

Probability Distribution Function Pratical Implementation

Probability Mass Funcation

```
In [11]: 1 import seaborn as sns
          2 import statistics
          3 import matplotlib.pyplot as plt
          4 import random
          5 import pandas as pd
```

```
# if Single Dice is Rolled for 1000 times
## Chances of getting numbers
```

```
In [2]: 1 L=[]
          2 for i in range(1000):
          3     L.append(random.randint(1,6))
```

```
In [5]: 1 L[:10]
```

```
Out[5]: [6, 6, 2, 4, 6, 5, 1, 6, 6, 1]
```

```
# if Two Dice is Rolled for 1000 times
## Chances of getting numbers
```

```
In [6]: 1 L = []
          2 for i in range(10000):
          3     a = random.randint(1,6)
          4     b = random.randint(1,6)
          5
          6     L.append(a + b)
```

```
In [9]: 1 L[:5]
```

```
Out[9]: [10, 8, 8, 5, 8]
```

```
In [12]: 1 pd.Series(L).value_counts()
```

```
Out[12]: 7      1641
          8      1410
          6      1387
          9      1109
          5      1093
          10      874
          4      841
          3      571
          11      568
          2      254
          12      252
          dtype: int64
```

```
In [13]: 1 pd.Series(L).value_counts().sum()
```

```
Out[13]: 10000
```

```
In [14]: 1 pd.Series(L).value_counts()/pd.Series(L).value_counts().sum()
```

```
Out[14]: 7      0.1641
8      0.1410
6      0.1387
9      0.1109
5      0.1093
10     0.0874
4      0.0841
3      0.0571
11     0.0568
2      0.0254
12     0.0252
dtype: float64
```

```
In [15]: 1 (pd.Series(L).value_counts()/pd.Series(L).value_counts().sum()).sort_index()
```

```
Out[15]: 2      0.0254
3      0.0571
4      0.0841
5      0.1093
6      0.1387
7      0.1641
8      0.1410
9      0.1109
10     0.0874
11     0.0568
12     0.0252
dtype: float64
```

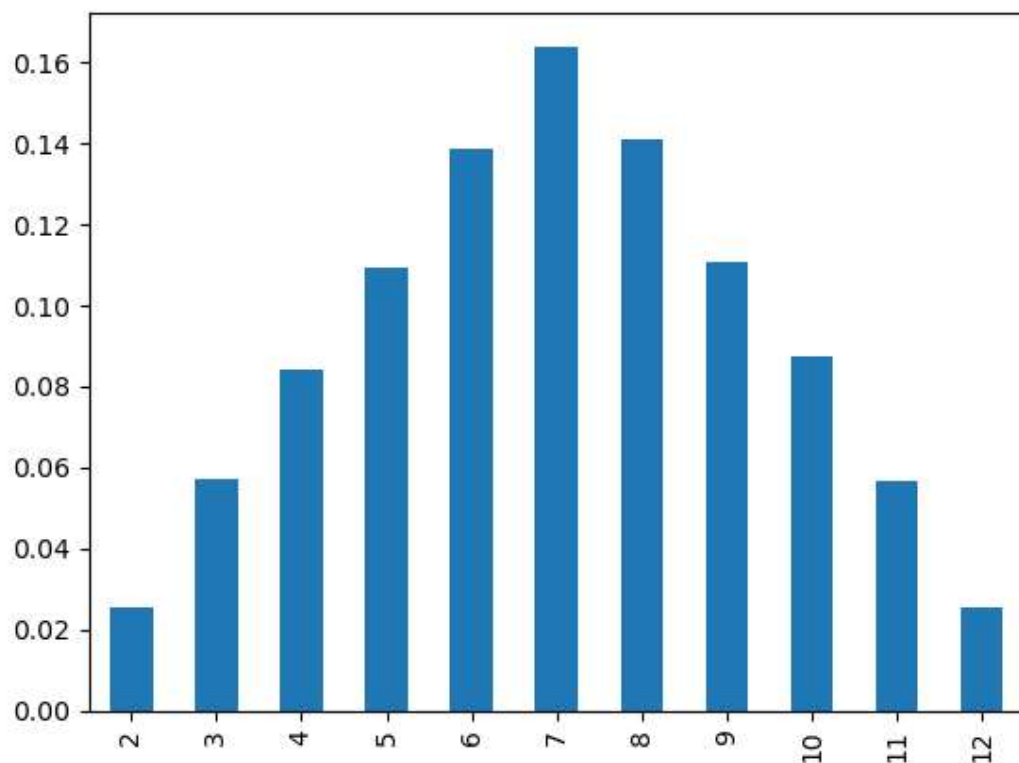
```
In [20]: 1 s = (pd.Series(L).value_counts()/pd.Series(L).value_counts().sum()).sort_index()
```

```
In [17]: 1 import numpy as np
2 np.cumsum(s)
```

```
Out[17]: 2      0.0254
3      0.0825
4      0.1666
5      0.2759
6      0.4146
7      0.5787
8      0.7197
9      0.8306
10     0.9180
11     0.9748
12     1.0000
dtype: float64
```

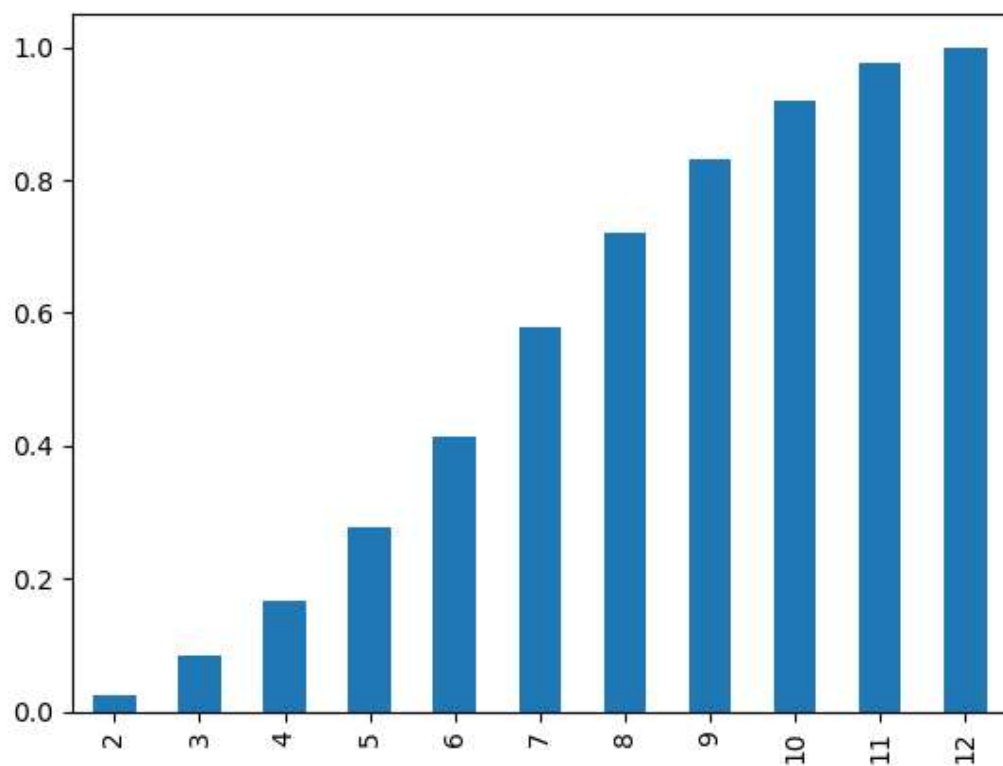
```
In [18]: 1 s.plot(kind='bar')
```

Out[18]: <Axes: >



```
In [19]: 1 np.cumsum(s).plot(kind='bar')
```

Out[19]: <Axes: >



```
In [ ]: 1
```

Probability Density Function

Parametric Density Estimation

```
In [35]: 1 from numpy.random import normal
          2
          3 sample = normal(loc=50, scale=5,size=1000)
```

```
In [36]: 1 sample
          55.53784104, 41.06298964, 46.33982049, 52.51954708, 45.0652093 ,
          48.62159399, 48.62449045, 47.56946958, 52.82961673, 41.6542016 ,
          44.4041443 , 46.52092917, 48.10721169, 50.39475177, 43.35602488,
          43.90552686, 59.61282112, 53.36249785, 55.5479493 , 51.63614222,
          52.40556726, 47.82609111, 49.68469971, 53.82622874, 49.26679037,
          51.35737457, 43.87720568, 54.28859473, 44.36625562, 43.92146206,
          49.25264189, 43.21754818, 55.25474992, 39.85602295, 50.27533111,
          46.7137934 , 50.90531791, 50.91395094, 32.875293 , 45.65092132,
          49.56763655, 57.36486996, 60.37704073, 46.68007223, 50.28255209,
          51.24170621, 55.91010554, 54.75528883, 45.82384594, 47.79932419,
          59.68112534, 52.6925637 , 48.90908203, 46.51047187, 52.27375947,
          45.14220822, 43.26596485, 52.06796534, 47.48795095, 44.74766323,
          56.70652037, 46.06053169, 60.07096964, 47.07644697, 52.45681659,
          49.30851503, 49.2073696 , 52.09024624, 55.80086697, 53.73898743,
          50.02330752, 55.25752193, 55.00881331, 56.18729407, 42.86896557,
          50.12429547, 43.74067699, 45.99803273, 49.26293691, 50.82826809,
          49.56305183, 54.00313897, 52.52011805, 52.19746119, 53.45362296,
          46.89485811, 50.98696345, 42.0030731 , 55.87449577, 46.69220442,
          47.7954326 , 57.67551738, 43.65415679, 52.09899674, 51.69887299,
          54.43743353, 45.08208384, 53.53044071, 48.41544376, 51.02919982,
```

```
In [37]: 1 sample.mean()
```

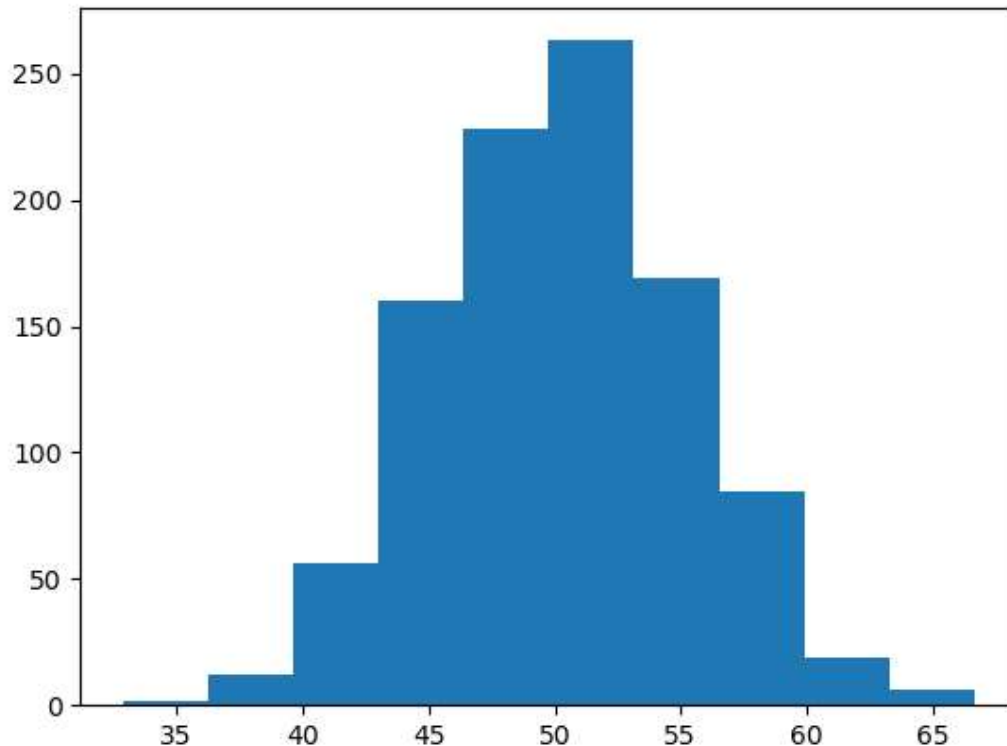
```
Out[37]: 50.19120634821261
```

```
In [38]: 1 sample.std()
```

```
Out[38]: 5.015112308145442
```

```
In [39]: 1 # plot histogram to understand the distribution of data
         2 plt.hist(sample,bins=10)
```

```
Out[39]: (array([ 2., 12., 56., 160., 228., 263., 169., 85., 19., 6.]),
          array([32.875293, 36.25253122, 39.62976944, 43.00700766, 46.38424587,
                49.76148409, 53.13872231, 56.51596053, 59.89319875, 63.27043697,
                66.64767519])),
          <BarContainer object of 10 artists>)
```



```
In [40]: 1 # calculate sample mean and sample std dev
         2 sample_mean = sample.mean()
         3 sample_std = sample.std()
```

```
In [41]: 1 # fit the distribution with the above parameters
         2
         3 from scipy.stats import norm
         4 dist = norm(60, 12)
```

```
In [42]: 1 values = np.linspace(sample.min(),sample.max(),100)
```

```
In [43]: 1 sample.max()
```

```
Out[43]: 66.6476751889642
```

```
In [44]: 1 sample.min()
```

```
Out[44]: 32.875292998372515
```

```
In [45]: 1 probabilities = [dist.pdf(value) for value in values]
```

```
In [48]: 1 import seaborn as sns
2         sns.distplot(sample)
```

C:\Users\DELL\AppData\Local\Temp\ipykernel_1036\1482356190.py:2: UserWarning:

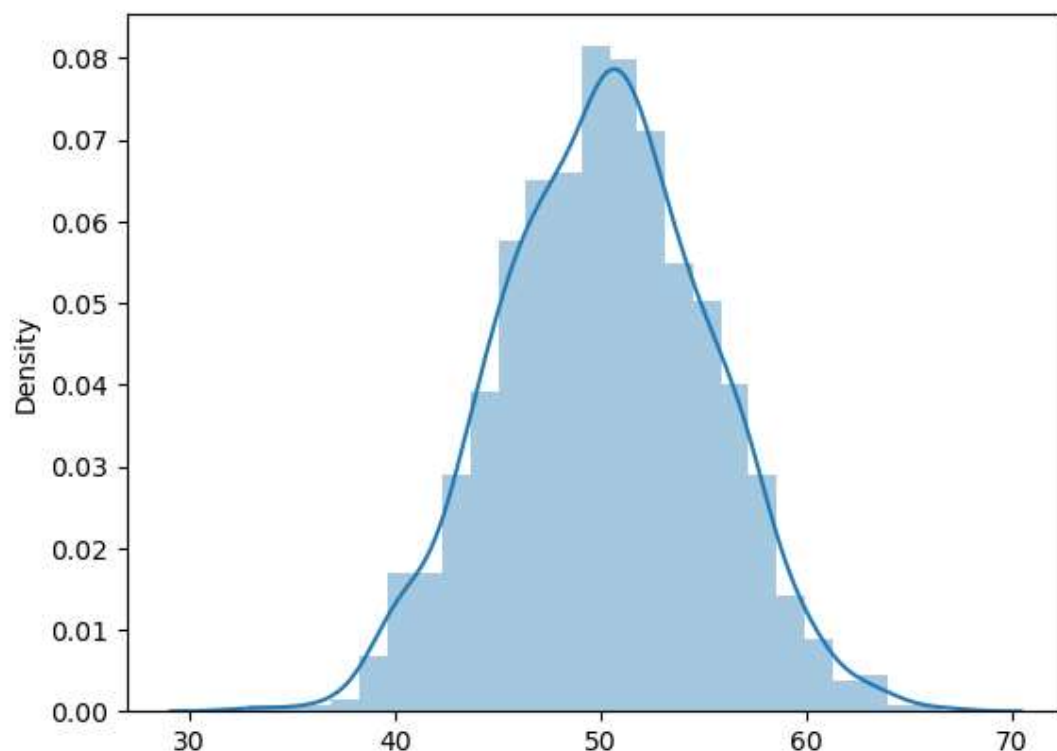
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(sample)
```

Out[48]: <Axes: ylabel='Density'>



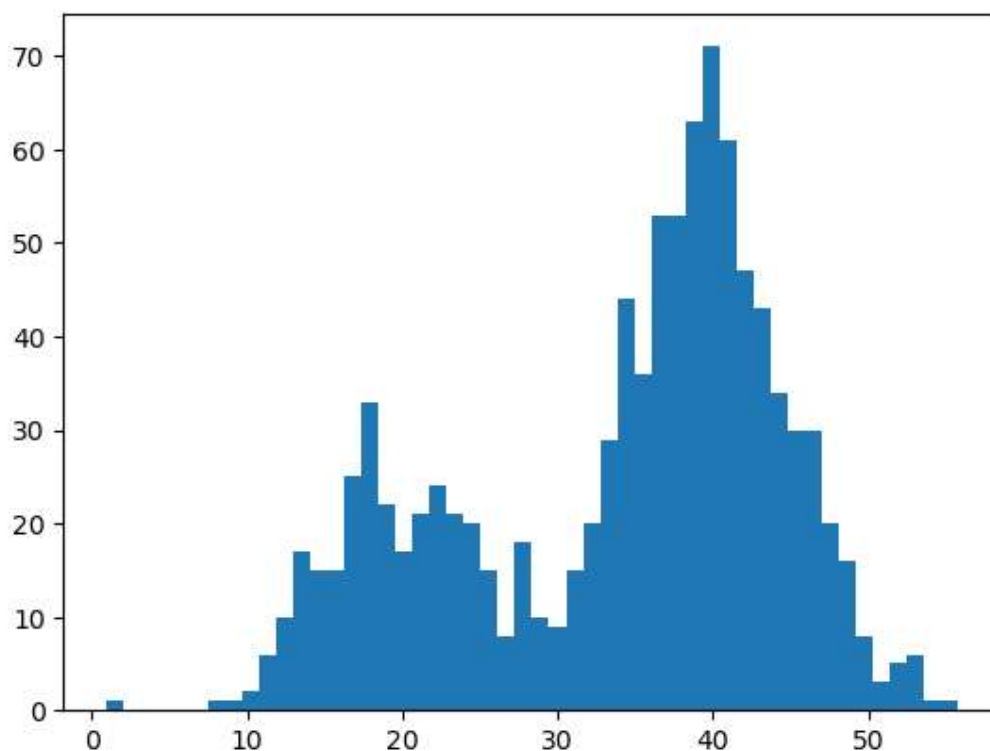
Non Parametric Density Estimation

```
In [50]: 1 # generate a sample
2         sample1 = normal(loc=20, scale=5, size=300)
3         sample2 = normal(loc=40, scale=5, size=700)
4         sample = np.hstack((sample1, sample2))
```



```
In [52]: 1 # plot histogram bins=50
         2 plt.hist(sample,bins=50)
```

```
Out[52]: (array([ 1.,  0.,  0.,  0.,  0.,  0.,  1.,  1.,  2.,  6., 10., 17., 15.,
                  15., 25., 33., 22., 17., 21., 24., 21., 20., 15.,  8., 18., 10.,
                  9., 15., 20., 29., 44., 36., 53., 53., 63., 71., 61., 47., 43.,
                  34., 30., 30., 20., 16.,  8.,  3.,  5.,  6.,  1.,  1.]),
          array([ 0.93130587,  2.02772532,  3.12414478,  4.22056423,  5.31698368,
                  6.41340314,  7.50982259,  8.60624205,  9.7026615 , 10.79908096,
                  11.89550041, 12.99191987, 14.08833932, 15.18475878, 16.28117823,
                  17.37759769, 18.47401714, 19.5704366 , 20.66685605, 21.76327551,
                  22.85969496, 23.95611441, 25.05253387, 26.14895332, 27.24537278,
                  28.34179223, 29.43821169, 30.53463114, 31.6310506 , 32.72747005,
                  33.82388951, 34.92030896, 36.01672842, 37.11314787, 38.20956733,
                  39.30598678, 40.40240623, 41.49882569, 42.59524514, 43.6916646 ,
                  44.78808405, 45.88450351, 46.98092296, 48.07734242, 49.17376187,
                  50.27018133, 51.36660078, 52.46302024, 53.55943969, 54.65585915,
                  55.7522786 ]),
          <BarContainer object of 50 artists>)
```



```
In [53]: 1 from sklearn.neighbors import KernelDensity
         2
         3 model = KernelDensity(bandwidth=5, kernel='gaussian')
         4
         5 # convert data to a 2D array
         6 sample = sample.reshape((len(sample), 1))
         7
         8 model.fit(sample)
```

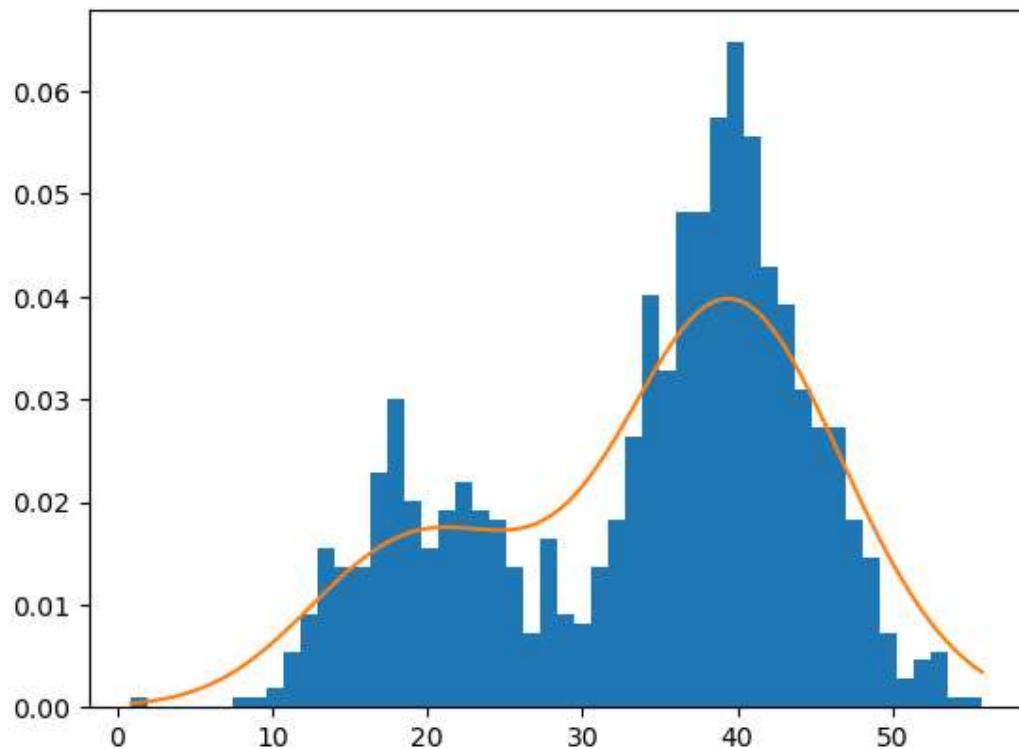
```
Out[53]: KernelDensity
          KernelDensity(bandwidth=5)
```



```
In [54]: 1 values = np.linspace(sample.min(), sample.max(), 100)
          2 values = values.reshape((len(values), 1))
```

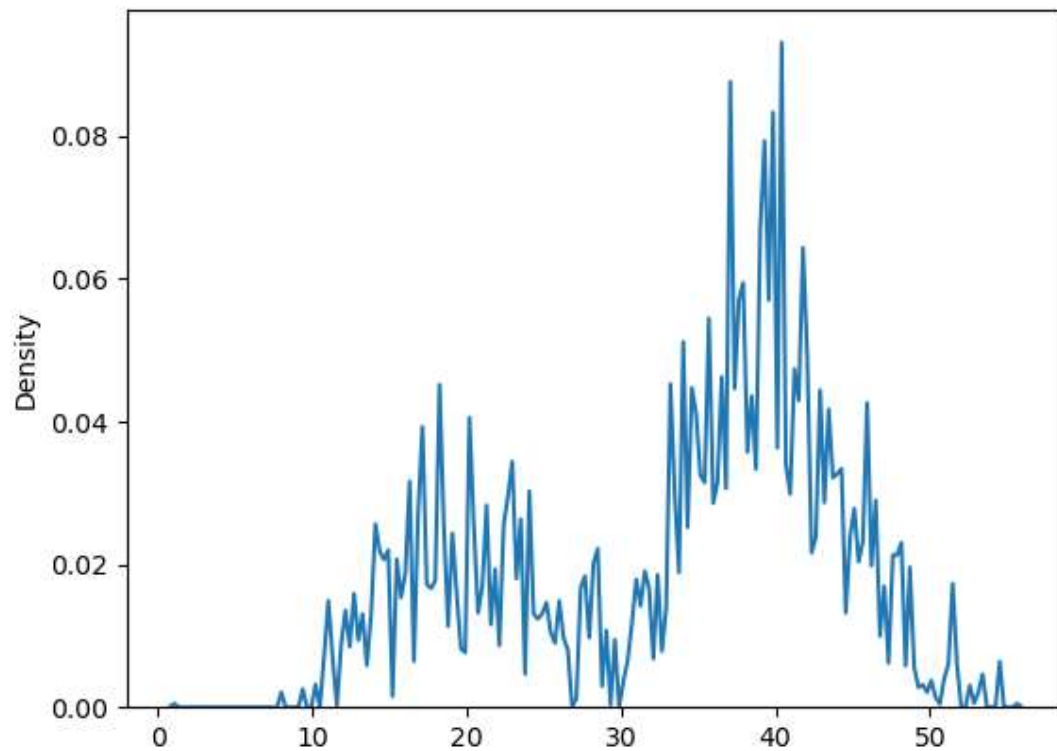
```
In [55]: 1 probabilities = model.score_samples(values)
          2 probabilities = np.exp(probabilities)
```

```
In [56]: 1 plt.hist(sample, bins=50, density=True)
          2 plt.plot(values[:,], probabilities)
          3 plt.show()
```



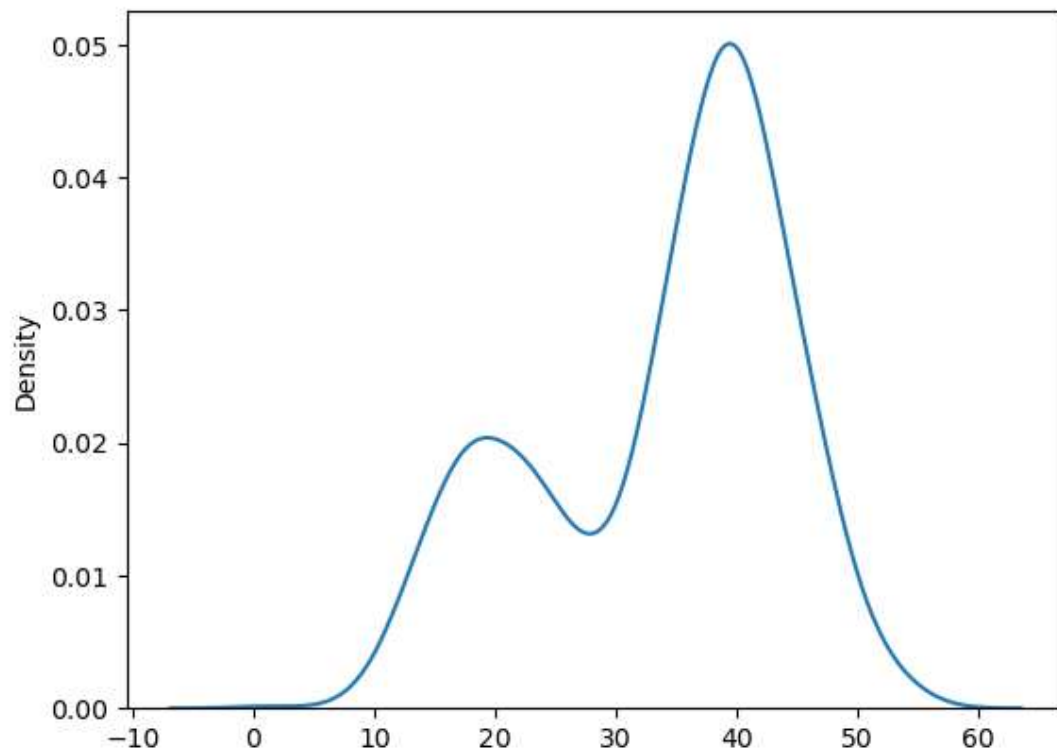
```
In [57]: 1 sns.kdeplot(sample.reshape(1000),bw_adjust=0.02)
```

```
Out[57]: <Axes: ylabel='Density'>
```



```
In [59]: 1 sns.kdeplot(sample.reshape(1000),bw_adjust=1)
```

```
Out[59]: <Axes: ylabel='Density'>
```



Performing PMF and PDF on iris Dataset

```
In [61]: 1 df = sns.load_dataset('iris')
```

```
In [62]: 1 df.head()
```

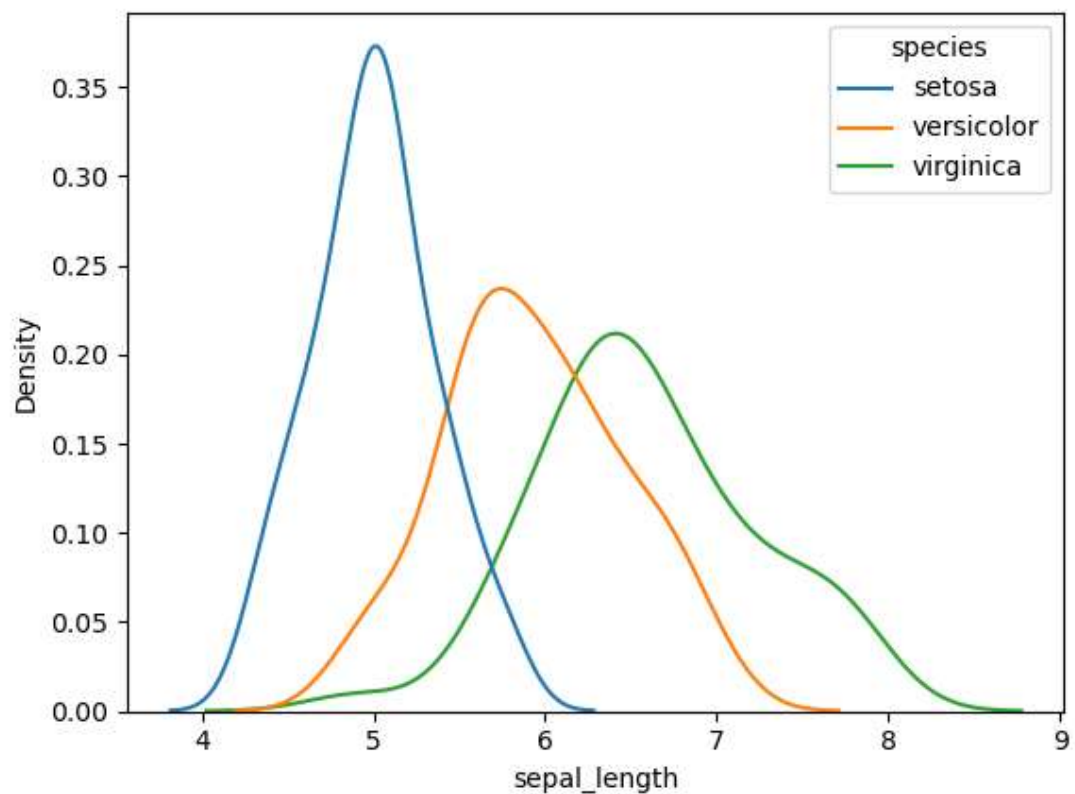
Out[62]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

```
In [ ]: 1 ## PDF on Continuous Variable
```

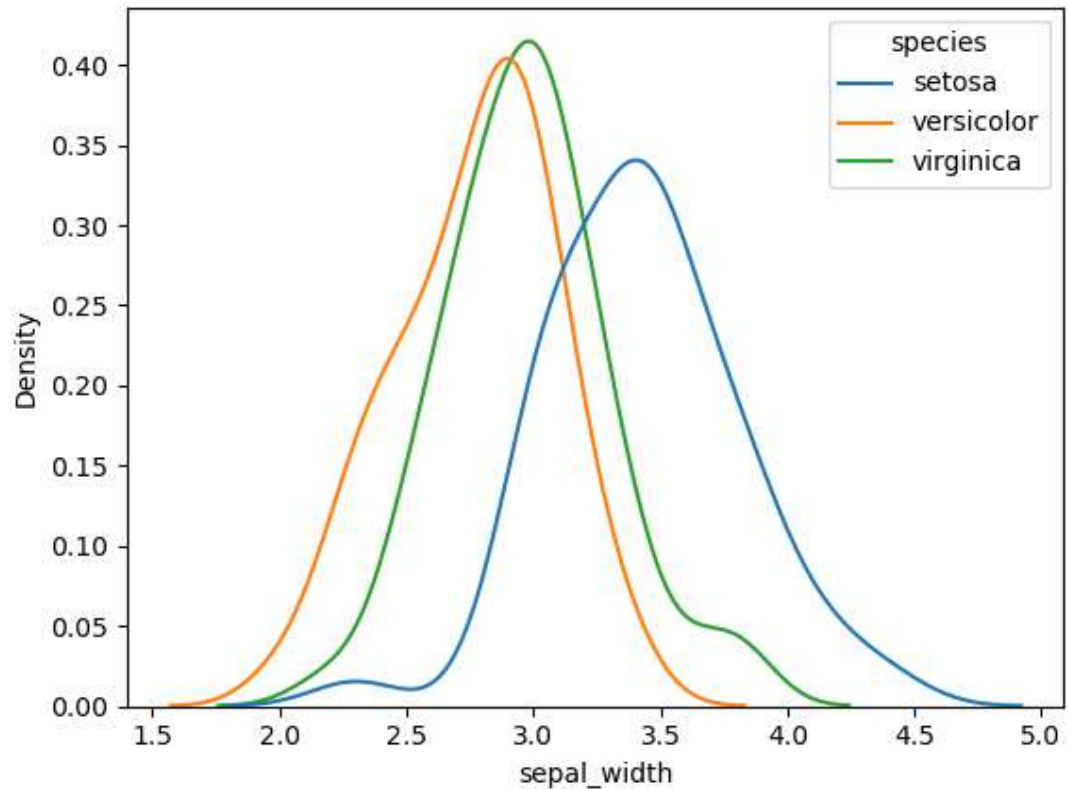
```
In [63]: 1 sns.kdeplot(data=df,x='sepal_length',hue='species')
```

Out[63]: <Axes: xlabel='sepal_length', ylabel='Density'>



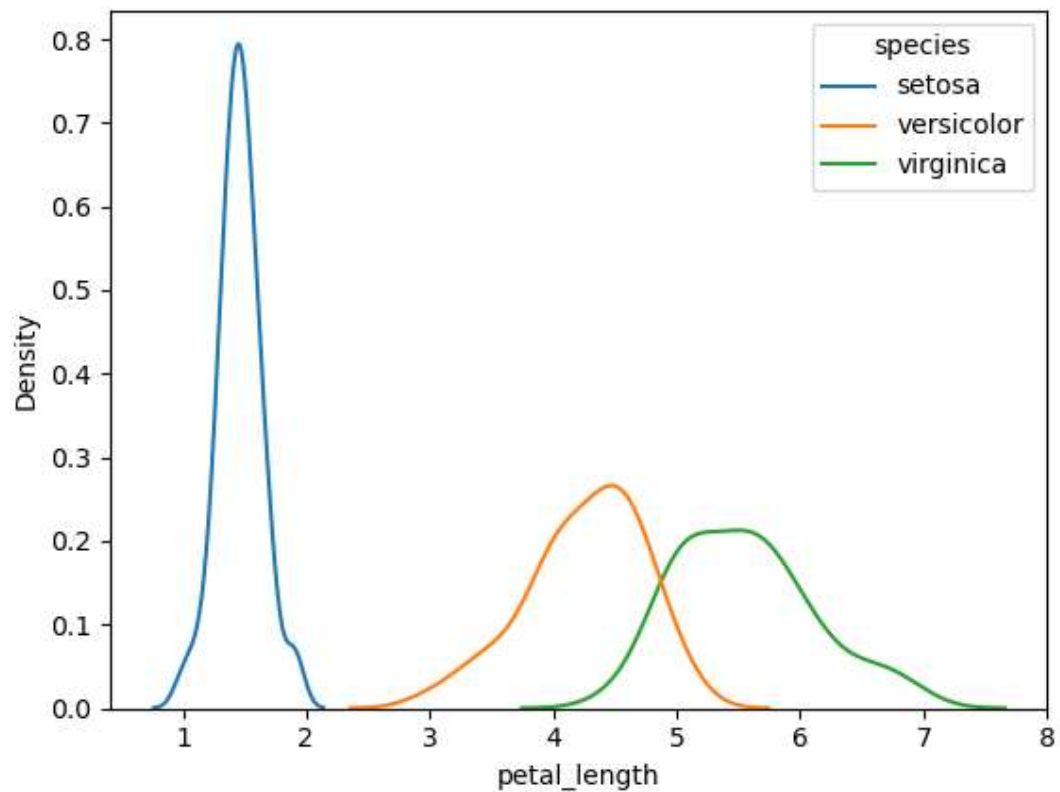
```
In [64]: 1 sns.kdeplot(data=df,x='sepal_width',hue='species')
```

```
Out[64]: <Axes: xlabel='sepal_width', ylabel='Density'>
```



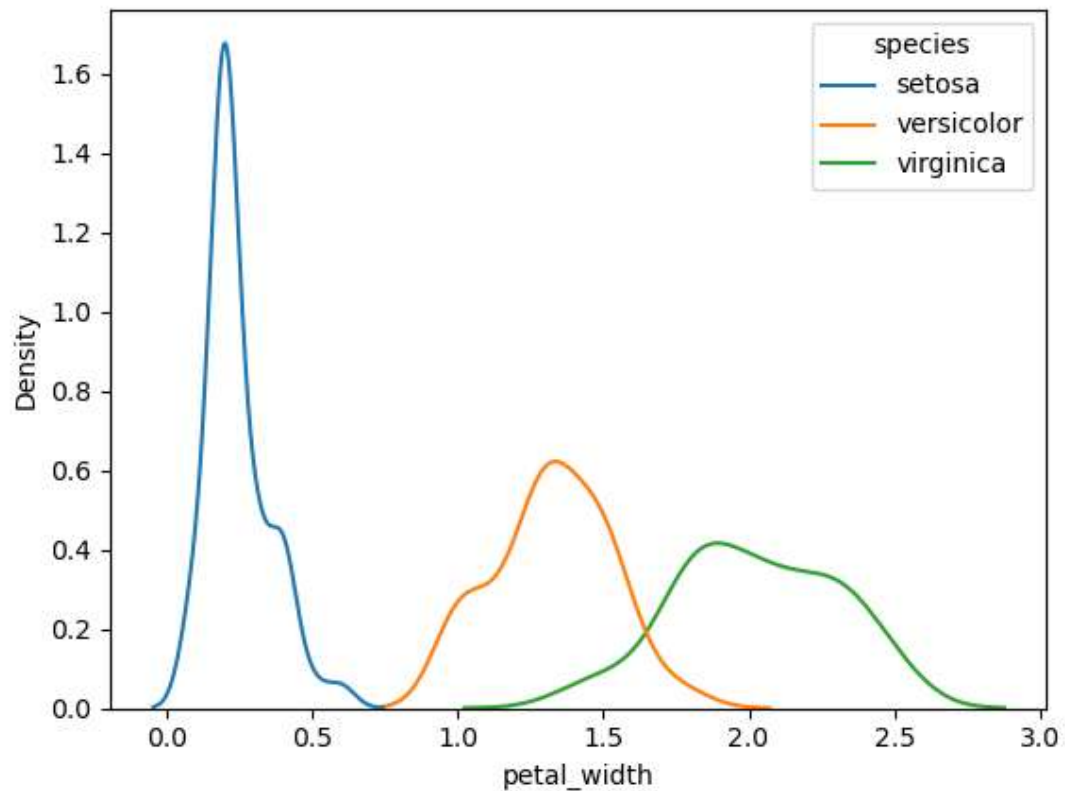
```
In [65]: 1 sns.kdeplot(data=df,x='petal_length',hue='species')
```

```
Out[65]: <Axes: xlabel='petal_length', ylabel='Density'>
```



```
In [66]: 1 sns.kdeplot(data=df,x='petal_width',hue='species')
```

```
Out[66]: <Axes: xlabel='petal_width', ylabel='Density'>
```



```
In [69]: 1 ## PMF on Descrete Variable
```

```
In [70]: 1 df.columns
```

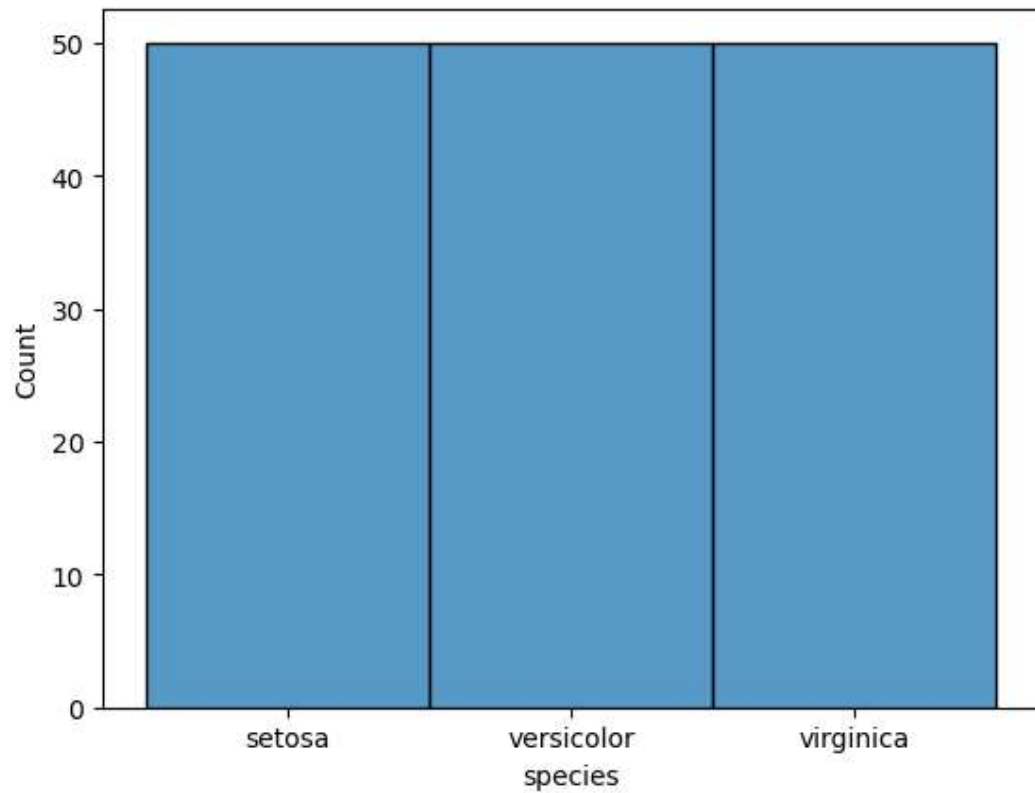
```
Out[70]: Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',  
              'species'],  
              dtype='object')
```

```
In [71]: 1 df['species']
```

```
Out[71]: 0      setosa  
         1      setosa  
         2      setosa  
         3      setosa  
         4      setosa  
         ...  
        145  virginica  
        146  virginica  
        147  virginica  
        148  virginica  
        149  virginica  
        Name: species, Length: 150, dtype: object
```

```
In [72]: 1 sns.histplot(df['species'])
```

```
Out[72]: <Axes: xlabel='species', ylabel='Count'>
```



```
In [ ]: 1
```