

## MADRAS INSTITUTE OFTECHNOLOGY ANNA UNIVERSITY



# DEPARTMENT OF INFORMATION TECHNOLOGY IT5612- DATA ANALYTICS AND CLOUD COMPUTING LABORATORY

## **RECORD**

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#### **BONAFIDE CERTIFICATE**

	Certified that the bona	afide record of the practical wo	ork done by
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Semester, I	3.Tech Information Te	echnology in the IT5612-Data	Analytics and
<b>Cloud Con</b>	nputing Laboratory du	uring the academic period from	n <mark>January 2023</mark>
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Date:		Co	ourse Instructor
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Internal Exa	aminer	Ext	ernal Examiner

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**EXP NO :**1.a **DATE:** 25-01-23

#### STUDY ON NUMPY, SCIPY, STATSMODELS AND PANDAS PACKAGES

#### Aim:

To study about the Python packages required to work with data analytics.

## **Theory:**

#### a. Pandas

Pandas is an open-source library that is made mainly for working with relational and labelled data. It provides various data structures and operations for manipulating data.

## **Functions in Pandas**

S. No	Description	Example
1.	read_csv(): This helps to read a csv [comma-separated-values] file into a pandas dataframe. It can also read files separated by delimiter	df = pd.read_csv ('data.csv')
2.	head(): The head(n) is used to return the first n rows of a dataset. By default, the function returns 5 rows of the dataframe	df.head(10)
3.	describe(): It is used to generate descriptive statistics of data in a pandas data frame or services	df.describe()
4.	memory_usage(): It returns a pandas series having the memory usage of each column in a dataframe	df.memory_usage(deep = = true)
5.	loc[]: It helps to access a group of rows and columns in a dataset	df.loc[0:4, ['Name', 'Age']]

## b. Numpy

Numpy is a Python library used for working with arrays. It also has functions for working in domain of linear algebra, Fourier transformation and matrices.

## **Functions in Numpy**

S. No	Description	Example
1.	numpy.concatenate(): This function is used to	res = np.concatenate((arr1,
	join two arrays of same size, either in a row-wise	arr2), axis =1)
	or column-wise way	
2.	numpy.char.add( ): It concatenates the data value	res = numpy.char.add(
	of two arrays, merge them and represent a new	['python'], ['java] )
	array as a result	
3.	numpy.median(): It calculates the median of an ordered array	med = np.median(a1)
4.	numpy.average(): It returns the average of all data values of the passed array	avg = np.average(a1)
5.	numpy reshape: This function allows us to change	res = a1.reshape(3, 2)
	the dimension of the array without hampering the	
	data value	

## c. Scipy

Scipy contains a variety of sub packages which help to solve the most common issue related to scientific computation. It is built on top of the numpy library.

## **Functions in Scipy**

S. No	Description	Example
1.	linalg.solve(): This gives the solutions of linear equations inn the matrix form	res = linalg.solve(a,b)
2.	linalg.inv(): This returns the inverse of the given matrix.	res = linalg.inv(a)
3.	linalg.det(): This returns the determinant of the given matrix.	res = linalg.det(ar)
4.	linalg.eig(): This returns the eigen values and eigen vectors of the given matrix.	eval, evec = linalg.eig(ar)
5.	special.logsumexp(x): This computes the log of sum of exponential input element	np.log( np.sum ( np.exp(a)))

#### d. Stats Model

It provides classes and functions for the estimation of many different statistical models for conducting statistical test and statistical data exploration

## **Functions in Stats Model:**

S. No	Description	Example
1.	get_rdataset : It is used to download any dataset we want	data = sm.datasets. get_rdataset("Tom","Chris").data
2.	add_constant(X): It is used to add a constant column to input dataset	res = sm.addconstant(res)
3.	OLS(y, x).fit(): It is a type of linear square method for estimating unknown parameters in linear regression	res = sm.OLS(y, x).fit()
4.	linear_rainbow(): The null hypothesis is the fit of the model using full sample. It is the same as using a central subset. Rainbow test has power against many different forms of non-linearity	print(sm.stats.linear_rainbow.— doc)

#### **Result:**

Thus, a study on the Python packages used to work with data analysis has been made.

**EXP NO :**1.b **DATE: 01-**02-23

## **READ DATA/DATASET**

#### Aim:

To write python programs to read and display content from text file, csv file and web.

## a. Reading from Text File

#### **CODE:**

#### **OUTPUT:**

'Welcome to Data Analytics Laboratory'

## b. Reading from CSV File

## **CODE:**

```
import csv
with open("sample.csv",'r') as file:
    reader=csv.reader(file)
    for i in reader:
        print(i)
```

## **OUTPUT:**

```
['Number', 'Square']
['1', '1']
['2', '4']
['3', '9']
['4', '16']
['5', '25']
```

#### c. Reading from Web:

#### **CODE:**

```
import requests
from bs4 import BeautifulSoup
r = requests.get("https://www.mitindia.edu/en/")
soup = BeautifulSoup (r.content, 'html.parser')
lines = soup.find_all('p')
for line in lines:
    print(line.text)
```

#### **OUTPUT:**

In 1949, Shri.C.Rajam, gave the newly independent India-Madras Institute of Technology, so that MIT could establish the strong t echnical base it needed to take its place in the world. It was the rare genius and daring of its founder that made MIT offer co urses like Aeronautical Engineering, Automobile Engineering, Electronics Engineering and Instrument Technology for the first ti me in our country. Now it also provides technical education in other engineering fields such as Rubber and Plastic Technology & Production Technology. It was merged with Anna University in the year 1978. Sixty years hence, while it continues to be a pione er in courses that it gave birth to, it is already renowned for producing the crème de la crème of the scientific community in more nascent courses such as Computer Science and Information Technology MIT has produced great scientist like Dr.A.P.J.Abdul K alam, versatile genius like Sujatha and many more. The broad-based education, coupled with practice-oriented training in their speciality, has enabled the students of MIT to handle with skill and success a wide variety of technical problems. The Madras I nstitute of Technology has developed into an important centre of engineering education and earned an excellent reputation both in India and abroad.MIT had received many awards which includes an award for the Best Overall Performance, awarded by Indian So ciety of Technical Education (ISTE) during the year 1999.

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#### **RESULT:**

Thus, the Python programs to read and display content from text file, CSV file, and web have been written and verified.

**EXP NO:** 2.a **DATE:** 01-02-23

#### **DESCRIPTIVE ANALYSIS**

#### Aim:

To explore the various commands for performing descriptive data analysis on the Iris Dataset.

## **Codes and Output:**

i. Import the necessary packages and dataset

```
import pandas as pd;

df=pd.read_csv("archive/city_hour.csv");

df=df[["PM2.5","PM10","NO2","SO2","O3","AQI_Bucket"]]
```

#### ii. Head

df.head()

	PM2.5	PM10	NO2	SO2	O3	AQI_Bucket
0	NaN	NaN	40.01	122.07	NaN	NaN
1	NaN	NaN	27.75	85.90	NaN	NaN
2	NaN	NaN	19.32	52.83	NaN	NaN
3	NaN	NaN	16.45	39.53	153.58	NaN
4	NaN	NaN	14.90	32.63	NaN	NaN

## iii. Getting Information about the dataset

## **Shape parameter**

df.shape

(707875, 6)

#### Info method

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 707875 entries, 0 to 707874
Data columns (total 6 columns):

#	Column	Non-Null Count	Dtype
0	PM2.5	562787 non-null	float64
1	PM10	411138 non-null	float64
2	NO2	590753 non-null	float64
3	502	577502 non-null	float64
4	03	578667 non-null	float64
5	AQI_Bucket	578795 non-null	object

dtypes: float64(5), object(1)

memory usage: 32.4+ MB

## **Describe method**

df.describe()

	PM2.5	PM10	NO2	<b>\$02</b>	03
count	562787.000000	411138.000000	590753.000000	577502.000000	578667.000000
mean	67.622994	119.075804	28.885157	14.038307	34.798979
std	74.730496	104.224752	29.162194	19.305540	29.806379
min	0.010000	0.010000	0.010000	0.010000	0.010000
25%	26.200000	52.380000	10.810000	4.880000	13.420000
50%	46.420000	91.500000	20.320000	8.370000	26.240000
75%	79.490000	147.520000	36.350000	14.780000	47.620000
max	999.990000	1000.000000	499.510000	199.960000	497.620000

## iv. Checking Missing Values

## isnull( ) method

df.isnull().sum()

PM2.5	145088
PM10	296737
NO2	117122
502	130373
03	129208
AQI_Bucket	129080

dtype: int64

## v. Checking Duplicates

## $drop\_duplicates(\ )$

dup=df.drop\_duplicates()
dup

	PM2.5	PM10	NO2	<b>SO2</b>	O3	AQI_Bucket
38289	25.91	84.90	58.41	140.10	47.69	Poor
38290	27.27	98.29	71.93	144.20	71.96	Poor
38291	27.27	102.68	66.79	124.37	88.85	Poor
38292	27.30	102.68	60.64	170.01	97.76	Poor
38294	23.84	121.94	70.40	31.13	94.66	Poor
707869	8.25	33.25	24.05	1.85	41.38	Satisfactory
707871	17.25	49.25	33.20	2.02	25.58	Satisfactory
707872	36.00	71.00	30.80	1.77	26.15	Good
707873	15.75	63.00	28.90	0.75	15.82	Good
707874	15.00	66.00	26.85	2.10	17.05	Good

347512 rows x 6 columns

## Series.value\_counts()

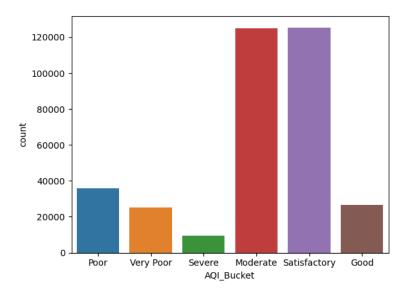
```
df.value\_counts("AQI\_Bucket")
```

AQI\_Bucket
Satisfactory 125448
Moderate 124879
Poor 35861
Good 26480
Very Poor 25241
Severe 9648
dtype: int64

#### vi. Data Visualization

## Data Visualization using countplot

```
import seaborn as sns
import matplotlib.pyplot as plt
sns.countplot(x='AQI_Bucket', data=df, )
plt.show()
```



## Data Visualization using histogram

```
fig, axes = plt.subplots(2, 2, figsize=(10,10))

axes[0,0].set_title("PM2.5")

axes[0,0].hist(df['PM2.5'], bins=10)

axes[0,1].set_title("PM10")

axes[0,1].hist(df['PM10'], bins=10);

axes[1,0].set_title("NO2")

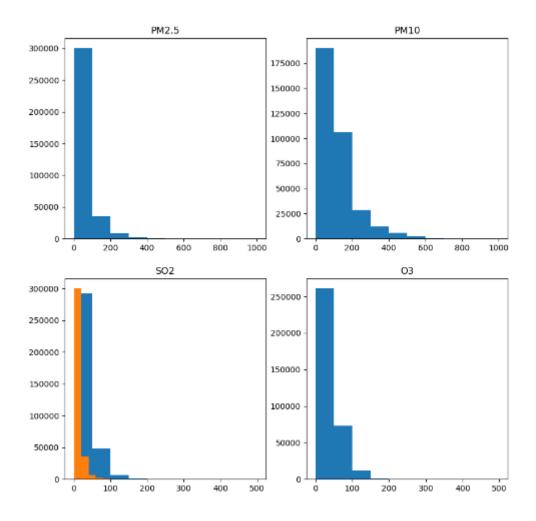
axes[1,0].hist(df['NO2'], bins=10);

axes[1,0].set_title("SO2")

axes[1,0].hist(df['SO2'], bins=10);

axes[1,1].set_title("O3")

axes[1,1].hist(df['O3'], bins=10);
```

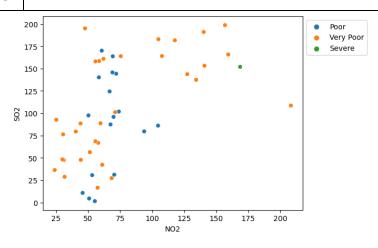


## Data Visualization using scatterplot

```
sns.scatterplot(x='NO2', y='SO2', hue='AQI_Bucket',
data=df.head(50), )
plt.legend(bbox_to_anchor=(1, 1), loc=2)
plt.show()
```

**EXP NO:** 2.b **DATE:** 08-02-23

## **UNIVARIATE ANALYSIS**



#### **Result:**

Thus, the various commands for performing descriptive data analysis on the Iris Dataset have been executed and verified.

#### Aim:

To perform univariate analysis on the Air Quality dataset.

## i. Frequency

```
import pandas as pd
df=pd.read_csv("archive/city_hour.csv")
df['AQI'].value_counts()
```

```
68.0
         3389
102.0
         3389
64.0
         3275
72.0
         3183
66.0
         3172
740.0
            1
661.0
            1
1157.0
           1
767.0
           1
8.0
            1
Name: AQI, Length: 884, dtype: int64
```

#### ii. Mean

```
df['O3'].mean()
```

36.05122673403284

#### iii. Median

```
df['O3'].mean()
```

27.96

#### iv. Mode

```
df['O3'].mode()
```

0 22.14

Name: 03, dtype: float64

#### v. Variance

848.4269834914085

#### vi. Standard Deviation

df['O3'].std()

29.127769971135937

#### vii. Skewness

from scipy.stats import skew
import numpy as np
import pylab as p
y1 = df['AQI']
print("Skewness = ", skew(y1))

Skewness = 2.958968947212698

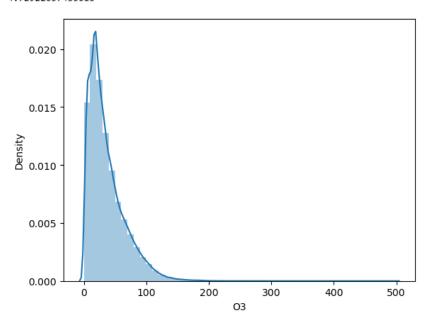
#### viii. Kurtosis

import seaborn as sns
sns.distplot(df['O3'], hist=True, kde=True)
df['O3'].kurt()

## **EXP NO:** 2.c **DATE:** 08-02-23

## **UNIVARIATE TIME SERIES ANALYSIS**

4.72922057433085



#### **Result:**

Thus, univariate analysis on the Air Quality dataset has been performed and verified.

## Aim:

To perform univariate time series analysis on the Air Quality dataset.

## i. Import Packages and Dataset

```
import pandas as pd
import numpy as np

df1=pd.read_csv("archive/time_series_dataset.csv",parse_dates=["Date"],
index_col=["Date"])

df1
```

#### ii. TIME SERIES FEATURES

## a. Mean in the period of 3 months

df1.resample('3M').mean()

#### Temperature

Date	
2023-01-31	23.000000
2023-04-30	24.666667
2023-07-31	24.000000
2023-10-31	27.000000

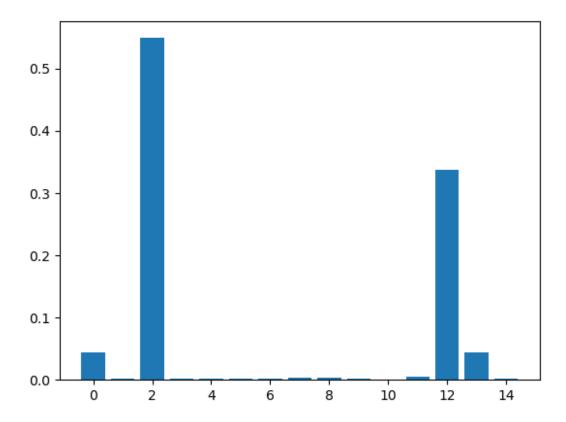
## **b.** Feature Importance

```
from sklearn.tree import DecisionTreeRegressor
import matplotlib.pyplot as plt

X, y = make_regression(n_samples=1000, n_features=15,
n_informative=5,
random_state=1)

model = DecisionTreeRegressor()

model.fit(X, y)
importance = model.feature_importances_
plt.bar([x for x in range(len(importance))], importance)
plt.show()
```



#### c. Forecast and Evaluate Data

```
pred = model.predict(X)
errors = abs(pred - y)
print('Mean Absolute Error:', round(np.mean(errors), 2), 'degrees.')
```

Mean Absolute Error: 0.0 degrees.

```
mape = 100 * (errors / y)
accuracy = 100 - np.mean(mape)
print('Accuracy:', round(accuracy, 2), '%.')
```

Accuracy: 100.0 %.

#### **Result:**

Thus, univariate time series analysis has been performed on the Air Quality dataset and has been verified.

**EXP NO:** 2.d **DATE:** 15-02-23

#### **BIVARIATE ANALYSIS**

#### Aim:

To perform bivariate analysis on the given dataset.

## i. Import necessary packages and dataset

import pandas as pd
import numpy as np;
from scipy.stats import pearsonr
df=pd.read\_csv("archive/city\_hour.csv")
df=df[["AQI\_Bucket"]]
encD=pd.get\_dummies(df,columns=['AQI\_Bucket'])

## ii. Pearson Correlation Coefficients and Interpretation

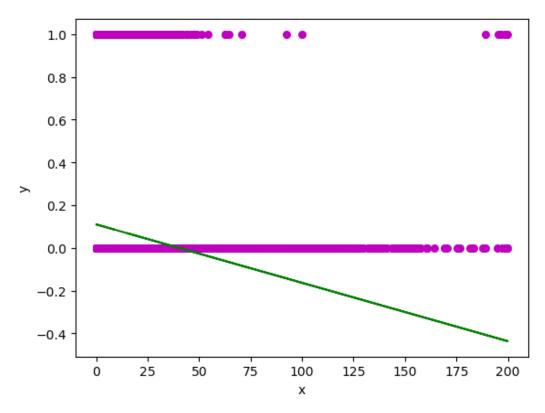
encD.corr()

	AQI_Bucket_Good	AQI_Bucket_Moderate	AQI_Bucket_Poor	AQI_Bucket_Satisfactory	AQI_Bucket_Severe	AQI_Bucket_Very Poor
AQI_Bucket_Good	1.000000	-0.238048	-0.103960	-0.197548	-0.045576	-0.084458
AQI_Bucket_Moderate	-0.238048	1.000000	-0.287591	-0.546489	-0.126078	-0.233641
AQI_Bucket_Poor	-0.103960	-0.287591	1.000000	-0.238662	-0.055061	-0.102035
AQI_Bucket_Satisfactory	-0.197548	-0.546489	-0.238662	1.000000	-0.104629	-0.193891
AQI_Bucket_Severe	-0.045576	-0.126078	-0.055061	-0.104629	1.000000	-0.044732
AQI_Bucket_Very Poor	-0.084458	-0.233641	-0.102035	-0.193891	-0.044732	1.000000

#### iii. Simple Linear Regression

```
import matplotlib.pyplot as plt
def estimate_coef(x, y):
n = np.size(x)
m_x = np.mean(x)
m_y = np.mean(y)
SS_xy = np.sum(y*x) - n*m_y*m_x
SS_x = np.sum(x*x) - n*m_x*m_x
b_1 = SS_xy / SS_xx
b_0 = m_y - b_1 * m_x
return (b_0, b_1)
def plot_regression_line(x, y, b):
plt.scatter(x, y, color = "m",
marker = "o", s = 30)
# predicted response vector
y_pred = b[0] + b[1]*x
plt.plot(x, y_pred, color = "g")
plt.xlabel('x')
plt.ylabel('y')
plt.show()
def main():
x = encD["SO2"]
y = encD["AQI_Bucket_Good"]
b = estimate\_coef(x, y)
print("Estimated coefficients:\nb_0 = \{\} \nb_1 = \{\}".format(b[0], b[1]))
plot_regression_line(x, y, b)
if __name__ == "__main__":
main()
```

Estimated coefficients: b\_0 = 0.1100817841186347 b\_1 = -0.002734473404634446



## iv. Chi-Squared Test

```
from scipy.stats import chi2_contingency
stat, p, dof, expected = chi2_contingency(encD["AQI"])
alpha = 0.05
print("p value is " + str(p))
if p <= alpha:
print('Dependent (reject H0)')
else:
print('Independent (H0 holds true)')
```

```
p value is 1.0
Independent (H0 holds true)
```

#### v. T-Test

```
from scipy.stats import ttest_ind

param1 = encD["O3"]

param2 = encD["AQI_Bucket_Good"]

stat, p = ttest_ind(param1, param2)

print("Statistics: %.3f, p = %.3f" % (stat,p))
```

Statistics: 471.017, p = 0.000

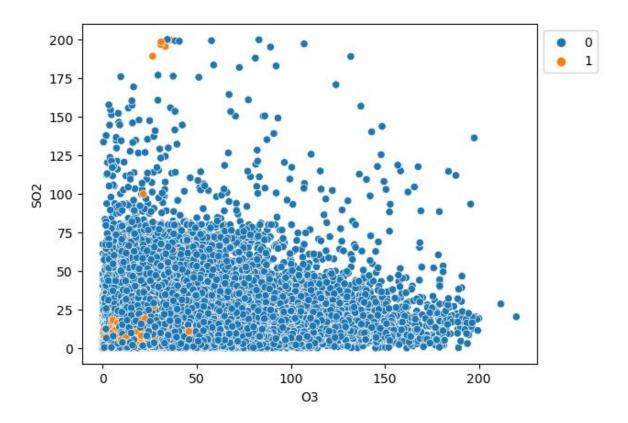
#### vi. Analysis of Variance

```
import scipy.stats as stats
fvalue, pvalue = stats.f_oneway(encD["O3"],
encD["AQI_Bucket_Good"],encD["AQI_Bucket_Moderate"],encD["AQI_Bucket
_Poor"],encD["AQI_Bucket_Satisfactory"],encD["AQI_Bucket_Severe"],encD["A
QI_Bucket_Very Poor"] )
print("fvalue: ", fvalue, ", pvalue: ", pvalue)
```

fvalue: 220625.20408816423 , pvalue: 0.0

## vii. Scatterplots

```
import seaborn as sns sns.scatterplot(x = "O3", y = "SO2", hue= "AQI\_Bucket\_Good", data = encD) \\ plt.legend(bbox\_to\_anchor = (1,1), loc = 2) \\ plt.show()
```



## **Result:**

Thus, bivariate analysis has been successfully carried out on the Air Quality dataset.

**EXP NO:** 2.e **DATE:** 15-02-23

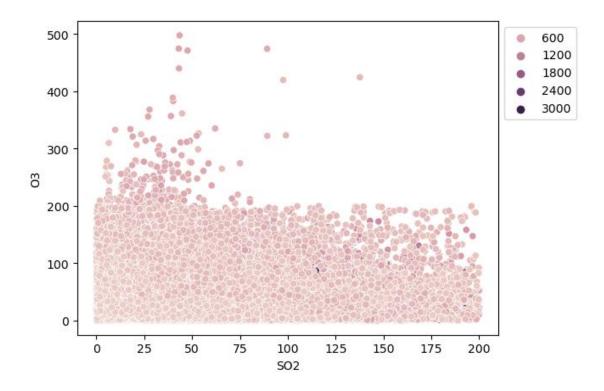
## **MULTIVARIATE ANALYSIS**

#### Aim:

To perform bivariate analysis on the given dataset.

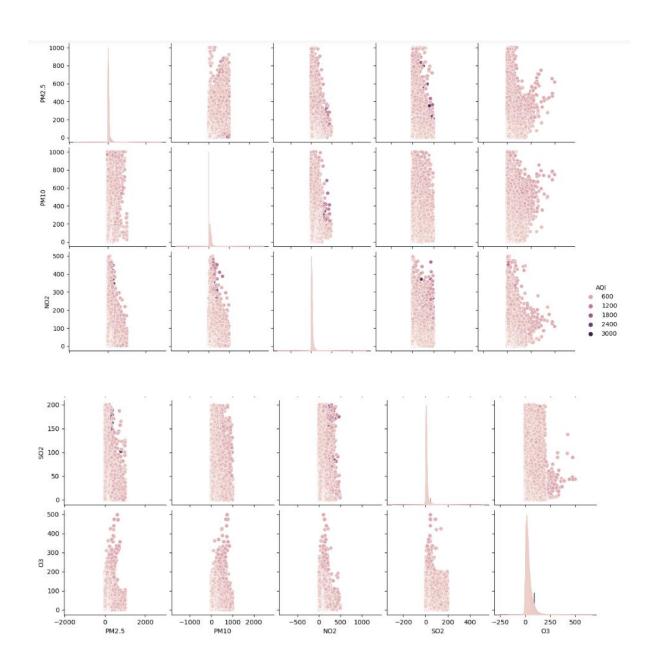
## i. Scatterplot

```
sns.scatterplot(x='SO2',y='O3',hue='AQI',data=df,)
plt.legend(bbox_to_anchor=(1,1),loc=2)
plt.show()
```



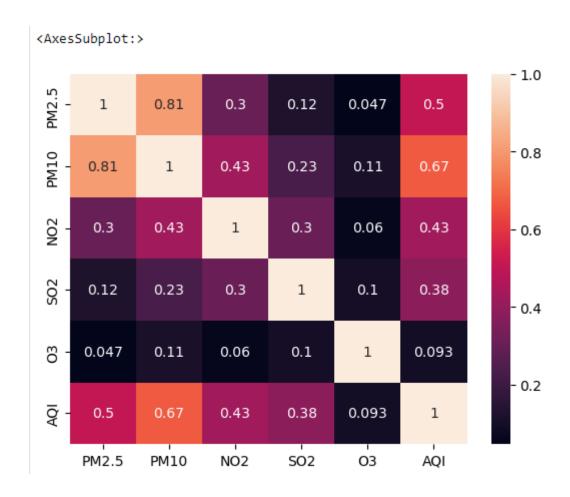
## ii. Pair Plots

sns.pairplot(df,hue='AQI')



## iii. Heat Map

sns.heatmap(df.corr(method="pearson"),annot=True)



## **Result:**

**EXP NO:** 3 **DATE:** 22-02-23

## **CLASSIFICATION OF NAÏVE BAYES**

Thus, multivariate analysis has been successfully carried out on the Air Quality dataset.

#### Aim:

To build models to perform Gaussian and Multinomial Naïve Bayes classification for the given dataset.

## **GAUSSIAN NAIVE BAYES**

#### i. Import necessary packages and dataset

```
import pandas as pd
import numpy as np
df=pd.read_csv("dataset.csv")
df.head()
```

	PM2.5	PM10	NO2	<b>SO2</b>	03	AQI
0	104.00	148.50	23.00	15.30	117.62	3.0
1	94.50	142.00	16.25	17.00	136.23	3.0
2	82.75	126.50	14.83	15.40	149.92	3.0
3	68.50	117.00	13.60	21.80	161.70	3.0
4	69.25	112.25	11.80	21.38	161.68	3.0

## ii. Exploratory Data Analysis

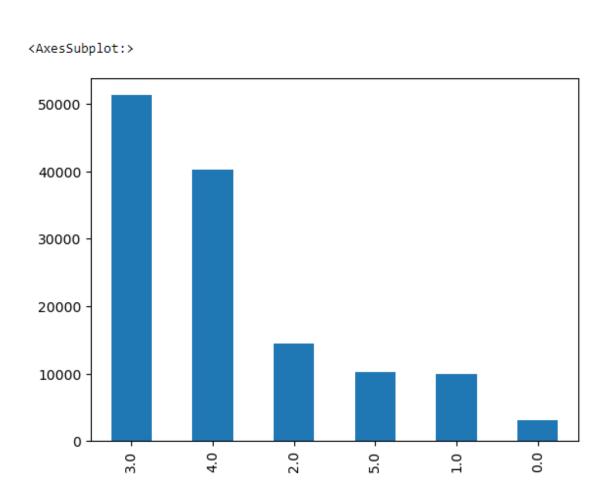
## a. Shape

df.shape

(129277, 6)

## b. Bar plot

import seaborn as sns
import matplotlib.pyplot as plt
df['AQI'].value\_counts().plot.bar()



## iii. Identify Numerical and Categorical Data

```
import numpy as np
numeric_data = df.select_dtypes(include=[np.number])
categorical_data = df.select_dtypes(exclude=[np.number])
print("Number of numerical variables: ", numeric_data.shape[1])
print("Numerical attributes: ", numeric_data.columns)

print("Number of categorical variables: ", categorical_data.shape[1])
print("Categorical attributes: ",categorical_data.columns)
```

```
Number of numerical variables: 6
Numerical attributes: Index(['PM2.5', 'PM10', 'N02', 'S02', '03', 'AQI'], dtype='object')
Number of categorical variables: 0
Categorical attributes: Index([], dtype='object')
```

## iv. Identify the Missing Values

df.isnull().sum()

PM2.5 0 PM10 0 NO2 0 SO2 0 O3 0 AQI 0 dtype: int64

#### v. Data preprocessing

```
from sklearn import preprocessing

#Feature scaling - min max scaling

from sklearn.preprocessing import MinMaxScaler

scaler = MinMaxScaler()

scaled = scaler.fit_transform(df)

encD=df

encD.head()
```

	PM2.5	PM10	NO2	<b>SO2</b>	О3	AQI
0	104.00	148.50	23.00	15.30	117.62	3.0
1	94.50	142.00	16.25	17.00	136.23	3.0
2	82.75	126.50	14.83	15.40	149.92	3.0
3	68.50	117.00	13.60	21.80	161.70	3.0
4	69.25	112.25	11.80	21.38	161.68	3.0

## vi. Split dataset into train and test set

```
cols=['PM2.5',"PM10","NO2","SO2","O3"]

X=encD[cols]

y=encD['AQI']

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4)
```

### vii. Train the model

```
from sklearn.naive_bayes import GaussianNB
gNBmodel = GaussianNB()
gNBmodel.fit(X_train,y_train)
```

GaussianNB()

# viii. Test model and display result

# ix. Accuracy Score

print('Accuracy of Gaussian Naive Bayes on test set: {:.2f}'.format
(gNBmodel.score(X\_test, y\_test)))

Accuracy of Gaussian Naive Bayes on test set: 0.61

# x. Check for overfitting and underfitting

```
from sklearn.metrics import mean_absolute_error

y_train_pred = gNBmodel.predict(X_train)

mae_train = mean_absolute_error(y_train, y_train_pred)

mae_test = mean_absolute_error(y_test, y_pred)

print(mae_train)

print(mae_test)

if mae_train < mae_test:

print("Overfitting is present")

else:

print("Underfitting is present")
```

```
0.4338498826805559
0.42874823538512113
Underfitting is present
```

### xi. Confusion matrix

```
from sklearn.metrics import confusion_matrix confusion_matrix(y_test, y_pred)
```

```
array([[ 666, 413, 108,
                         34,
                                12,
                                       1],
     [ 352, 1760, 1217,
                        635,
                               57,
                                       6],
        34, 791, 1852, 2730, 309,
                                      12],
        21, 197, 1061, 13716, 5271,
                                    400],
            5, 19, 2311, 10000, 3638],
         2,
                    0,
                          11, 687, 3383]], dtype=int64)
          0,
               0,
```

# xii. Analyse model performance visually

from sklearn.metrics import ConfusionMatrixDisplay

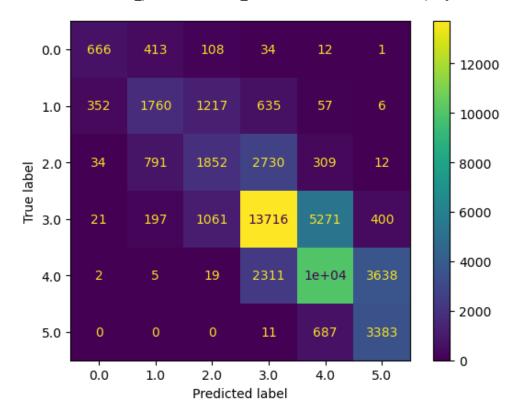
cm = confusion\_matrix(y\_test, y\_pred, labels=gNBmodel.classes\_)

disp = ConfusionMatrixDisplay(confusion\_matrix=cm,

display\_labels=gNBmodel.classes\_)

disp.plot()

<sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x2124a9cbeb0>



# xiii. Classification report

from sklearn.metrics import classification\_report print(classification\_report(y\_test, y\_pred))

	precision	recall	f1-score	support
0.0	0.62	0.54	0.58	1234
1.0	0.56	0.44	0.49	4027
2.0	0.44	0.32	0.37	5728
3.0	0.71	0.66	0.68	20666
4.0	0.61	0.63	0.62	15975
5.0	0.45	0.83	0.59	4081
accuracy			0.61	51711
macro avg	0.56	0.57	0.55	51711
weighted avg	0.61	0.61	0.60	51711

# **MULTINOMIAL NAÏVE BAYES**

# i. Train the model

 $\label{eq:mnaive_bayes import MultinomialNB} $$ mNBmodel = MultinomialNB() $$ mNBmodel.fit(X_train,y_train)$ 

MultinomialNB()

# ii. Test model and display result

```
y_pred = mNBmodel.predict(X_test)
y_pred
```

```
array([3., 4., 4., ..., 2., 2., 3.])
```

# iii. Accuracy Score

```
print('Accuracy of Multinomial Naive Bayes on test
set:{:.2f}'.format(mNBmodel.score(X_test, y_test)))
```

Accuracy of Multinomial Naive Bayes on test set:0.31

# iv. Checking for Overfitting and Underfitting

```
from sklearn.metrics import mean_absolute_error

y_train_pred = mNBmodel.predict(X_train)

mae_train = mean_absolute_error(y_train, y_train_pred)

mae_test = mean_absolute_error(y_test, y_pred)

print(mae_train)

print(mae_test)

if mae_train < mae_test:

print("Overfitting is present")

else:

print("Underfitting is present")
```

<sup>0.9832787561560478</sup> 

<sup>0.9877008760225097</sup> 

# v. Confusion Matrix

from sklearn.metrics import confusion\_matrix confusion\_matrix(y\_test, y\_pred)

# vi. Classification Report

from sklearn.metrics import classification\_report print(classification\_report(y\_test, y\_pred))

	precision	recall	f1-score	support
0.0	0.19	0.56	0.29	1234
1.0	0.17	0.28	0.21	4027
2.0	0.16	0.19	0.17	5728
3.0	0.47	0.28	0.35	20666
4.0	0.43	0.30	0.36	15975
5.0	0.20	0.52	0.29	4081
accuracy			0.31	51711
macro avg	0.27	0.36	0.28	51711
weighted avg	0.37	0.31	0.32	51711

# **Result:**

**EXP NO: 4 DATE:** 22-02-23

# **SVM CLASSIFICATION**

Thus, the models to perform Gaussian and Multinomial Naïve Bayes classification for the given dataset have been built successfully.

### Aim:

To build models to perform classification using SVM classifier for the given dataset.

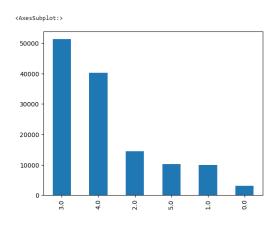
# **Codes and Output:**

import pandas as pd import numpy as np df=pd.read\_csv("dataset.csv")

#### **Exploratory Data Analysis** i.

df.shape (707875, 6)

import seaborn as sns import matplotlib.pyplot as plt df['AQI'].value\_counts().plot.bar()



# ii. Explore Missing Values in Variables

df.isnull().sum()

PM2.5 0 PM10 0 NO2 0 SO2 0 O3 0 AQI 0 dtype: int64

# iii. Split data into separate training and test set

 $from \ sklearn.model\_selection \ import \ train\_test\_split$ 

cols=['PM2.5',"PM10","NO2","SO2","O3"]

X=df[cols]

y=df['AQI']

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.2, random\_state=40)

# iv. Run SVM with default Hyperparameter

```
from sklearn.svm import SVC

from sklearn.metrics import accuracy_score

svc=SVC()

svc.fit(X_train,y_train)

y_pred=svc.predict(X_test)

print('Model accuracy score with default hyperparameters: {0:0.4f}'.

format(accuracy_score(y_test, y_pred)))
```

Model accuracy score with default hyperparameters: 0.6655

# v. Compare the Train-Set and Test-Set Accuracy

```
from sklearn.svm import SVC

from sklearn.metrics import accuracy_score

svc=SVC()

svc.fit(X_train,y_train)

y_pred=svc.predict(X_test)

print('Model accuracy score with default hyperparameters: {0:0.4f}'.

format(accuracy_score(y_test, y_pred)))
```

Model accuracy score with default hyperparameters: 0.6655

# vi. Check for Overfitting and Underfitting

```
from sklearn.metrics import mean_absolute_error

y_train_pred = svc.predict(X_train)

mae_train = mean_absolute_error(y_train, y_train_pred)

mae_test = mean_absolute_error(y_test, y_pred)

print(mae_train)

print(mae_test)

if mae_train < mae_test:

print("Overfitting is present")

else:

print("Underfitting is present")
```

# vii. Confusion Matrix

from sklearn.metrics import confusion\_matrix confusion\_matrix(y\_test, y\_pred)

# viii. Classification report

# **EXP NO: 5**

**DATE:** 01-03-23

#### I OCICTIC DECDESSION

from sklearn.metrics import classification\_report
print(classification\_report(y\_test, y\_pred))

	precision	recall	f1-score	support
0.0	0.81	0.49	0.61	615
1.0	0.65	0.56	0.60	1988
2.0	0.54	0.35	0.42	2942
3.0	0.69	0.78	0.73	10093
4.0	0.67	0.73	0.70	8158
5.0	0.66	0.46	0.54	2060
accuracy			0.67	25856
macro avg	0.67	0.56	0.60	25856
weighted avg	0.66	0.67	0.66	25856

### **Result:**

Thus, the models to perform classification using SVM classifier for the Air Quality dataset have been built successfully.

# Aim:

To perform logistic regression on the Air Quality dataset.

# **Codes and Output:**

i. Import necessary packages and dataset

```
import pandas as pd
import numpy as np
df=pd.read_csv("dataset.csv")
```

#### Split data into separate training and test set ii.

```
from sklearn.model_selection import train_test_split
cols=['PM2.5',"PM10","NO2","SO2","O3"]
X=df[cols]
y=df['AQI']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=40)
```

#### iii. Train the model

```
from sklearn.linear_model import LogisticRegression
lr_model = LogisticRegression()
lr_model.fit(X_train, y_train)
y_pred = lr_model.predict(X_test)
```

#### iv. **Accuracy Score**

```
from sklearn.metrics import accuracy_score
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
```

Accuracy: 0.4583462252475248

### v. Confusion Matrix

```
from sklearn.metrics import confusion_matrix confusion_matrix(y_test, y_pred)
```

```
array([[ 51, 374,
                    8, 187,
                                     0],
                                1,
                                     0],
      [ 122, 837,
                   94, 981,
                              1,
        31, 570,
                  89, 2181,
                              42,
                                     0],
        32, 528, 177, 8694,
                            832,
                                     0],
         5, 113, 73, 5614, 2178,
                                     5],
         0,
              2,
                   4, 657, 1371,
                                    2]], dtype=int64)
```

# vi. Classification report

from sklearn.metrics import classification\_report print(classification\_report(y\_test, y\_pred))

```
precision
                      recall f1-score support
       0.0
                0.21
                       0.08
                                 0.12
                                           621
       1.0
                0.35
                       0.41
                                 0.38
                                          2035
       2.0
                0.20
                        0.03
                                 0.05
                                          2913
                        0.85
                0.47
       3.0
                                 0.61
                                         10263
       4.0
                0.49
                         0.27
                                  0.35
                                          7988
       5.0
                0.29
                         0.00
                                  0.00
                                          2036
   accuracy
                                  0.46
                                          25856
                0.33
                         0.27
                                  0.25
                                          25856
  macro avg
weighted avg
                0.42
                         0.46
                                  0.39
                                          25856
```

# vii. Analyse model performance visually

```
from sklearn.metrics import ConfusionMatrixDisplay

cm = confusion_matrix(y_test, y_pred,
labels=lr_model.classes_)

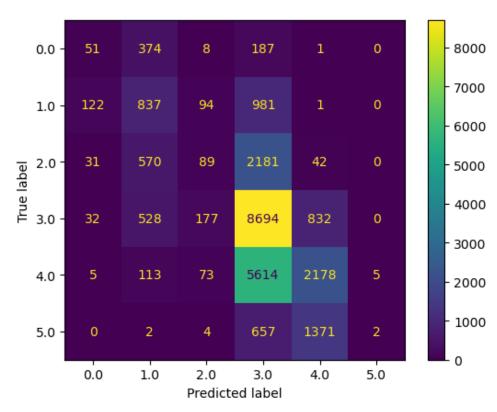
disp = ConfusionMatrixDisplay(confusion_matrix=cm,
display_labels=lr_model.classes_)

disp.plot()
```

**EXP NO:** 6 **DATE:** 01-03-23

# **MULTIPLE REGRESSION**

<sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x14519a28940>



# **Result:**

Thus, logical regression has been carried out on the Air Quality dataset successfully.

# Aim:

To build a multiple regression model on the Air Quality dataset.

# i. Import necessary packages and dataset

```
import pandas as pd
import numpy as np
df=pd.read_csv("dataset.csv")
```

# ii. Data preprocessing

```
from sklearn import preprocessing
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
scaled = scaler.fit_transform(df)
encD=df
encD.head()
```

	PM2.5	PM10	NO2	<b>SO2</b>	O3	AQI
0	104.00	148.50	23.00	15.30	117.62	3.0
1	94.50	142.00	16.25	17.00	136.23	3.0
2	82.75	126.50	14.83	15.40	149.92	3.0
3	68.50	117.00	13.60	21.80	161.70	3.0
4	69.25	112.25	11.80	21.38	161.68	3.0

# iii. Split dataset into train and test set

```
cols=['PM2.5',"PM10","NO2","SO2","O3"]

X=encD[cols]

y=encD['AQI']

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4)
```

### iv. Train the model

```
from sklearn.linear_model import LinearRegression
linear_regression = LinearRegression()
linear_regression.fit(X_train, y_train)
```

LinearRegression()

# v. Test model and display result

### **Result:**

Thus, multiple regression has been carried out on the Air Quality dataset successfully.

**EXP NO :** 7.a **DATE:** 08-03-23

### **OPENSTACK**

### Aim:

To understand OpenStack deployment, its implementation and its applications.

# Theory:

OpenStack is a cloud operating system that controls large pools of compute, storage and networking resources throughout a data-centre, all managed and provisioned through APIs with common authentication mechanisms.

A dashboard is also available, giving administrators control while empowering their users to provide resources through a web interface. It began in 2010 as a joint project of Ra despace hosting and NASA. It was managed by the OpenStack Foundation, a non-profit entity.

# Working of OpenStack

It is essentially a series of commands known as scripts. These scripts are bundled into packages called projects that relay tasks that create cloud environments. To create these environments, OpenStack relies on

- Virtualization that creates a layer of virtual resources abstracted from hardware
- A base as that carries out commands given by OpenStack scripts

OpenStack uses virtualized resources to build clouds. It doesn't execute commands, rather delays them to base OS.

# **Components of OpenStack**

OpenStack's architecture is made up of numerous open source projects. These are used to set up OpenStack's undercloud and overcloud. Overcloud is used by cloud users and undercloud is used by system admins. Underclouds contain the core components system admins need to set up and manage end user's environments called overclouds.

There are 6 stable, core services that handle computing, networking, storage, identity and images. These make up the infrastructure of OpenStack that allows the rest of the projects to handle dashboarding, orchestration, etc. The 6 components are:

**Nova:** It is a full management tool that helps compute resource-handling, scheduling, creation, deletion, etc.

**Neutron:** It connects the networks across other services

**Swift:** It is a highly fault-tolerant object storage service that stores and retrieves unstructured data objects

**Cinder:** Provides persistent block storage

**Keystone:** Authenticates and authorises all services

Glance: Stores and retrieves VM disc images from various locations

# **Deployment of OpenStack**

It is mostly deployed as infrastructure-as-a-Service (IaaS) in both public and private clouds where virtual servers and other resources are made available to users. The software platform consists of interrelated components control diverse, multivendor hardware. Lifecycle management tools and packaging are used to help instances maintain the lifecycle of deployments. Frameworks for the above are Tripleo, OpenStack-helm, kolla-ansible, kayole, etc.

# **Challenges in Implementation**

- Installation challenges
- Documentation
- Upgrading OpenStack
- Long Term Support

# **Applications of OpenStack**

- Private clouds and Public clouds
- Network function virtualization
- Containers

### **Result:**

**EXP NO:** 7.b **DATE:** 08-03-23

# **OPENSTACK INSTALLATION**

Thus, OpenStack deployment, its implementation and its applications have been studied.

### Aim:

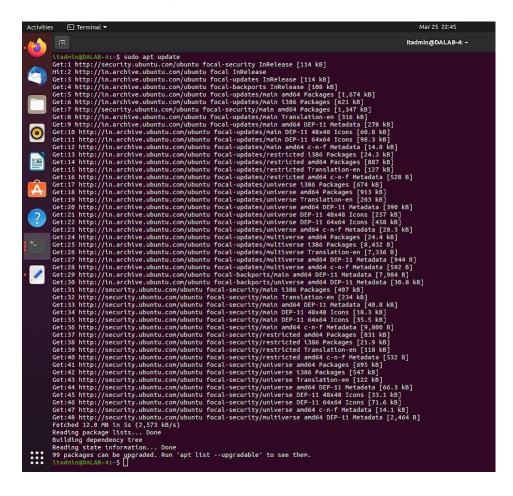
To install OpenStack and implement Infrastructure-as-a-service using it.

# **Installation of OpenStack in Ubuntu**

- 1. Login to sudo ("superuser do") user
- 2. Open Terminal
- 3. In terminal, type:

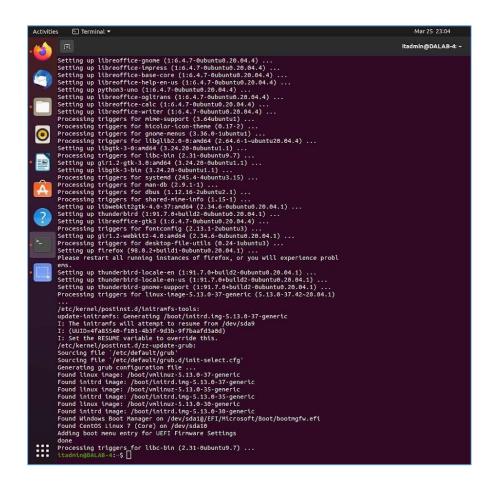
sudo apt update

This downloads package information from all configured sources.



4. Upgrade everything in the system, all the packages, and the kernel to the latest versions assupported by the repositories:

sudo apt -y upgrade



5. Upgrade existing packages, installs new dependencies that are not in the system, and deletesthose that are not needed:

sudo apt -y dist-upgrade

```
ttadmin@DALAB-4:-$ sudo apt -y dist-upgrade
[sudo] password for itadmin:
Reading package lists... Done
Building dependency tree
Reading state information... Done
Calculating upgrade... Done
The following packages were automatically installed and are no longer required:
    libfwupdplugini linux-headers-5.13.0-30-generic
    linux-hwe-5.13-headers-5.13.0-30 linux-image-5.13.0-30-generic
    linux-modules-5.13.0-30-generic linux-modules-extra-5.13.0-30-generic
Use 'sudo apt autoremove' to remove them.
    upgraded, 0 newly installed, 0 to remove and 0 not upgraded.
ttadmin@DALAB-4:-$
```

6. Add user called stack to run DevStack. It should be run as non-root user with sudo enabled:

sudo useradd -s /bin/bash -d /opt/stack -m stack

```
base) student@DALAB-4:~/Desktop$ sudo useradd -s /bin/bash -d /opt/stack -m stack sudo] password for student: student is not in the sudoers file. This incident will be reported. [base) student@DALAB-4:~/Desktop$ su - itadmin assword: sudoent@DALAB-4:~$ sudo useradd -s /bin/bash -d /opt/stack -m stack [sudo] password for itadmin: stadmin@DALAB-4:~$
```

7. User should have sudo privileges to make changes to the system:

echo "stack ALL=(ALL) NOPASSWD: ALL" | sudo tee /etc/sudoers.d/stack

```
itadmin@DALAB-4:~$ echo "stack ALL=(ALL) NOPASSWD: ALL" | sudo tee /etc/sudoers.d/stack
stack ALL=(ALL) NOPASSWD: ALL
itadmin@DALAB-4:~$ ■
```

8. Log in to the stack once user is created:

```
sudo su - stack
```

```
itadmin@DALAB-4:~$ sudo su - stack
stack@DALAB-4:~$ sudo su -
```

9. Install git:

sudo apt -y install git

```
Etack@DALAB-4:-$ sudo apt -y install git
Reading package lists... Done
Building dependency tree
Reading state information... Done
The following packages were automatically installed and are no longer required:
libfwupdplugin1 linux-headers-5.13.0-30-generic
linux-hwe-5.13-headers-5.13.0-30-generic linux-modules-5.13.0-30-generic linux-modules-5.13.0-30-generic linux-modules-5.13.0-30-generic linux-modules-5.13.0-30-generic linux-modules-5.13.0-30-generic linux-modules-5.13.0-30-generic linux-modules-6.13.0-30-generic linux-modules-6.13.
```

10. Download devstack from its repository into system:

git clone https://git.openstack.org/openstack-dev/devstack

```
stack@DALAB-4:—$ git clone https://github.com/openstack-dev/devstack.git
Cloning into 'devstack'...
remote: Enumerating objects: 48499, done.
remote: Counting objects: 100% (1725/1725), done.
remote: Compressing objects: 100% (686/686), done.
remote: Total 48499 (delta 1177), reused 1444 (delta 1033), pack-reused 46774
Receiving objects: 100% (48499/48499), 15.48 MiB | 4.31 MiB/s, done.
Resolving deltas: 100% (33808/33808), done.
```

11. Download devstack setup configurations files for it. Need to navigate devstack folder by running:

cd devstack
nano local.conf
vi local.conf

```
stack@DALAB-4:~$ cd devstack
stack@DALAB-4:~/devstack$ nano local.conf
stack@DALAB-4:~/devstack$ vi local.conf
```

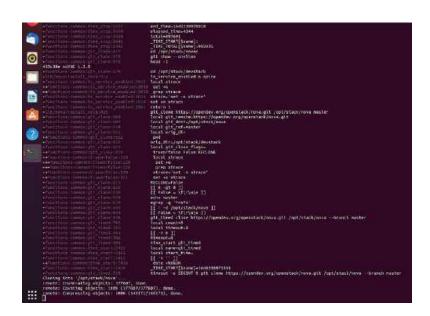
# 12. Add following inside local.conf:

[[local|localrc]]
# Password for KeyStone, Database, RabbitMQ and
Service
ADMIN\_PASSWORD=StrongAdminSecret
DATABASE\_PASSWORD=\$ADMIN\_PASSWORD
RABBIT\_PASSWORD=\$ADMIN\_PASSWORD
SERVICE\_PASSWORD=\$ADMIN\_PASSWORD

After setting above, in local.conf, press Esc key and type :wq to write/save and quit

# 13. Run following to setup Openstack on system:

./stack.sh



After installation, terminal:

# 14.Browse URL on browser: http://localhost/dashboard

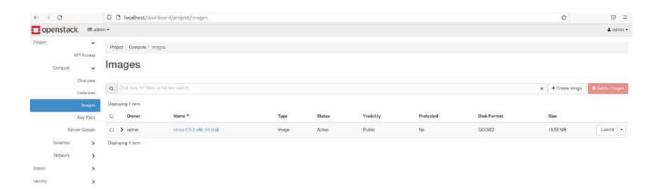


**EXP NO:** 8

# **CREATION OF VM INSTANCES IN AWS**

**DATE:** 15-03-23

15.Enter credentials. Log in as admin using username 'admin' and password as given in local.conf file:



# **Result:**

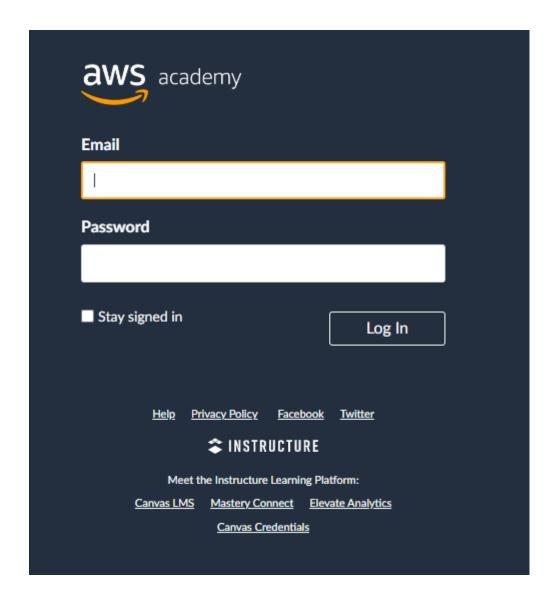
Thus, OpenStack has been successfully installed.

### Aim:

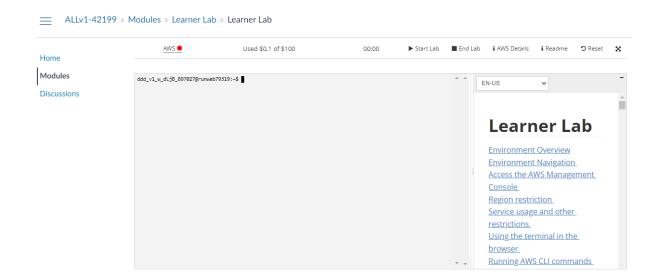
To deploy a VM in AWS and execute a simple application on it.

# **DEPLOYING VM IN AWS:**

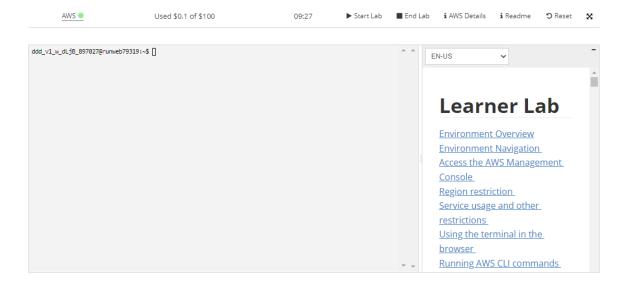
1. Login to aws academy using credentials.



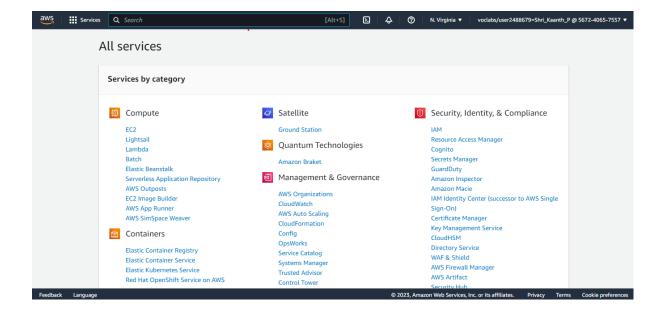
2. In the dashboard, select Modules → Learner Lab. The Learner lab will be opened.



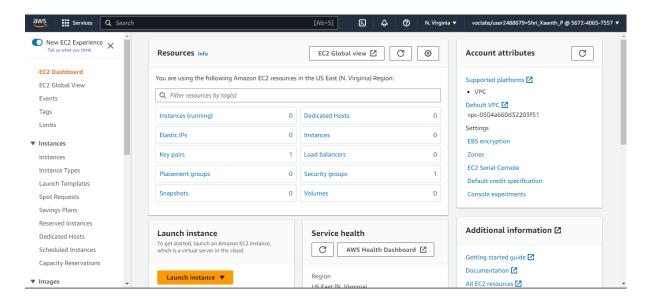
3. The red signal indicates that the lab has not started yet. To start the learner lab, click Start Lab. Wait for some time until the lab starts. The yellow signal indicates that the lab is being started. After the lab starts, the red signal changes to green, indicating the lab has started.



4. Click on AWS to open AWS Management Console. In the following page, click on EC2.



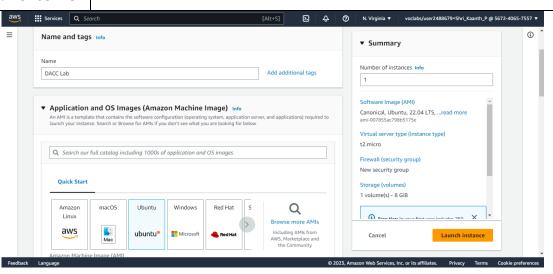
5. Click on **Instances** to view the EC2 instances created. To create a new instance, click **Launch Instance** 



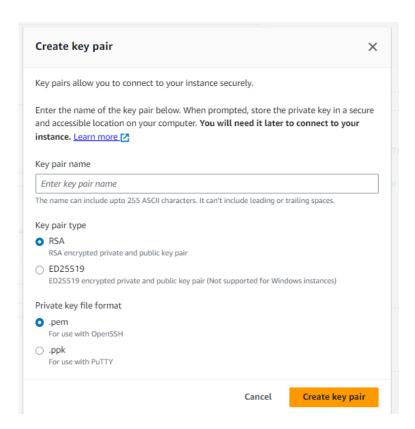
6. Select name, Application and OS image, appropriate netwoork settings and VM storage.

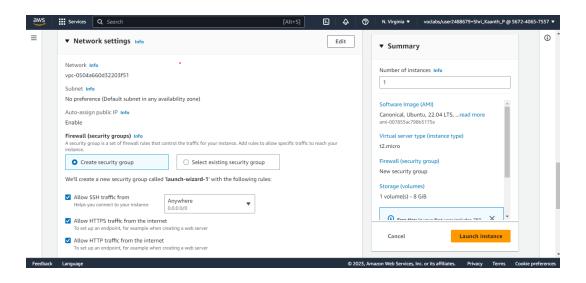
**EXP NO:** 09 **DATE:** 15-03-23

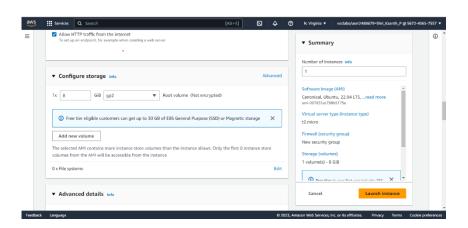
### **MONGODB BASICS**



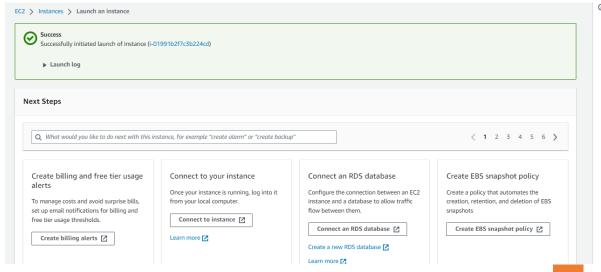
For key-pair attribute, click on **Create new key pair**. Enter the name of the key pair and select .pem for using in OpenSSH and click **Create key pair**. A .pem file will be downloaded.



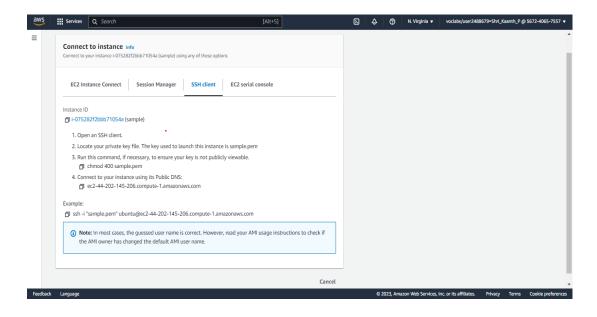




7. Click **Launch instance** to create the new instance. Your EC2 instance is created. After creating EC2 instance, click **Connect to insance.** 



8. Go to **SSH client** section and copy the example given.



9. Now, open command prompt using start menu. Then go to the directory where the .pem file is saved. By default, it is saved in **Downloads** folder.

```
Microsoft Windows [Version 10.0.22621.1413]
(c) Microsoft Corporation. All rights reserved.

C:\Users\shrik>cd Downloads

C:\Users\shrik\Downloads>
```

10. Paste the command that is copied from the SSH client section after creating the instance and press Enter.

ssh -i "dacc lab keypair.pem" ubuntu@ec2-54-147-58-180.compute-1.amazonaws.com

# You've logged in to your virtual machine.

```
C:\Users\shrik\Downloads>ssh -i "sample.pem" ubuntu@ec2-44-202-145-206.compute-1.amazonaws.com
Welcome to Ubuntu 22.04.2 LTS (GNU/Linux 5.15.0-1031-aws x86_64)
* Documentation: https://help.ubuntu.com
 * Management:
                  https://landscape.canonical.com
 * Support:
                  https://ubuntu.com/advantage
 System information as of Wed Mar 29 04:23:27 UTC 2023
 System load: 0.0
                                 Processes:
                                                        97
 Usage of /: 20.4% of 7.57GB Users logged in:
 Memory usage: 22%
                                 IPv4 address for eth0: 172.31.95.203
 Swap usage:
 * Introducing Expanded Security Maintenance for Applications.
   Receive updates to over 25,000 software packages with your
  Ubuntu Pro subscription. Free for personal use.
     https://ubuntu.com/aws/pro
Expanded Security Maintenance for Applications is not enabled.
0 updates can be applied immediately.
Enable ESM Apps to receive additional future security updates.
See https://ubuntu.com/esm or run: sudo pro status
Last login: Wed Mar 29 03:22:19 2023 from 106.203.91.247
To run a command as administrator (user "root"), use "sudo <command>".
See "man sudo_root" for details.
ubuntu@ip-172-31-95-203:~$
```

11.Enter the command below to create and edit a new file. Enter the file content and press Ctrl+c.

```
cat > sample.txt
```

```
ubuntu@ip-172-31-31-114:~$ cat > dacc.txt
Welcome to DACC Laboratory
^C
ubuntu@ip-172-31-31-114:~$
```

12. Enter the command below to display the file content.

```
cat dacc.txt
```

```
ubuntu@ip-172-31-31-114:~$ cat dacc.txt
Welcome to DACC Laboratory
```

# **Result:**

Thus, VM has been deployed in OpenStack and a simple application has been executed.

**EXP NO:** 09 **DATE:** 29-03-23

### **MONGODB BASICS**

# 1. Study on MongoDB

### Aim:

To study about the basics of MongoDB and its operations.

# Theory:

MongoDB is a general-purpose document database designed for modern application development and for the cloud. Its scale-out architecture allows you to meet the increasing demand for your system by adding more nodes to share the load.

MongoDB is a cross-platform, document-oriented database that provides, high performance, high availability, and easy scalability. MongoDB works on concept of collection and document.

### **Document**

MongoDB stores data as JSON documents. The document data model maps naturally to objects in application code, making it simple for developers to learn and use. A document is a set of key-value pairs. Documents have dynamic schema. Dynamic schema means that documents in the same collection do not need to have the same set of fields or structure, and common fields in a collection's documents may hold different types of data. Documents can be nested to express hierarchical relationships and to store structures such as arrays.

### Collection

Collection is a group of MongoDB documents. It is the equivalent of an RDBMS table. A collection exists within a single database. Collections do not enforce a schema. Documents within a collection can have different fields. Typically, all documents in a collection are of similar or related purpose.

id is a 12 bytes hexadecimal number which assures the uniqueness of every document. The first 4 bytes for the current timestamp, next 3 bytes for machine id, next 2 bytes for process id of MongoDB server and remaining 3 bytes are simple incremental VALUE.

# Sample document

```
{
    __id: ObjectId("6418761848304d22ba39626b"),
    emp_name: 'Ram',
    emp_id: 1001,
    age: 25,
    dept: 'Sales'
}
```

# **Operations**

### 1. Use

To create and use a database. If the database exists, it returns existing database

use employee

local> use employee switched to db employee

# 2. Create

To create/insert a new document to a collection

# insertOne( )

```
db.data.insertOne({emp_name:"Ram",emp_id:1001,age:25, dept:"Sales"})
```

```
employee> db.data.insertOne({emp_name:"Ram",emp_id:1001,age:25,dept:"Sales"})
{
   acknowledged: true,
   insertedId: ObjectId("6418761848304d22ba39626b")
}
```

# insertMany()

db.data.insertMany([{emp\_name:"Sam",emp\_id:1002,age:30,dept:"Accounts"},{emp\_name:"Raju",emp\_id:1003,age:35,dept:"Management"},{emp\_name:"Balu",emp\_id:1004,age:40,dept:"Production"}])

### 3. Read

To retrieve documents from a collection

```
db.data.find()
```

```
employee> db.data.find()
  {
    _id: ObjectId("6418761848304d22ba39626b"),
    emp_name: 'Ram',
    emp_id: 1001,
    age: 25,
    dept: 'Sales'
    _id: ObjectId("641876be48304d22ba39626c"),
    emp_name: 'Sam',
    emp_id: 1002,
    age: 30,
    dept: 'Accounts'
    _id: ObjectId("641876be48304d22ba39626d"),
    emp_name: 'Raju',
    emp_id: 1003,
    dept: 'Management',
    age: 35
    _id: ObjectId("641876be48304d22ba39626e"),
    emp_id: 1004,
    emp_name: 'Balu',
    age: 40,
dept: 'Production'
```

## 4. Update

To modify existing documents in a collection

```
db.data.updateOne({emp_id:1003},{$set:{age:50}})
```

```
employee> db.data.updateOne({emp_id:1003},{$set:{age:50}})
{
   acknowledged: true,
   insertedId: null,
   matchedCount: 1,
   modifiedCount: 0
}
employee> db.data.find({emp_id:1003})
[
   {
    _id: ObjectId("641876be48304d22ba39626d"),
     emp_name: 'Raju',
     emp_id: 1003,
     dept: 'Management',
     age: 50
}
```

### 5. Delete

To remove documents from a collection

```
db.data.deleteMany({age:{$lt:38}})
```

```
employee> db.data.updateOne({emp_id:1003},{$set:{age:50}})
{
    acknowledged: true,
    insertedId: null,
    matchedCount: 1,
    upsertedCount: 0
}
employee> db.data.find({emp_id:1003})
[
    {
        id: ObjectId("641876be48304d22ba39626d"),
        emp_name: 'Raju',
        emp_id: 1003,
        dept: 'Management',
        age: 50
}
```

#### **Result:**

**EXP NO:** 10 **DATE:** 12-04-23

### **HADOOP INSTALLATION**

A student management system has been designed using MongoDB and CRUD operations have been performed on it successfully.

### Aim:

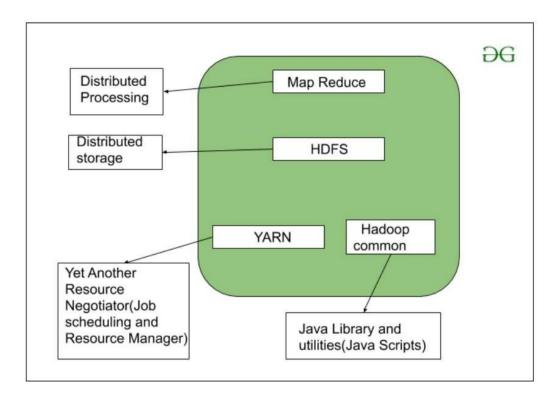
To study about the architecture of Hadoop and its components.

## **Theory:**

Hadoop is a framework written in Java that utilizes a large cluster of commodity hardware to maintain and store big size data. Hadoop works on MapReduce Programming Algorithm that was introduced by Google.

The Hadoop Architecture Mainly consists of 4 components:

- MapReduce
- HDFS(Hadoop distributed File System)
- YARN(Yet Another Resource Framework)
- Common Utilities or Hadoop Common



### **Components of Hadoop Architecture**

## 1. MapReduce

MapReduce is an Algorithm or a data structure that is based on the YARN framework. The major feature of MapReduce is to perform the distributed processing in parallel in a Hadoop cluster which Makes Hadoop fast. MapReduce has 2 main tasks which are divided phase-wise. In first phase, Map is utilized and in next phase Reduce is utilized.

The Input is provided to the Map() function and then its output is used as an input to the Reduce() function to receive the final output. The Input is a set of Data. The Map() function breaks data blocks into tuples that are key-value pairs. These key-value pairs are now sent as input to the Reduce(). The Reduce() function then combines the broken key-value pairs based on its key value and form set of tuples. Some operation like sorting, summation, etc. is performed which is then sent to the final Output Node which generates the final output.

#### 2. HDFS

The Hadoop Distributed File System (HDFS) is a distributed file system for Hadoop. It contains a master/slave architecture. This architecture consist of a single NameNode performs the role of master, and multiple DataNodes performs the role of a slave. Both NameNode and DataNode are capable enough to run on commodity machines. The Java language is used to develop HDFS. So any machine that supports Java language can easily run the NameNode and DataNode software.

The NameNode manages the file system namespace by executing an operation like the opening, renaming and closing the files. The HDFS cluster contains several DataNodes each of which contains multiple data blocks that store data. The DataNodes read and write requests from the file system's clients.

#### 3. YARN

YARN is a Framework on which MapReduce works. YARN performs 2 operations that are Job scheduling and Resource Management. The Purpose of Job scheduler is to divide a big task into small jobs so that each job can be assigned to various slaves in a Hadoop cluster and processing can be maximized. Job Scheduler also keeps track of which job is important, which job has more priority, dependencies between the jobs and all the other information like job timing, etc. And the use of resource manager is to manage all the resources that are made available for running a Hadoop cluster.

### 4. Hadoop Common

Hadoop common or Common utilities are nothing but Java library and Java files or the java scripts that are needed for all the other components present in a Hadoop cluster. These utilities are used by HDFS, YARN, and MapReduce for running the cluster. Hadoop Common

INSTALL AND CONFIGURE HADOOP IN ITS TWO OPERATING MODES:

HADOOP SINGLE-NODE CLUSTER INSTALLATION

- 1. Download the Java 8 Package. Save this file in your home directory.
- 2. Extract the Java Tar File.

```
tar -xvf jdk-8u101-linux-i586.tar.gz
```

3. Download the Hadoop 2.7.3 Package.

https://archive.apache.org/dist/hadoop/core/hadoop-2.7.3/hadoop-2.7.3.tar.gz

4. Extract the Hadoop tar File.

```
tar -xvf hadoop-2.7.3.tar.gz
```

Add the Hadoop and Java paths in the bash file (.bashrc).
 Open. bashrc file. Now, add Hadoop and Java Path as shown below.

vi .bashrc

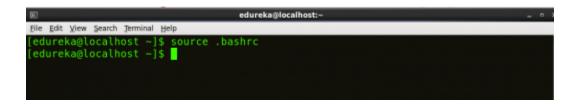
```
# User specific aliases and functions

export HADOOP HOME=$HOME/hadoop-2.7.3
export HADOOP CONF DIR=$HOME/hadoop-2.7.3/etc/hadoop
export HADOOP MAPRED HOME=$HOME/hadoop-2.7.3
export HADOOP COMMON HOME=$HOME/hadoop-2.7.3
export HADOOP HDFS HOME=$HOME/hadoop-2.7.3
export YARN HOME=$HOME/hadoop-2.7.3
export YARN HOME=$HOME/hadoop-2.7.3
export PATH=$PATH:$HOME/hadoop-2.7.3/bin

# Set JAVA_HOME
export JAVA_HOME=/home/edureka/jdk1.8.0_101
export PATH=/home/edureka/jdk1.8.0_101/bin:$PATH
```

Then, save the bash file and close it. For applying all these changes to the current Terminal, execute the source command.

source .bashrc



To make sure that Java and Hadoop have been properly installed on your system and can be accessed through the Terminal, execute the java - version and hadoop version commands.

java -version

```
edureka@localhost:- _ ○

File Edit View Search Terminal Help

[edureka@localhost ~]$ java -version
java version "1.8.0_101"

Java(TM) SE Runtime Environment (build 1.8.0_101-b13)

Java HotSpot(TM) 64-Bit Server VM (build 25.101-b13, mixed mode)

[edureka@localhost ~]$
```

hadoop version

```
edureka@localhost-- _ = :

Eile Edit View Search Terminal Help

[edureka@localhost ~]$ hadoop version

Hadoop 2.7.3

Subversion https://git-wip-us.apache.org/repos/asf/hadoop.git -r baa91f7c6bc9cb92be

5982de4719c1c8af91ccff

Compiled by root on 2016-08-18T01:41Z

Compiled with protoc 2.5.0

From source with checksum 2e4ce5f957ea4db193bce3734ff29ff4

This command was run using /home/edureka/hadoop-2.7.3/share/hadoop/common/hadoop-common-2.7.3.jar

[edureka@localhost ~]$ [
```

6. Edit the Hadoop Configuration files.

```
cd hadoop-2.7.3/etc/hadoop/
```

7. Open core-site.xml and edit the property mentioned below inside configuration tag:

8. Open hdfs-site.xml and edit the property mentioned below inside configuration tag:

vi hdfs-site.xml

```
<?xml version="1.0" encoding="UTF-8"?>
<?xml-stylesheet type="text/xs1" href="configuration.xs1"?>
<configuration>
cproperty>
<name>dfs.replication</name>
<value>1</value>
</property>
cproperty>
<name>dfs.permission</name>
<value>false</value>
</property>
</pro>
```

9. Open mapred-site.xml and edit the property mentioned below inside configuration tag:

10. Open yarn-site.xml and edit the property mentioned below inside configuration tag:

</configuration>

11. Edit hadoop-env.sh and add the Java Path as mentioned below:

vi hadoop-env.sh

The java implementation to use.

kport JAVA\_HOME=/home/edureka/jdk1.8.0\_101

12.Go to Hadoop home directory and format the NameNode.

cd

cd Hadoop-2.7.3

bin/Hadoop namenode -format

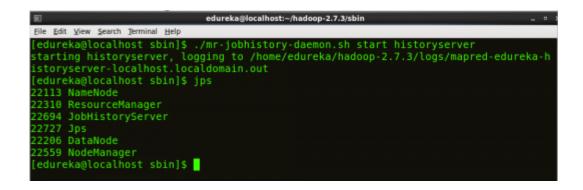
13. Once the NameNode is formatted, go to hadoop-2.7.3/sbin directory and start all the daemons.

cd hadoop-2.7.3/sbin

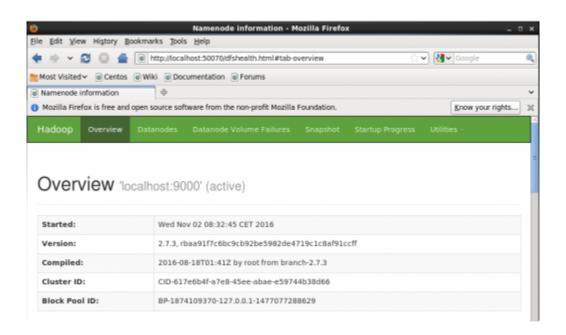
./start-all.sh

14. To check that all the Hadoop services are up and running, run the below command.

jps



15. Now open the Mozilla browser and go to localhost: 50070/dfshealth.html to check the NameNode interface.



Thus, Single-node Hadoop cluster is installed.

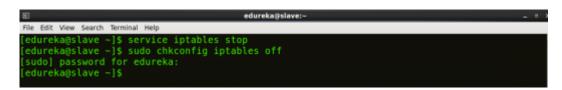
### HADOOP MULTI -NODE CLUSTER INSTALLATION

1. Check the IP address of all machines.

ip addr show

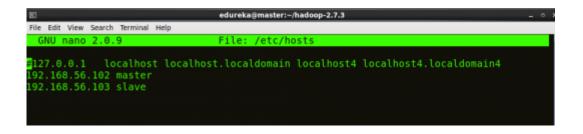
2. Disable the firewall restrictions.

service iptables stop sudo chkconfig iptables off



3. Open hosts file to add master and data node with their respective IP addresses.

sudo nano /etc/hosts



4. Restart the sshd service.

service sshd restart

5. Create the SSH Key in the master node. (Press enter button when it asks you to enter a filename to save the key).

ssh-keygen -t rsa -P ""

```
[edureka@master ~]$ ssh-keygen -t rsa -P ""
Generating public/private rsa key pair.
Enter file in which to save the key (/home/edureka/.ssh/id_rsa):
Your identification has been saved in /home/edureka/.ssh/id_rsa.
Your public key has been saved in /home/edureka/.ssh/id_rsa.pub.
 he key fingerprint is:
9:95:df:04:b3:63:d8:36:7a:ef:dc:20:c2:1f:33:44 edureka@master
he key's randomart image is:
---[ RSA 2048]----+
  edureka@master ~]$
```

6. Copy the generated ssh key to master node's authorized keys.

cat \$HOME/.ssh/id\_rsa.pub >> HOME/.ssh/authorized\_keys

7. Copy the master node's ssh key to slave's authorized keys.

ssh-copy-id -i \$HOME/.ssh/id\_rsa.pub edureka@slave

- 8. Download the Java 8 Package. Save this file in your home directory.
- 9. Extract the Java Tar File.

tar -xvf jdk-8u101-linux-i586.tar.gz

10. Download the Hadoop 2.7.3 Package.

https://archive.apache.org/dist/hadoop/core/hadoop-2.7.3/hadoop-2.7.3.tar.gz

11.Extract the Hadoop tar File.

tar -xvf hadoop-2.7.3.tar.gz

12. Add the Hadoop and Java paths in the bash file (.bashrc).

Open. bashrc file. Now, add Hadoop and Java Path as shown below.

vi .bashrc

Then, save the bash file and close it. For applying all these changes to the current Terminal, execute the source command.

```
source .bashrc
```

```
E edureka@localhost:~ _ □

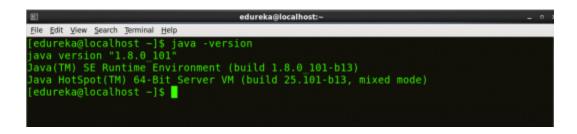
File Edit View Search Jerminal Help

[edureka@localhost ~]$ source .bashrc

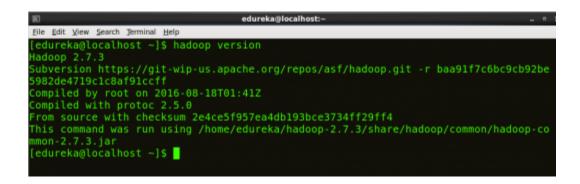
[edureka@localhost ~]$
```

To make sure that Java and Hadoop have been properly installed on your system and can be accessed through the Terminal, execute the java - version and hadoop version commands.

java -version

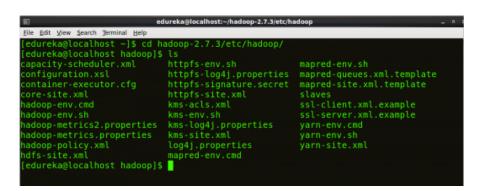


hadoop version



Edit the Hadoop Configuration files.

cd hadoop-2.7.3/etc/hadoop/ ls



13. Create masters file and edit as follows in both master and slave machines

as sudo gedit masters below:

14. Edit slaves file in master machine as follows:

sudo gedit /home/edureka/hadoop-2.7.3/etc/hadoop/slaves

15. Edit slaves file in slave machine as follows:

sudo gedit /home/edureka/hadoop-2.7.3/etc/hadoop/slaves

16. Edit core-site.xml on both master and slave machines as follows:

sudo gedit /home/edureka/hadoop-2.7.3/etc/hadoop/core-site.xml

17.Edit hdfs-site.xml on master machine as follows:

sudo gedit /home/edureka/hadoop-2.7.3/etc/hadoop/hdfs-site.xml

18. Edit core-site.xml on slave machine as follows:

sudo gedit /home/edureka/hadoop-2.7.3/etc/hadoop/core-site.xml

19. Copy mapred-site from the template in configuration folder and the edit mapred-site.xml on both master and slave machines as follows:

cp mapred-sitw.xml.template mapred-site.xml sudo gedit /home/edureka/hadoop-2.7.3/etc/hadoop/mapred-site.xml

20.Edit yarn-site.xml on both master and slave machines as follows:

sudo gedit /home/edureka/hadoop-2.7.3/etc/hadoop/yarn-site.xml

21. Format the namenode (Only on master machine).

hadoop namenode -format

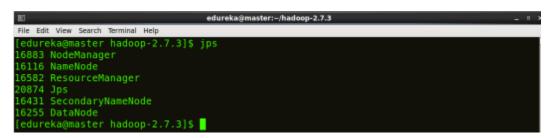
22. Start all daemons (Only on master machine).

./sbin/start-all.sh

23. Check all the daemons running on both master and slave machines.

jps

### On master:



### On slave:

```
Rile Edit View Search Terminal Help

[edureka@slave ~| $ jps

8869 Jps

8618 DataNode

8735 NodeManager

[edureka@slave ~| $ |
```

24. Open the browser and go to master:50070/dfshealth.html on your master machine, this will give you the NameNode interface.

for forward for a city	34.47 GB		
Configured Capacity:	34.47 GB		
DFS Used:	48 KB (0%)		
Non DFS Used:	17.15 GB		
DFS Remaining:	17.32 GB (50.25%)		
Block Pool Used:	48 KB (0%)		
DataNodes usages% (Min/Median/Max/stdDev):	0.00% / 0.00% / 0.00% / 0.00%		
Live Nodes	2 (Decommissioned: 0)		
Dead Nodes	0 (Decommissioned: 0)		
Decommissioning Nodes	0		
Total Datanode Volume Failures	0 (0 B)		
Number of Under-Replicated Blocks	0		
Number of Blocks Pending Deletion	0		

Thus, Multi-node Hadoop cluster is installed.

# **RESULT:**

Thus, Hadoop has been successfully installed.

**VISUALIZATION TOOLS IN PYTHON** 

**EXP NO:** 11 **DATE:** 19-04-23

### Aim:

To study about the visualization tools in python that are used for data analytics.

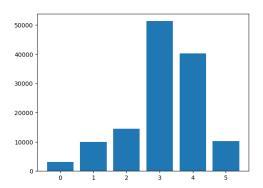
### IMPORTING PACKAGES AND DATASET

```
import pandas as pd
import numpy as np
df = pd.read_csv("dataset.csv")
df.head()
```

	PM2.5	PM10	NO2	SO2	O3	AQI
0	104.00	148.50	23.00	15.30	117.62	3.0
1	94.50	142.00	16.25	17.00	136.23	3.0
2	82.75	126.50	14.83	15.40	149.92	3.0
3	68.50	117.00	13.60	21.80	161.70	3.0
4	69.25	112.25	11.80	21.38	161.68	3.0

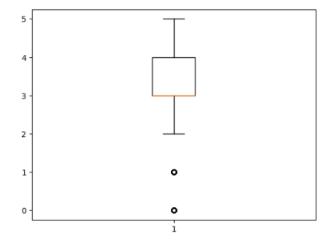
### 1. BARPLOT

import matplotlib.pyplot as plt
plot = df['AQI'].value\_counts()
plt.bar(plot.index,plot.values)



## 2. BOXPLOT

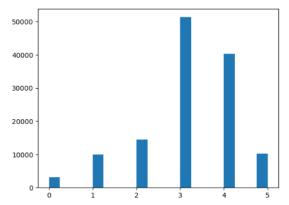
plt.boxplot(df['AQI'])



### 3. HISTOGRAM

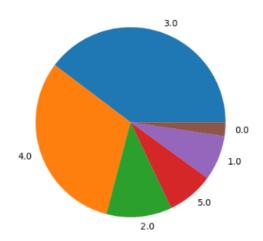
plt.hist(df['AQI'],bins=20)

```
(array([ 3047., 0., 0., 0., 9896., 0., 0., 0., 0., 14425., 0., 0., 0., 51330., 0., 0., 0., 0., 40336., 0., 0., 10243.]), array([0., 0.25, 0.5, 0.75, 1., 1.25, 1.5, 1.75, 2., 2.25, 2.5, 2.75, 3., 3.25, 3.5, 3.75, 4., 4.25, 4.5, 4.75, 5.]), <a href="mailto:abcooks are also be a second as a secon
```



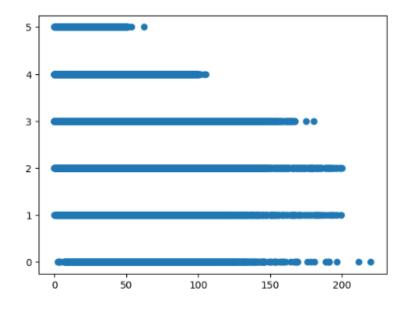
## 4. PIECHART

plt.pie(df['AQI'].value\_counts(),labels=df['AQI'].value\_counts().index)



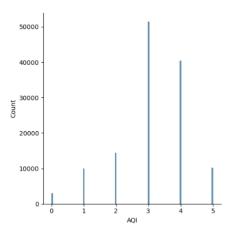
## 5. SCATTERPLOT

plt.scatter(df['O3'],df['AQI'])



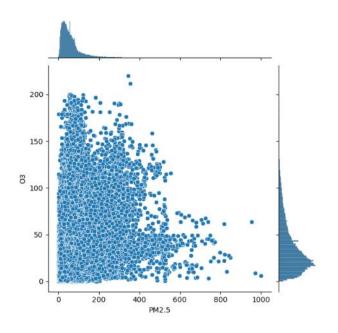
# 6. SEABORN

import seaborn as sns sns.displot(df['AQI'])



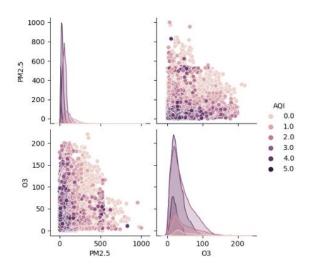
# 7. JOINTPLOT

sns.jointplot(x='PM2.5',y='O3',data=df)



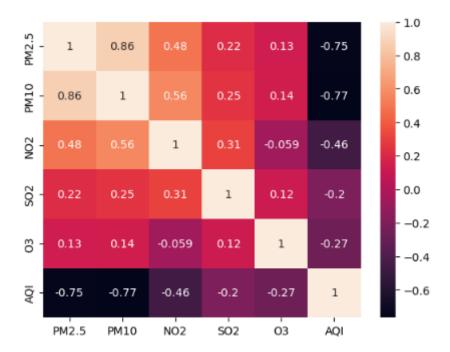
# 8. PAIRPLOT

sns.pairplot(df, vars=['PM2.5', 'O3'], hue='AQI')



## 9. HEATMAP

sns.heatmap(df.corr(), annot = True)



# **RESULT:**

Thus, a study on the visualization tools in Python used for data analysis has been made.