# LABNov20

November 27, 2023

#### Lab November 20th

## Julia Verdickt

```
[67]: import numpy as np import matplotlib.pyplot as plt import pandas as pd
```

# 1 Question 1

# 1.1 Question 1(a)

1) a) Stochastic Matrix

b) simulate starting from

Xo=1, see code

## 1.2 Question 1(b)

```
[68]: stochastic_matrix = np.array([[2/10, 7/10, 1/10], [2/10, 5/10, 3/10], [2/10,4/

4/10]])

      nodes = np.array([1,2,3])
      X = [1]
      X_{temp} = 1
      np.random.seed(123)
      for i in range (20):
          if X_temp == 1:
              X_temp = np.random.choice(nodes, p = stochastic_matrix[0])
              X.append(X_temp)
          elif X_temp == 2:
              X_temp = np.random.choice(nodes, p = stochastic_matrix[1])
              X.append(X_temp)
          else:
              X_temp = np.random.choice(nodes, p = stochastic_matrix[2])
              X.append(X_temp)
```

```
[69]: X
```

[69]: [1, 2, 2, 2, 2, 3, 2, 3, 3, 2, 2, 2, 3, 2, 1, 2, 3, 1, 1, 2, 2]

# 2 Question 2

## 2.1 Question 2(a)

$$\begin{bmatrix} 1 & -\frac{1}{4} & -\frac{1}{4} & 0 \\ 0 & -0.325 & 0.575 \\ 0 & 0.325 & -0.575 \\ 0 & 0.325 & -0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0.575 \\ 0 & 0 & 0$$

$$dot(P-\lambda I) = \left(\frac{1}{5} - \lambda\right) det \begin{vmatrix} \frac{1}{2} - \lambda & \frac{3}{10} \\ \frac{2}{5} & \frac{2}{5} - \lambda \end{vmatrix} - \left(\frac{1}{5}\right) det \begin{vmatrix} \frac{7}{10} & \frac{1}{10} \\ \frac{2}{5} & \frac{2}{5} - \lambda \end{vmatrix} + \left(\frac{1}{5}\right) det \begin{vmatrix} \frac{7}{10} & \frac{1}{10} \\ \frac{1}{2} - \lambda & \frac{3}{10} \end{vmatrix} = \left(\frac{1}{5} - \lambda\right) \left[\left(\frac{1}{2} - \lambda\right)\left(\frac{2}{5} - \lambda\right) - \frac{3}{25}\right] - \left(\frac{1}{5}\left(\frac{2}{5} - \lambda\right) - \frac{1}{10}\left(\frac{1}{2} - \lambda\right)\right) \left[\frac{17}{50} - \frac{2}{50}\right] + \left[\frac{21}{100} - \frac{1}{10}\left(\frac{1}{2} - \lambda\right)\right] \left[\frac{17}{50} - \frac{2}{50}\right] = \left(\frac{1}{5} - \lambda\right) \left[\lambda^2 - \frac{7}{10}\lambda + \frac{2}{25}\right] - \left[\frac{1}{5}\left(\frac{1}{50} - \frac{7}{10}\lambda\right) + \frac{1}{5}\left(\frac{4}{25} + \frac{1}{10}\lambda\right)\right]$$

$$= -\lambda^{3} + \frac{11}{10}\lambda^{2} - \frac{13}{50}\lambda + \frac{2}{125} - \frac{6}{125}$$

$$+ \frac{7}{50}\lambda + \frac{4}{125} + \frac{1}{50}\lambda$$

$$= -\lambda^{3} + \frac{11}{10}\lambda^{2} - \frac{1}{10}\lambda$$

$$= -\lambda^{3} + \frac{11}{10}\lambda^{2} - \frac{1}{10}\lambda$$

$$= \lambda^{3} - \frac{11}{10}\lambda^{2} + \frac{1}{10}\lambda$$

$$= 0.1, 0.1 \quad \text{eigen values}$$

$$\Rightarrow D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0.1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0.2 & 0.7 & 0.1 & 0 \\ 0 & 0.2 & 0.7 & 0.1 & 0 \\ 0 & -0.2 & 0.2 & 0 \\ 0 & -0.3 & 0.3 & 0 \end{bmatrix}$$

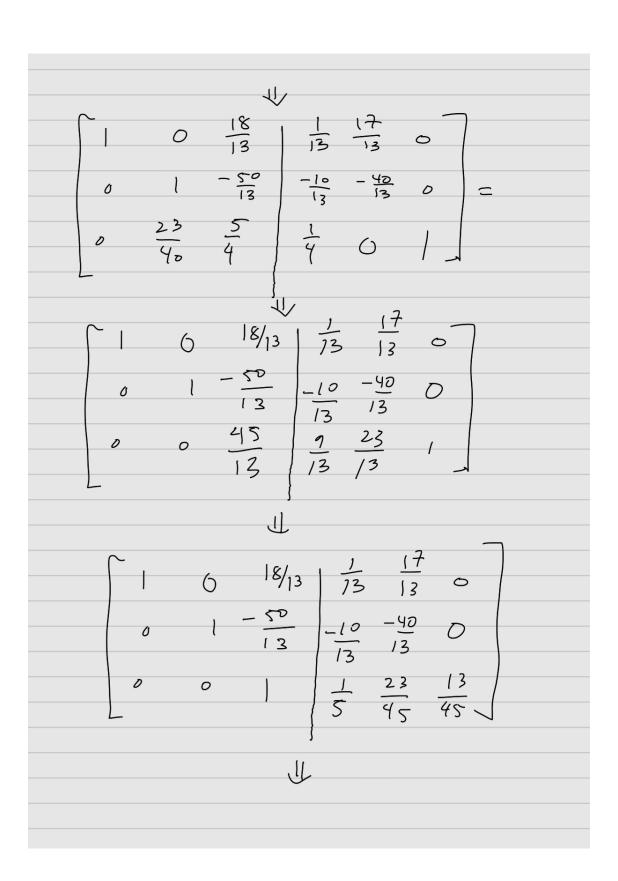
$$= \begin{bmatrix} 1 & 3.5 & 0.5 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

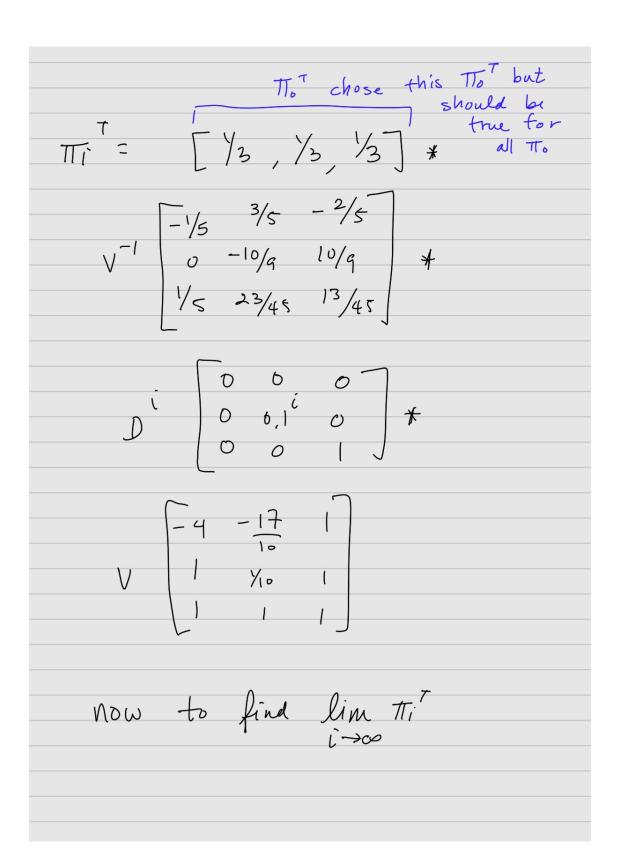
$= \begin{bmatrix} 1 & 0 & 4 &   & 0 \\ 0 & 1 & -1 &   & 0 \\ 0 & 0 & 0 &   & 0 \end{bmatrix} \overrightarrow{X_i} = \begin{bmatrix} -4 \\ 1 \end{bmatrix}$
$P - T = \begin{bmatrix} -0.8 & 0.7 & 0.1 & 0 \\ 0.2 & -0.5 & 0.3 & 0 \\ 0.2 & 0.4 & -0.6 & 0 \end{bmatrix}$
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
$= \begin{bmatrix} 1 & -\frac{7}{8} & -\frac{7}{8} & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ $= \begin{bmatrix} 1 & 0 & -\frac{7}{8} & -\frac{7}{8} & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ $= \begin{bmatrix} 1 & 0 & -\frac{7}{8} & -\frac{7}{8} & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ $= \begin{bmatrix} 1 & 0 & -\frac{7}{8} & -\frac{7}{8} & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 &$

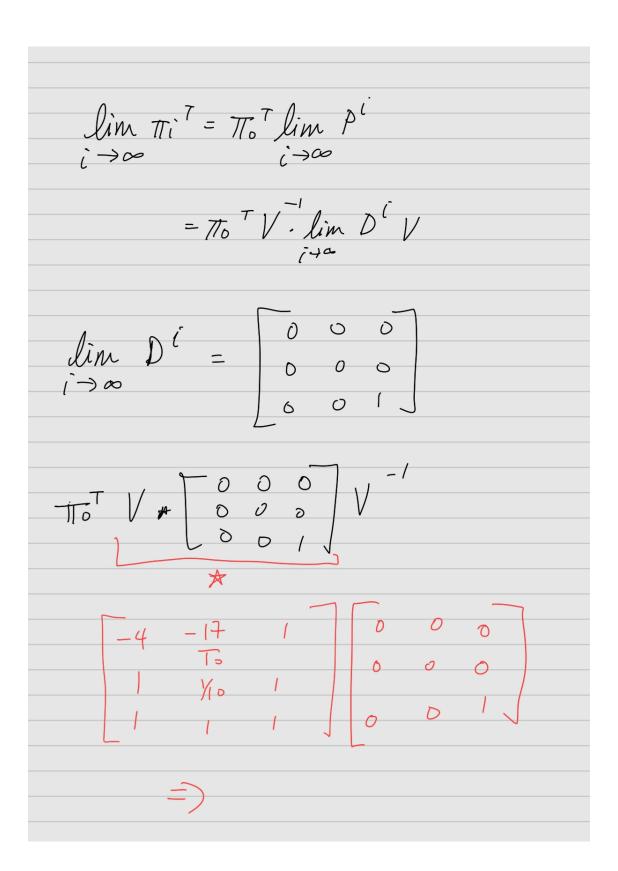
	$P = 0.1 \times 0.1 \times 0.1 \times 0.1 \times 0.2 \times 0.4 \times 0.3 \times 0.2 \times 0.4 \times 0.3 \times 0.3 \times 0.2 \times 0.4 \times 0.3 \times $
$= \begin{bmatrix} 1 & 0 & 1 & -0.1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ $= \begin{bmatrix} 1 & 0 & 1.7 & 0 \\ 0 & 1 & -0.1 & 0 \end{bmatrix}$ $= \begin{bmatrix} -17 & -17 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ $= \begin{bmatrix} 1 & 0 & 1.7 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ $= \begin{bmatrix} 1 & 0 & 1.7 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ $= \begin{bmatrix} 1 & 0 & 1.7 & 0 \\ 0 & 1 & 0 \end{bmatrix}$	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	o 1 -0.1 °  o igen vector
	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$

So we have P= V D V-1 where o 10 and D = 0 0 0.1 V = 1 1/10 1 Find V-1 0 1 0 J

D 0	$-\frac{17}{40}$ $-\frac{13}{40}$ $\frac{23}{40}$		1 - 4	0	0
	17-40	- <del>1</del>	1-4	6	
0	23 40	-50 13 514 18 13	13 - 4	0	
0	1 23 40	- <u>50</u>	- <u>10</u>	13 - 40 13	
		$\Psi$			







$$V + \lim_{i \to \infty} D^{i} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

$$V \lim_{i \to \infty} D^{i} V^{-1} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -\frac{1}{5} & \frac{3}{5} & -\frac{3}{5} \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -\frac{1}{5} & \frac{3}{5} & \frac{3}{5} \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -\frac{1}{5} & \frac{1}{5} & \frac{1}{5} \\ \frac{1}{5} & \frac{23}{45} & \frac{13}{45} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{5} & \frac{23}{45} & \frac{13}{45} \\ \frac{1}{5} & \frac{23}{45} & \frac{13}{45} \end{bmatrix} = \frac{13}{15}$$

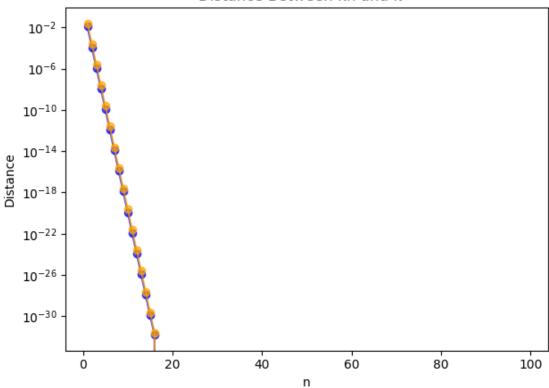
$$= \begin{bmatrix} \frac{1}{5} & \frac{23}{45} & \frac{13}{45} \\ \frac{1}{5} & \frac{23}{45} & \frac{13}{45} \end{bmatrix} = \frac{1}{15} = \frac{1}{15}$$

THIS IS TRUE FOR ALL TIOT
I'll generate the plot next of i against $  \Pi i - \Pi \infty   _2^2$

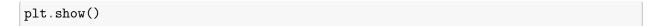
#### 2.2 Question 2(b) code analytical solution above

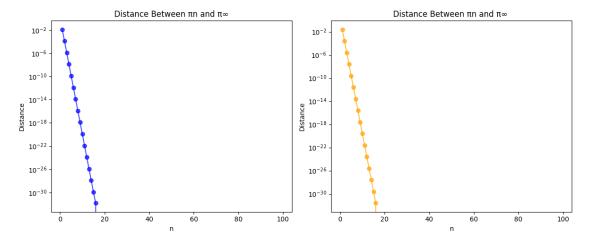
```
[70]: #defining two different 0
      pi 0 1 = np.array([1/3,1/3,1/3])
      pi_0_2 = np.array([1,0,0])
      #defining V matrix and D matrix for diagonalization
      V_{\text{matrix}} = \text{np.array}([[-4, 1, 1], [-17/10, 1/10, 1], [1,1,1]])
      #didn't match my analytical solution unless I too transpose for some reason
      V_{\text{matrix}} inv = np.array([[(-1/5), 0, (1/5)], [3/5, -10/9, 23/45], [-2/5,10/9,13/
      diag_matrix = np.array([[0, 0, 0], [0, 0.1, 0], [0,0,1]])
      #defining \omega
      pi_inf = np.array([9/45,23/45,13/45])
[92]: #calculating the distance
      def calc diff(pi 0):
          norms_sq = []
          for i in range(1,100):
              pi_i = pi_0 @ ((V_matrix @ (diag_matrix**i)) @ V_matrix_inv)
              norm_sq = np.sum((pi_i - pi_inf)**2)
              enum_norm_sq = (i, norm_sq)
              norms_sq.append(enum_norm_sq)
          return(norms_sq)
[94]: graphing_df1 = pd.DataFrame(calc_diff(pi_0_1))
      graphing_df2 = pd.DataFrame(calc_diff(pi_0_2))
[95]: #layering plots to see difference
      plt.plot(graphing df1[0], graphing df1[1], 'o-', color='blue', alpha=0.7)
      plt.plot(graphing_df2[0], graphing_df2[1], 'o-', color='orange', alpha=0.7)
      plt.title('Distance Between n and w')
      plt.xlabel('n')
      plt.ylabel('Distance')
      plt.yscale('log')
      plt.tight_layout()
      plt.show()
```

#### Distance Between πn and π∞



```
[96]: fig, axes = plt.subplots(1, 2, figsize=(12, 5)) # 1 row, 2 columns for side by
      ⇔side histograms
      # Histogram for pi_0_1
      axes[0].plot(graphing_df1[0], graphing_df1[1], 'o-', color='blue', alpha=0.7)
      axes[0].set_title('Distance Between n and w')
      axes[0].set_xlabel('n')
      axes[0].set_ylabel('Distance')
      axes[0].set yscale('log')
      # Histogram for pi_0_2
      axes[1].plot(graphing_df2[0], graphing_df2[1], 'o-', color='orange', alpha=0.7)
      axes[1].set_title('Distance Between n and o ')
      axes[1].set_xlabel('n')
      axes[1].set_ylabel('Distance')
      axes[1].set_yscale('log')
      # Show plot
      plt.tight_layout()
```





They both converge regardless of the starting point!!!

## 3 Question 3

#### 3.1 Question 3(a)

```
[76]: def chain_simulator(start):
    X_temp = start[0]
    while X_temp != 3:
        if X_temp == 1:
            X_temp = np.random.choice(nodes, p = stochastic_matrix[0])
            start.append(X_temp)
        elif X_temp == 2:
            X_temp = np.random.choice(nodes, p = stochastic_matrix[1])
            start.append(X_temp)
        else:
            break
    return start, len(start)-1
```

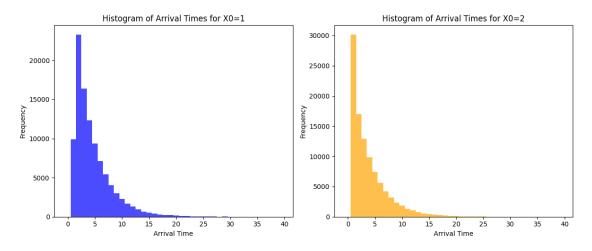
```
[77]: np.random.seed(123)

arrival_times_s1 = []
for i in range(100000):
    _, time = chain_simulator(start=[1])
    arrival_times_s1.append(time)
```

```
[78]: np.random.seed(123)
arrival_times_s2 = []
```

```
for i in range(100000):
    _, time = chain_simulator(start=[2])
    arrival_times_s2.append(time)
```

```
[79]: fig, axes = plt.subplots(1, 2, figsize=(12, 5)) # 1 row, 2 columns for side by
       ⇔side histograms
      # Histogram for XO=1
      axes[0].hist(arrival_times_s1, bins=np.arange(-0.5,40.5,1), color='blue',_
       \Rightarrowalpha=0.7)
      axes[0].set_title('Histogram of Arrival Times for X0=1')
      axes[0].set xlabel('Arrival Time')
      axes[0].set_ylabel('Frequency')
      # Histogram for XO=2
      axes[1].hist(arrival_times_s2, bins=np.arange(-0.5,40.5,1), color='orange',__
       \Rightarrowalpha=0.7)
      axes[1].set_title('Histogram of Arrival Times for X0=2')
      axes[1].set_xlabel('Arrival Time')
      axes[1].set_ylabel('Frequency')
      # Show plot
      plt.tight_layout()
      plt.show()
```



What are the mean arrival times calculated numerically?

```
[80]: mu_T1 = np.mean(arrival_times_s1)
mu_T1
```

[80]: 4.61251

```
[81]: mu_T2 = np.mean(arrival_times_s2)
mu_T2
```

[81]: 3.84046

# 3.2 Question 3(b)

We can see that the analyticaly solution found below matches quite closely to the numerical solution found above

3)
b) 
$$\mu_{i} = \mathbb{E}[T_{i}] = 1 + \sum_{j=1}^{3} p_{ij} \mu_{j}$$
 $\mu_{b} = \mathbb{E}[T_{b}] = 0$ 

$$\mu_{1} = 1 + p_{11} \mu_{1} + p_{12} \mu_{2} + p_{13} \mu_{3}$$

$$= 1 + p_{11} \mu_{1} + p_{12} \mu_{2}$$

$$\mu_{2} = 1 + p_{21} \mu_{1} + p_{22} \mu_{2} + p_{13} \mu_{3}$$

$$= 1 + p_{21} \mu_{1} + p_{22} \mu_{2}$$

$$1 = 0.2$$

$$1 = 0.2$$

$$1 = 0.2$$

$$2 = 0.5$$

$$0.2 = 0.5$$

$$0.2 = 0.4 = 0.4$$

$$p_{22} = 0.5$$

$$\frac{1 \text{ inear system}}{p_{21}} = 1 + 0.2 \mu_{1} + 0.5 \mu_{2}$$

$$0.4 \mu_{1} = 1 + 0.7 \mu_{2}$$

$$\mu_{1} = 1.25 + 0.875 \mu_{2}$$

$0.5\mu_2 = 1 + 0.2 (1.25 + 0.875 \mu_2)$
11 - 2 - n (1/1 - n - n + 2 + 1/1 - 1)
Mz = 2 + 0.4 (1.25 + 0.875 Uz)
Uz = 2+ 0.5 + 0.35Uz
u - 2.5 50
$N_2 = \frac{2.5}{0.45} = \frac{50}{13} \approx 3.844$
0.65
11. = 1.25 + 2×2c 50
$\mu_1 = 1.25 + 0.875.\frac{50}{13}$
7
$= 1.25 + \frac{7}{8}.\frac{50}{13}$
= <u>60</u> ~ 4.6154
13