

## CS 132 – Spring 2020, Assignment 9

### Answer A

First lets multiply the two to find  $Ax$ :

$$\begin{bmatrix} 3 & 6 & 7 \\ 3 & 3 & 7 \\ 5 & 6 & 5 \end{bmatrix} \times \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 - 12 + 7 \\ 3 - 6 + 7 \\ 5 - 12 + 5 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \\ -2 \end{bmatrix}$$

Hence we can see that  $x$  is in fact an eigenvector of  $A$ , and it's eigenvalue is  $-2$ .

### Answer B

If we first organize ourselves to put this into  $Ax = \lambda x$ , then to find  $x$  we have that:  $Ax - \lambda x = 0 \equiv x(A - I\lambda) = 0$

First we solve  $A - I\lambda$ :

$$\begin{bmatrix} 1 & 2 & 2 \\ 3 & -2 & 1 \\ 0 & 1 & 1 \end{bmatrix} - \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix} = \begin{bmatrix} -2 & 2 & 2 \\ 3 & -5 & 1 \\ 0 & 1 & -2 \end{bmatrix}$$

Now we solve  $(A - I\lambda)x = 0$ :

$$\begin{bmatrix} -2 & 2 & 2 & 0 \\ 3 & -5 & 1 & 0 \\ 0 & 1 & -2 & 0 \end{bmatrix} \equiv \begin{bmatrix} 1 & -1 & -1 & 0 \\ 0 & 1 & -2 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

We know that  $\lambda$  is a eigenvalue of  $A$  if  $x(A - I\lambda) = 0$  has a non-trivial solution, which happens if it has a free variable.

Since we've already proved that already, we can easily say that it is infact a eigenvalue of  $A$ .

Now we only have to find a vector that fulfill the previously stated solution, so if we look at that previous matrix, we realize that any eigenvector must have the following form:

$$\begin{bmatrix} 3x_3 \\ 2x_3 \\ x_3 \end{bmatrix}$$

Let  $x_3 = 1$  then we have that one possible eigenvector for this is:  $\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$

---

**Answer C**

If we first organize ourselves to put this into  $Ax = \lambda x$ , then to find  $x$  we have that:  $Ax - \lambda x = 0 \equiv x(A - I\lambda) = 0$   
First we solve  $A - I\lambda$ :

$$\begin{bmatrix} 10 & -9 \\ 4 & -2 \end{bmatrix} - \begin{bmatrix} 4 & 0 \\ 0 & 4 \end{bmatrix} = \begin{bmatrix} 6 & -9 \\ 4 & -6 \end{bmatrix}$$

Now we solve  $(A - I\lambda)x = 0$ :

$$\begin{bmatrix} 6 & -9 & 0 \\ 4 & -6 & 0 \end{bmatrix} \equiv \begin{bmatrix} 1 & -3/2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

With that, we know that any eigenvector will follow the basic condition:  $\begin{bmatrix} 1.5x_2 \\ x_2 \end{bmatrix}$ .

As such, the basis for this eigenspace is:  $\begin{bmatrix} 1.5 \\ 1 \end{bmatrix}$

**Answer D**

a. ☐ True

As per the definition of eigenvectors.

b. ☐ True

As if they corresponded to the same eigenvalue, due to the definition of how eigenvalues/eigenvector relate to each other, the vectors would be linearly dependent.

c. ☐ True

As the stochastic matrix to A with a steady state vector  $q$ , satisfies the equation:  $Aq = 1q$ .

d. ☐ True

As per theorem 1 of this chapter, the Eigenvalues of a triangular matrix are in its main diagonal.

e. ☐ True

As the eigenspace of A will be a combination of vectors, from which we can build another matrix B for which said vectors are the nullspace of.

---

**Answer E**

Any non-invertible 2x2 matrix will only have a single eigenvalue, 0 so an example of one would be:

$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

**Answer F**

The determinant of this matrix (or the characteristic polynomial) is:

$$(5-\lambda)^2 - 3^2 = 16 - 10\lambda + \lambda^2$$

If we solve this equation we get that  $\lambda = 2 | \lambda = 8$

**Answer G**

The determinant of this matrix (or the characteristic polynomial) is:

$$(5-\lambda) \times (3-\lambda) - (-3)(-4) = 3 - 8\lambda + \lambda^2$$

If we solve this equation we get that  $\lambda = -7.6 | \lambda = 7.6$

**Answer H**

b. ☐ False

As per the Theorem 3,  $\det A^T = \det A$

c. ☐ True

As per the definition above.

d. ☐ True

As no matter how many row replacements you end up doing, the eigenvalue calculation ends up with the same result.

---

**Answer I****a.**

Since  $v_1$  is already the steady state vector of  $A$ , we first need to discover the characteristic polynomial of:

$$\begin{bmatrix} .6 & .3 \\ .4 & .7 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} .6 - \lambda & .3 \\ .4 & .7 - \lambda \end{bmatrix}$$

So we get that:  $(.6 - \lambda) \times (.7 - \lambda) - .3 \times .4 = 0 \equiv .3 - 1.3\lambda + \lambda^2 = 0$

If we solve this we get  $\lambda = 1 | \lambda = 0.3$ , since  $v_1$  uses the first choice, then we must calculate the eigenvector for .3

To calculate this, we will use the equation  $(A - I\lambda)x = 0$ :

$$\begin{bmatrix} .3 & .3 & 0 \\ .4 & .4 & 0 \end{bmatrix} \equiv \begin{bmatrix} .3 & .3 & 0 \\ 0 & 0 & 0 \end{bmatrix} \equiv \begin{cases} x_1 = -x_2 \\ x_2 = free \end{cases}$$

As such, the basis for  $R^2$  made of  $v_1 v_2$ :  $\begin{bmatrix} -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 3/7 \\ 4/7 \end{bmatrix}$

**b.**

$$v_1 + cv_2 = x_0 \equiv \begin{bmatrix} 3/7 \\ 4/7 \end{bmatrix} + \begin{bmatrix} 0.5/7 \\ -0.5/7 \end{bmatrix} = \begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix}$$

**c.**

$$x_1 = A(x_0) = A(v_1 + cv_2) = A(v_1) + cA(v_2) = 1 \times v_1 + c \times 0.3 \times v_2$$

$$x_2 = A(x_1) = A(v_1 + 0.3cv_2) = A(v_1) + 0.3cA(v_2) = v_1 + 0.3^2 \times c \times v_2$$

In general:  $x_k = v_1 + (0.3)^k \times c \times v_2$

As  $k$  increases,  $x_k$  will tend to  $v_1$ , as  $0.3^k$  will tend to 0.

## Answer J

As we can see, in both cases, the eigenvalues are the same, but the eigenvectors are not.

```
In [16]: import numpy as np

A = np.random.randint(11, size=(4,4))
At = A.transpose()

(u1, v1) = np.linalg.eig(A)
(u2, v2) = np.linalg.eig(At)

print(A)
print(At)
print()
print(np.round(u1,2))
print(np.round(u2,2))
print()
print(np.round(v1,2))
print(np.round(v2,2))

[[10  8  6  6]
 [ 3  5  2  2]
 [ 8  1  4  5]
 [ 2  8  0  1]]
[[10  3  8  2]
 [ 8  5  1  8]
 [ 6  2  4  0]
 [ 6  2  5  1]]

[18.57+0.j  1.77+0.j -0.17+1.j -0.17-1.j]
[18.57+0.j  1.77+0.j -0.17+1.j -0.17-1.j]

[[-0.78+0.j  -0.37+0.j  -0.29+0.19j -0.29-0.19j]
 [-0.28+0.j   0.17+0.j  -0.04+0.05j -0.04-0.05j]
 [-0.52+0.j  -0.49+0.j  -0.26-0.43j -0.26+0.43j]
 [-0.22+0.j   0.77+0.j   0.78+0.j   0.78-0.j  ]]
[[-0.61+0.j   0.31+0.j  -0.29+0.23j -0.29-0.23j]
 [-0.61+0.j  -0.95+0.j   0.83+0.j   0.83-0.j  ]
 [-0.34+0.j   0.02+0.j   0.1  -0.3j   0.1  +0.3j ]
 [-0.37+0.j   0.07+0.j  -0.26-0.09j -0.26+0.09j]]
```

```

In [18]: import numpy as np

A = np.random.randint(11, size=(5,5))
At = A.transpose()

(u1, v1) = np.linalg.eig(A)
(u2, v2) = np.linalg.eig(At)

print(A)
print(At)
print()
print(np.round(u1,2))
print(np.round(u2,2))
print()
print(np.round(v1,2))
print(np.round(v2,2))

[[ 0  1  7  1  6]
 [ 4  3  3 10  6]
 [ 3 10  2  6  3]
 [ 1  9  5 10  5]
 [10  2 10  3  9]]
[[ 0  4  3  1 10]
 [ 1  3 10  9  2]
 [ 7  3  2  5 10]
 [ 1 10  6 10  3]
 [ 6  6  3  5  9]]

[26.72+0.j    6.92+0.j   -3.85+0.j   -2.89+2.61j -2.89-2.61j]
[26.72+0.j    6.92+0.j   -3.85+0.j   -2.89+2.61j -2.89-2.61j]

[[-0.26+0.j   -0.29+0.j   -0.56+0.j   -0.02+0.31j -0.02-0.31j]
 [-0.45+0.j    0.12+0.j    0.3 +0.j    0.25-0.33j  0.25+0.33j]
 [-0.41+0.j    0.26+0.j   -0.26+0.j   -0.58+0.j   -0.58-0.j  ]
 [-0.54+0.j    0.53+0.j   -0.3 +0.j   -0.2 +0.22j -0.2 -0.22j]
 [-0.52+0.j   -0.74+0.j    0.66+0.j    0.54-0.14j  0.54+0.14j]]
[[ 0.32+0.j    0.44+0.j    0.65+0.j   -0.24+0.07j -0.24-0.07j]
 [ 0.44+0.j   -0.19+0.j    0.55+0.j   -0.66+0.j   -0.66-0.j  ]
 [ 0.44+0.j    0.44+0.j   -0.07+0.j   -0.02-0.43j -0.02+0.43j]
 [ 0.53+0.j   -0.69+0.j   -0.32+0.j    0.42+0.27j  0.42-0.27j]
 [ 0.48+0.j    0.31+0.j   -0.42+0.j    0.28+0.02j  0.28-0.02j]]

```

In [ ]:

---

**Answer K****a.**

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

**b.**

In order to get the eigenvalues of A, we must first find the characteristic equation of A, which would be:

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} 1-\lambda & 1 \\ 1 & -\lambda \end{bmatrix}$$

So we get that:  $(1-\lambda) \times (-\lambda) - 1 = \lambda^2 - \lambda - 1 = 0$

If we solve this we get that  $\lambda = -0.618$  |  $\lambda = 1.618$ , now we will solve  $(A - \lambda I)x = 0$ , and find the eigenvectors:

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} - \begin{bmatrix} -0.618 & 0 \\ 0 & -0.618 \end{bmatrix} = \begin{bmatrix} 1.618 & 1 \\ 1 & 0.618 \end{bmatrix} \equiv \begin{bmatrix} 0 & 0 \\ 1 & 0.618 \end{bmatrix} \equiv \begin{cases} x_1 = -0.618x_2 \\ x_2 = \text{free} \end{cases} \equiv \begin{bmatrix} -0.618 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} - \begin{bmatrix} 1.618 & 0 \\ 0 & 1.618 \end{bmatrix} = \begin{bmatrix} -0.618 & 1 \\ 1 & -1.618 \end{bmatrix} \equiv \begin{bmatrix} 0 & 0 \\ 1 & -1.618 \end{bmatrix} \equiv \begin{cases} x_1 = 1.618x_2 \\ x_2 = \text{free} \end{cases} \equiv \begin{bmatrix} 1.618 \\ 1 \end{bmatrix}$$

Eigenvalues:  $-0.618$  |  $1.618$

Basis for the eigenspace:  $\begin{bmatrix} 1.618 \\ 1 \end{bmatrix}, \begin{bmatrix} -0.618 \\ 1 \end{bmatrix}$

---

**c.**

Now that we know that the basis for the eigenspace (let it be S) is:

$$\begin{bmatrix} 1.618 & -0.618 \\ 1 & 1 \end{bmatrix}$$

We can get the diagonalization matrix easily, let it be D:

$$\begin{bmatrix} 1.618 & 0 \\ 0 & 0.618 \end{bmatrix}$$

Since we know that  $x_0$ , will be made from our S matrix multiplied by some scalar we have that:

$$x_0 = Sc \equiv x_0 S^{-1} = c, \text{ so now we need } S^{-1}:$$

$$S^{-1} = \frac{1}{\det(S)} \times \begin{bmatrix} 1 & -0.618 \\ -1 & 1.618 \end{bmatrix} \equiv \begin{bmatrix} 0.447 & -0.276 \\ -0.447 & 0.724 \end{bmatrix}$$

Finally we can calculate c and get:

$$c = \begin{bmatrix} 0.447 & -0.276 \\ -0.447 & 0.724 \end{bmatrix} \times \begin{bmatrix} 0 \\ 1 \end{bmatrix} \equiv c = \begin{bmatrix} -0.276 \\ 0.724 \end{bmatrix}$$

**d.**

With all that we've gather before we know that:

$$x_k = S \times D^k \times c$$



---

**e.**

```
In [6]: import numpy as np

def getFib(x):
    return np.round(((1.618 ** (x-3)) * (0.724) * (1.618)) + ((-0.618 ** (x-3)) * (0.276) * (-0.618)),4)

print("F(40) = ", getFib(40))
print("F(50) = ", getFib(50))
print("F(60) = ", getFib(60))

F(40) = 63229860.3597
F(50) = 7775125279.0733
F(60) = 956076334208.8616
```

---

**d.**

If we go back to our equation, we know that:  $x_k = S \times D^k \times c$ , but since, when  $k$  approached infinity,  $D^k$  approaches 0, then as the sequence grows, it will tend towards  $Sc$

## Answer J

```
In [7]: import numpy as np

A = np.array([[0, 0, 1/10, 1/10, 1/10, 1/10],
              [1/7, 1, 0, 0, 0, 0],
              [1/10, 1/10, 0, 0, 1/10, 1/10],
              [0, 0, 1/7, 1, 0, 0],
              [1/10, 1/10, 1/10, 1/10, 0, 0],
              [0, 0, 0, 0, 1/7, 1]])

In [10]: #A
x, v = np.linalg.eig(A)

print(np.real(x))

[ 0.16575188  1.03424812 -0.08685593 -0.08685593  0.98685593  0.98685593]

In [11]: #B
a = np.amax(x)
print("The largest eigenvalue is: ", a, " which means that the array is unstable.")

The largest eigenvalue is:  1.0342481186734473  which means that the array is unstable.

In [12]: #C
startVal = [1, 0, 0, 0, 0, 0]
print(v.dot(startVal))

[-0.56906708  0.09744738 -0.56906708  0.09744738 -0.56906708  0.09744738]

In [18]: #D
b = np.real(v[1])
print("Let the eigenvalue a=", biggest, "and the vector b=", b)
print()
print("xk = b * a^k * ", v.dot(startVal)[1])

Let the eigenvalue a= 1.0342481186734473 and the vector b= [ 0.09744738 -0.56144158 -0.10640568  0.04097612 -0.81306232 -0.2854
8524]

xk = b * a^k *  0.09744738430733318

In [27]: #E
def getWorms(k):
    return np.round(b.dot(((a ** k) * v.dot(startVal)[1])),4)

print("F(100) = ", getWorms(100))
print("F(250) = ", getWorms(250))
print("F(500) = ", getWorms(500))
print("F(750) = ", getWorms(750))

F(100) = [ 7.50571552e+18 -4.32440628e+19 -8.19571268e+18  3.15611459e+18
-6.26247128e+19 -2.19890047e+19]
F(250) = [ 1.66789878e+50 -9.60957279e+50 -1.82122799e+50  7.01342818e+49
-1.39162858e+51 -4.88633416e+50]
F(500) = [ 2.92953716e+102 -1.68704826e+103 -3.19884823e+102  1.23185523e+102
-2.44428959e+103 -8.58247372e+102]
F(750) = [ 5.14550885e+154 -2.96457689e+155 -5.61853324e+154  2.16365988e+154
-4.29320846e+155 -1.50744613e+155]
```