

DEPARTMENT OF INFORMATION TECHNOLOGY

**BIG DATA ANALYTICS
LAB MANUAL**

IV B.TECH I SEMESTER



**DEPARTMENT
OF
INFORMATION TECHNOLOGY**

MALLA REDDY COLLEGE OF ENGINEERING & TECHNOLOGY

(Autonomous institution –UGC , Govt of India)

Affiliated to JNTUH & Approved by AICTE, New Delhi

2025-2026

DEPARTMENT OF INFORMATION TECHNOLOGY

VISION

- To achieve high quality in technical education that provides the skills and attitude to adapt to the global needs of the Information Technology sector, through academic and research excellence.

MISSION

- To equip the students with the cognizance for problem solving and to improve the teaching learning pedagogy by using innovative techniques.
- To strengthen the knowledge base of the faculty and students with motivation towards possession of effective academic skills and relevant research experience.
- To promote the necessary moral and ethical values among the engineers, for the betterment of the society.

PROGRAMME EDUCATIONAL OBJECTIVES (PEOs)

PEO1 – ANALYTICAL SKILLS

- ❖ To facilitate the graduates with the ability to visualize, gather information, articulate, analyze, solve complex problems, and make decisions. These are essential to address the challenges of complex and computation intensive problems increasing their productivity.

PEO2 – TECHNICAL SKILLS

- ❖ To facilitate the graduates with the technical skills that prepare them for immediate employment and pursue certification providing a deeper understanding of the technology in advanced areas of computer science and related fields, thus encouraging to pursue higher education and research based on their interest.

PEO3 – SOFT SKILLS

- ❖ To facilitate the graduates with the soft skills that include fulfilling the mission, setting goals, showing self-confidence by communicating effectively, having a positive attitude, get involved in team-work, being a leader, managing their career and their life.

PEO4 – PROFESSIONAL ETHICS

- ❖ To facilitate the graduates with the knowledge of professional and ethical responsibilities by paying attention to grooming, being conservative with style, following dress codes, safety codes, and adapting themselves to technological advancements.

PROGRAM SPECIFIC OUTCOMES (PSOs)

After the completion of the course, B. Tech Information Technology, the graduates will have the following Program Specific Outcomes:

1. **Fundamentals and critical knowledge of the Computer System:** Able to understand the working principles of the computer System and its components, Apply the knowledge to build, asses, and analyze the software and hardware aspects of it.
2. **The comprehensive and applicative knowledge of Software Development:** Comprehensive skills of Programming Languages, Software process models, methodologies, and able to plan, develop, test, analyze, and manage the software and hardware intensive systems in heterogeneous platforms individually or working in teams.
3. **Applications of Computing Domain & Research:** Able to use the professional, managerial, interdisciplinary skill set, and domain specific tools in development processes, identify the research gaps, and provide innovative solutions to them.

PROGRAM OUTCOMES (POs)

Engineering Graduates should possess the following:

1. **Engineering knowledge:** Apply the knowledge of mathematics, science, engineering fundamentals, and an engineering specialization to the solution of complex engineering problems.
2. **Problem analysis:** Identify, formulate, review research literature, and analyze complex engineering problems reaching substantiated conclusions using first principles of mathematics, natural sciences, and engineering sciences.
3. **Design / development of solutions:** Design solutions for complex engineering problems and design system components or processes that meet the specified needs with appropriate consideration for the public health and safety, and the cultural, societal, and environmental considerations.
4. **Conduct investigations of complex problems:** Use research-based knowledge and research methods including design of experiments, analysis and interpretation of data, and synthesis of the information to provide valid conclusions.
5. **Modern tool usage:** Create, select, and apply appropriate techniques, resources, and modern engineering and IT tools including prediction and modeling to complex engineering activities with an understanding of the limitations.
6. **The engineer and society:** Apply reasoning informed by the contextual knowledge to assess societal, health, safety, legal and cultural issues and the consequent responsibilities relevant to the professional engineering practice.
7. **Environment and sustainability:** Understand the impact of the professional engineering solutions in societal and environmental contexts, and demonstrate the knowledge of, and need for sustainable development.
8. **Ethics:** Apply ethical principles and commit to professional ethics and responsibilities and norms of the engineering practice.
9. **Individual and team work:** Function effectively as an individual, and as a member or leader in diverse teams, and in multidisciplinary settings.
10. **Communication:** Communicate effectively on complex engineering activities with the engineering community and with society at large, such as, being able to comprehend and write effective reports and design documentation, make effective presentations, and give and receive clear instructions.
11. **Project management and finance:** Demonstrate knowledge and understanding of the engineering and management principles and apply these to one's own work, as a member and leader in a team, to manage projects and in multi-disciplinary environments.
12. **Life- long learning:** Recognize the need for, and have the preparation and ability to engage in independent and life-long learning in the broadest context of technological change.



MALLA REDDY COLLEGE OF ENGINEERING & TECHNOLOGY

Maisammaguda, Dhulapally Post, Via Hakimpet, Secunderabad – 500100

DEPARTMENT OF INFORMATION TECHNOLOGY

GENERAL LABORATORY INSTRUCTIONS

1. Students are advised to come to the laboratory at least 5 minutes before (to the starting time), those who come after 5 minutes will not be allowed into the lab.
2. Plan your task properly much before to the commencement, come prepared to the lab with the synopsis / program / experiment details.
3. Student should enter into the laboratory with:
 - a. Laboratory observation notes with all the details (Problem statement, Aim, Algorithm, Procedure, Program, Expected Output, etc.,) filled in for the lab session.
 - b. Laboratory Record updated up to the last session experiments and other utensils (if any) needed in the lab.
 - c. Proper Dress code and Identity card.
4. Sign in the laboratory login register, write the TIME-IN, and occupy the computer system allotted to you by the faculty.
5. Execute your task in the laboratory, and record the results / output in the lab observation note book, and get certified by the concerned faculty.
6. All the students should be polite and cooperative with the laboratory staff, must maintain the discipline and decency in the laboratory.
7. Computer labs are established with sophisticated and high-end branded systems, which should be utilized properly.
8. Students / Faculty must keep their mobile phones in SWITCHED OFF mode during the lab sessions. Misuse of the equipment, misbehaviors with the staff and systems etc., will attract severe punishment.
9. Students must take the permission of the faculty in case of any urgency to go out; if anybody found loitering outside the lab / class without permission during working hours will be treated seriously and punished appropriately.
10. Students should LOG OFF/ SHUT DOWN the computer system before he/she leaves the lab after completing the task (experiment) in all aspects. He/she must ensure the system / seat is kept properly.

HEAD OF THE DEPARTMENT

PRINCIPAL

MALLA REDDY COLLEGE OF ENGINEERING AND TECHNOLOGY

IV Year B.TECH -IT-I-SEM

L/T/P/C

-/-/2/1

(R22A0590) BIG DATA ANALYTICS LAB

COURSE OBJECTIVES:

The objectives of this course are

1. To Understand Configuration of various big data Frame Works.
2. To learn various visualization techniques to explore data
3. To implement Map Reduce programs for processing big data.
4. To realize storage of big data using MongoDB.
5. To implement clustering techniques using SPARK and to analyze data using machine learning techniques.

List of Experiments

1. Install, configure and run python, numPy and Pandas.
2. Install, configure and run Hadoop and HDFS.
3. Visualize data using basic plotting techniques in Python.
4. Implement NoSQL Database Operations: CRUD operations, Arrays using MongoDB.
5. Implement Functions: Count – Sort – Limit – Skip – Aggregate using MongoDB.
6. Implement word count / frequency programs using MapReduce.
7. Implement a MapReduce program that processes a dataset.
8. Implement clustering techniques using SPARK.
9. Implement an application that stores big data in MongoDB / Pig using Hadoop / R.

Course Outcomes:

Upon completion of the course, the students should be able to:

1. Proficient in Configuration of various big data Frame Works.
2. Apply various visualization techniques to explore data.
3. Demonstrate data base operations using MongoDB.
4. Build and apply Map-Reduce & NoSQL Concepts.
5. Frame clustering techniques using SPARK.
6. Perform data analysis with machine learning methods.

BIG DATA ANALYTICS LAB

Table of Contents

S.No	Name of the Experiment	Page No.
1.	Install, configure and run python, numPy and Pandas.	1
2.	Install, configure and run Hadoop and HDFS.	16
3.	Visualize data using basic plotting techniques in Python.	28
4.	Implement NoSQL Database Operations: CRUD operations, Arrays using MongoDB.	32
5.	Implement Functions: Count – Sort – Limit – Skip – Aggregate using MongoDB.	37
6.	Implement word count / frequency programs using MapReduce.	43
7.	Implement a MapReduce program that processes a dataset.	44
8.	Implement clustering techniques using SPARK.	45
9.	Implement an application that stores big data in MongoDB / Pig using Hadoop / R.	46

EXPERIMENT: 1

Install, configure and run python, numpy and pandas.

PROGRAM:

AIM: To Installing and Running Applications On python, numpy and pandas.

How to Install Anaconda on Windows?

Anaconda is an open-source software that contains Jupyter, spyder, etc that are used for large data processing, data analytics, heavy scientific computing. Anaconda works for R and python programming language. Spyder(sub-application of Anaconda) is used for python. Opencv for python will work in spyder. Package versions are managed by the package management system called conda.

To begin working with Anaconda, one must get it installed first. Follow the below instructions to Download and install Anaconda on your system:

Download and install Anaconda:

Head over to anaconda.com and install the latest version of Anaconda. Make sure to download the “Python 3.7 Version” for the appropriate architecture.

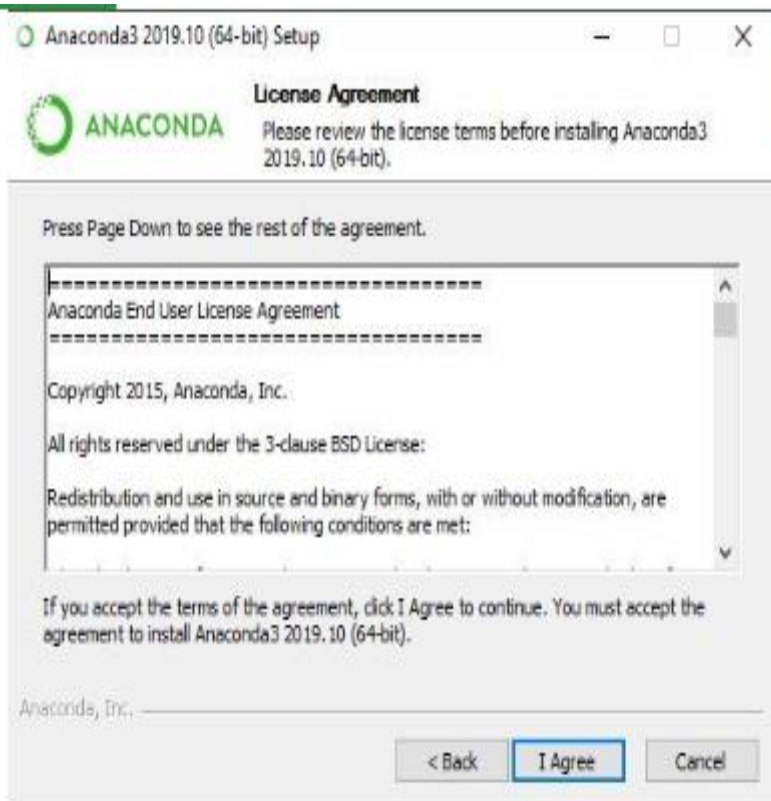


Begin with the installation process:

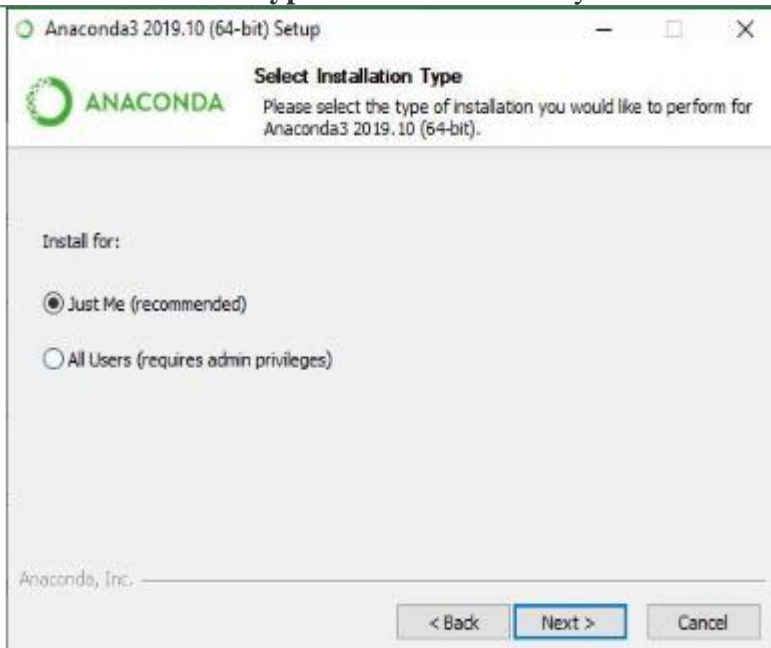
- **Getting Started**



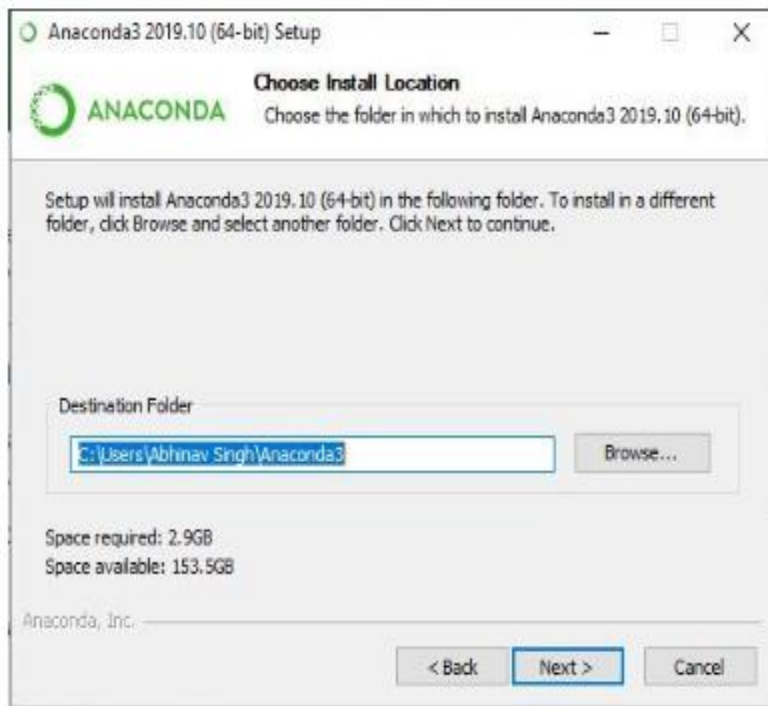
Getting through the License Agreement:



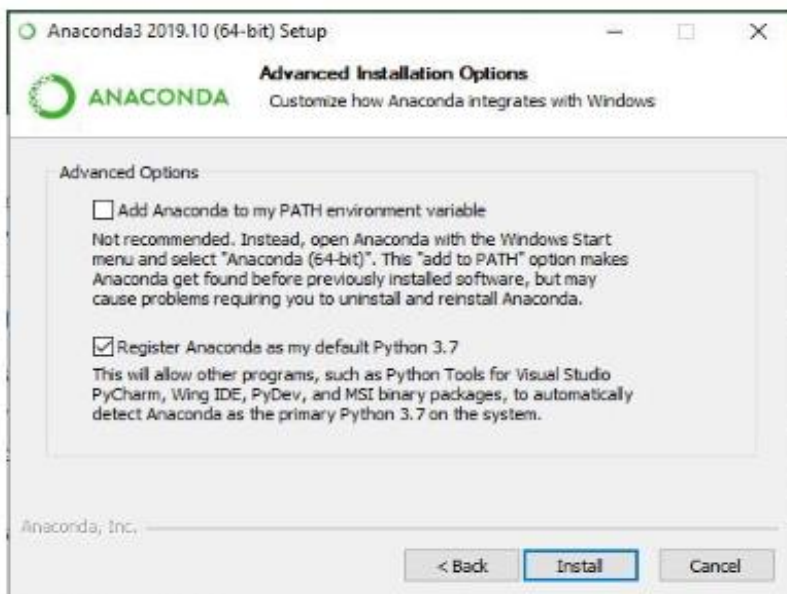
Select Installation Type: Select **Just Me** if you want the software to be used by a single User



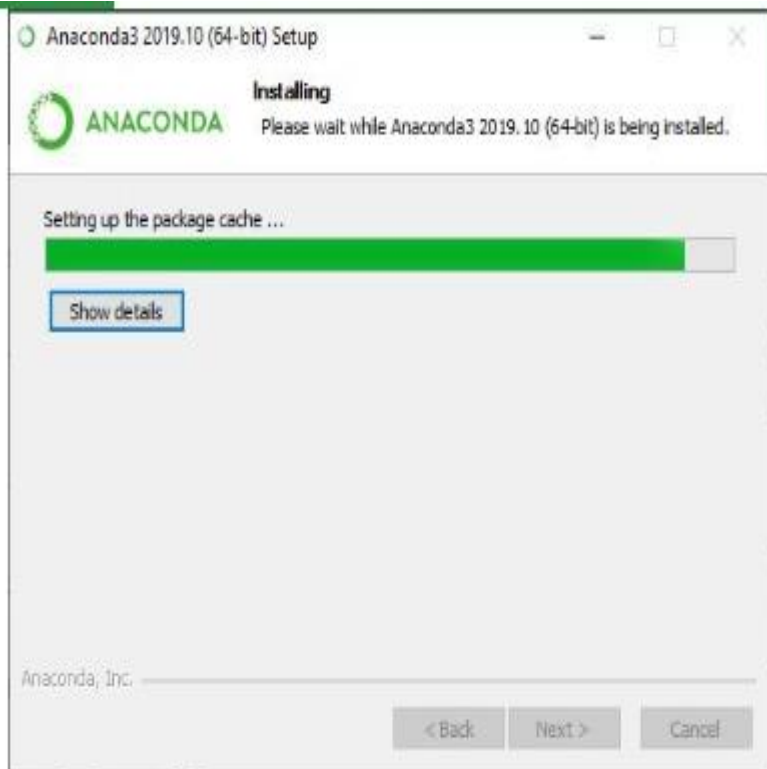
Choose Installation Location:



Advanced Installation Option:



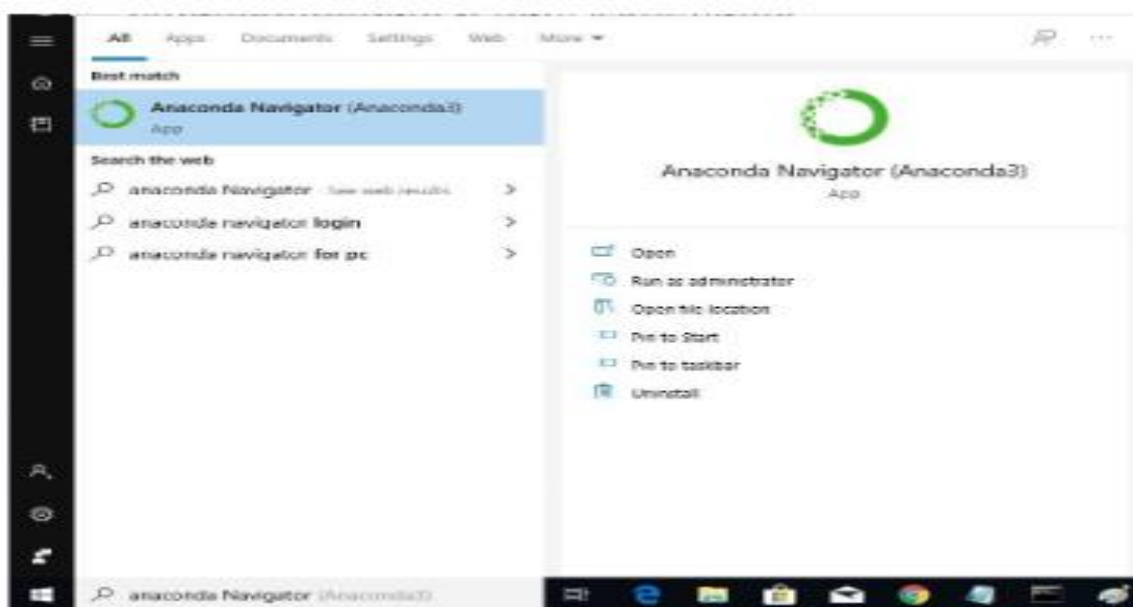
Getting through the Installation Process:

**Recommendation to Install Pycharm:****Finishing up the Installation:**



Working with Anaconda:

Once the installation process is done, Anaconda can be used to perform multiple operations. To Begin using anaconda search for anaconda navigator from the start menu in windows.





```
#import pandas in jupyter notebook
import pandas
```

```
#loading the dataset which is excel file
dataset = pandas.read_csv("crime.csv")
```

```
#displaying the data
dataset
```

	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
0	1965	18073000	836	2320	28182	27464	183443	58452
1	1966	18258000	882	2439	30098	29142	196127	64368
2	1967	18336000	996	2665	40202	31261	219157	83775
3	1968	18113000	1185	2527	59857	34946	250918	104877
4	1969	18321000	1324	2902	64754	36890	248477	115400
5	1970	18190740	1444	2875	81149	39145	267474	125674
6	1971	18391000	1823	3225	97682	42318	273704	127658
7	1972	18366000	2026	4199	86391	45926	239886	105081
8	1973	18265000	2040	4852	80795	47781	246246	112328
9	1974	18111000	1919	5240	86814	51454	271824	104095
10	1975	18120000	1996	5099	93499	54593	301996	116274
11	1976	18084000	1969	4663	95718	54638	318919	133504
12	1977	17924000	1919	5272	84703	57193	309735	133669
13	1978	17748000	1820	5168	83785	58484	292956	119264
14	1979	17649000	2092	5394	93471	60949	308302	124343
15	1980	17506690	2228	5405	112273	60329	360925	133041
16	1981	17594000	2166	5479	120344	60189	350422	136849
17	1982	17659000	2013	5159	107843	59818	295245	137880
18	1983	17667000	1958	5296	94783	59452	249115	127861
19	1984	17735000	1786	5599	89900	64872	222956	115392
20	1985	17783000	1683	5706	89706	68270	219633	106537
21	1986	17772000	1907	5415	91360	76528	217010	113247
22	1987	17825000	2016	5537	89721	82417	216826	125329
23	1988	17898000	2244	5479	97434	91239	218060	153898
24	1989	17950000	2246	5242	103983	91571	211130	171007
25	1990	17990455	2605	5368	112380	92105	208813	187591
26	1991	18058000	2571	5085	112342	90186	204499	181287
27	1992	18119000	2397	5152	108154	87608	193548	168922
28	1993	18197000	2420	5008	102122	85802	181709	151949
29	1994	18169000	2016	4700	86617	82100	164650	128873
30	1995	18136000	1550	4290	72492	74351	146562	102596
31	1996	18185000	1353	4174	61822	64857	129828	89900

```
import pandas as pd
dataset1 = pd.read_csv("crime.csv")
dataset1
```


	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
0	1965	18073000	836	2320	28182	27464	183443	58452
1	1966	18258000	882	2439	30098	29142	196127	64368
2	1967	18336000	996	2665	40202	31261	219157	83775
3	1968	18113000	1185	2527	59857	34946	250918	104877
4	1969	18321000	1324	2902	64754	36890	248477	115400
5	1970	18190740	1444	2875	81149	39145	267474	125674
6	1971	18391000	1823	3225	97682	42318	273704	127658
7	1972	18366000	2026	4199	86391	45926	239886	105081
8	1973	18265000	2040	4852	80795	47781	246246	112328
9	1974	18111000	1919	5240	86814	51454	271824	104095
10	1975	18120000	1996	5099	93499	54593	301996	116274
11	1976	18084000	1969	4663	95718	54638	318919	133504
12	1977	17924000	1919	5272	84703	57193	309735	133669
13	1978	17748000	1820	5168	83785	58484	292956	119264
14	1979	17649000	2092	5394	93471	60949	308302	124343
15	1980	17506690	2228	5405	112273	60329	360925	133041
16	1981	17594000	2166	5479	120344	60189	350422	136849
17	1982	17659000	2013	5159	107843	59818	295245	137880
18	1983	17667000	1958	5296	94783	59452	249115	127861
19	1984	17735000	1786	5599	89900	64872	222956	115392
20	1985	17783000	1683	5706	89706	68270	219633	106537
21	1986	17772000	1907	5415	91360	76528	217010	113247
22	1987	17825000	2016	5537	89721	82417	216826	125329
23	1988	17898000	2244	5479	97434	91239	218060	153898
24	1989	17950000	2246	5242	103983	91571	211130	171007
25	1990	17990455	2605	5368	112380	92105	208813	187591
26	1991	18058000	2571	5085	112342	90186	204499	181287
27	1992	18119000	2397	5152	108154	87608	193548	168922
28	1993	18197000	2420	5008	102122	85802	181709	151949
29	1994	18169000	2016	4700	86617	82100	164650	128873
30	1995	18136000	1550	4290	72492	74351	146562	102596
31	1996	18185000	1353	4174	61822	64857	129828	89900

dataset1.head()

	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
0	1965	18073000	836	2320	28182	27464	183443	58452
1	1966	18258000	882	2439	30098	29142	196127	64368
2	1967	18336000	996	2665	40202	31261	219157	83775
3	1968	18113000	1185	2527	59857	34946	250918	104877
4	1969	18321000	1324	2902	64754	36890	248477	115400

dataset1.tail()

	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
42	2007	19297729	801	2926	31094	45094	64857	28030
43	2008	19467789	836	2799	31789	42122	65537	25096
44	2009	19541453	781	2582	28141	43606	62769	21871
45	2010	19395206	868	2797	28630	44197	65839	20639
46	2011	19465197	774	2752	28396	45568	65397	19311

```
dataset1.head(10)
```

	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
0	1965	18073000	836	2320	28182	27464	183443	58452
1	1966	18258000	882	2439	30098	29142	196127	64368
2	1967	18336000	996	2665	40202	31261	219157	83775
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6	1971	18391000	1823	3225	97682	42318	273704	127658
7	1972	18366000	2026	4199	86391	45926	239886	105081
8	1973	18265000	2040	4852	80795	47781	246246	112328
9	1974	18111000	1919	5240	86814	51454	271824	104095

```
dataset1.tail(10)
```

	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
37	2002	19134293	909	3885	36653	53583	76700	47366
38	2003	19212425	934	3775	35790	48987	75453	45204
39	2004	19280727	889	3608	33506	46911	70696	41002
40	2005	19315721	874	3636	35179	46150	68034	35736
41	2006	19306183	921	3169	34489	45387	68565	32134
42	2007	19297729	801	2926	31094	45094	64857	28030
43	2008	19467789	836	2799	31789	42122	65537	25096
44	2009	19541453	781	2582	28141	43606	62769	21871
45	2010	19395206	868	2797	28630	44197	65839	20639
46	2011	19465197	774	2752	28396	45568	65397	19311

```
type(dataset1)
```

```
pandas.core.frame.DataFrame
```

```
pandas.core.frame.DataFrame
```

```
#to find any null values in the last 5 rows
```

```
dataset1.isnull().tail()
```

	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
42	False	False	False	False	False	False	False	False
43	False	False	False	False	False	False	False	False
44	False	False	False	False	False	False	False	False
45	False	False	False	False	False	False	False	False
46	False	False	False	False	False	False	False	False

```
#to makesure that no null values exists
```

```
dataset1.notnull().tail()
```


	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
42	True	True	True	True	True	True	True	True
43	True	True	True	True	True	True	True	True
44	True	True	True	True	True	True	True	True
45	True	True	True	True	True	True	True	True
46	True	True	True	True	True	True	True	True

#displays the number of null values in each column
dataset1.isnull().sum()

```
Year          0
Population    0
Murder        0
Rape          0
Robbery       0
Assault       0
Burglary      0
CarTheft      0
dtype: int64
```

#helps to find null values with respect to ROBBERY column
dataset1[dataset1.Robbery.isnull()]

Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
------	------------	--------	------	---------	---------	----------	----------

dataset1.shape
(47, 8)

#helps to find how many times values in a particular column has repeated
dataset1['Robbery'].value_counts()

```
94783      1
34489      1
86814      1
86617      1
80795      1
97434      1
108154     1
120344     1
56094      1
28182      1
31094      1
30098      1
91360      1
59857      1
35790      1
36555      1
40202      1
83785      1
```

#consolidated value counts for all the columns in the dataset
for col in dataset1.columns:
 display(dataset1[col].value_counts())



1983	1
1995	1
2004	1
2003	1
2002	1
2001	1
2000	1
1999	1
1998	1
1997	1
1996	1
1994	1
2006	1
1993	1
1992	1
1991	1
1990	1
1989	1
1988	1

#helps to find number of rows in the dataset

```
dataset_length=len(dataset1)
```

```
dataset_length
```

```
47
```

#helps to find number of columns in the dataset

```
dataset_col=len(dataset1.columns)
```

```
dataset_col
```

```
8
```

#helps to find the summary of numerical columns

```
dataset1.describe()
```

	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
count	47.000000	4.700000e+01	47.000000	47.000000	47.000000	47.000000	47.000000	47.000000
mean	1988.000000	1.834426e+07	1549.978723	4200.425532	70429.297872	58022.234043	189119.829787	97573.553191
std	13.711309	6.024504e+05	590.454265	1096.569507	30204.823764	17455.534367	90256.257143	46707.064488
min	1965.000000	1.750669e+07	774.000000	2320.000000	28141.000000	27464.000000	62769.000000	19311.000000
25%	1976.500000	1.793700e+07	922.500000	3197.000000	36604.000000	45477.500000	90581.500000	56246.000000
50%	1988.000000	1.816900e+07	1683.000000	4199.000000	81149.000000	57193.000000	208813.000000	106537.000000
75%	1999.500000	1.868373e+07	2016.000000	5241.000000	94141.000000	64864.500000	250016.500000	128367.000000
max	2011.000000	1.954145e+07	2605.000000	5706.000000	120344.000000	92105.000000	360925.000000	187591.000000

#helps to describe individual column

```
dataset1.Murder.describe()
```

```
count      47.000000
mean      1549.978723
std        590.454265
min        774.000000
25%        922.500000
50%       1683.000000
75%       2016.000000
max       2605.000000
Name: Murder, dtype: float64
```

```
dataset1.skew()
```

```
Year          0.000000
Population    0.795669
Murder        0.059733
Rape          -0.237130
Robbery       -0.134085
Assault       0.464637
Burglary      -0.020278
CarTheft      -0.129653
dtype: float64
```

```
dataset1.var()
```

```
Year          1.880000e+02
Population    3.629465e+11
Murder        3.486362e+05
Rape          1.202465e+06
Robbery       9.123314e+08
Assault       3.046957e+08
Burglary      8.146192e+09
CarTheft      2.181550e+09
dtype: float64
```

```
dataset1.kurtosis()
```

```
Year          -1.200000
Population    -0.692220
Murder        -1.513564
Rape          -1.471445
Robbery       -1.527674
Assault       -0.482013
Burglary      -1.186281
CarTheft      -0.951036
dtype: float64
```

```
print(dataset1.dtypes)
```

```
Year          int64
Population    int64
Murder        int64
Rape          int64
Robbery       int64
Assault       int64
Burglary      int64
CarTheft      int64
dtype: object
```

NUMPY

Numpy is the core library for scientific and numerical computing in Python. It provides high performance multi dimensional array object and tools for working with arrays.

Numpy main object is the multidimensional array, it is a table of elements (usually numbers) all of the same type indexed by a positive integers.

In Numpy dimensions are called as axes.

Numpy is fast, convenient and occupies less memory when compared to python list.

```
import numpy
arr = numpy.array([1, 2, 3, 4, 5])
print(arr)
[1 2 3 4 5]
```

NumPy is usually imported under the np alias.

```
import numpy as np
```

Now the NumPy package can be referred to as np instead of numpy.

```
import numpy as np
arr = np.array([1, 2, 3, 4, 5])
print(arr)
[1 2 3 4 5]
```

Checking NumPy Version

The version string is stored under `__version__` attribute.

```
import numpy as np
print(np.__version__)
1.18.1
```

Create a NumPy ndarray Object

NumPy is used to work with arrays. The array object in NumPy is called ndarray.

We can create a NumPy ndarray object by using the `array()` function.

```
import numpy as np
arr = np.array([1, 2, 3, 4, 5])
print(arr)
print(type(arr))
[1 2 3 4 5]
<class 'numpy.ndarray'>
```

type(): This built-in Python function tells us the type of the object passed to it. Like in above code it shows that `arr` is `numpy.ndarray` type.

To create an ndarray, we can pass a list, tuple or any array-like object into the `array()` method, and it will be converted into an ndarray:

Use a tuple to create a NumPy array:

```
import numpy as np
arr = np.array((1, 2, 3, 4, 5))
print(arr)
[1 2 3 4 5]
```

Dimensions in Arrays

A dimension in arrays is one level of array depth (nested arrays).

nested array: are arrays that have arrays as their elements.

0-D Arrays

0-D arrays, or Scalars, are the elements in an array. Each value in an array is a 0-D array.

```
#Create a 0-D array with value 42
import numpy as np
arr = np.array(42)
print(arr)
42
```

1-D Arrays

These are the most common and basic arrays.

```
#Create a 1-D array containing the values 1,2,3,4,5:
import numpy as np
arr = np.array([1, 2, 3, 4, 5])
print(arr)
[1 2 3 4 5]
```

2-D Arrays

An array that has 1-D arrays as its elements is called a 2-D array.

These are often used to represent matrix or 2nd order tensors.

```
#Create a 2-D array containing two arrays with the values 1,2,3 and 4,5,6:
import numpy as np
arr = np.array([[1, 2, 3], [4, 5, 6]])
print(arr)
[[1 2 3]
 [4 5 6]]
```

3-D arrays

An array that has 2-D arrays (matrices) as its elements is called 3-D array.

These are often used to represent a 3rd order tensor.

```
#Create a 3-D array with two 2-D arrays, both containing two arrays with the values 1,2,3 and 4,5,6:
import numpy as np
arr = np.array([[[1, 2, 3], [4, 5, 6]], [[1, 2, 3], [4, 5, 6]]])
print(arr)
[[[1 2 3]
  [4 5 6]]
 [[1 2 3]
  [4 5 6]]]
```

Check Number of Dimensions?

NumPy Arrays provides the `ndim` attribute that returns an integer that tells us how many dimensions the array have.

#Check how many dimensions the arrays have:

```
import numpy as np
a = np.array(42)
b = np.array([1, 2, 3, 4, 5])
c = np.array([[1, 2, 3], [4, 5, 6]])
d = np.array([[[[1, 2, 3], [4, 5, 6]], [[1, 2, 3], [4, 5, 6]]]])
print(a.ndim)
print(b.ndim)
print(c.ndim)
print(d.ndim)

0
1
2
3
```

#Create an array with 5 dimensions and verify that it has 5 dimensions:

```
import numpy as np
arr = np.array([1, 2, 3, 4], ndmin=5)
print(arr)
print('number of dimensions :', arr.ndim)

[[[[[1 2 3 4]]]])
number of dimensions : 5
```

NumPy Array Indexing

Access Array Elements

Array indexing is the same as accessing an array element.

You can access an array element by referring to its index number.

The indexes in NumPy arrays start with 0, meaning that the first element has index 0, and the second has index 1 etc.

#Get the first element from the following array:

```
import numpy as np
arr = np.array([1, 2, 3, 4])
print(arr[0])
```

#Get the second element from the following array.

```
import numpy as np
arr = np.array([1, 2, 3, 4])
print(arr[1])
```

#Get third and fourth elements from the following array and add them.

```
import numpy as np
arr = np.array([1, 2, 3, 4])
print(arr[2] + arr[3])
```

```
1
2
7
```

Access 2-D Arrays

To access elements from 2-D arrays we can use comma separated integers representing the dimension and the index of the element.

Think of 2-D arrays like a table with rows and columns, where the dimension represents the row and the index represents the column.

#Access the element on the first row, second column:

```
import numpy as np
arr = np.array([[1,2,3,4,5], [6,7,8,9,10]])
print('2nd element on 1st row: ', arr[0, 1])
2nd element on 1st row:  2
```

#Access the element on the 2nd row, 5th column:

```
import numpy as np
arr = np.array([[1,2,3,4,5], [6,7,8,9,10]])
print('5th element on 2nd row: ', arr[1, 4])
```

```
5th element on 2nd row:  10
```

EXPERIMENT: 2**Install, Configure and Run Hadoop and HDFS****PROGRAM:**

AIM: To Installing and Running Applications On Hadoop and HDFS.

HADOOP INSTALATION IN WINDOWS**1. Prerequisites****Hardware Requirement**

* RAM — Min. 8GB, if you have SSD in your system then 4GB RAM would also work.

* CPU — Min. Quad core, with at least 1.80GHz

2. JRE 1.8 — Offline installer for JRE

3. Java Development Kit — 1.8

4. A Software for Un-Zipping like 7Zip or Win Rar

* I will be using a 64-bit windows for the process, please check and download the version supported by your system x86 or x64 for all the software.

5. Download Hadoop zip

* I am using Hadoop-2.9.2, you can use any other STABLE version for hadoop.

Index of /dist/hadoop/core/hadoop-2.9.2

Name	Last modified	Size	Description
Parent Directory	-	-	-
hadoop-2.9.2-src.tar.gz	2020-07-03 04:37	37M	
hadoop-2.9.2-src.tar.gz.asc	2020-07-03 04:36	801	
hadoop-2.9.2-src.tar.gz.mds	2020-07-03 04:36	1.0K	
hadoop-2.9.2.tar.gz	2020-07-03 04:38	349M	
hadoop-2.9.2.tar.gz.asc	2020-07-03 04:37	801	
hadoop-2.9.2.tar.gz.mds	2020-07-03 04:36	1.0K	

Fig. 1:- Download Hadoop 2.9.2

Once we have Downloaded all the above software, we can proceed with next steps in installing the Hadoop.

2. Unzip and Install Hadoop

After Downloading the Hadoop, we need to Unzip the hadoop-2.9.2.tar.gz file.

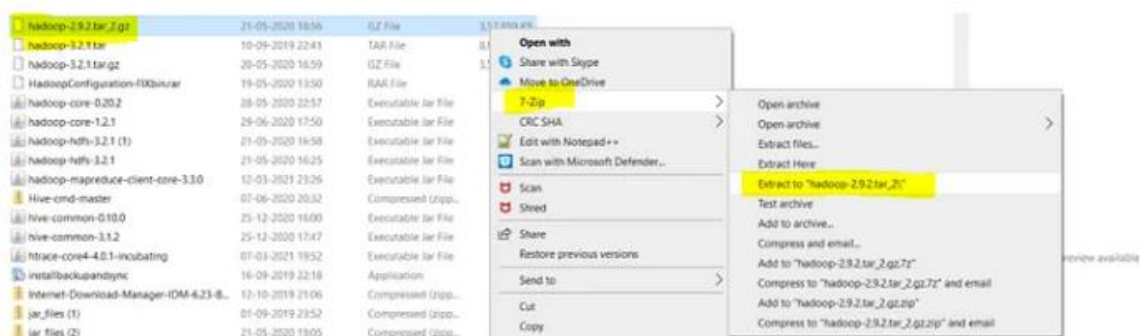


Fig. 2:- Extracting Hadoop Step-1

Once extracted, we would get a new file hadoop-2.9.2.tar.

Now, once again we need to extract this tar file.

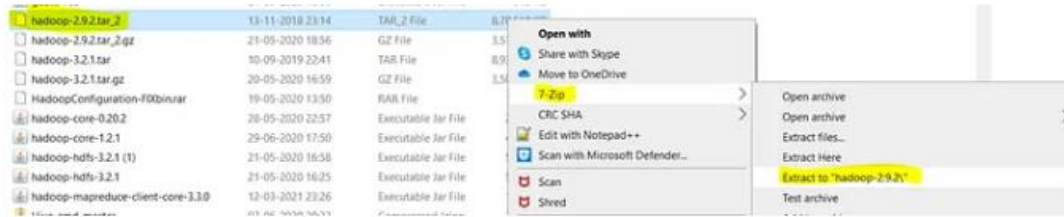


Fig. 3:- Extracting Hadoop Step-2

Now we can organize our Hadoop installation, we can create a folder and move the final extracted file in it. For Eg. :-

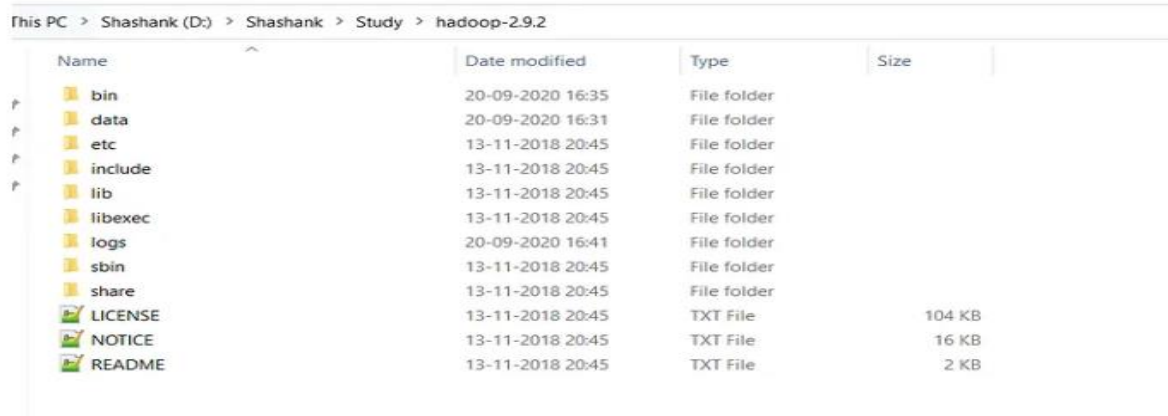


Fig. 4:- Hadoop Directory

Please note while creating folders, DO NOT ADD SPACES IN BETWEEN THE FOLDER NAME.(it can cause issues later)

I have placed my Hadoop in D: drive you can use C: or any other drive also.

3. Setting Up Environment Variables

Another important step in setting up a work environment is to set your Systems environment variable.

To edit environment variables, go to Control Panel > System > click on the “Advanced system settings” link

Alternatively, We can Right click on This PC icon and click on Properties and click on the “Advanced system settings” link

Or, easiest way is to search for Environment Variable in search bar and there you GO... □

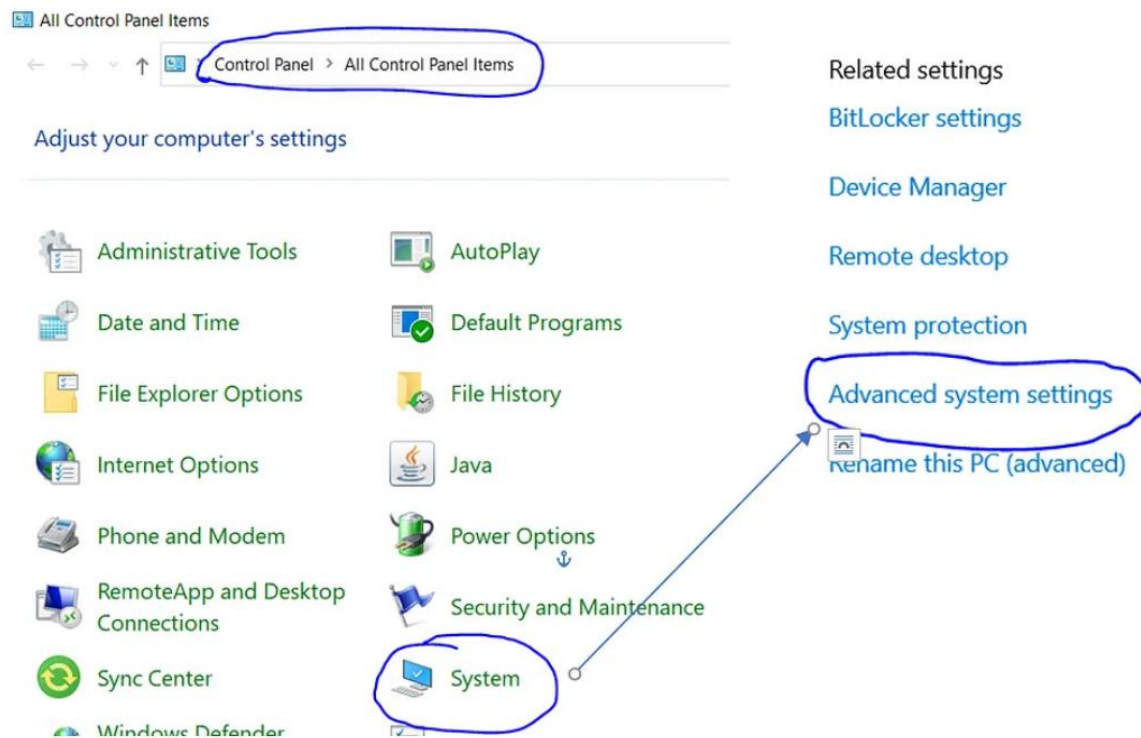


Fig. 5:- Path for Environment Variable

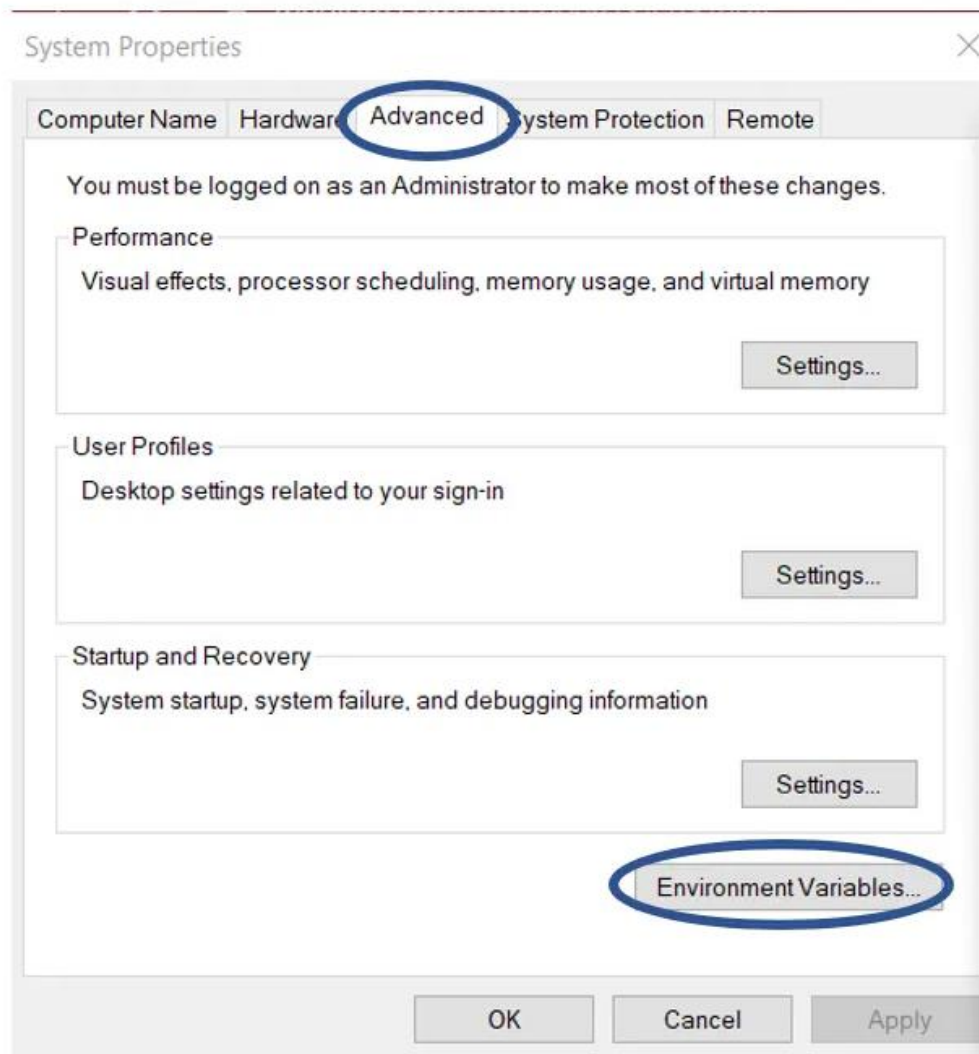


Fig. 6:- Advanced System Settings Screen

3.1 Setting JAVA_HOME

Open environment Variable and click on “New” in “User Variable”

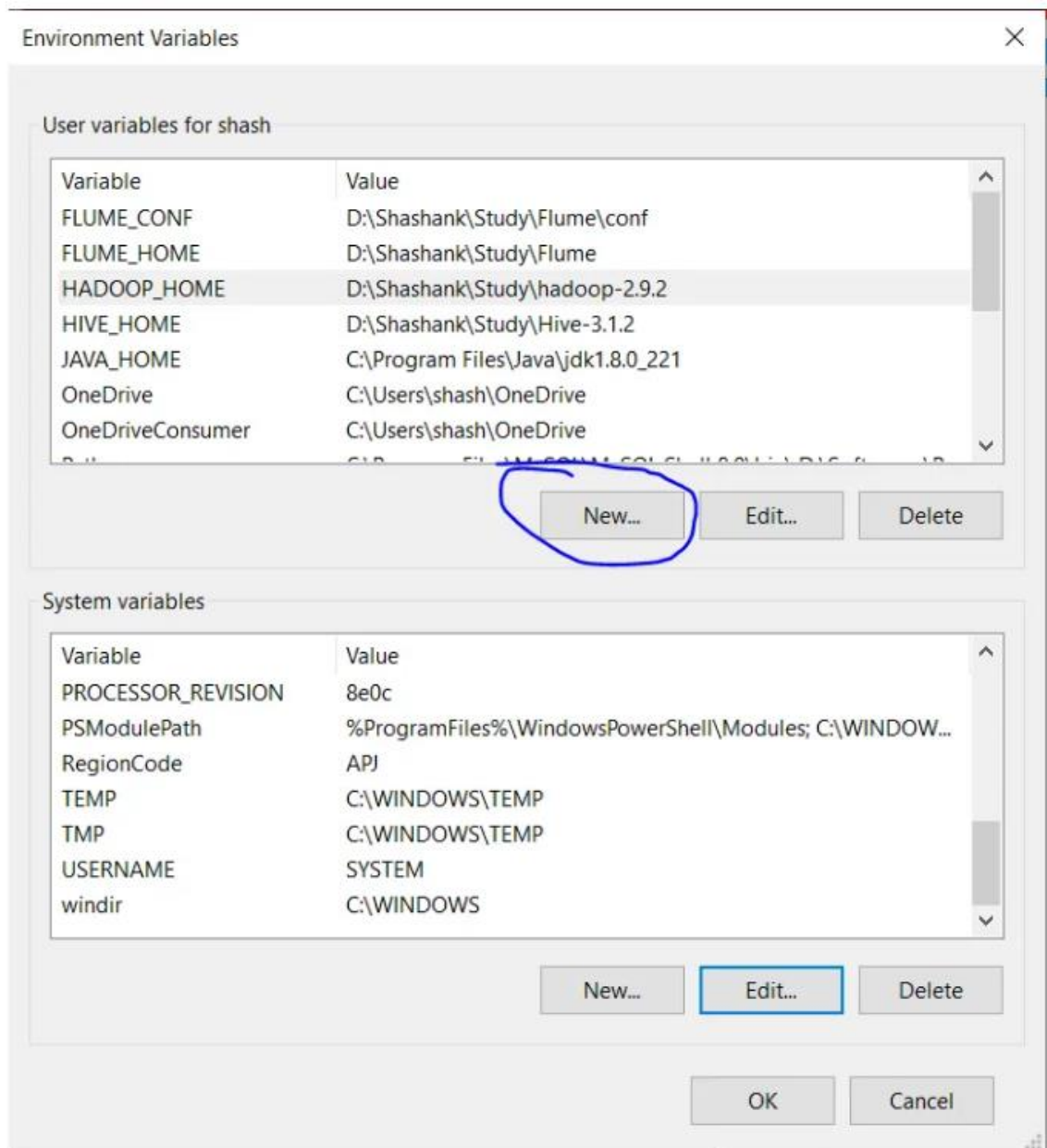


Fig. 7:- Adding Environment Variable

On clicking “New”, we get below screen.

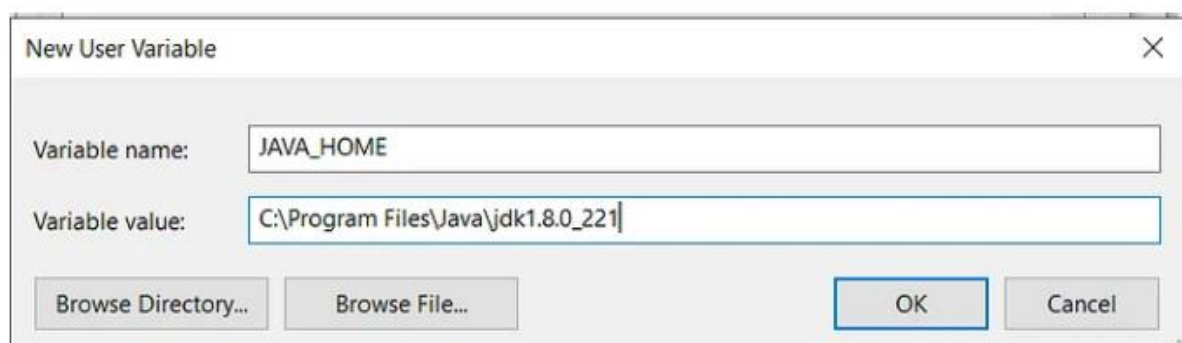


Fig. 8:- Adding JAVA_HOME

Now as shown, add JAVA_HOME in variable name and path of Java(jdk) in Variable Value. Click OK and we are half done with setting JAVA_HOME.

3.2 Setting HADOOP_HOME

Open environment Variable and click on “New” in “User Variable”

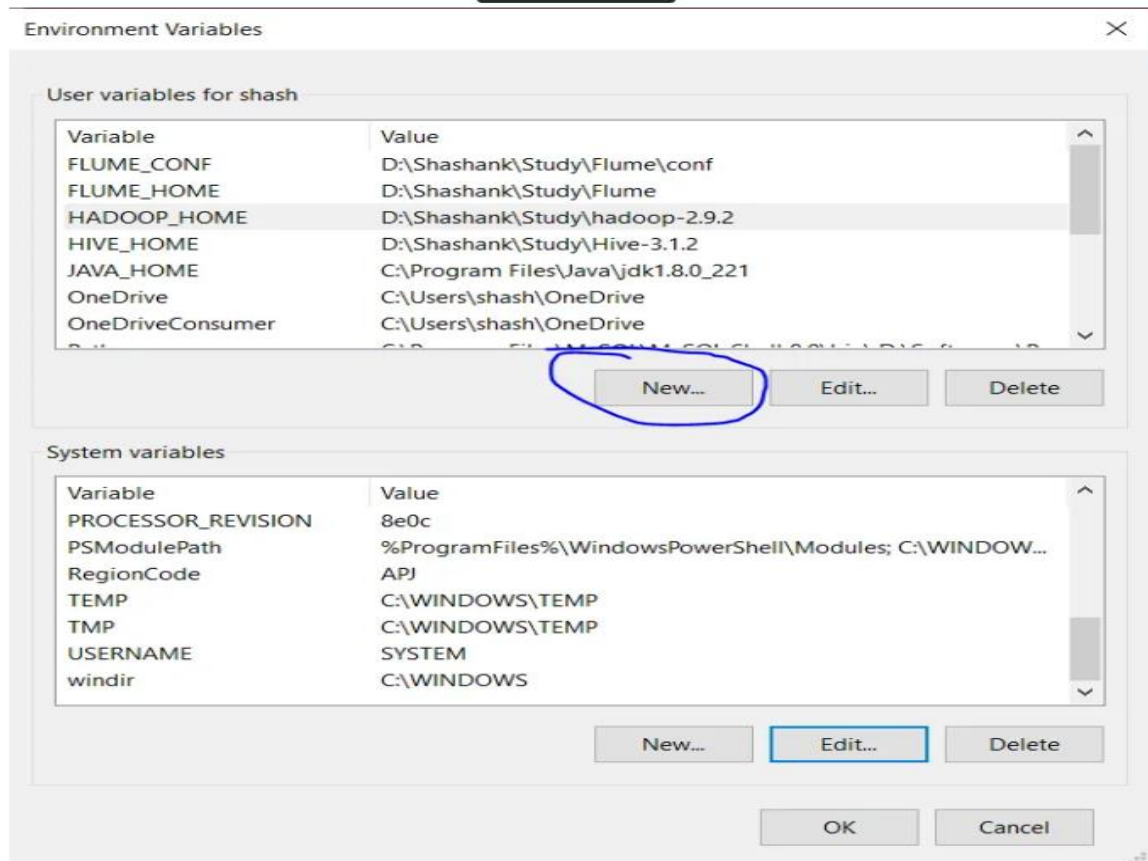


Fig. 9:- Adding Environment Variable

On clicking “New”, we get below screen.

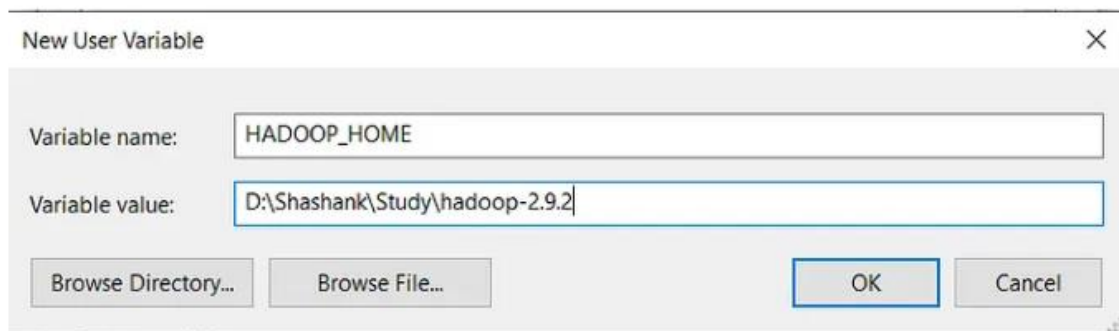


Fig. 10:- Adding HADOOP_HOME

Now as shown, add HADOOP_HOME in variable name and path of Hadoop folder in Variable Value.

Click OK and we are half done with setting HADOOP_HOME.

Note:- If you want the path to be set for all users you need to select “New” from System Variables.

3.3 Setting Path Variable

Last step in setting Environment variable is setting Path in System Variable.

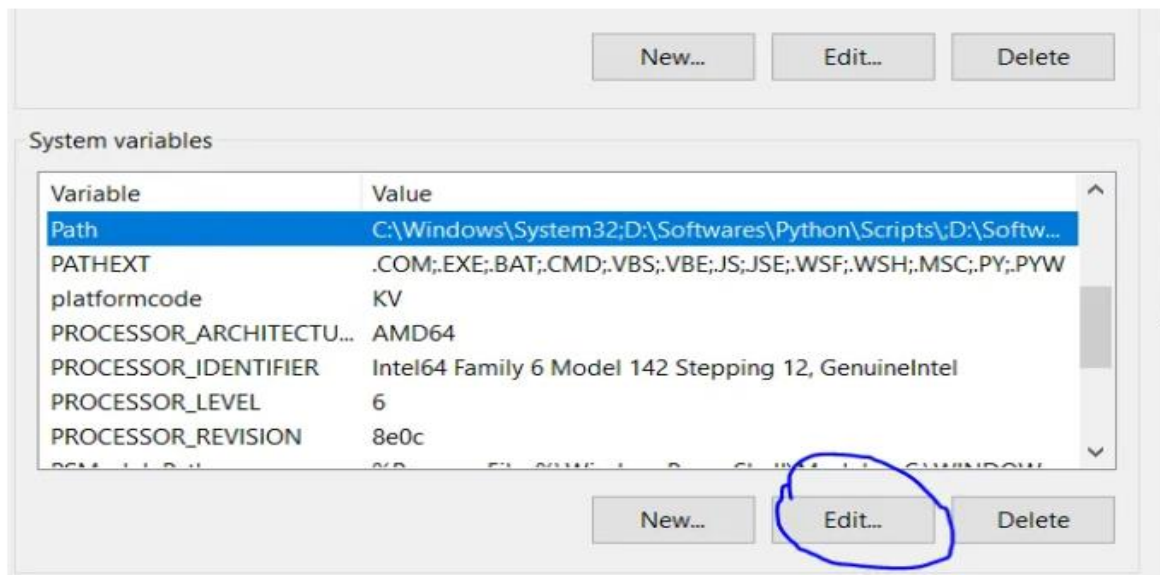


Fig. 11:- Setting Path Variable

Select Path variable in the system variables and click on “Edit”.

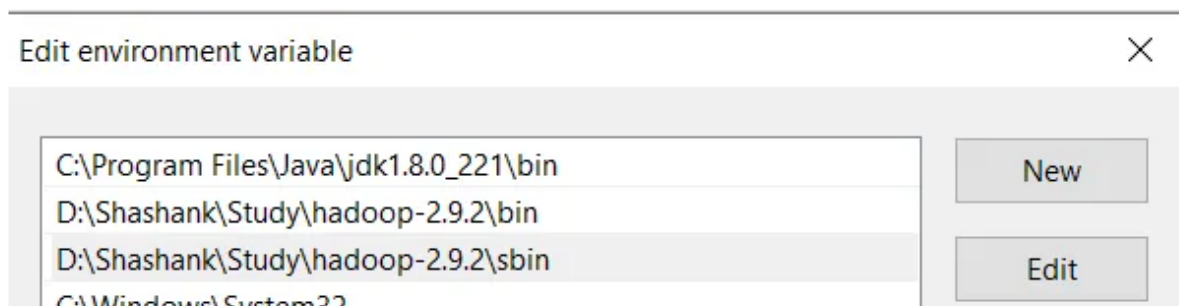


Fig. 12:- Adding Path

Now we need to add these paths to Path Variable one by one:-

- * %JAVA_HOME%\bin
- * %HADOOP_HOME%\bin
- * %HADOOP_HOME%\sbin

Click OK and OK. & we are done with Setting Environment Variables.

3.4 Verify the Paths

Now we need to verify that what we have done is correct and reflecting.

Open a NEW Command Window

Run following commands

```
echo %JAVA_HOME%
echo %HADOOP_HOME%
echo %PATH%
```

4. Editing Hadoop files

Once we have configured the environment variables next step is to configure Hadoop. It has 3 parts:-

4.1 Creating Folders

We need to create a folder data in the hadoop directory, and 2 sub folders namenode and datanode

PC > Shashank (D:) > Shashank > Study > hadoop-2.9.2

Name	Date modified	Type	Size
bin	20-09-2020 16:35	File folder	
data	20-09-2020 16:31	File folder	
etc	13-11-2018 20:45	File folder	
include	13-11-2018 20:45	File folder	
lib	13-11-2018 20:45	File folder	
libexec	13-11-2018 20:45	File folder	
logs	20-09-2020 16:41	File folder	
sbin	13-11-2018 20:45	File folder	
share	13-11-2018 20:45	File folder	
LICENSE	13-11-2018 20:45	TXT File	104 KB
NOTICE	13-11-2018 20:45	TXT File	16 KB
README	13-11-2018 20:45	TXT File	2 KB

Fig. 13:- Creating Data Folder

Create DATA folder in the Hadoop directory

PC > Shashank (D:) > Shashank > Study > hadoop-2.9.2 > data

Name	Date modified	Type	Size
datanode	25-12-2020 15:34	File folder	
namenode	25-12-2020 15:34	File folder	

Fig. 14:- Creating Sub-folders

Once DATA folder is created, we need to create 2 new folders namely, namenode and datanode inside the data folder

These folders are important because files on HDFS resides inside the datanode.

4.2 Editing Configuration Files

Now we need to edit the following config files in hadoop for configuring it :-

(We can find these files in Hadoop -> etc -> hadoop)

- * core-site.xml
- * hdfs-site.xml
- * mapred-site.xml
- * yarn-site.xml
- * hadoop-env.cmd

4.2.1 Editing core-site.xml

Right click on the file, select edit and paste the following content within <configuration> </configuration> tags.

Note:- Below part already has the configuration tag, we need to copy only the part inside it.

```
<configuration>
<property>
  <name>fs.defaultFS</name>
  <value>hdfs://localhost:9000</value>
</property>
</configuration>
```

4.2.2 Editing hdfs-site.xml

Right click on the file, select edit and paste the following content within

<configuration></configuration>tags.

Note:- Below part already has the configuration tag, we need to copy only the part inside it.

Also replace PATH~1 and PATH~2 with the path of namenode and datanode folder that we created recently(step 4.1).

```
<configuration>
<property>
  <name>dfs.replication</name>
  <value>1</value>
</property>
<property>
  <name>dfs.namenode.name.dir</name>
  <value>C:\hadoop\data\namenode</value>
</property>
<property>
  <name>dfs.datanode.data.dir</name>
  <value>C:\hadoop\data\datanode</value>
</property>
</configuration>
```

4.2.3 Editing mapred-site.xml

Right click on the file, select edit and paste the following content within <configuration>

</configuration> tags.

Note:- Below part already has the configuration tag, we need to copy only the part inside it.

```
<configuration>
<property>
  <name>mapreduce.framework.name</name>
  <value>yarn</value>
</property>
</configuration>
```

4.2.4 Editing yarn-site.xml

Right click on the file, select edit and paste the following content within <configuration>

</configuration> tags.

Note:- Below part already has the configuration tag, we need to copy only the part inside it.

```
<configuration>
<property>
  <name>yarn.nodemanager.aux-services</name>
  <value>mapreduce_shuffle</value>
</property>
<property>
  <name>yarn.nodemanager.auxservices.mapreduce.shuffle.class</name>
  <value>org.apache.hadoop.mapred.ShuffleHandler</value>
</property>
</configuration>
```

4.2.5 Verifying hadoop-env.cmd

Right click on the file, select edit and check if the JAVA_HOME is set correctly or not.

We can replace the JAVA_HOME variable in the file with your actual JAVA_HOME that we configured in the System Variable.

set JAVA_HOME=%JAVA_HOME%

OR

set JAVA_HOME="C:\Program Files\Java\jdk1.8.0_221"

4.3 Replacing bin

Last step in configuring the hadoop is to download and replace the bin folder.

- ## 5. Testing Setup

We are done with the setting up the Hadoop in our System.

Now we need to check if everything works smoothly...

5.1 Formatting Namenode

Before starting hadoop we need to format the namenode for this we need to start a NEW Command Prompt and run below command

```
hadoop namenode -format
```

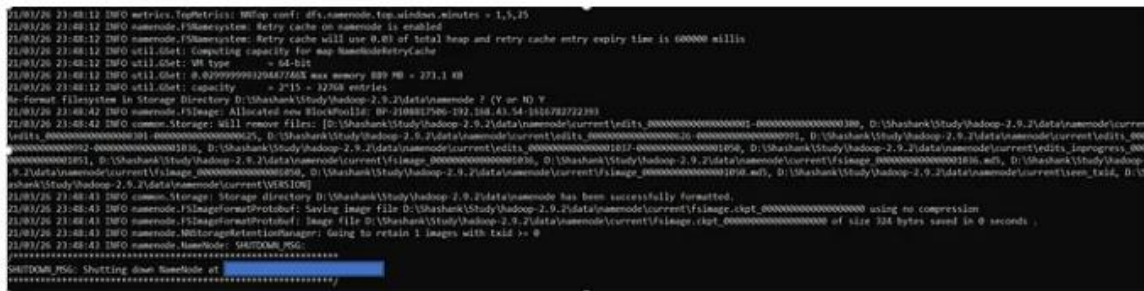


Fig. 15:- Formatting Namenode

Note:- This command formats all the data in namenode. So, its advisable to use only at the start and do not use it every time while starting hadoop cluster to avoid data loss.

5.2 Launching Hadoop

Now we need to start a new Command Prompt remember to run it as administrator to avoid permission issues and execute below commands

```
start-all.cmd
```

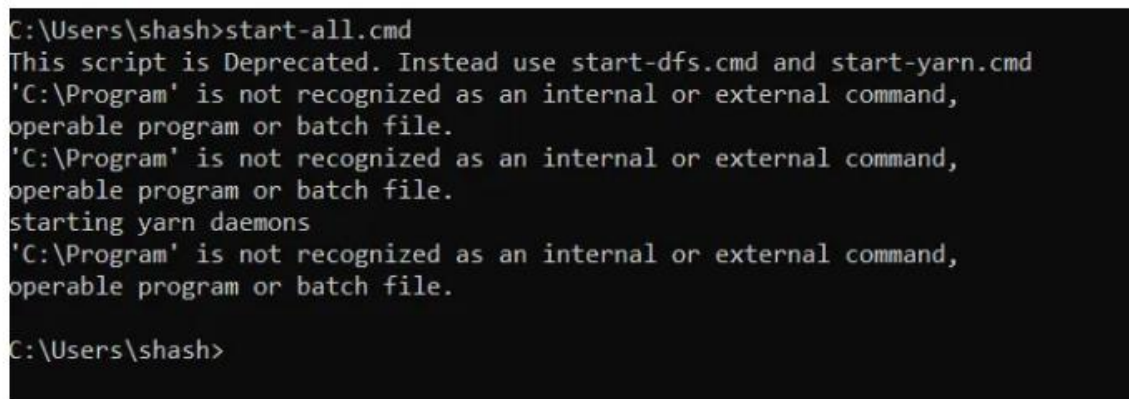


Fig. 16:- start-all.cmd

This will open 4 new cmd windows running 4 different Daemons of hadoop:-

- * Namenode
- * Datanode
- * ResourceManager
- * Nodemanager

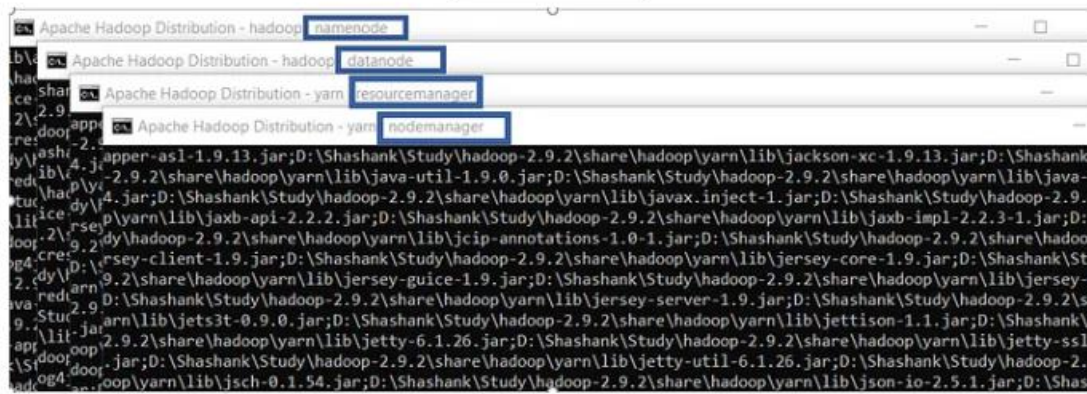


Fig. 17:- Hadoop Daemons

Note:- We can verify if all the daemons are up and running using jps command in new cmd window.

6. Running Hadoop (Verifying Web UIs)

6.1 Namenode

Open localhost:50070 in a browser tab to verify namenode health.

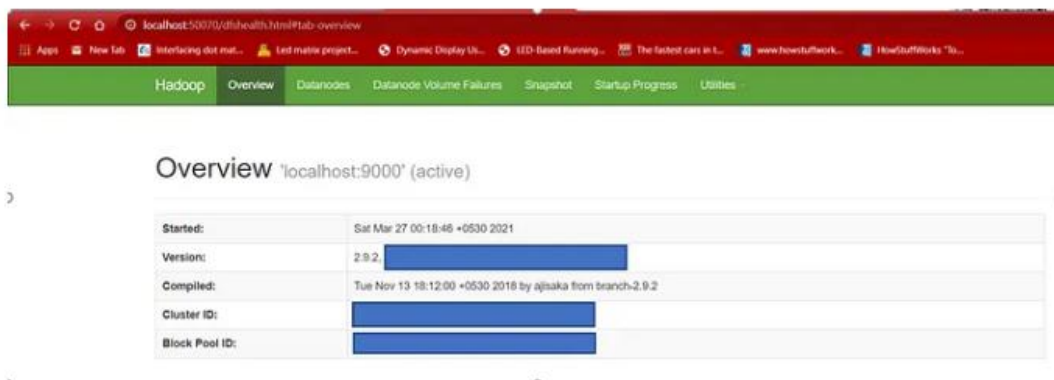


Fig. 18:- Namenode Web UI

6.2 Resourcemanager

Open localhost:8088 in a browser tab to check resourcemanager details.

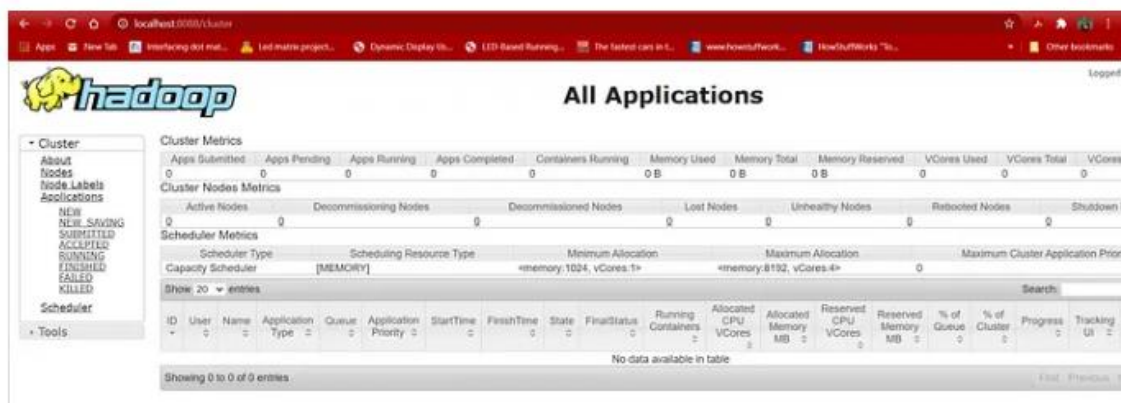


Fig. 19:- Resourcemanager Web UI

6.3 Datanode

Open localhost:50075 in a browser tab to checkout datanode.

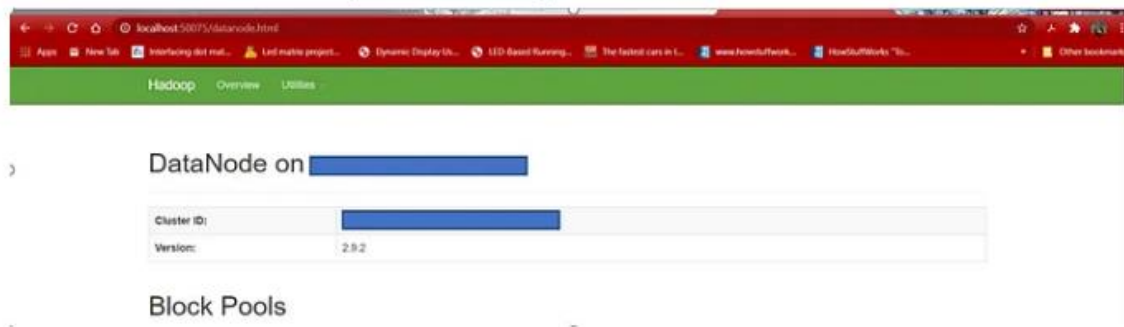


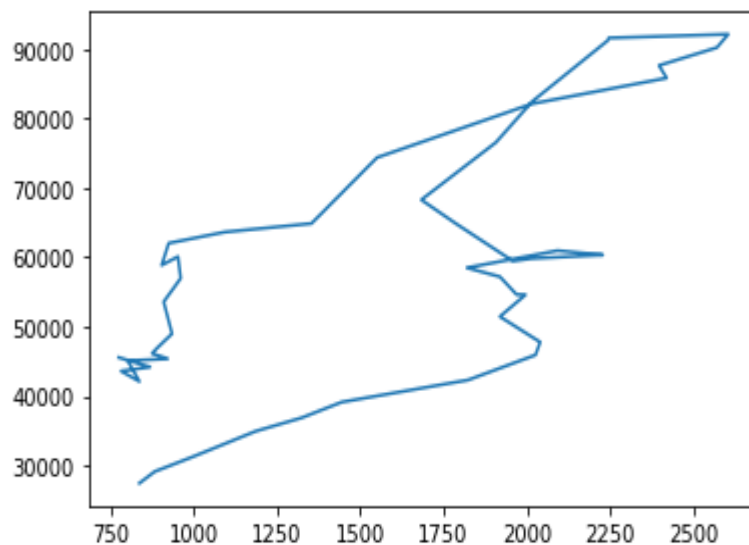
Fig. 20:- Datanode Web UI

EXPERIMENT: 3**Visualize Data Using Basic Plotting Techniques In Python.****PROGRAM:****AIM:** To create an application that takes the Visualize Data Using Basic Plotting Techniques.

```
import pandas as pb
import matplotlib.pyplot as plt
import seaborn as sns
crime=pb.read_csv('crime.csv')
crime
```

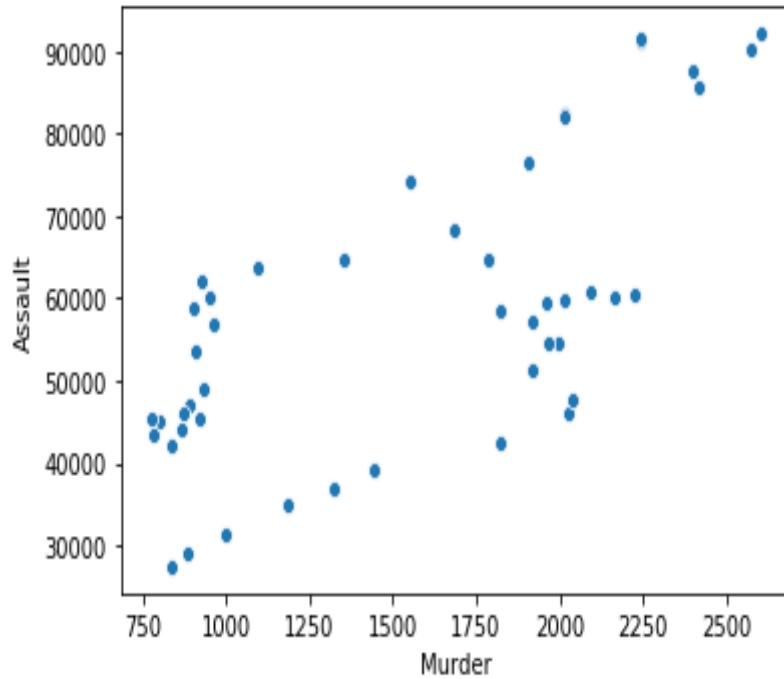
	Year	Population	Murder	Rape	Robbery	Assault	Burglary	CarTheft
0	1965	18073000	836	2320	28182	27464	183443	58452
1	1966	18258000	882	2439	30098	29142	196127	64368
2	1967	18336000	996	2665	40202	31261	219157	83775
3	1968	18113000	1185	2527	59857	34946	250918	104877
4	1969	18321000	1324	2902	64754	36890	248477	115400
5	1970	18190740	1444	2875	81149	39145	267474	125674
6	1971	18391000	1823	3225	97682	42318	273704	127658
7	1972	18366000	2026	4199	86391	45926	239886	105081
8	1973	18265000	2040	4852	80795	47781	246246	112328
9	1974	18111000	1919	5240	86814	51454	271824	104095
10	1975	18120000	1996	5099	93499	54593	301996	116274
11	1976	18084000	1969	4663	95718	54638	318919	133504
12	1977	17924000	1919	5272	84703	57193	309735	133669
13	1978	17748000	1820	5168	83785	58484	292956	119264
14	1979	17649000	2092	5394	93471	60949	308302	124343
15	1980	17506690	2228	5405	112273	60329	360925	133041
16	1981	17594000	2166	5479	120344	60189	350422	136849
17	1982	17659000	2013	5159	107843	59818	295245	137880
18	1983	17667000	1958	5296	94783	59452	249115	127861
19	1984	17735000	1786	5599	89900	64872	222956	115392
20	1985	17783000	1683	5706	89706	68270	219633	106537

```
plt.plot(crime.Murder,crime.Assault);
```

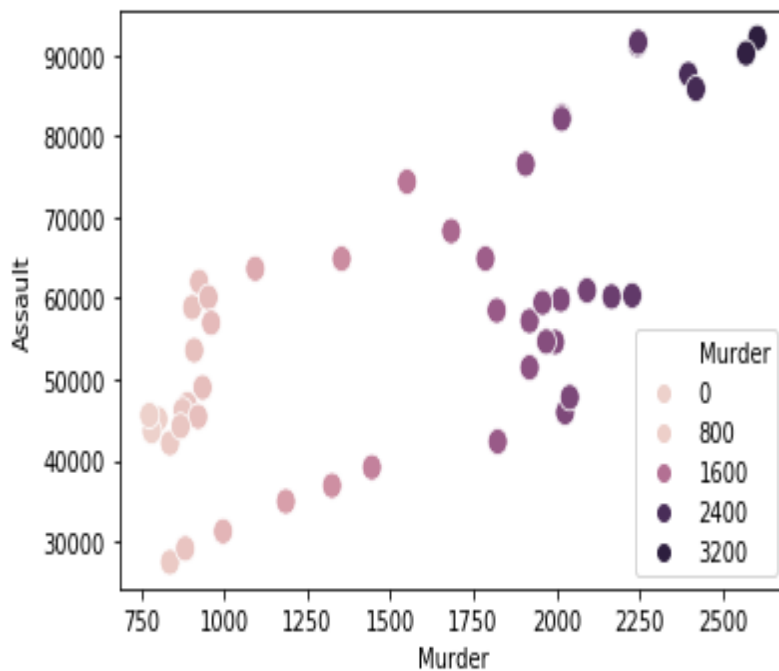


```
import seaborn as sns
```

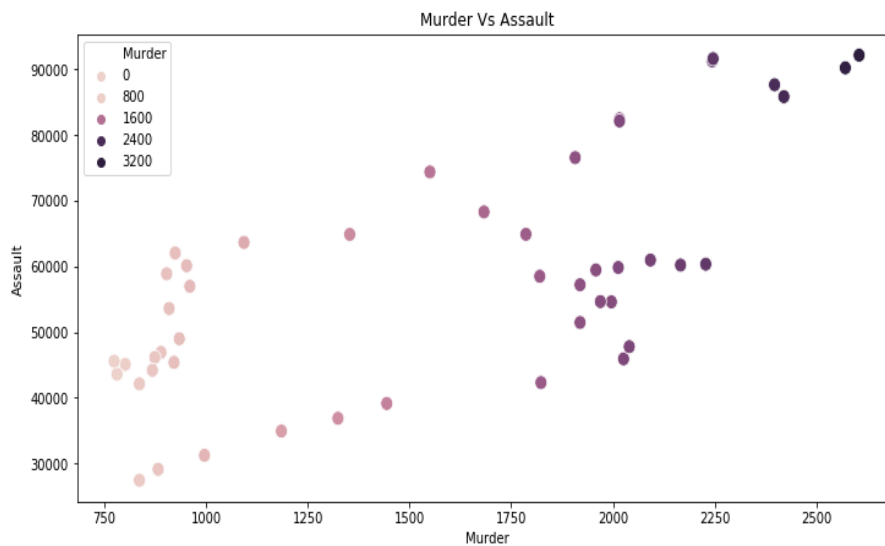
```
sns.scatterplot(crime.Murder,crime.Assault);
```



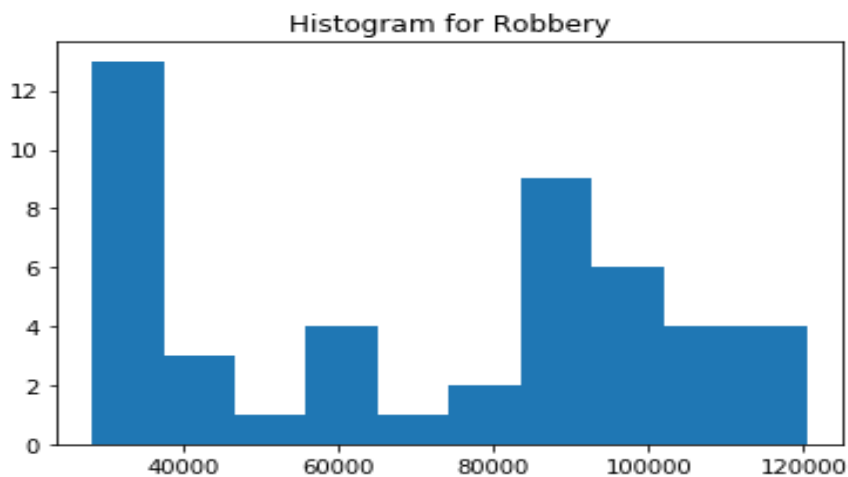
```
sns.scatterplot(crime.Murder,crime.Assault,hue=crime.Murder,s=100);
```



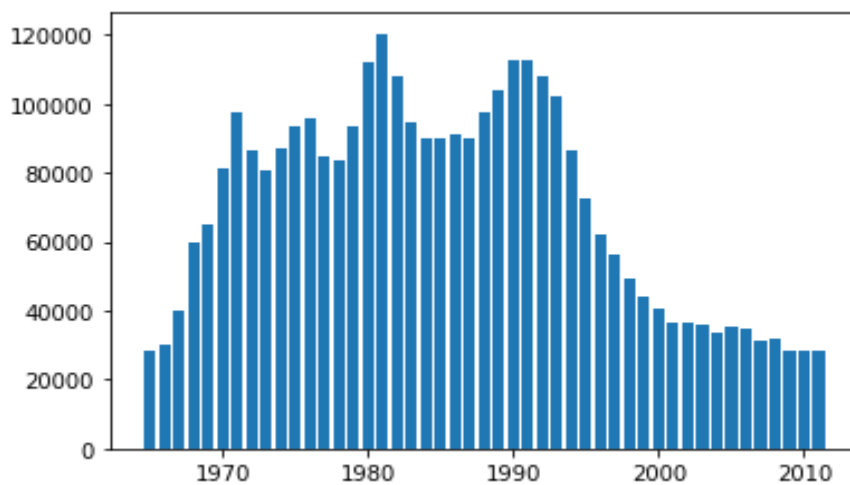
```
plt.figure(figsize=(12,6))  
plt.title('Murder Vs Assault')  
sns.scatterplot(crime.Murder,crime.Assault,hue=crime.Murder,s=100);
```



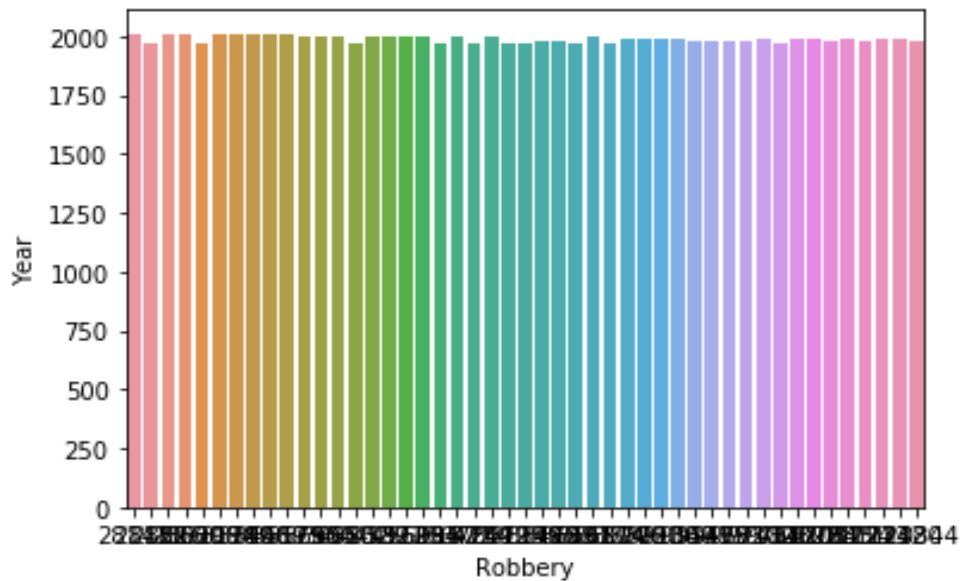
```
plt.title('Histogram for Robbery')
plt.hist(crime.Robbery);
```



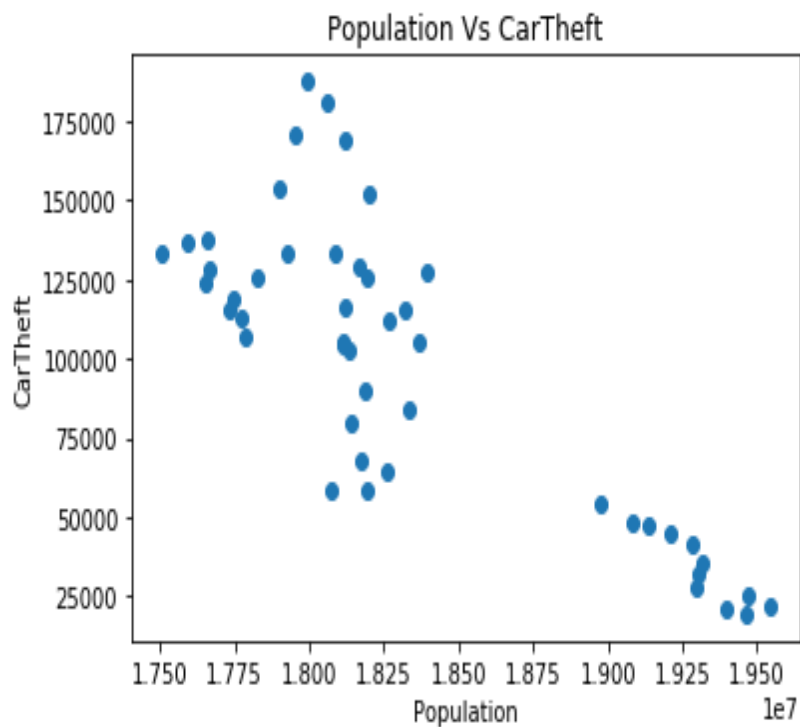
```
plt.bar(crime_bar.index,crime_bar.Robbery);
```



```
sns.barplot('Robbery','Year',data=crime);
```



```
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
data=pd.read_csv('crime.csv')
x=data.Population
y=data.CarTheft
plt.scatter(x,y)
plt.xlabel('Population')
plt.ylabel('CarTheft')
plt.title('Population Vs CarTheft')
plt.show();
```



EXPERIMENT: 4**Implement no sql Database Operations: Crud Operations, Arrays Using MONGODB.****PROGRAM:****AIM:** To Create a operations for crud and arrays without no sql database.**TITLE: Basic CRUD operations in MongoDB.**

CRUD operations refer to the basic Insert, Read, Update and Delete operations.

Inserting a document into a collection (Create)

➤ The command `db.collection.insert()` will perform an insert operation into a collection of a document. ➤ Let us insert a document to a student collection. You must be connected to a database for doing any insert. It is done as follows:

```
db.student.insert({
  regNo: "3014",
  name: "Test Student",
  course: { courseName: "MCA", duration: "3 Years" },
  address: {
    city: "Bangalore",
    state: "KA",
    country: "India" } })
```

An entry has been made into the collection called student.

```
db.student.insert(
{
  regNo: "3014",
  name: "Test Student",
  course: {
    courseName: "MCA",
    duration: "3 Years"
  },
  address: {
    city: "Bangalore",
    state: "KA",
    country: "India"
  }
})

C:\windows\system32\cmd.exe - mongo
> db.student.insert(
...
{
  regNo: "3014",
  name: "Test Student",
  course: {
    courseName: "MCA",
    duration: "3 Years"
  },
  address: {
    city: "Bangalore",
    state: "KA",
    country: "India"
  }
}
...
)
WriteResult({ "nInserted" : 1 })
>
```

Querying a document from a collection (Read)

To retrieve (Select) the inserted document, run the below command. The `find()` command will retrieve all the documents of the given collection.`db.collection_name.find()`

➤ If a record is to be retrieved based on some criteria, the `find()` method should be called passing parameters, then the record will be retrieved based on the attributes specified.

`db.collection_name.find({"fieldname": "value"})`

➤ For Example: Let us retrieve the record from the student collection where the attribute `regNo` is 3014 and the query for the same is as shown below:

`db.students.find({"regNo": "3014"})`

```
C:\windows\system32\cmd.exe - mongo
> db.student.find()
{ "_id" : ObjectId("5780db48e30f7f8faae88a83"), "regNo" : "3014", "name" : "Test Student", "course" : { "courseName" : "MCA", "duration" : "3 Years" }, "address" : { "city" : "Bangalore", "state" : "KA", "country" : "India" } }
>
```


Updating a document in a collection (Update) In order to update specific field values of a collection in MongoDB, run the below query. `db.collection_name.update()`

➤ `update()` method specified above will take the fieldname and the new value as argument to update a document.

➤ Let us update the attribute name of the collection student for the document with regNo 3014.

```
db.student.update({
```

```
  "regNo": "3014"
```

```
},
```

```
$set:
```

```
{
```

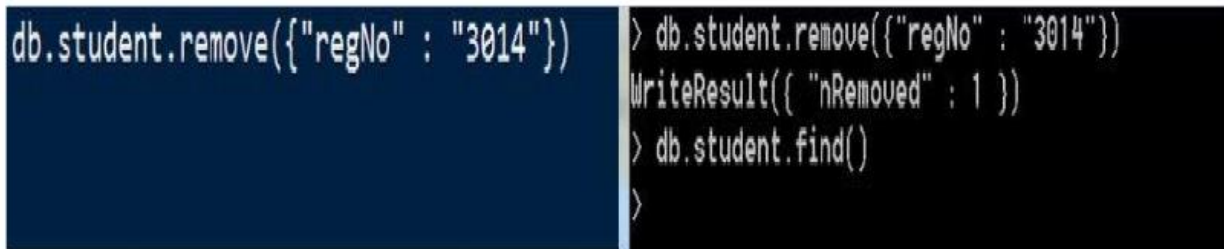
```
  "name": "Viraj"
```

```
})
```

Removing an entry from the collection (Delete)

➤ Let us now look into the deleting an entry from a collection. In order to delete an entry from a collection, run the command as shown below : `db.collection_name.remove({"fieldname":"value"})`

➤ For Example : `db.student.remove({"regNo":"3014"})`



```
db.student.remove({"regNo" : "3014"})
```

```
> db.student.remove({"regNo" : "3014"})
WriteResult({ "nRemoved" : 1 })
> db.student.find()
>
```

Note that after running the `remove()` method, the entry has been deleted from the student collection.

Working with Arrays in MongoDB

1. Introduction

In a MongoDB database, data is stored in collections and a collection has documents. A document has fields and values, like in a JSON. The field types include scalar types (string, number, date, etc.) and composite types (arrays and objects). In this article we will look at an example of using the array field type.

The example is an application where users create blog posts and write comments for the posts. The relationship between the posts and comments is One-to-Many; i.e., a post can have many comments. We will consider a collection of blog posts with their comments. That is a post document will also store the related comments. In MongoDB's document model, a 1:N relationship data can be stored within a collection; this is a de-normalized form of data. The related data is stored together and can be accessed (and updated) together. The comments are stored as an array; an array of comment objects.

A sample document of the blog posts with comments:

```
{
  "_id" : ObjectId("5ec55af811ac5e2e2aafb2b9"),
  "name" : "Working with Arrays",
  "user" : "Database Rebel",
  "desc" : "Maintaining an array of objects in a document",
  "content" : "some content ...",
  "created" : ISODate("2020-05-20T16:28:55.468Z"),
  "updated" : ISODate("2020-05-20T16:28:55.468Z"),
  "tags" : [ "mongodb", "arrays" ],
  "comments" : [
    {
      "user" : "DB Learner",
```

```
"content" : "Nice post.",
"updated" : ISODate("2020-05-20T16:35:57.461Z")
}
]
```

In an application, a blog post is created, comments are added, queried, modified or deleted by users. In the example, we will write code to create a blog post document, and do some CRUD operations with comments for the post.

2. Create and Query a Document

Let's create a blog post document. We will use a database called as blogs and a collection called as posts. The code is written in mongoshell (an interactive JavaScript interface to MongoDB). Mongo shell is started from the command line and is connected to the MongoDB server. From the shell:

use blogs

NEW_POST =

```
{
  name: "Working with Arrays",
  user: "Database Rebel",
  desc: "Maintaining an array of objects in a document",
  content: "some content...",
  created: ISODate(),
  updated: ISODate(),
  tags: [ "mongodb", "arrays" ]
}
```

db.posts.insertOne(NEW_POST)

Returns a result { "acknowledged" : true, "insertedId" : ObjectId("5ec55af811ac5e2e2aafb2b9") } indicating that a new document is created. This is a common acknowledgement when you perform a write operation. When a document is inserted into a collection for the first time, the collection gets created (if it doesn't exist already). The insertOne method inserts a document into the collection.

Now, let's query the collection :

db.posts.findOne()

```
{
  "_id" : ObjectId("5ec55af811ac5e2e2aafb2b9"),
  "name" : "Working with Arrays",
  "user" : "Database Rebel",
  "desc" : "Maintaining an array of objects in a document",
  "content" : "some content...",
  "created" : ISODate("2020-05-20T16:28:55.468Z"),
  "updated" : ISODate("2020-05-20T16:28:55.468Z"),
  "tags" : [
    "mongodb",
    "arrays"
  ]
}
```

The findOne method retrieves one matching document from the collection. Note the scalar fields name (string type) and created (date type), and the array field tags. In the newly inserted document there are no comments, yet.

3. Add an Array Element

Let's add a comment for this post, by a user "DB Learner":

NEW_COMMENT = {

```

user: "DB Learner",
text: "Nice post, can I know more about the arrays in MongoDB?",
updated: ISODate()
}
db.posts.updateOne(
  { _id : ObjectId("5ec55af811ac5e2e2aafb2b9") },
  { $push: { comments: NEW_COMMENT } }
)

```

Returns: { "acknowledged" : true, "matchedCount" : 1, "modifiedCount" : 1 }

The updateOne method updates a document's fields based upon the specified condition. \$push is an array update operator which adds an element to an array. If the array doesn't exist, it creates an array field and then adds the element.

Let's query the collection and confirm the new comment visually, using the findOne method:

```

{
  "_id" : ObjectId("5ec55af811ac5e2e2aafb2b9"),
  "name" : "Working with Arrays",
  ...
  "comments" : [
    {
      "user" : "DB Learner",
      "text" : "Nice post, can I know more about the arrays in MongoDB?",
      "updated" : ISODate("2020-05-20T16:35:57.461Z")
    }
  ]
}

```

Note the comments array field has comment objects as elements. Let's add one more comment using the same \$push update operator. This new comment (by user "Database Rebel") is appended to the comments array:

```

"comments" : [
  {
    "user" : "DB Learner",
    "text" : "Nice post, can I know more about the arrays in MongoDB?",
    "updated" : ISODate("2020-05-20T16:35:57.461Z")
  },
  {
    "user" : "Database Rebel",
    "text" : "Thank you, please look for updates",
    "updated" : ISODate("2020-05-20T16:48:25.506Z")
  }
]

```

4. Update an Array Element

Let's update the comment posted by "Database Rebel" with modified text field :

NEW_CONTENT = "Thank you, please look for updates - updated the post".

```

db.posts.updateOne(
  { _id : ObjectId("5ec55af811ac5e2e2aafb2b9"), "comments.user": "Database Rebel" },
  { $set: { "comments.$.text": NEW_CONTENT } }
)

```

The \$set update operator is used to change a field's value. The positional \$ operator identifies an element in an array to update without explicitly specifying the position of the element in the array. The first matching element is updated. The updated comment object:

```
"comments" : [
```

```
{
  "user" : "Database Rebel",
  "text" : "Thank you, please look for updates - updated",
  "updated" : ISODate("2020-05-20T16:48:25.506Z")
}
```

5. Delete an Array Element

The user changed his mind and wanted to delete the comment, and then add a new one.

```
db.posts.updateOne(
  { _id : ObjectId("5ec55af811ac5e2e2aafb2b9") },
  { $pull: { comments: { user: "Database Rebel" } } }
)
```

The \$pull update operator removes elements from an array which match the specified condition - in this case { comments: { user: "Database Rebel" } }.

A new comment is added to the array after the above delete operation, with the following text:

"Thank you for your comment. I have updated the post with CRUD operations on an array field".

6. Add a New Field to all Objects in the Array

Let's add a new field likes for all the comments in the array.

```
db.posts.updateOne(
  { "_id" : ObjectId("5ec55af811ac5e2e2aafb2b9") },
  { $set: { "comments.$[].likes": 0 } }
)
```

The all positional operator \$[] specifies that the update operator \$set should modify all elements in the specified array field. After the update, all comment objects have the likes field, for example:

```
{
  "user" : "DB Learner",
  "text" : "Nice post, can I know more about the arrays in MongoDB?",
  "updated" : ISODate("2020-05-20T16:35:57.461Z"),
  "likes" : 0
}
```

7. Update a Specific Array Element Based on a Condition

First, let's add another new comment using the \$push update operator:

```
NEW_COMMENT = {
  user: "DB Learner",
  text: "Thanks for the updates!",
  updated: ISODate()
}
```

Note the likes field is missing in the input document. We will update this particular comment in the comments array with the condition that the likes field is missing.

```
db.posts.updateOne(
  { "_id" : ObjectId("5ec55af811ac5e2e2aafb2b9") },
  { $inc: { "comments.$[ele].likes": 1 } },
  { arrayFilters: [ { "ele.user": "DB Learner", "ele.likes": { $exists: false } } ] }
)
```

The likes field is updated using the \$inc update operator (this increments a field's value, or if not exists adds the field and then increments). The filtered positional operator \$[<identifier>] identifies the array elements that match the arrayFilters conditions for an update operation.

EXPERIMENT: 5

Implement Functions: Count – Sort – Limit – Skip – Aggregate Using MONGODB.

PROGRAM:

AIM: To create function operations for sort, limit, skip and aggregate.

1. COUNT

How do you get the number of Debit and Credit transactions? One way to do it is by using count() function as below

```
> db.transactions.count({cr_dr : "D"});
```

or

```
> db.transactions.find({cr_dr : "D"}).length();
```

But what if you do not know the possible values of cr_dr upfront. Here Aggregation framework comes to play. See the below Aggregate query.

```
> db.transactions.aggregate([
  {
    $group : {
      _id : '$cr_dr', // group by type of transaction
      // Add 1 for each document to the count for this type of
transaction
      count : {$sum : 1}
    }
  }
]);
```

And the result is

```
{
  "_id" : "C",
  "count" : 3
}
{
  "_id" : "D",
  "count" : 5
}
```

2. SORT

Definition

\$sort

Sorts all input documents and returns them to the pipeline in sorted order.

The

\$sort

stage has the following prototype form:

```
{ $sort: { <field1>: <sort order>, <field2>: <sort order> ... } }
```

\$sort

takes a document that specifies the field(s) to sort by and the respective sort order. <sort order> can have one of the following values:

Value

Description

1

Sort ascending.

-1

Sort descending.

```
{ $meta: "textScore" }
```

Sort by the computed textScore metadata in descending order. See

Text Score Metadata Sort

for an example.

If sorting on multiple fields, sort order is evaluated from left to right. For example, in the form above, documents are first sorted by <field1>. Then documents with the same <field1> values are further sorted by <field2>.

Behavior

Limits

You can sort on a maximum of 32 keys.

Sort Consistency

MongoDB does not store documents in a collection in a particular order. When sorting on a field which contains duplicate values, documents containing those values may be returned in any order.

If consistent sort order is desired, include at least one field in your sort that contains unique values. The easiest way to guarantee this is to include the `_id` field in your sort query.

Consider the following restaurant collection:

```
db.restaurants.insertMany( [
  { "_id" : 1, "name" : "Central Park Cafe", "borough" : "Manhattan"},
  { "_id" : 2, "name" : "Rock A Feller Bar and Grill", "borough" :
"Queens"},
  { "_id" : 3, "name" : "Empire State Pub", "borough" : "Brooklyn"},
  { "_id" : 4, "name" : "Stan's Pizzeria", "borough" : "Manhattan"},
  { "_id" : 5, "name" : "Jane's Deli", "borough" : "Brooklyn"},
] )
```

The following command uses the
\$sort
stage to sort on the borough field:

```
db.restaurants.aggregate(  
  [  
    { $sort : { borough : 1 } }  
  ]  
)
```

In this example, sort order may be inconsistent, since the borough field contains duplicate values for both Manhattan and Brooklyn. Documents are returned in alphabetical order by borough, but the order of those documents with duplicate values for borough might not be the same across multiple executions of the same sort. For example, here are the results from two different executions of the above command:

```
{ "_id" : 3, "name" : "Empire State Pub", "borough" : "Brooklyn" }  
{ "_id" : 5, "name" : "Jane's Deli", "borough" : "Brooklyn" }  
{ "_id" : 1, "name" : "Central Park Cafe", "borough" : "Manhattan" }  
{ "_id" : 4, "name" : "Stan's Pizzeria", "borough" : "Manhattan" }  
{ "_id" : 2, "name" : "Rock A Feller Bar and Grill", "borough" : "Queens"  
}  
{ "_id" : 5, "name" : "Jane's Deli", "borough" : "Brooklyn" }  
{ "_id" : 3, "name" : "Empire State Pub", "borough" : "Brooklyn" }  
{ "_id" : 4, "name" : "Stan's Pizzeria", "borough" : "Manhattan" }  
{ "_id" : 1, "name" : "Central Park Cafe", "borough" : "Manhattan" }  
{ "_id" : 2, "name" : "Rock A Feller Bar and Grill", "borough" : "Queens"  
}
```

While the values for borough are still sorted in alphabetical order, the order of the documents containing duplicate values for borough (i.e. Manhattan and Brooklyn) is not the same.

To achieve a consistent sort, add a field which contains exclusively unique values to the sort. The following command uses the
\$sort

stage to sort on both the borough field and the _id field:

```
db.restaurants.aggregate(  
  [  
    { $sort : { borough : 1, _id: 1 } }  
  ]  
)
```

Since the `_id` field is always guaranteed to contain exclusively unique values, the returned sort order will always be the same across multiple executions of the same sort.

Examples

Ascending/Descending Sort

For the field or fields to sort by, set the sort order to 1 or -1 to specify an ascending or descending sort respectively, as in the following example:

```
db.users.aggregate(
  [
    { $sort : { age : -1, posts: 1 } }
  ]
)
```

This operation sorts the documents in the `users` collection, in descending order according by the `age` field and then in ascending order according to the value in the `posts` field.

3. LIMIT

\$sort

Sorts all input documents and returns them to the pipeline in sorted order.

The `$sort` stage has the following prototype form:

```
{ $sort: { <field1>: <sort order>, <field2>: <sort order> ... } }
```

`$sort` takes a document that specifies the field(s) to sort by and the respective sort order. `<sort order>` can have one of the following values:

Value	Description
1	Sort ascending.
-1	Sort descending.
{ \$meta: "textScore"	Sort by the computed textScore metadata in descending order. See Text Score Metadata Sort for an example.

If sorting on multiple fields, sort order is evaluated from left to right. For example, in the form above, documents are first sorted by `<field1>`. Then documents with the same `<field1>` values are further sorted by `<field2>`.

Behavior

Limits

You can sort on a maximum of 32 keys.

Sort Consistency

MongoDB does not store documents in a collection in a particular order. When sorting on a field which contains duplicate values, documents containing those values may be returned in any order.

If consistent sort order is desired, include at least one field in your sort that contains unique values. The easiest way to guarantee this is to include the `_id` field in your sort query.

Consider the following restaurant collection:

```
db.restaurants.insertMany( [
  { "_id" : 1, "name" : "Central Park Cafe", "borough" : "Manhattan" },
  { "_id" : 2, "name" : "Rock A Feller Bar and Grill", "borough" : "Queens" },
  { "_id" : 3, "name" : "Empire State Pub", "borough" : "Brooklyn" },
  { "_id" : 4, "name" : "Stan's Pizzeria", "borough" : "Manhattan" },
  { "_id" : 5, "name" : "Jane's Deli", "borough" : "Brooklyn" },
])
```

The following command uses the `$sort` stage to sort on the `borough` field:

```
db.restaurants.aggregate(
[
  { $sort : { borough : 1 } }
]
)
```

In this example, sort order may be inconsistent, since the `borough` field contains duplicate values for both Manhattan and Brooklyn. Documents are returned in alphabetical order by borough, but the order of those documents with duplicate values for borough might not be the same across multiple executions of the same sort. For example, here are the results from two different executions of the above command:

```
{ "_id" : 3, "name" : "Empire State Pub", "borough" : "Brooklyn" }
{ "_id" : 5, "name" : "Jane's Deli", "borough" : "Brooklyn" }
{ "_id" : 1, "name" : "Central Park Cafe", "borough" : "Manhattan" }
{ "_id" : 4, "name" : "Stan's Pizzeria", "borough" : "Manhattan" }
{ "_id" : 2, "name" : "Rock A Feller Bar and Grill", "borough" : "Queens" }

{ "_id" : 5, "name" : "Jane's Deli", "borough" : "Brooklyn" }
{ "_id" : 3, "name" : "Empire State Pub", "borough" : "Brooklyn" }
{ "_id" : 4, "name" : "Stan's Pizzeria", "borough" : "Manhattan" }
{ "_id" : 1, "name" : "Central Park Cafe", "borough" : "Manhattan" }
{ "_id" : 2, "name" : "Rock A Feller Bar and Grill", "borough" : "Queens" }
```

While the values for `borough` are still sorted in alphabetical order, the order of the documents containing duplicate values for `borough` (i.e. Manhattan and Brooklyn) is not the same.

To achieve a *consistent sort*, add a field which contains exclusively unique values to the sort. The following command uses the `$sort` stage to sort on both the `borough` field and the `_id` field:

```
db.restaurants.aggregate(
[
  { $sort : { borough : 1, _id: 1 } }
]
)
```

Since the `_id` field is always guaranteed to contain exclusively unique values, the returned sort order will always be the same across multiple executions of the same sort.

Examples

Ascending/Descending Sort

For the field or fields to sort by, set the sort order to 1 or -1 to specify an ascending or descending sort respectively, as in the following example:

```
db.users.aggregate(  
  [  
    { $sort : { age : -1, posts: 1 } }  
  ]  
)
```

4. SKIP

`$skip`

Skips over the specified number of documents that pass into the stage and passes the remaining documents to the next stage in the pipeline.

The

`$skip`

stage has the following prototype form:

```
{ $skip: <positive 64-bit integer> }
```

EXPERIMENT: 6**Implement Word Count/ Frequency Programs Using Map Reduce.****PROGRAM:**

AIM: To count a given number using map reduce functions.

Hadoop Streaming API for helping us passing data between our Map and Reduce code
from collections import defaultdict

```
# Sample input data
data = [
    "Hello world",
    "MapReduce is a programming model",
    "Hello MapReduce",
    "MapReduce example"
]

# Initialize a dictionary to hold word counts
word_counts = defaultdict(int)

# Map Function
def map_function(data):
    for line in data:
        words = line.split()
        for word in words:
            yield (word.lower(), 1)

# Reduce Function
def reduce_function(mapped_data):
    for word, count in mapped_data:
        word_counts[word] += count

# Map phase
mapped_data = list(map_function(data))

# Shuffle and Sort (usually handled by the MapReduce framework)

# Reduce phase
reduce_function(mapped_data)

# Output the word counts
for word, count in word_counts.items():
    print(f"{word}: {count}")
```

OUTPUT:

EXPERIMENT: 7

Implement a MapReduce Program that process a dataset.

AIM:

To create process dataset using map reduce functions.

PROGRAM:

```
import csv
import pickle

# Read data from data.csv and perform mapping
mapped_data = []

with open('crime.csv', 'r') as csvfile:
    reader = csv.reader(csvfile)
    for row in reader:
        for word in row: # Split each line into words
            mapped_data.append((word, 1)) # Emit (word, 1) for each word

# Store the mapped data in shuffled.pkl
with open('shuffled.pkl', 'wb') as output_file:
    pickle.dump(mapped_data, output_file)

import pickle
from collections import defaultdict

# Read the mapped data from shuffled.pkl
with open('shuffled.pkl', 'rb') as input_file:
    mapped_data = pickle.load(input_file)

# Perform reducing (word count) operation
word_counts = defaultdict(int)
for word, count in mapped_data:
    word_counts[word] += count

# Print the word counts
for word, count in word_counts.items():
    print(f"{word}: {count}")
```

EXPERIMENT: 8**Implement Clustering Techniques using SPARK.**

AIM: To create a clustering using SPARK.

PROGRAM:

```
from pyspark.sql import SparkSession
from pyspark.ml.clustering import KMeans
from pyspark.ml.feature import VectorAssembler
from pyspark.ml.evaluation import ClusteringEvaluator

# Start Spark session
spark = SparkSession.builder.appName("IrisKMeans").getOrCreate()

# Load the dataset
data = spark.read.csv("iris.csv", header=True, inferSchema=True)

# Print schema to confirm structure
data.printSchema()
data.show(5)

# Prepare features by excluding the label column
feature_cols = data.columns
if "label" in feature_cols:
    feature_cols.remove("label")
elif "species" in feature_cols: # Common case for Iris
    feature_cols.remove("species")

# Assemble features
assembler = VectorAssembler(inputCols=feature_cols, outputCol="features")
assembled_data = assembler.transform(data)

# Train KMeans
kmeans = KMeans(k=3, seed=1)
model = kmeans.fit(assembled_data)

# Predictions
predictions = model.transform(assembled_data)
predictions.select("prediction", "features").show(5)

# Evaluate with Silhouette Score
evaluator = ClusteringEvaluator()
silhouette = evaluator.evaluate(predictions)
print(f"Silhouette Score = {silhouette}")

# Optional: Show cluster centers
centers = model.clusterCenters()
print("Cluster Centers:")
for i, center in enumerate(centers):
    print(f"Cluster {i}: {center}")
```

EXPERIMENT : 9

Implement an application that stores Big Data in MONGODB / PIG Using Hadoop / R.

AIM: To design a application to stores data in mongdob using hadoop.

PROGRAM:

```
pip install pymongo
```

```
import pymongo
```

```
# Connect to MongoDB
```

```
client = pymongo.MongoClient("mongodb://localhost:27017/")
```

```
db = client["bigdata_db"]
```

```
collection = db["bigdata_collection"]
```

```
# Simulated big data
```

```
big_data = [{"_id": i, "data": f"Data point {i}"} for i in range(1, 1000000)]
```

```
# Insert big data into MongoDB
```

```
collection.insert_many(big_data)
```

```
# Query data from MongoDB
```

```
result = collection.find_one({"_id": 1})
```

```
print("Retrieved data from MongoDB:")
```

```
print(result)
```

```
# Close MongoDB connection
```

```
client.close()
```

Signature of the Faculty