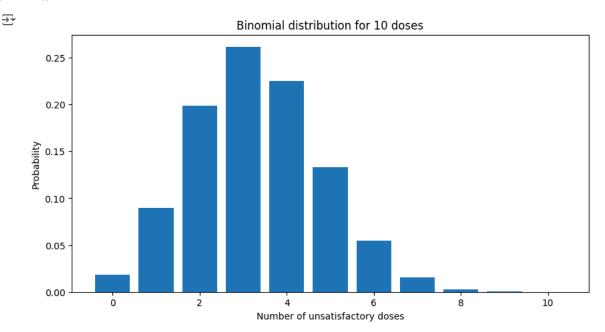
→ Business Context

A leading pharmaceutical company has tested five batches of a vaccine. With 300,000 doses already administered, the sixth batch of 60,000 doses needs quality assurance testing for effectiveness and curing time. Previous data shows each dose is twice as likely to be satisfactory as unsatisfactory. This test is to ensure the sixth batch's quality, not a clinical trial. Objective Analyze random samples from the sixth batch to infer its quality and curing time. Your tasks include:

Task 1: Analyze 10 randomly selected doses to determine:

- · Probability distribution of unsatisfactory doses
- Probability that exactly 3 out of 10 doses are unsatisfactory
- · Probability that at most 3 out of 10 doses are unsatisfactory
- · Probability that more than 8 out of 10 doses are satisfactory

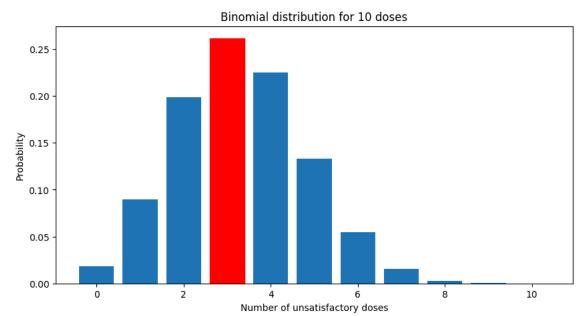
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import binom
import scipy.stats as stats
n=10
p=0.33
k=np.arange(0,11)
binomial_prob=binom.pmf(k=k,n=n,p=p)
binomial_prob
⇒ array([1.82283780e-02, 8.97815635e-02, 1.98993465e-01, 2.61364552e-01,
            2.25280640e-01, 1.33150945e-01, 5.46515074e-02, 1.53816609e-02,
            2.84101573e-03, 3.10956945e-04, 1.53157899e-05])
plt.figure(figsize=(10,5))
plt.bar(k,binomial_prob)
plt.title('Binomial distribution for 10 doses')
plt.xlabel('Number of unsatisfactory doses')
plt.ylabel('Probability')
plt.show()
```



```
# Probability that exactly 3 out of 10 doses are unsatisfactory
plt.figure(figsize=(10,5))
barl=plt.bar(k,binomial_prob)
plt.title('Binomial distribution for 10 doses')
plt.xlabel('Number of unsatisfactory doses')
```

plt.ylabel('Probability');
barl[3].set_color('r')





binom.pmf(k=3,n=n,p=p)

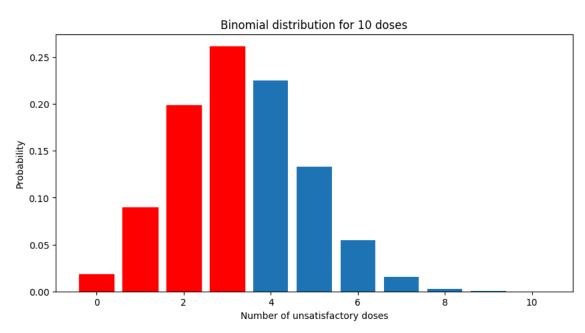
→ 0.2613645515525908

 ${\tt binomial_prob[3]}$

0.2613645515525908

```
# Probability that at most 3 out of 10 doses are unsatisfactory
plt.figure(figsize=(10,5))
barl=plt.bar(k,binomial_prob)
plt.title('Binomial distribution for 10 doses')
plt.xlabel('Number of unsatisfactory doses')
plt.ylabel('Probability');
for i in range(0,4):
    barl[i].set_color('r')
plt.show()
```



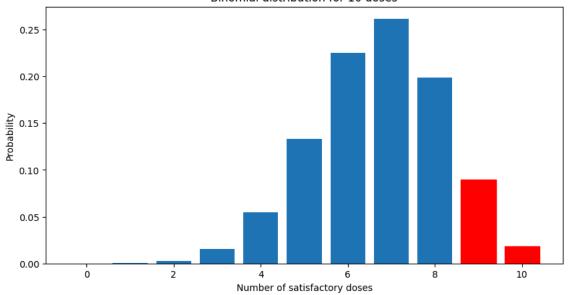


sum(binomial_prob[:4])

```
→ 0.5683679584925132
prob_atmost3=binom.cdf(k=3,n=n,p=p)
prob_atmost3
→ 0.5683679584925142
# Probability that more than 8 out of 10 doses are satisfactory
p_satisfactory=1-.33
n=10
k=np.arange(0,11)
bi_prob_satis=binom.pmf(k=k,n=n,p=p_satisfactory)
bi_prob_satis
array([1.53157899e-05, 3.10956945e-04, 2.84101573e-03, 1.53816609e-02,
            5.46515074e-02, 1.33150945e-01, 2.25280640e-01, 2.61364552e-01,
            1.98993465e-01, 8.97815635e-02, 1.82283780e-02])
plt.figure(figsize=(10,5))
barl=plt.bar(k,bi_prob_satis)
plt.title('Binomial distribution for 10 doses')
plt.xlabel('Number of satisfactory doses')
plt.ylabel('Probability');
barl[9].set_color('r')
barl[10].set_color('r')
```



Binomial distribution for 10 doses



prob_atmost1=binom.cdf(k=8,n=n,p=1-p)
1-prob_atmost1

→ 0.10800994155329091

Task 2: For 20 doses requested by the New York City administration:

- · Probability that at least 11 out of 20 doses are unsatisfactory
- · Probability that at most 5 out of 20 doses are unsatisfactory
- · Probability that at least 13 out of 20 doses are satisfactory

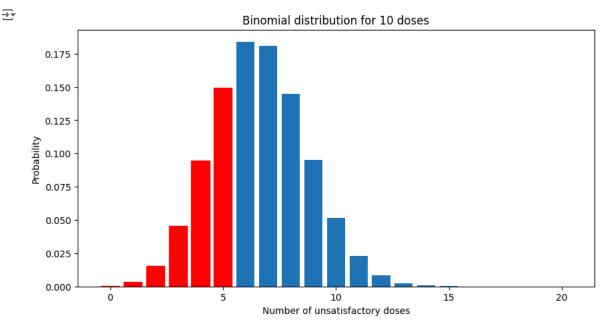
Double-click (or enter) to edit

```
n=20
p=.33
k=np.arange(0,21)
k
```

```
→ array([ 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16,
             17, 18, 19, 20])
binomial_prob=binom.pmf(k=k,n=n,p=p)
binomial_prob
⇒ array([3.32273766e-04, 3.27314456e-03, 1.53153854e-02, 4.52603926e-02,
            9.47428368e-02, 1.49326023e-01, 1.83871596e-01, 1.81127244e-01, 1.44969380e-01, 9.52037718e-02, 5.15805510e-02, 2.30957691e-02,
             8.53164605e-03, 2.58594094e-03, 6.36836201e-04, 1.25466237e-04,
             1.93115010e-05, 2.23803172e-06, 1.83719022e-07, 9.52510246e-09,
             2.34573419e-10])
# Probability that at least 11 out of 20 doses are unsatisfactory
plt.figure(figsize=(10,5))
barl=plt.bar(k,binomial prob)
plt.title('Binomial distribution for 10 doses')
plt.xlabel('Number of unsatisfactory doses')
plt.ylabel('Probability');
for i in range(11,21):
    barl[i].set_color('r')
plt.show()
```



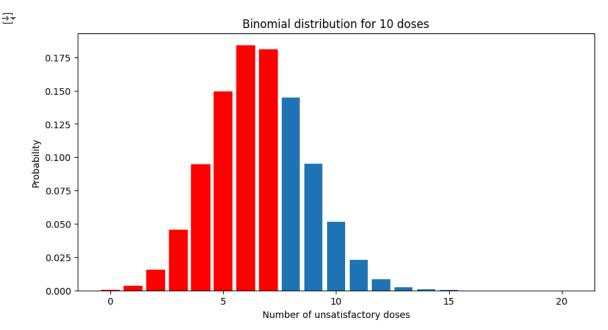
0.175 - 0.150 - 0.125 - 0.075 - 0.050 - 0.025 - 0.000 - 0.025 - 0.000



prob_atmost5=binom.cdf(k=5,n=n,p=p)
prob_atmost5

0.30825005639386527

```
# Probability that at least 13 out of 20 doses are satisfactory
plt.figure(figsize=(10,5))
barl=plt.bar(k,binomial_prob)
plt.title('Binomial distribution for 10 doses')
plt.xlabel('Number of unsatisfactory doses')
plt.ylabel('Probability');
for i in range(0,8):
    barl[i].set_color('r')
plt.show()
```



prob_atmost6=binom.cdf(k=7,n=n,p=p)
prob_atmost6

→ 0.6732488959678964

Task 3: Analyze the time of effect for 50 doses given to volunteers.

- Probability that the time of effect is less than 11.5 hours
- Probability that the time of effect is more than 10 hours
- Calculate the 90th percentile of the time of effect
- · Use dataset doses.csv.

Probability that the time of effect is less than 11.5 hours
df=pd.read_csv('doses.csv')
df.head()

_		drug_serial_number	time_of_effect
	0	672	5.8
	1	895	17.3
	2	518	16.7
	3	448	13.1
	4	402	13.6

len(df[df['time_of_effect'] <11.5])/len(df)*100</pre>

→ 34.0

df.info()

```
cclass 'pandas.core.frame.DataFrame'>
RangeIndex: 50 entries, 0 to 49
Data columns (total 2 columns):
# Column Non-Null Count Dtype
-------
0 drug_serial_number 50 non-null int64
1 time_of_effect 50 non-null float64
dtypes: float64(1), int64(1)
memory usage: 928.0 bytes
```

mean=df['time_of_effect'].mean()
mean

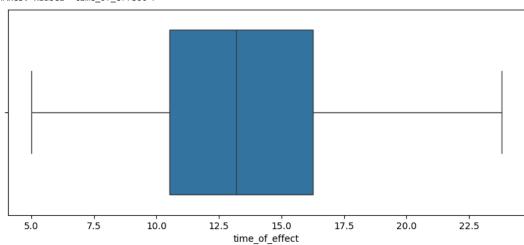
→ 13.442

std=df['time_of_effect'].std()
std

4.7455280775371955

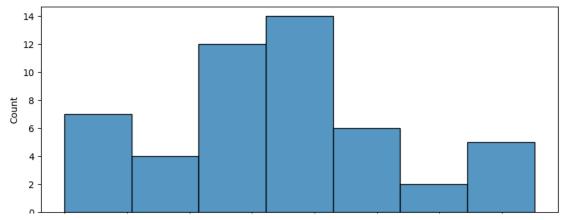
plt.figure(figsize=(10,4))
sns.boxplot(df['time_of_effect'],orient='h')

```
→ <Axes: xlabel='time_of_effect'>
```



```
plt.figure(figsize=(10,4))
sns.histplot(df['time_of_effect'])
```

→ <Axes: xlabel='time_of_effect', ylabel='Count'>



df['time_of_effect'].mean(), df['time_of_effect'].median()

→ (13.442, 13.2)

 $\verb|stats.norm.cdf(11.5,df['time_of_effect'].mean(),df['time_of_effect'].std()||$

→ 0.3411864031732611

stats.norm.cdf(11.5,mean,std)

→ 0.3411864031732611

1-stats.norm.cdf(10,mean,std)

0.7658704228920478

from scipy.stats import norm
norm.cdf(50,45,1.7)

→ 0.9983651589936532

norm.ppf(0.998365,45,1.7)

49.999948792243856

norm.ppf(0.90,mean,std)

→ 19.52363893710365