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In [ ]: ''' 1) b) Simulate one single realization of the chain, that is, starting from  $X_0$  =
the value of  $X_i$  using the probabilities defined by the process'''.
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In [9]: import numpy as np
import math
import matplotlib.pyplot as plt
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```
In [22]: trans_matrix = np.array([
    [0.2, 0.7, 0.1],
    [0.2, 0.5, 0.3],
    [0.2, 0.4, 0.4]
])

initial_state = 1
state_list = [initial_state]

for _ in range(20):
    initial_state = np.random.choice([1, 2, 3], p=trans_matrix[initial_state-1])
    state_list.append(initial_state)

state_list
```

```
Out[22]: [1, 2, 2, 1, 2, 2, 3, 2, 2, 2, 3, 2, 3, 1, 2, 3, 2, 2, 3, 2, 3]
```

```
In [ ]:
```

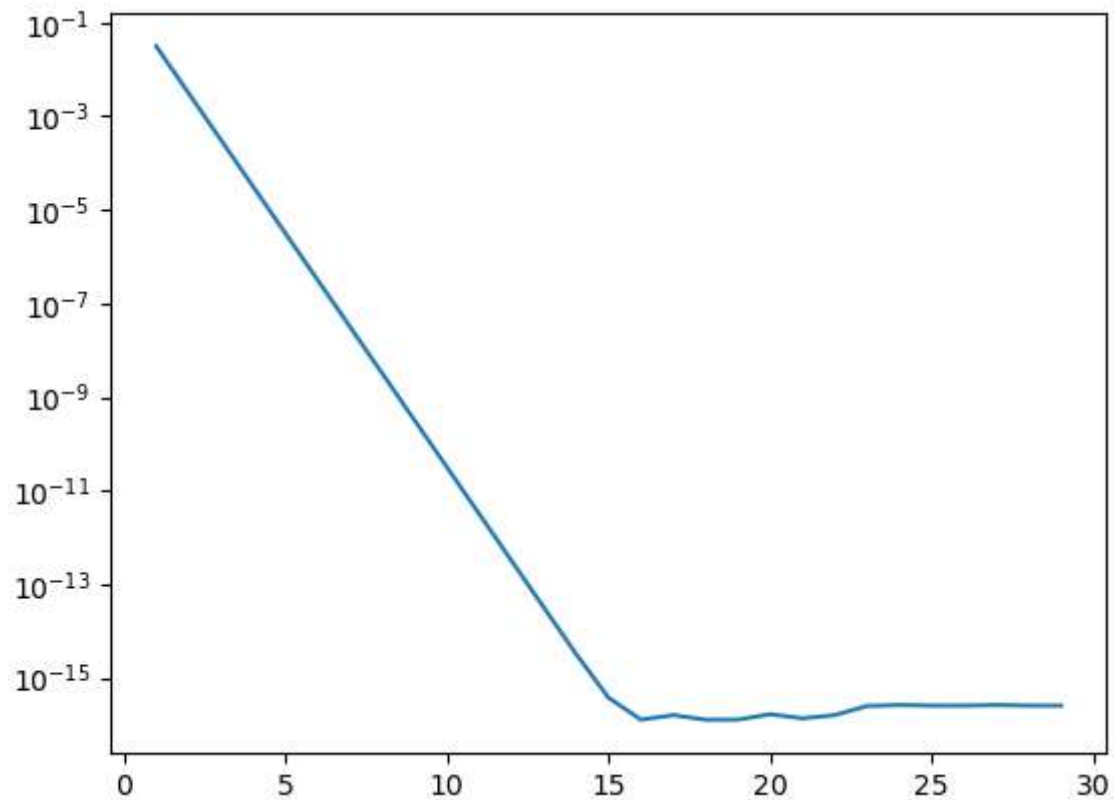
```
In [ ]: '''2) b) Starting now from an initial probability distribution  $\pi_0$  on the nodes, com
0 P i the probability distribution at time i. Show that  $\pi_i \rightarrow \pi_\infty$ 
1
and make plot of i vs  $\|\pi_i - \pi_\infty\|^2$ 
2. Generate this plot for at least two different
initial conditions  $\pi_0$  and compare.'''
```

```
In [25]: import numpy as np
import matplotlib.pyplot as plt

target_distribution = np.array([1/5, 23/45, 13/45])
initial_distribution = np.array([1/3, 1/3, 1/3])
iteration_axis = []
distance_values = []
transpose_initial_distribution = np.transpose(initial_distribution)

for iteration_step in range(1, 30):
    iteration_axis.append(iteration_step)
    iteration_matrix = np.linalg.matrix_power(transpose_initial_distribution, iteration_step)
    current_distribution = np.dot(initial_distribution, iteration_matrix)
    distance = math.dist(current_distribution, target_distribution)
    distance_values.append(distance)

plt.plot(iteration_axis, distance_values)
plt.yscale("log")
plt.show()
```



In []:

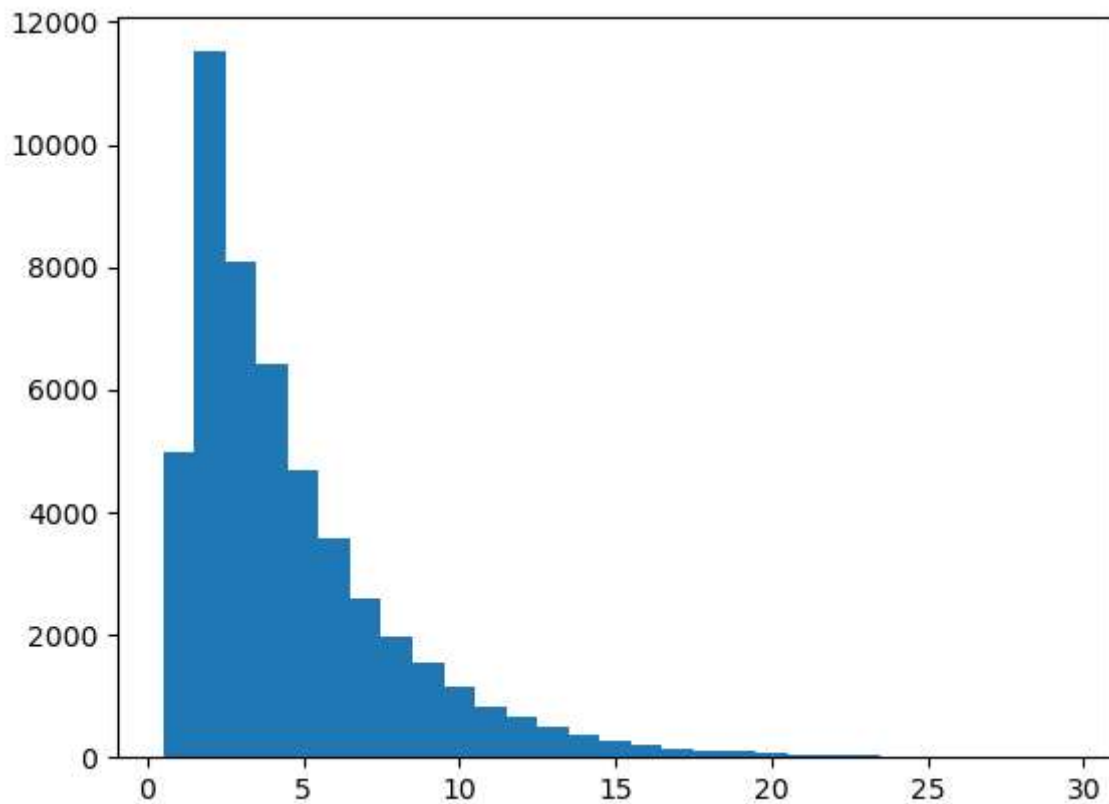
In []: '''3 Absorbing state. Consider now that node 3 is an absorbing state and we want to estimate the waiting time until the process arrives at $X_i = 3$ from any other node. a) Starting from each one of $X_0 = 1$ and $X_0 = 2$, run multiple simulation of the Markov chain (Problem 1, part b) until $X_i = 3$ and store the arrival time until this happens. Make a histogram of the arrival time for both $X_0 = 1$ and $X_0 = 2$ and compute the mean.

```
In [40]: import numpy as np
import matplotlib.pyplot as plt

arrival_durations = []

for _ in range(1, 50000):
    current_position = 1
    position_history = []
    while current_position != 3:
        current_position = np.random.choice([1, 2, 3], p=transition_matrix[current_position])
        position_history.append(current_position)
    arrival_durations.append(len(position_history))

np.mean(arrival_durations)
plt.hist(arrival_durations, bins=np.arange(0.5, 30, 1))
plt.show()
np.mean(arrival_durations)
```



Out[40]: 4.617032340646813

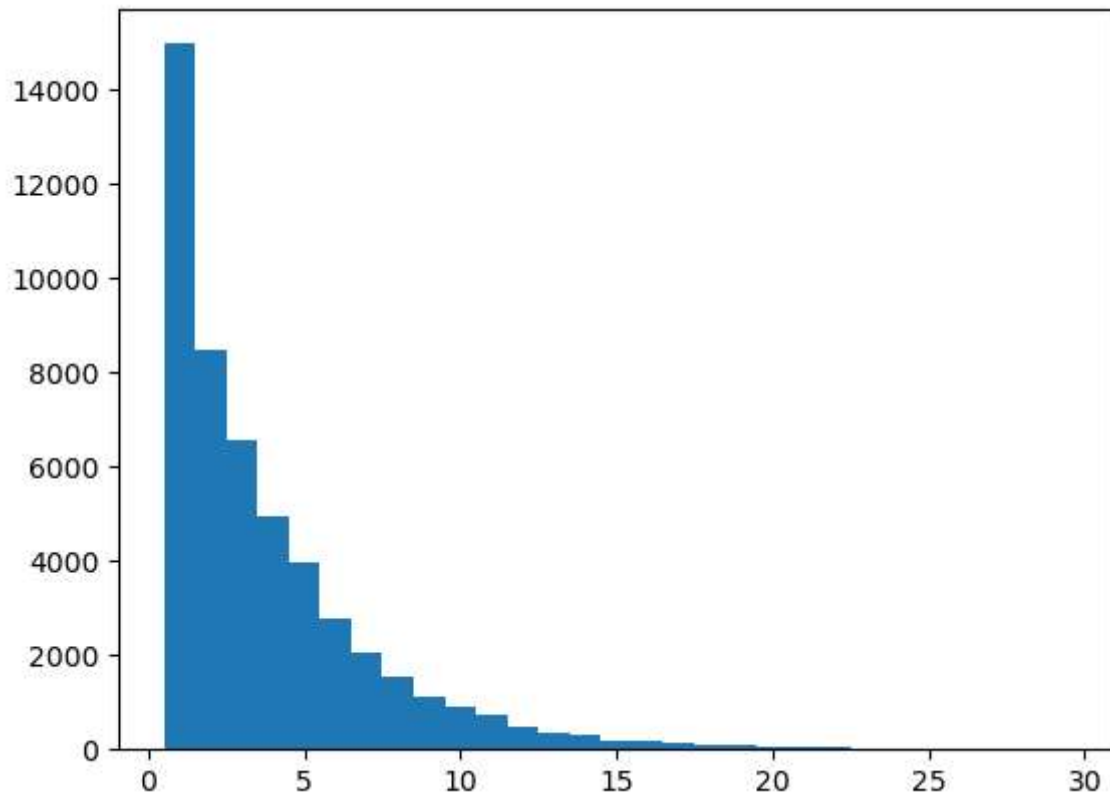
In []:

```
In [41]: import numpy as np
import matplotlib.pyplot as plt

arrival_durations = []

for _ in range(1, 50000):
    current_position = 2
    position_history = []
    while current_position != 3:
        current_position = np.random.choice([1, 2, 3], p=transition_matrix[current_position])
        position_history.append(current_position)
    arrival_durations.append(len(position_history))

np.mean(arrival_durations)
plt.hist(arrival_durations, bins=np.arange(0.5, 30, 1))
plt.show()
np.mean(arrival_durations)
```



Out[41]: 3.847076941538831

In []:

In []:

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