Koushik Sahu 118CS0597 Machine Learning Lab - 1

Code:

Problem 1:

```
from pathlib import Path
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
class Config:
  iris datapath = Path('../data/iris.data')
iris dataframe = pd.read csv(Config.iris datapath,
                 names=['sepal length', 'sepal width', 'petal length', 'petal width', 'class'])
iris dataframe.head()
classes = list(iris dataframe['class'].unique())
print(f'Classes in iris dataset: {classes}')
iris class = classes[2]
print(f'Class picked for the problem: {iris class}')
iris virginica df = iris dataframe[iris dataframe['class'] == iris class].reset index(drop=True)
iris_virginica_df.head()
# lets pick the attribute petal length for this problem
iris_attr = iris_virginica_df.columns[2]
petal length column = iris dataframe[iris attr]
print(f'Attribute picked for this problem: {iris attr}')
def evaluate mean(vals):
  count = vals.shape[0]
  summation = 0
  for i in vals:
    summation += i
  mean = summation / count
  return mean
def evaluate std(vals):
  count = vals.shape[0]
  mean = evaluate_mean(vals)
  variance = 0
```

```
for i in vals:
    variance += (i-mean)**2
  variance /= count
  std = np.sqrt(variance)
  return std
# calculating mean
petal length mean = evaluate mean(petal length column)
print(f'Mean of petal length: {petal length mean}')
# calculating standard deviation
petal length std = evaluate std(petal length column)
print(f'Standard deviation of petal length: {petal length std}')
def normal distribution(x, mean, std):
  return 1/(std*np.sqrt(np.pi)) * np.power(np.e, (-1/2)*(((x-mean)/std)**2))
# ploting normal distribution
low_x = -5
high x = 15
plot pt count = int(1e3)
delta = (high x-low x) / plot pt count
norm distr values = list()
plot_pts = list()
while low x \le high x:
  plot pts.append(low x)
  norm distr values.append(normal distribution(low x, petal length mean,
petal length std))
  low x += delta
plt.plot(plot pts, norm distr values)
plt.xlabel('x-axis')
plt.ylabel('y-axis')
plt.show()
# finding minimum and maximum
min_petal_length = min(petal_length_column)
max petal length = max(petal length column)
def uniform distribution(x, min val, max val):
  if min val < x and x < max val:
    return 1 / (max val-min val)
  return 0
# ploting normal distribution
low x = -5
```

```
high x = 15
plot_pt_count = int(1e3)
delta = (high x-low x) / plot pt count
uniform distr values = list()
plot_pts = list()
while low_x <= high_x:
  plot_pts.append(low_x)
  uniform distr values.append(uniform distribution(low x, min petal length,
max_petal_length))
  low_x += delta
plt.plot(plot pts, uniform distr values)
plt.xlabel('x-axis')
plt.ylabel('y-axis')
plt.show()
def exponential distribution(x, \lambda):
  if x \ge 0:
     return \lambda * np.power(np.e, -\lambda*x)
  return 0
# ploting exponential distribution
low x = -5
high x = 30
plot_pt_count = int(1e3)
delta = (high x-low x) / plot pt count
\lambda = 1/petal length mean
expo_distr_values = list()
plot pts = list()
while low x \le high x:
  plot pts.append(low x)
  expo distr values.append(exponential distribution(low x, \lambda))
  low x += delta
plt.plot(plot_pts, expo_distr_values)
plt.xlabel('x-axis')
plt.ylabel('y-axis')
plt.show()
def factorial(x):
  if x==1 or x==0:
     return 1
  return x*factorial(x-1)
def poisson distribution(x, \lambda):
  return (np.power(np.e, -\lambda)*np.power(\lambda, x)) / factorial(np.floor(x))
```

```
# ploting exponential distribution
low x = 0
high x = 30
plot pt count = int(1e3)
delta = (high_x-low_x) / plot_pt_count
\lambda = 1/petal length mean
poisson distr values = list()
plot pts = list()
while low x \le high x:
  plot pts.append(low x)
  poisson distribution(low x, \lambda))
  low x += delta
plt.plot(plot pts, poisson distr values)
plt.xlabel('x-axis')
plt.ylabel('y-axis')
plt.show()
# using function written in problem 1
petal length mean = evaluate mean(petal length column)
print(f'Mean petal length: {petal length mean}')
```

Problem 2:

```
# Median
def evaluate median(vals):
  n = vals.shape[0]
  if n%2==1:
    return vals.iloc[n//2]
  return (vals.iloc[n//2] + vals[(n+1)//2]) / 2
petal length median = evaluate median(petal length column)
print(f'Median petal length: {petal length median}')
petal length std = evaluate std(petal length column)
print(f'Standard deviation petal length: {petal length std}')
petal_length_variance = petal_length_std**2
print(f'petal length: {petal length variance}')
petal_length_skewness = (3*(petal_length_median-petal_length_mean)) / petal_length_std
print(f'Skewness petal length: {petal length skewness}')
count = petal length column.shape[0]
kurtosis = 0
```

```
for i in petal_length_column:
    kurtosis += ((i-petal_length_mean)/petal_length_std)**4

kurtosis /= count

print(f'Kurtosis petal length: {kurtosis}')
```

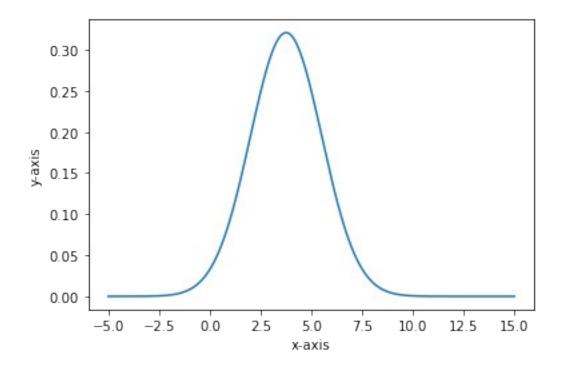
Problem 3:

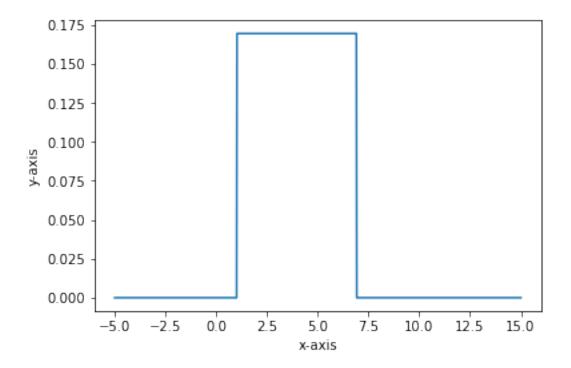
```
mean, variance = 0, 1

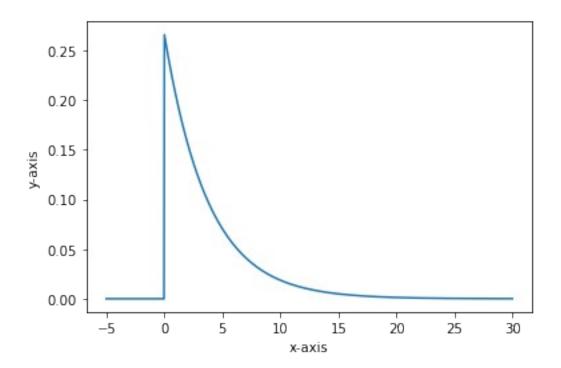
data = np.random.normal(mean, variance, 1000)
count, bins, ignored = plt.hist(data, 20, density=1)
plt.plot(bins,
    1/(variance * np.sqrt(2 * np.pi)) * np.exp( - (bins - mean)**2 / (2 * variance**2) ),
    linewidth=2,
    color='r')

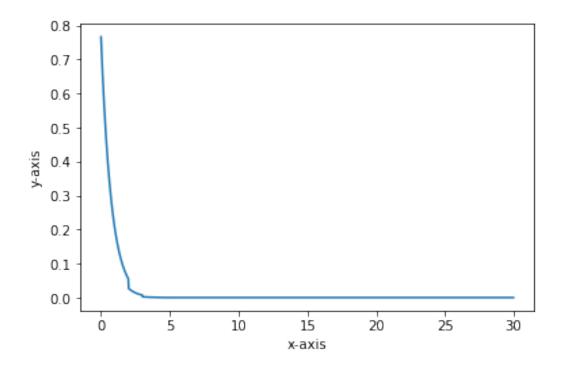
plt.title('Normal Distribution')
plt.show()
```

Output: Problem 1:









Mean petal length: 3.75866666666693

Median petal length: 4.4

Standard deviation petal length: 1.7585291834055201

petal length variance: 3.09242488888888854 Skewness petal length: 1.0940961447532118 Kurtosis petal length: 1.6046406978602903

