1.0517
                      1.0490
                               1.0501
                                          0.4672
                                                     0.2317
                                                                0.3011
```

Ignoring rounding error, these numbers agree with the results I obtained in part (i).