

## ✓ Markov Chains

### ✓ Prompt

Original Prompt can be found [here](#). A copy of the prompt along with the completed exercise can be found under [Applications](#).

### ✓ Summary

Every day for lunch you have either a \$9 sandwich (state 1), a \$7 burrito (state 2), or a \$5 pizza (state 3). Suppose your lunch choices from one day to the next follow a Markov chain with transition matrix. You start out with eating pizza on Monday (Day 0).

$$\mathbf{P} = \begin{bmatrix} 0 & 0.5 & 0.5 \\ 0.1 & 0.4 & 0.5 \\ 0.2 & 0.3 & 0.5 \end{bmatrix}$$

$T$  = the first day you have a sandwich.

$V$  = the number of days you have a burrito this five-day work week.

$X_n$  = the cost of your lunch on day  $n$ .

$W = X_0 + \dots + X_4$  = the total lunch cost for this five-day work week.

### ✓ Application

```
pip install symbulate
```

```
Requirement already satisfied: symbulate in /usr/local/lib/python3.10/dist-packages (0.5.7)
Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from symbulate) (1.23.5)
Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-packages (from symbulate) (1.11.4)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-packages (from symbulate) (3.7.1)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (1.2.0)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (4.47.2)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (1.4.5)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (23.2)
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (9.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (3.1.1)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib->symbulate) (2.8.2)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.7->matplotlib->symbulate) (1
```

```
from symbulate import *
%matplotlib inline
```

### ✓ 1.

Approximate the marginal distribution, along with the expected value and standard deviation, of each of the following:

$X_4$       $T$       $V$       $W$

```
#Set up markov chain
states = [9, 7, 5]

TransitionMatrix = [[0, 0.5, 0.5],
                    [0.1, 0.4, 0.5],
                    [0.2, 0.3, 0.5]]

InitialDistribution = [0, 0, 1] # pizza on the Monday

X = MarkovChain(TransitionMatrix, InitialDistribution, states)

#define any other simulation parameters
n_sims = 10000
days_in_week = 5 # Friday is day 4
```

$X_4$

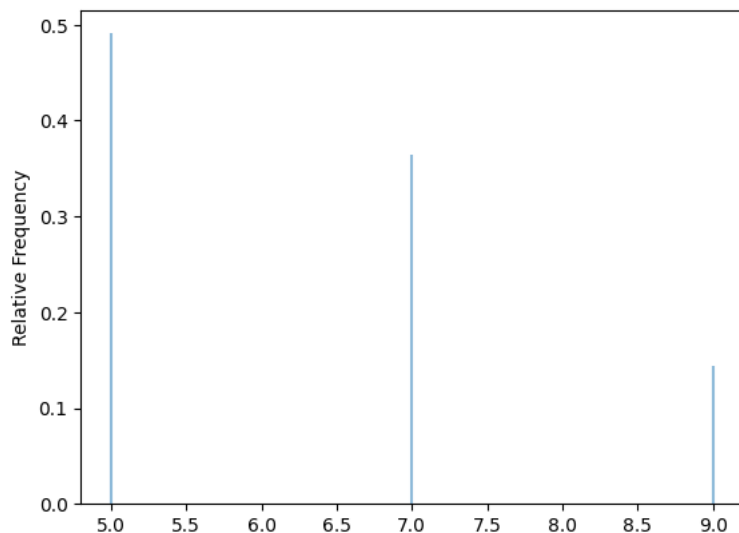
```
#run simulation and generate results
results = X.sim(10000)
results[4].tabulate()
```

	Value	Frequency
5		4908
7		3650
9		1442
<b>Total</b>		<b>10000</b>

```
results[4].tabulate(normalize=True)
```

	Value	Relative Frequency
5		0.4908
7		0.365
9		0.1442
<b>Total</b>		<b>1.0</b>

```
results[4].plot()
```



From the plot we can see the distribution of the cost of lunch on Friday. Half the time we are spending \$5 on pizza , while closely behind that we are spending \$7 on burritos. It is every so often that we splurge on a \$9 sandwich.

```
#calculate and present expected value
ev = results[4].mean()

print("The expected value is ${:.2f}".format(ev))

The expected value is $6.31
```

```
#calculate and present standard deviation
std_dev = results[4].std()

print("The standard deviation is ${:.2f}".format(std_dev))

    The standard deviation is $1.44
```

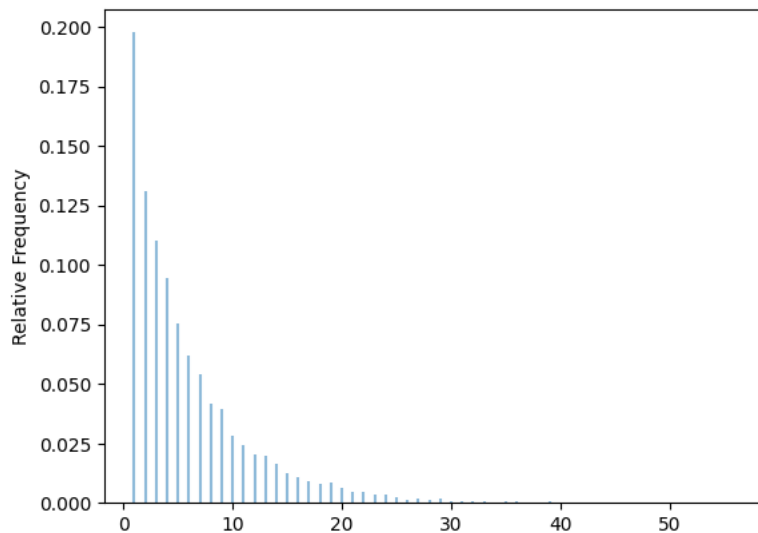
$T$

```
def first_sando(x):
    day = 1;
    while(x[day] != 9):
        day += 1
    return day

T = results.apply(first_sando)
T.tabulate()
```

	Value	Frequency
1		1976
2		1312
3		1102
4		943
5		756
6		620
7		539
8		418
9		395
10		284
11		243
12		202
13		196
14		163
15		123
16		106
17		90
18		80
19		83
...		...
56		1
<b>Total</b>		<b>10000</b>

T.plot()



From this plot we can see that the distribution of first sandwich day is skewed. We are very likely to have a sandwich within the first few days, but it is possible to go a few weeks without having the first one.

```
#calculate and present expected value
ev = T.mean()

print("The expected value is {:.2f} days".format(ev))

    The expected value is 6.00 days

#calculate and present standard deviation
std_dev = T.std()

print("The standard deviation is {:.2f} days".format(std_dev))

    The standard deviation is 5.75 days
```

$V$

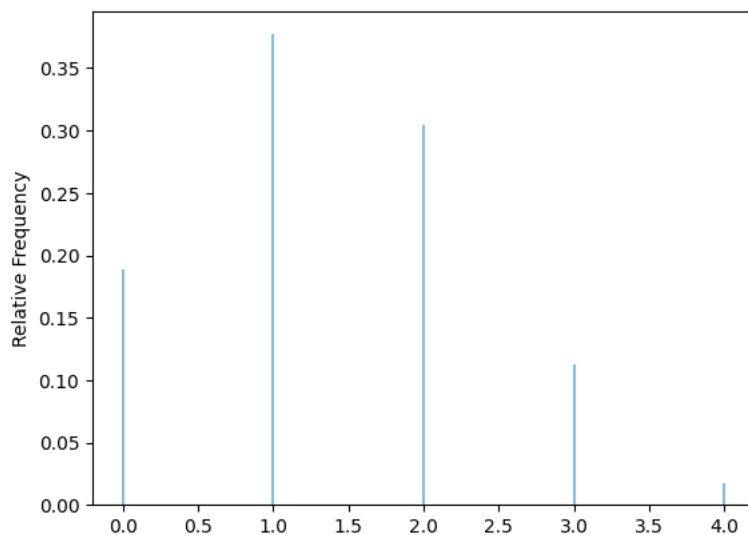
```
def burrito_count(x):
    count = 0
    for day in range(1,days_in_week):
        if x[day] == 7:
            count +=1

    return count
```

```
V = results.apply(burrito_count)
V.tabulate()
```

	Value	Frequency
0		1886
1		3770
2		3045
3		1123
4		176
<b>Total</b>		<b>10000</b>

```
V.plot()
```



This plot of our burrito count distribution is showing how it is likely we will have at least 1 or 2 burritos in this week but less likely we will have all burritos after one day of pizza or no burritos at all.

```
#calculate and present expected value
ev = V.mean()

print("The expected value is {:.2f} burritos".format(ev))

    The expected value is 1.39 burritos
```

```
#calculate and present standard deviation
std_dev = V.std()

print("The standard deviation is {:.2f} burritos".format(std_dev))

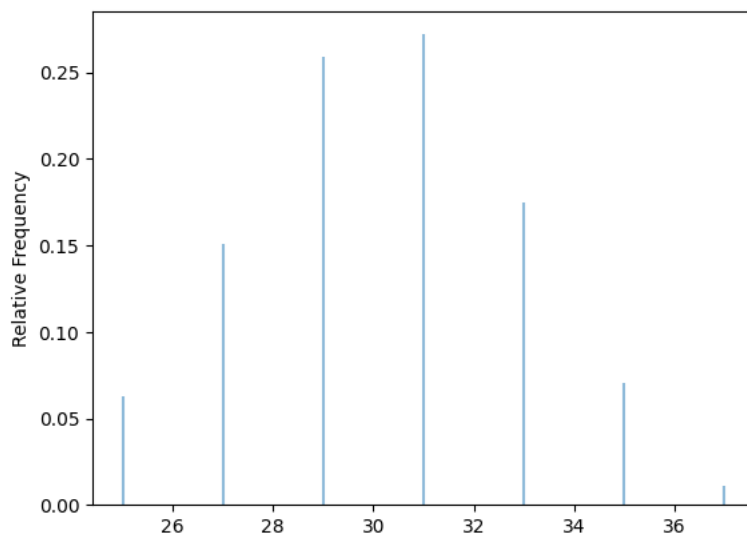
The standard deviation is 0.97 burritos
```

$W$

```
W = (X[0] + X[1] + X[2] + X[3] + X[4]).sim(10000)
W.tabulate()
```

Value	Frequency
25	624
27	1509
29	2588
31	2718
33	1750
35	702
37	109
<b>Total</b>	<b>10000</b>

```
W.plot()
```



This distribution shows how the cost of lunches this week is likely to hover around \$29-\$39.

```
#calculate and present expected value
ev = W.mean()

print("The expected value is {:.2f} for lunch this week".format(ev))

The expected value is $30.20 for lunch this week
```

```
#calculate and present standard deviation
std_dev = W.std()

print("The standard deviation is {:.2f} for lunch this week".format(std_dev))

The standard deviation is $2.70 for lunch this week
```

✓ 2.

Approximate the joint distribution, along with the correlation, of each of the following:

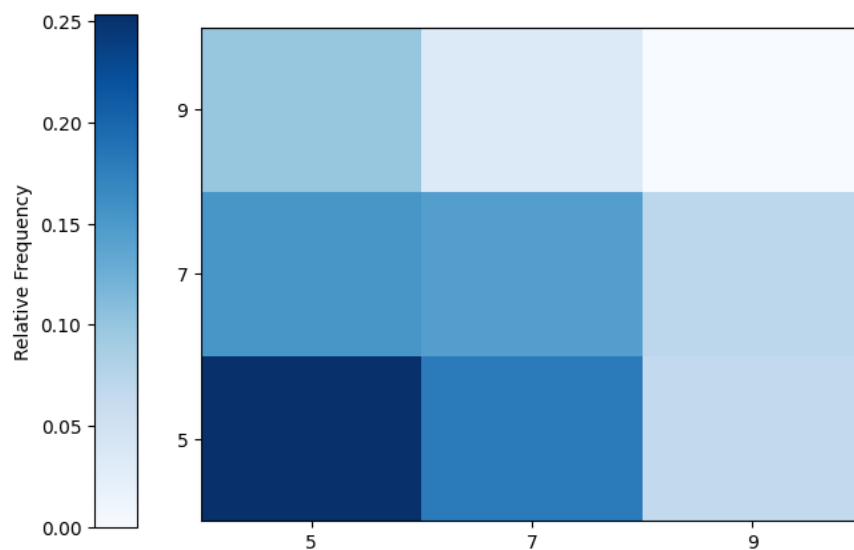
$X_4$  and  $X_5$   
 $T$  and  $V$

$T$  and  $W$  $W$  and  $V$  $X_4$  and  $X_5$ 

```
result = (X[4] & X[5]).sim(10000)
result.tabulate(normalize = True)
```

	Value	Relative Frequency
(5, 5)		0.2533
(5, 7)		0.1529
(5, 9)		0.0993
(7, 5)		0.1796
(7, 7)		0.1438
(7, 9)		0.0339
(9, 5)		0.0661
(9, 7)		0.0711
<b>Total</b>		<b>1.0000000000000002</b>

```
result.plot(type="tile")
```



This tile plot shows us how the cost of lunch varies from Friday to next Monday jointly.

```
correlation = result.corr()
```

```
print("The correlation between the price of lunch of the two days is {:.4f}".format(correlation))
```

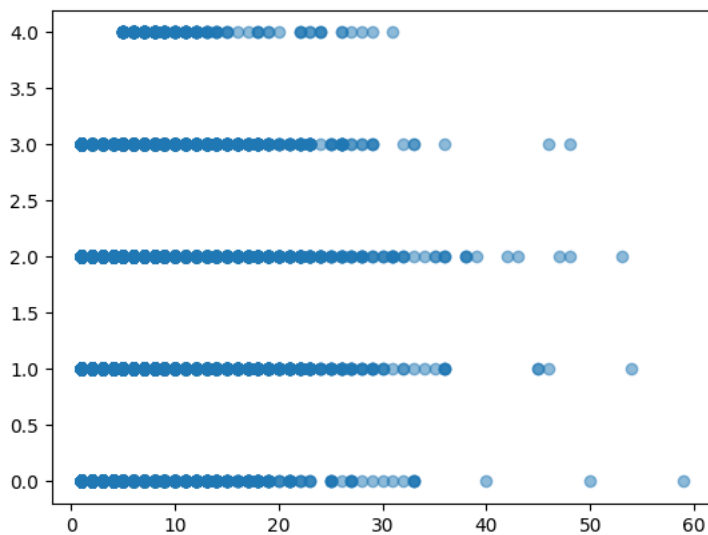
```
The correlation between the price of lunch of the two days is -0.0931
```

 $T$  and  $V$ 

```
T_V = (X.apply(first_sando) & X.apply(burrito_count)).sim(10000)
T_V.tabulate()
```

Value	Frequency
(1, 0)	460
(1, 1)	836
(1, 2)	568
(1, 3)	167
(2, 0)	353
(2, 1)	521
(2, 2)	334
(2, 3)	62
(3, 0)	252
(3, 1)	493
(3, 2)	269
(3, 3)	66
(4, 0)	253
(4, 1)	374
(4, 2)	211
(4, 3)	45
(5, 0)	115
(5, 1)	266
(5, 2)	202
...	...
(59, 0)	1
<b>Total</b>	<b>10000</b>

T\_V.plot()



Here we can see how the joint distribution of first sandwich day and burrito count behave. They do not seem to be strongly related.

```
correlation = T_V.corr()
```

```
print("The correlation between the first day of sandwich and burrito count is {:.4f}".format(correlation))
```

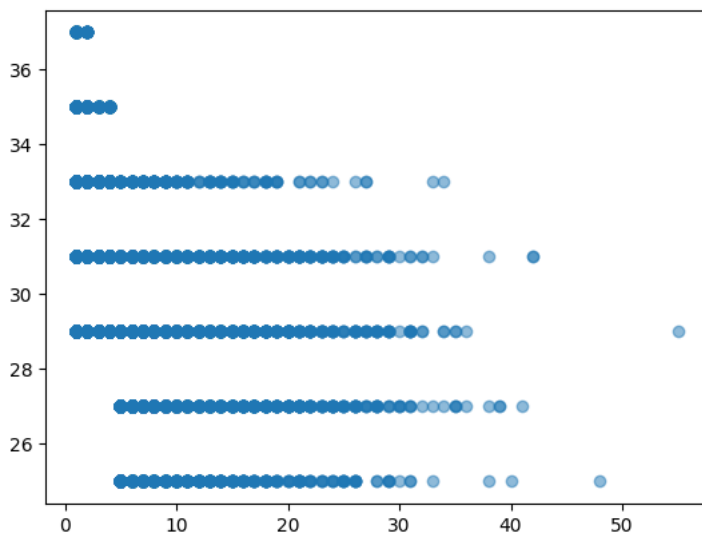
The correlation between the first day of sandwich and burrito count is 0.1969

$T$  and  $W$

```
T_W = (X.apply(first_sando) & (X[0] + X[1] + X[2] + X[3] + X[4])).sim(10000)
T_W.tabulate()
```

Value	Frequency
(1, 29)	245
(1, 31)	546
(1, 33)	658
(1, 35)	455
(1, 37)	69
(2, 29)	251
(2, 31)	471
(2, 33)	420
(2, 35)	148
(2, 37)	19
(3, 29)	271
(3, 31)	463
(3, 33)	292
(3, 35)	63
(4, 29)	246
(4, 31)	377
(4, 33)	236
(4, 35)	58
(5, 25)	141
...	...
(55, 29)	1
<b>Total</b>	<b>10000</b>

T\_W.plot()



Above we see how first sandwich and lunch cost are distributed. We can clearly see there is a relationship between the two. since sandwiches are more expensive, it is likely that when we have them the cost of that weeks lunch will also be significantly more.

```
correlation = T_W.corr()
```

```
print("The correlation between the first day of sandwich and weekly lunch cost is {:.4f}".format(correlation))
```

The correlation between the first day of sandwich and weekly lunch cost is -0.4754

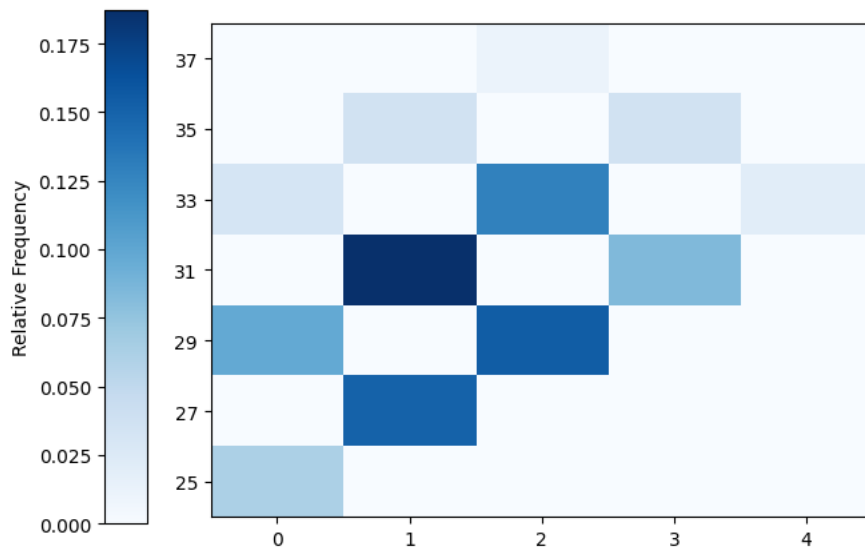
$W$  and  $V$

```
V_W = (X.apply(burrito_count) & (X[0] + X[1] + X[2] + X[3] + X[4])).sim(10000)
V_W.tabulate()
```



	Value	Frequency
(0, 25)		617
(0, 29)		975
(0, 33)		317
(1, 27)		1511
(1, 31)		1872
(1, 35)		364
(2, 29)		1549
(2, 33)		1286
(2, 37)		111
(3, 31)		838
(3, 35)		360
(4, 33)		200
<b>Total</b>		<b>10000</b>

```
V_W.plot(type="tile")
```



The tile plot above shows how burrito count and weekly cost vary jointly. We can see that as the burritos we consume increases, the cost for the week also increases.

```
correlation = V_W.corr()

print("The correlation between burrito count and weekly lunch cost is {:.4f}".format(correlation))

The correlation between burrito count and weekly lunch cost is 0.4584
```

✓ 3.

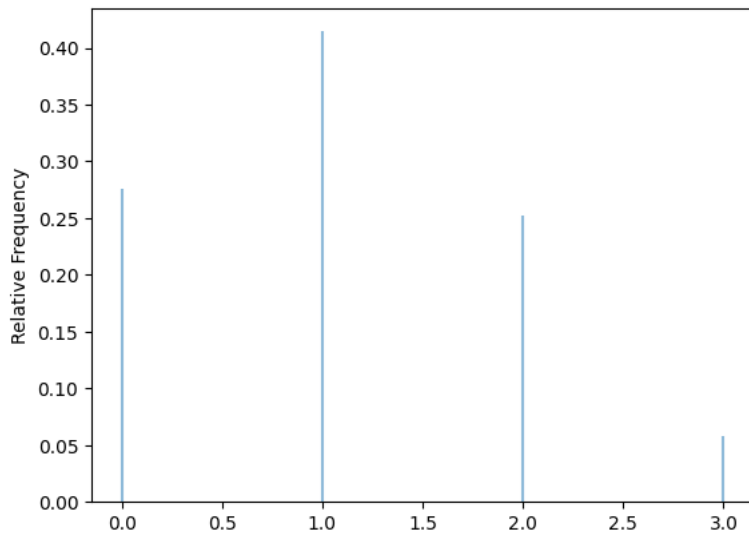
Approximate the conditional distribution of  $V$  given  $T = 4$ , along with its (conditional) mean and standard deviation.

$V$

```
T_4 = ((X[0] != 9) & (X[1] != 9) & (X[2] != 9) & (X[3] != 9) & (X[4] == 9))
V_T4 = (X.apply(burrito_count) | T_4).sim(10000)
V_T4.tabulate()
```

	Value	Frequency
0		2755
1		4144
2		2517
3		584
<b>Total</b>		<b>10000</b>

```
V_T4.plot()
```



This distribution shows the likely amount of burritos to be consumed during the week given we know we will eat a sandwich on Friday.

$\mu$

```
#calculate and present expected value
ev = V_T4.mean()

print("The expected value is {:.2f} burritos given Friday is sandwich day.".format(ev))

    The expected value is 1.09 burritos given Friday is sandwich day.
```

$\sigma$

```
#calculate and present standard deviation
std_dev = V_T4.std()

print("The standard deviation is {:.2f} burritos given Friday is sandwich day.".format(std_dev))

    The standard deviation is 0.87 burritos given Friday is sandwich day.
```

✓ 4.

Your choice. Choose at least one other joint, conditional, or marginal distribution to investigate. You can work with  $X_n, T, V, W$ , but you are also welcome to define other random variables in this context. You can also look at time frames other than a single week.

Double-click (or enter) to edit

```
W_V2 = ((X[0] + X[1] + X[2] + X[3] + X[4]) | (X.apply(burrito_count) == 2)).sim(10000)
W_V2.tabulate()
```

	Value	Frequency
	29	5380
	33	4275
	37	345
	<b>Total</b>	<b>10000</b>

```
W_V2.plot()
```

