

# Machine Learning

## Lab 2

S Shyam Sundaram

19BCE1560

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### EXERCISE 1: NUMPY

#### CREATING ARRAYS

##### **Code**

```
import numpy as np
arr1 = np.array([1,2,3,4,5,6])
print("1D array: ", arr1)
arr2 = np.array([[1,2,3],[4,5,6],[7,8,9]])
print("2D array: ", "\n",arr2)
# Creating a rank 1 Array
arr = np.array([1, 2, 3])
print("Array with Rank 1: \n",arr)
# Creating a rank 2 Array
arr = np.array([[1, 2, 3],
                [4, 5, 6]])
print("Array with Rank 2: \n", arr)
# Creating an array from tuple
arr = np.array((1, 3, 2))
print("\nArray created using "
      "passed tuple:\n", arr)
#Using 'arange'
arr8 = np.arange(1,10,3)
print("Array created with arange: \t\n",arr8)
#Using linspace
arr9 = np.linspace(1,10,50)
print("Array created using linspace requesting 50 elements within 1 to 10:\t\n",arr9)
```

##### **Output**

```
In [1]: import numpy as np

In [2]: arr1 = np.array([1,2,3,4,5,6])
print("1D array: ", arr1)
arr2 = np.array([[1,2,3],[4,5,6],[7,8,9]])
print("2D array: ", "\n",arr2)

1D array:  [1 2 3 4 5 6]
2D array:
[[1 2 3]
 [4 5 6]
 [7 8 9]]
```

```
In [3]: # Creating a rank 1 Array
arr = np.array([1, 2, 3])
print("Array with Rank 1: \n",arr)
```

```
Array with Rank 1:
[1 2 3]
```

```
In [4]: # Creating a rank 2 Array
arr = np.array([[1, 2, 3],
                [4, 5, 6]])
print("Array with Rank 2: \n", arr)
```

```
Array with Rank 2:
[[1 2 3]
 [4 5 6]]
```

```
In [5]: # Creating an array from tuple
arr = np.array((1, 3, 2))
print("\nArray created using "
      "passed tuple:\n", arr)
```

```
Array created using passed tuple:
[1 3 2]
```

## UNDERSTANDING ATTRIBUTES OF ARRAYS

### Code

```
print("About arr2:")
print("Type\t\t:",type(arr2))
print("Datatype\t:",arr2.dtype)
print("Shape\t\t:", arr2.shape)
print("Size\t\t:",arr2.size)
print("itemsize\t:",arr2.itemsize)
print("No. of dim\t:",arr2.ndim)
print("No. of bytes\t:", arr2.nbytes)
```

### Output

```
In [8]: print("About arr2:")
print("Type\t\t:",type(arr2))
print("Datatype\t:",arr2.dtype)
print("Shape\t\t:", arr2.shape)
print("Size\t\t:",arr2.size)
print("itemsize\t:",arr2.itemsize)
print("No. of dim\t:",arr2.ndim)
print("No. of bytes\t:", arr2.nbytes)
```

```
About arr2:
Type          : <class 'numpy.ndarray'>
Datatype      : int32
Shape         : (3, 3)
Size          : 9
itemsize      : 4
No. of dim    : 2
No. of bytes  : 36
```

## SPECIAL ARRAYS

### Code

```
arr3 = np.zeros((5,2), dtype=int)
arr4 = np.ones((3,4),dtype=float)
arr5 = np.eye(4,3)
arr6 = np.random.rand(3,2)
arr7 = np.random.randint(7,size=(2,6))
print("Zero Array \t:\n",arr3)
print("Arrays with unit values\t:\n",arr4)
print("Identity matrix\t:\n",arr5)
print("Random array\t\n",arr6)
print("Random integer array\t:\n" , arr7)
```

### Output

```
In [9]: arr3 = np.zeros((5,2), dtype=int)
arr4 = np.ones((3,4),dtype=float)
arr5 = np.eye(4,3)
arr6 = np.random.rand(3,2)
arr7 = np.random.randint(7,size=(2,6))
print("Zero Array \t:\n",arr3)
print("Arrays with unit values\t:\n",arr4)
print("Identity matrix\t:\n",arr5)
print("Random array\t\n",arr6)
print("Random integer array\t:\n" , arr7)

Zero Array      :
[[0 0]
 [0 0]
 [0 0]
 [0 0]
 [0 0]]
Arrays with unit values :
[[1. 1. 1. 1.]
 [1. 1. 1. 1.]
 [1. 1. 1. 1.]]
Identity matrix :
[[1. 0. 0.]
 [0. 1. 0.]
 [0. 0. 1.]
 [0. 0. 0.]]
Random array
[[0.78050112 0.94526287]
 [0.90809064 0.48067856]
 [0.77292278 0.71222528]]
Random integer array :
[[5 3 6 1 2 1]
 [3 4 3 4 3 0]]
```

## SLICING

### Code

```
arr10 = np.array([[1,2,3,4,5],[6,7,8,9,10],[11,12,13,14,15],[16,17,18,19,20]])
print(arr10)
print("Row 2:",arr10[1,:])
print("Column 2:",arr10[:,3])
print("Elements 7,8,12,13 :\n", arr10[1:3,1:3])
print("All elements: \n",arr10[:,:])
print("Strides:\n",arr10[0::2,1:3])
```

## Output

```
In [12]: arr10 = np.array([[1,2,3,4,5],[6,7,8,9,10],[11,12,13,14,15],[16,17,18,19,20]])
print(arr10)
print("Row 2:",arr10[1,:])
print("Column 2:",arr10[:,3])
print("Elements 7,8,12,13 :\n", arr10[1:3,1:3])
print("All elements: \n",arr10[:])
print("Strides:\n",arr10[0::2,1:3])

[[ 1  2  3  4  5]
 [ 6  7  8  9 10]
 [11 12 13 14 15]
 [16 17 18 19 20]]
Row 2: [ 6  7  8  9 10]
Column 2: [ 4  9 14 19]
Elements 7,8,12,13 :
[[ 7  8]
 [12 13]]
All elements:
[[ 1  2  3  4  5]
 [ 6  7  8  9 10]
 [11 12 13 14 15]
 [16 17 18 19 20]]
Strides:
[[ 2  3]
 [12 13]]
```

## MASKING

### Code

```
arr11 = np.array([1,2,3,4,5,6,7,8,9])
mask = np.array([0,1,1,0,1,0,1,0,0],dtype=bool)
print(arr11[mask])
```

### Output

```
In [13]: arr11 = np.array([1,2,3,4,5,6,7,8,9])
mask = np.array([0,1,1,0,1,0,1,0,0],dtype=bool)
print(arr11[mask])

[2 3 5 7]
```

## SCALAR OPERATIONS

### Code

```
arr13 = np.array([1,2,3])
arr14 = np.array([4,5,6])
arr15 = arr13+arr14
arr16 = arr13 - arr14
print("Summation:\t",arr15)
print("Difference:\t",arr16)
arr16+=5
print("Previous output after adding 5:",arr16)
```

## Output

```
In [14]: arr13 = np.array([1,2,3])
arr14 = np.array([4,5,6])
arr15 = arr13+arr14
arr16 = arr13 - arr14
print("Summation:\t",arr15)
print("Difference:\t",arr16)

Summation:      [5 7 9]
Difference:     [-3 -3 -3]

In [15]: arr16+=5
print("Previous output after adding 5:",arr16)

Previous output after adding 5: [2 2 2]
```

## TRIGONOMETRIC OPERATIONS

### Code

```
arr17 = np.array([15,30,45,90])
result7 = np.sin(arr17)
print("Sin values:\t",result7)
result8 = np.log(arr17)
print("Log value:\t",result8)
```

### Output

```
In [16]: arr17 = np.array([15,30,45,90])
result7 = np.sin(arr17)
print("Sin values:\t",result7)
result8 = np.log(arr17)
print("Log value:\t",result8)

Sin values:      [ 0.65028784 -0.98803162  0.85090352  0.89399666]
Log value:       [2.7080502  3.40119738 3.80666249 4.49980967]
```

## EXERCISE 2: PANDAS

### CREATING DATAFRAMES FROM LIST

#### Code

```
import pandas as pd
# list of strings
lst = ['Soon', 'For', 'Good', 'is',
       'portal', 'for', 'Good']
# Calling DataFrame constructor on list
df = pd.DataFrame(lst)
print(df)
```

#### Output

```
In [18]: # List of strings
lst = ['Soon', 'For', 'Good', 'is',
       'portal', 'for', 'Good']

In [19]: # Calling DataFrame constructor on list
df = pd.DataFrame(lst)
print(df)
```

	0
0	Soon
1	For
2	Good
3	is
4	portal
5	for
6	Good

### CREATING DATAFRAME FROM DICTIONARY

#### Code

```
# initialise data of lists.
data = {'Name':['Tom', 'nick', 'krish', 'jack'],
        'Age':[20, 21, 19, 18]}

# Create DataFrame
df = pd.DataFrame(data)
print(df)

# Define a dictionary containing employee data
data = {'Name':['Jai', 'Princi', 'Gaurav', 'Anuj'],
        'Age':[27, 24, 22, 32],
        'Address':['Delhi', 'Kanpur', 'Allahabad', 'Kannauj'],
        'Qualification':['Msc', 'MA', 'MCA', 'Phd']}

# Convert the dictionary into DataFrame
df = pd.DataFrame(data)

# select two columns
print(df[['Name', 'Qualification']])
```

## Output

```
In [21]: # initialise data of Lists.
data = {'Name':['Tom', 'nick', 'krish', 'jack'],
        'Age':[20, 21, 19, 18]}

# Create DataFrame
df = pd.DataFrame(data)
print(df)
```

	Name	Age
0	Tom	20
1	nick	21
2	krish	19
3	jack	18

```
In [22]: # Define a dictionary containing employee data
data = {'Name':['Jai', 'Princi', 'Gaurav', 'Anuj'],
        'Age':[27, 24, 22, 32],
        'Address':['Delhi', 'Kanpur', 'Allahabad', 'Kannauj'],
        'Qualification':['Msc', 'MA', 'MCA', 'Phd']}

# Convert the dictionary into DataFrame
df = pd.DataFrame(data)

# select two columns
print(df[['Name', 'Qualification']])
```

	Name	Qualification
0	Jai	Msc
1	Princi	MA
2	Gaurav	MCA
3	Anuj	Phd

## CREATE DATAFRAME FROM CSV FILE

### Code

```
data = pd.read_csv("nba.csv", index_col="Name")
# retrieving row by loc method
first = data.loc["Avery Bradley"]
second = data.loc["R.J. Hunter"]
print(first, "\n\n", second)
```

### Output

```
In [25]: # making data frame from csv file
data = pd.read_csv("nba.csv", index_col="Name")
```

```
In [26]: # retrieving row by loc method
first = data.loc["Avery Bradley"]
second = data.loc["R.J. Hunter"]
print(first, "\n\n", second)
```

Team	Boston Celtics
Number	0
Position	PG
Age	25
Height	2-Jun
Weight	180
College	Texas
Salary	7730337.0

Name: Avery Bradley, dtype: object

Team	Boston Celtics
Number	28
Position	SG
Age	22
Height	5-Jun
Weight	185
College	Georgia State
Salary	1148640.0

Name: R.J. Hunter, dtype: object

## RETRIEVAL OF DATA FROM DATAFRAME

### Code

# retrieving columns by indexing operator

```
first = data["Age"]
```

```
print(first)
```

# retrieving rows by iloc method

```
row2 = data.iloc[3]
```

```
print(row2)
```

### Output

```
In [27]: # retrieving columns by indexing operator
first = data["Age"]
print(first)
```

```
Name
Avery Bradley    25
Jae Crowder      25
John Holland     27
R.J. Hunter      22
Jonas Jerebko    29
..
Trey Lyles       20
Shelvin Mack     26
Raul Neto        24
Tibor Pleiss     26
Jeff Withey      26
Name: Age, Length: 457, dtype: int64
```

```
In [29]: # retrieving rows by iloc method
row2 = data.iloc[3]
print(row2)
```

```
Team      Boston Celtics
Number          28
Position          SG
Age           22
Height        5-Jun
Weight         185
College    Georgia State
Salary      1148640.0
Name: R.J. Hunter, dtype: object
```

## HANDLING MISSING VALUES

### Code

# dictionary of lists

```
dict = {'First Score':[100, 90, np.nan, 95],
        'Second Score': [30, 45, 56, np.nan],
        'Third Score':[np.nan, 40, 80, 98]}
```

# creating a dataframe from list

```
df = pd.DataFrame(dict)
```

# using isnull() function

```
df.isnull()
```



# filling missing value using fillna()

df.fillna(0)

# dictionary of lists

```
dict = {'First Score':[100, 90, np.nan, 95],  
        'Second Score': [30, np.nan, 45, 56],  
        'Third Score':[52, 40, 80, 98],  
        'Fourth Score':[np.nan, np.nan, np.nan, 65]}
```

# creating a dataframe from dictionary

df = pd.DataFrame(dict)

df.dropna() #drop rows with NA

## Output

In order to check missing values in Pandas DataFrame, we use a function isnull() and notnull(). Both function help in checking whether a value is NaN or not. These function can also be used in Pandas Series in order to find null values in a series.

```
In [30]: # dictionary of lists  
dict = {'First Score':[100, 90, np.nan, 95],  
        'Second Score': [30, 45, 56, np.nan],  
        'Third Score':[np.nan, 40, 80, 98]}  
  
# creating a dataframe from list  
df = pd.DataFrame(dict)
```

```
In [31]: # using isnull() function  
df.isnull()
```

```
Out[31]:
```

	First Score	Second Score	Third Score
0	False	False	True
1	False	False	False
2	True	False	False
3	False	True	False

In order to fill null values in a datasets, we use fillna(), replace() and interpolate() function these function replace NaN values with some value of their own. All these function help in filling a null values in datasets of a DataFrame. Interpolate() function is basically used to fill NA values in the dataframe but it uses various interpolation technique to fill the missing values rather than hard-coding the value.

```
In [32]: # filling missing value using fillna()  
df.fillna(0)
```

```
Out[32]:
```

	First Score	Second Score	Third Score
0	100.0	30.0	0.0
1	90.0	45.0	40.0
2	0.0	56.0	80.0
3	95.0	0.0	98.0

```
In [33]: # dictionary of lists  
dict = {'First Score':[100, 90, np.nan, 95],  
        'Second Score': [30, np.nan, 45, 56],  
        'Third Score':[52, 40, 80, 98],  
        'Fourth Score':[np.nan, np.nan, np.nan, 65]}  
  
# creating a dataframe from dictionary  
df = pd.DataFrame(dict)
```

```
In [35]: df.dropna() #drop rows with NA
```

```
Out[35]:
```

	First Score	Second Score	Third Score	Fourth Score
3	95.0	56.0	98	65.0

## ITERATING OVER ROWS

### Code

#### # dictionary of lists

```
dict = {'name': ["aparna", "pankaj", "sudhir", "Geeku"],
        'degree': ["MBA", "BCA", "M.Tech", "MBA"],
        'score': [90, 40, 80, 98]}
```

#### # creating a dataframe from a dictionary

```
df = pd.DataFrame(dict)
print(df)
```

#### # iterating over rows using iterrows() function

```
for i, j in df.iterrows():
    print(i, j)
    print()
```

### Output

In order to iterate over rows, we can use three function `iteritems()`, `iterrows()`, `itertuples()`. These three function will help in iteration over rows.

```
In [36]: # dictionary of lists
dict = {'name': ["aparna", "pankaj", "sudhir", "Geeku"],
        'degree': ["MBA", "BCA", "M.Tech", "MBA"],
        'score': [90, 40, 80, 98]}

# creating a dataframe from a dictionary
df = pd.DataFrame(dict)

print(df)
```

	name	degree	score
0	aparna	MBA	90
1	pankaj	BCA	40
2	sudhir	M.Tech	80
3	Geeku	MBA	98

```
In [37]: # iterating over rows using iterrows() function
for i, j in df.iterrows():
    print(i, j)
    print()
```

```
0 name      aparna
  degree      MBA
  score       90
  Name: 0, dtype: object
```

```
1 name      pankaj
  degree      BCA
  score       40
  Name: 1, dtype: object
```

```
2 name      sudhir
  degree    M.Tech
  score       80
  Name: 2, dtype: object
```

```
3 name      Geeku
  degree      MBA
  score       98
  Name: 3, dtype: object
```

## ITERATING OVER COLUMNS

### Code

```
dict = {'name':["aparna", "pankaj", "sudhir", "Geeku"],
        'degree': ["MBA", "BCA", "M.Tech", "MBA"],
        'score':[90, 40, 80, 98]}
```

```
# creating a dataframe from a dictionary
df = pd.DataFrame(dict)
print(df)
# creating a list of dataframe columns
columns = list(df)
```

```
for i in columns:
    # printing the third element of the column
    print (df[i][2])
```

### Output

In order to iterate over columns, we need to create a list of dataframe columns and then iterating through that list to pull out the dataframe columns.

```
In [39]: dict = {'name':["aparna", "pankaj", "sudhir", "Geeku"],
                'degree': ["MBA", "BCA", "M.Tech", "MBA"],
                'score':[90, 40, 80, 98]}

# creating a dataframe from a dictionary
df = pd.DataFrame(dict)
print(df)
```

	name	degree	score
0	aparna	MBA	90
1	pankaj	BCA	40
2	sudhir	M.Tech	80
3	Geeku	MBA	98

```
In [40]: # creating a list of dataframe columns
columns = list(df)

for i in columns:
    # printing the third element of the column
    print (df[i][2])
```

```
sudhir
M.Tech
80
```

## WORKING WITH NBA DATASET

### READING THE CSV AND DISPLAY THE HEAD AND TAIL

#### Code

```
df2 = pd.read_csv("nba.csv")
df2.head()
df2.tail()
```

#### Output

```
In [8]: df2 = pd.read_csv("nba.csv")
df2.head()
```

```
Out[8]:
```

	Name	Team	Number	Position	Age	Height	Weight	College	Salary
0	Avery Bradley	Boston Celtics	0	PG	25	2-Jun	180	Texas	7730337.0
1	Jae Crowder	Boston Celtics	99	SF	25	6-Jun	235	Marquette	6796117.0
2	John Holland	Boston Celtics	30	SG	27	5-Jun	205	Boston University	NaN
3	R.J. Hunter	Boston Celtics	28	SG	22	5-Jun	185	Georgia State	1148640.0
4	Jonas Jerebko	Boston Celtics	8	PF	29	10-Jun	231	NaN	5000000.0

```
In [9]: df2.tail()
```

```
Out[9]:
```

	Name	Team	Number	Position	Age	Height	Weight	College	Salary
452	Trey Lyles	Utah Jazz	41	PF	20	10-Jun	234	Kentucky	2239800.0
453	Shelvin Mack	Utah Jazz	8	PG	26	3-Jun	203	Butler	2433333.0
454	Raul Neto	Utah Jazz	25	PG	24	1-Jun	179	NaN	900000.0
455	Tibor Pleiss	Utah Jazz	21	C	26	3-Jul	256	NaN	2900000.0
456	Jeff Withey	Utah Jazz	24	C	26	Jul-00	231	Kansas	947276.0

### CHECKING OUT THE TYPE AND INFORMATION REGARDING THE DATA FRAME

#### Code

```
print("Type:\n",type(df2))
print("Information about the dataframe:\n")
df2.info(verbose=True)
print("Shape of dataframe:",df2.shape)
df2.drop_duplicates()
print(df2.shape)
print("Columns of dataframe:\t",df2.columns)
print(df2.describe())
```

## Output

```
In [19]: print("Type:\n",type(df2))
print("Information about the dataframe:\n")
df2.info(verbose=True)

Type:
<class 'pandas.core.frame.DataFrame'>
Information about the dataframe:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 457 entries, 0 to 456
Data columns (total 9 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   Name        457 non-null   object
1   Team        457 non-null   object
2   Number      457 non-null   int64
3   Position    457 non-null   object
4   Age         457 non-null   int64
5   Height      457 non-null   object
6   Weight      457 non-null   int64
7   College     373 non-null   object
8   Salary      446 non-null   float64
dtypes: float64(1), int64(3), object(5)
memory usage: 32.3+ KB
```

## DROPPING DUPLICATE ROWS AND DISPLAYING A DESCRIPTION OF THE DATASET

### Code

```
print("Shape of dataframe:",df2.shape)
df2.drop_duplicates()
print(df2.shape)
print("Columns of dataframe:\t",df2.columns)
print(df2.describe())
```

## Output

```
In [11]: print("Shape of dataframe:",df2.shape)
df2.drop_duplicates()
print(df2.shape)
```

```
Shape of dataframe: (457, 9)
(457, 9)
```

```
In [45]: print("Columns of dataframe:\t",df2.columns)
```

```
Columns of dataframe:   Index(['Name', 'Team', 'Number', 'Position', 'Age', 'Height', 'Weight',
    'College', 'Salary'],
    dtype='object')
```

```
In [46]: print(df2.describe())
```

	Number	Age	Weight	Salary
count	457.000000	457.000000	457.000000	4.460000e+02
mean	17.678337	26.938731	221.522976	4.842684e+06
std	15.966090	4.404016	26.368343	5.229238e+06
min	0.000000	19.000000	161.000000	3.088800e+04
25%	5.000000	24.000000	200.000000	1.044792e+06
50%	13.000000	26.000000	220.000000	2.839073e+06
75%	25.000000	30.000000	240.000000	6.500000e+06
max	99.000000	40.000000	307.000000	2.500000e+07

## EXERCISE 3: PANDAS (CHARACTER DEATH DATA)

### REMOVING THE HEADER

#### Code

```
df2 = pd.read_csv('character-deaths.csv',header=None)
df2.head()
```

#### Output

Reading the CSV without using the 1st row as header or column names.

```
In [4]: df2 = pd.read_csv('character-deaths.csv',header=None)
df2.head()
```

```
Out[4]:
```

	0	1	2	3	4	5	6	7	8	9	10	11	12
0	Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD
1	Addam Marbrand	Lannister	NaN	NaN	NaN	56	1	1	1	1	1	1	0
2	Aegon Frey (Jinglebell)	None	299	3	51	49	1	1	0	0	1	0	0
3	Aegon Targaryen	House Targaryen	NaN	NaN	NaN	5	1	1	0	0	0	0	1
4	Adrack Humble	House Greyjoy	300	5	20	20	1	1	0	0	0	0	1

### ADDING THE HEADER BACK

#### Code

```
header=df2.iloc[0]
df2=df2[1:]
df2.columns=header
df2.head()
```

#### Output

Using the first row of the dataframe as column names

```
In [5]: header=df2.iloc[0]
df2=df2[1:]
df2.columns=header
df2.head()
```

```
Out[5]:
```

	Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD
1	Addam Marbrand	Lannister	NaN	NaN	NaN	56	1	1	1	1	1	1	0
2	Aegon Frey (Jinglebell)	None	299	3	51	49	1	1	0	0	1	0	0
3	Aegon Targaryen	House Targaryen	NaN	NaN	NaN	5	1	1	0	0	0	0	1
4	Adrack Humble	House Greyjoy	300	5	20	20	1	1	0	0	0	0	1
5	Aemon Costayne	Lannister	NaN	NaN	NaN	NaN	1	1	0	0	1	0	0

## DISPLAYING THE FIRST 10 ROWS

### Code

```
df2[:10]
```

### Output

```
In [7]: df2[:10]
```

```
Out[7]:
```

	Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD
1	Addam Marbrand	Lannister	NaN	NaN	NaN	56	1	1	1	1	1	1	0
2	Aegon Frey (Jinglebell)	None	299	3	51	49	1	1	0	0	1	0	0
3	Aegon Targaryen	House Targaryen	NaN	NaN	NaN	5	1	1	0	0	0	0	1
4	Adrack Humble	House Greyjoy	300	5	20	20	1	1	0	0	0	0	1
5	Aemon Costayne	Lannister	NaN	NaN	NaN	NaN	1	1	0	0	1	0	0
6	Aemon Estermont	Baratheon	NaN	NaN	NaN	NaN	1	1	0	1	1	0	0
7	Aemon Targaryen (son of Maekar I)	Night's Watch	300	4	35	21	1	1	1	0	1	1	0
8	Aenys Frey	None	300	5	NaN	59	0	1	1	1	1	0	1
9	Aeron Greyjoy	House Greyjoy	NaN	NaN	NaN	11	1	1	0	1	0	1	0
10	Aethan	Night's Watch	NaN	NaN	NaN	0	1	0	0	0	1	0	0

## SAVING THE MODIFIED DATAFRAME AS CSV (SAVING THE FIRST 10 ROWS RETRIEVED)

### Code

```
df_temp=df2[:10]  
df_temp.to_csv("result.csv",index=False)
```

### Output

pandasnotepad1 exe 2.txt	11-08-2021 14:19	Text Document	12 KB
pandasnotepad2 exe 3.txt	11-08-2021 14:19	Text Document	46 KB
result.csv	12-08-2021 21:00	Microsoft Excel C...	1 KB

## OPEN AN XLS FILE AND SAVE IT AS CSV

### Code

```
df3=pd.read_excel("result2.xls")
df3
df3.to_csv("result3.csv",index=False)
```

### Output

#### Open an xls file and save it as CSV

```
In [9]: df3=pd.read_excel("result2.xls")
df3
```

```
Out[9]:
```

	Unnamed: 0	Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FFC	DwD
0	1	Addam Marbrand	Lannister	NaN	NaN	NaN	56.0	1	1	1	1	1	1	0
1	2	Aegon Frey (Jinglebell)	None	299.0	3.0	51.0	49.0	1	1	0	0	1	0	0
2	3	Aegon Targaryen	House Targaryen	NaN	NaN	NaN	5.0	1	1	0	0	0	0	1
3	4	Adrack Humble	House Greyjoy	300.0	5.0	20.0	20.0	1	1	0	0	0	0	1
4	5	Aemon Costayne	Lannister	NaN	NaN	NaN	NaN	1	1	0	0	1	0	0
5	6	Aemon Estermont	Baratheon	NaN	NaN	NaN	NaN	1	1	0	1	1	0	0
6	7	Aemon Targaryen (son of Maekar I)	Night's Watch	300.0	4.0	35.0	21.0	1	1	1	0	1	1	0
7	8	Aenys Frey	None	300.0	5.0	NaN	59.0	0	1	1	1	1	0	1
8	9	Aeron Greyjoy	House Greyjoy	NaN	NaN	NaN	11.0	1	1	0	1	0	1	0
9	10	Aethan	Night's Watch	NaN	NaN	NaN	0.0	1	0	0	0	1	0	0

```
In [10]: df3.to_csv("result3.csv",index=False)
```

## SETTING COLUMN AS AN INDEX

### Code

```
df=df2.copy(deep=True)
df.set_index("Gender",inplace=True)
df
```

### Output

```
In [11]: df=df2.copy(deep=True)
df.set_index("Gender",inplace=True)
df
```

```
Out[11]:
```

	Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Nobility	GoT	CoK	SoS	FFC	DwD
Gender												
1	Addam Marbrand	Lannister	NaN	NaN	NaN	56	1	1	1	1	1	0
1	Aegon Frey (Jinglebell)	None	299	3	51	49	1	0	0	1	0	0
1	Aegon Targaryen	House Targaryen	NaN	NaN	NaN	5	1	0	0	0	0	1
1	Adrack Humble	House Greyjoy	300	5	20	20	1	0	0	0	0	1
1	Aemon Costayne	Lannister	NaN	NaN	NaN	NaN	1	0	0	1	0	0
...	...	...	...	...	...	...	...	...	...	...	...	...
1	Zollo	None	NaN	NaN	NaN	21	0	0	0	1	0	0
1	Yurkhaz zo Yunzak	None	300	5	59	47	0	0	0	0	0	1
1	Yezzan Zo Qaggaz	None	300	5	57	25	1	0	0	0	0	1
1	Torwynd the Tame	Wildling	300	5	73	73	0	0	0	1	0	0
1	Talbert Serry	Tyrell	300	4	29	29	1	0	0	0	1	0

917 rows x 12 columns



## INFORMATION ABOUT THE DATAFRAME

### Code

```
df2.info  
type(df2)
```

### Output

```
In [12]: df2.info  
Out[12]: <bound method DataFrame.info of 0  
1      Addam Marbrand      Lannister      NaN      NaN  
2      Aegon Frey (Jinglebell)      None      299      3  
3      Aegon Targaryen      House Targaryen      NaN      NaN  
4      Adrack Humble      House Greyjoy      300      5  
5      Aemon Costayne      Lannister      NaN      NaN  
..      ...      ...      ...      ...  
913      Zollo      None      NaN      NaN  
914      Yurkhaz zo Yunzak      None      300      5  
915      Yezzan Zo Qaggaz      None      300      5  
916      Torwynd the Tame      Wildling      300      5  
917      Talbert Serry      Tyrell      300      4  
  
0      Death Chapter Book Intro Chapter Gender Nobility GoT CoK SoS FfC DwD  
1      NaN      56      1      1      1      1      1      0  
2      51      49      1      1      0      0      1      0      0  
3      NaN      5      1      1      0      0      0      0      1  
4      20      20      1      1      0      0      0      0      1  
5      NaN      NaN      1      1      0      0      1      0      0  
..      ...      ...      ...      ...      ...      ...      ...      ...  
913      NaN      21      1      0      0      0      1      0      0  
914      59      47      1      0      0      0      0      0      1  
915      57      25      1      1      0      0      0      0      1  
916      73      73      1      0      0      0      1      0      0  
917      29      29      1      1      0      0      0      1      0  
  
[917 rows x 13 columns]>  
  
In [13]: type(df2)  
Out[13]: pandas.core.frame.DataFrame
```

## APPENDING DATA TO DATAFRAME

### Code

```
df=df2[-10:]  
df2=df2.append(df) #appending the last 10 rows of the dataframe to itself, thus reatng duplicate  
rows  
df2[-20:]  
df2.shape
```

## Output

```
In [14]: df=df2[-10:]
df2=df2.append(df) #appending the last 10 rows of the dataframe to itself, thus reatng duplicate rows
df2[-20:]
```

```
Out[14]:
```

	Name	Allegiances	Death Year	Book of Death	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FFC	DwD
908	Yohn Royce	Arryn	NaN	NaN	NaN	29	1	1	1	0	0	1	0
909	Yoren	Night's Watch	299	2	19	13	1	0	1	1	0	0	0
910	Young Henly	Night's Watch	299	3	55	55	1	0	0	0	1	0	0
911	Ysilla	House Targaryen	NaN	NaN	NaN	8	1	0	0	0	0	0	1
912	Zei	Stark	NaN	NaN	NaN	64	0	0	0	0	1	0	0
913	Zollo	None	NaN	NaN	NaN	21	1	0	0	0	1	0	0
914	Yurkhaz zo Yunzak	None	300	5	59	47	1	0	0	0	0	0	1
915	Yezzan Zo Qaggaz	None	300	5	57	25	1	1	0	0	0	0	1
916	Torwynd the Tame	Wildling	300	5	73	73	1	0	0	0	1	0	0
917	Talbert Serry	Tyrell	300	4	29	29	1	1	0	0	0	1	0
908	Yohn Royce	Arryn	NaN	NaN	NaN	29	1	1	1	0	0	1	0
909	Yoren	Night's Watch	299	2	19	13	1	0	1	1	0	0	0
910	Young Henly	Night's Watch	299	3	55	55	1	0	0	0	1	0	0
911	Ysilla	House Targaryen	NaN	NaN	NaN	8	1	0	0	0	0	0	1
912	Zei	Stark	NaN	NaN	NaN	64	0	0	0	0	1	0	0
913	Zollo	None	NaN	NaN	NaN	21	1	0	0	0	1	0	0
914	Yurkhaz zo Yunzak	None	300	5	59	47	1	0	0	0	0	0	1
915	Yezzan Zo Qaggaz	None	300	5	57	25	1	1	0	0	0	0	1
916	Torwynd the Tame	Wildling	300	5	73	73	1	0	0	0	1	0	0
917	Talbert Serry	Tyrell	300	4	29	29	1	1	0	0	0	1	0

```
In [15]: df2.shape
```

```
Out[15]: (927, 13)
```

## DROPPING DUPLICATE ROWS (THE REPEATING ROWS ADDED ABOVE)

### Code

```
df2.drop_duplicates(inplace=True)
df2.shape
```

### Output

```
In [16]: df2.drop_duplicates(inplace=True)
df2.shape
```

```
Out[16]: (917, 13)
```

## RENAMING COLUMN NAMES (RENAMING 'BOOK OF DEATH' AS 'BOD')

### Code

```
print("Column names: ",df2.columns)
df2.rename(columns={'Book of Death':'BoD'},inplace=True)
print(df2.columns)
```

### Output

```
In [17]: print("Column names: ",df2.columns)
Column names: Index(['Name', 'Allegiances', 'Death Year', 'Book of Death', 'Death Chapter',
                    'Book Intro Chapter', 'Gender', 'Nobility', 'GoT', 'CoK', 'SoS', 'FfC',
                    'DwD'],
                    dtype='object', name=0)

In [18]: df2.rename(columns={'Book of Death':'BoD'},inplace=True)
print(df2.columns)
Index(['Name', 'Allegiances', 'Death Year', 'BoD', 'Death Chapter',
      'Book Intro Chapter', 'Gender', 'Nobility', 'GoT', 'CoK', 'SoS', 'FfC',
      'DwD'],
      dtype='object', name=0)
```

## DESCRIBING THE DATAFRAME

### Code

```
df2.describe()
```

### Output

```
In [19]: df2.describe()
Out[19]:
```

	Name	Allegiances	Death Year	BoD	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD
count	917	917	305	307	299	905	917	917	917	917	917	917	917
unique	916	21	4	5	71	79	2	2	2	2	2	2	2
top	Myles	None	299	3	34	0	1	0	0	0	0	0	0
freq	2	253	156	97	16	41	760	487	667	593	528	667	656

## COUNTING UNIQUE VALUES

### Code

```
print("No. of chapters talking about death:\t",df2['Death Chapter'].count())
print("No. of unique chapters talking about death:\t",df2['Death Chapter'].nunique())
print("Unique chapters talking about death:\t",df2['Death Chapter'].unique())
```

### Output

```
In [20]: print("No. of chapters talking about death:\t",df2['Death Chapter'].count())
print("No. of unique chapters talking about death:\t",df2['Death Chapter'].nunique())
print("Unique chapters talking about death:\t",df2['Death Chapter'].unique())

No. of chapters talking about death:      299
No. of unique chapters talking about death:      71
Unique chapters talking about death:  [nan '51' '20' '35' '56' '4' '46' '10' '34' '47' '41' '21' '33' '39' '52'
'37' '24' '31' '27' '62' '65' '7' '49' '50' '42' '30' '66' '64' '29' '0'
'53' '58' '76' '63' '55' '14' '61' '12' '68' '69' '80' '36' '1' '19' '72'
'59' '43' '70' '75' '11' '3' '60' '26' '44' '67' '45' '18' '23' '57' '25'
'16' '17' '48' '9' '2' '6' '32' '77' '74' '40' '38' '73']
```

## ACCESSING ROWS AND COLUMNS (USING LOC AND ILOC)

### Code

```
print("Row 0 with iloc[0]: \n",df2.iloc[0]) #Row 0 with iloc[0]
print("Rows 4 and 5 with columns 1,2 and 3\n",df2.iloc[[4,5],[1,2,3]]) #columns 1,2,3 of rows 4 and 5

print("First five values of 'Allegiances' column\t:\n",df2.iloc[0:5,0])

df2.set_index('Name',inplace=True)
print("\n\nUsing loc method:\t",df2.loc['Adrack Humble','Death Year'])
```

### Output

iloc

```
In [21]: print("Row 0 with iloc[0]: \n",df2.iloc[0]) #Row 0 with iloc[0]
print("Rows 4 and 5 with columns 1,2 and 3\n",df2.iloc[[4,5],[1,2,3]]) #columns 1,2,3 of rows 4 and 5
```

```
Row 0 with iloc[0]:
0
Name                Addam Marbrand
Allegiances          Lannister
Death Year           NaN
BoD                 NaN
Death Chapter       NaN
Book Intro Chapter   56
Gender              1
Nobility            1
GoT                 1
CoK                 1
SoS                 1
FfC                 1
DwD                 0
Name: 1, dtype: object
Rows 4 and 5 with columns 1,2 and 3
0 Allegiances Death Year BoD
5 Lannister      NaN NaN
6 Baratheon      NaN NaN
```

```
In [22]: print("First five values of 'Allegiances' column\t:\n",df2.iloc[0:5,0])
```

```
First five values of 'Allegiances' column      :
1                Addam Marbrand
2  Aegon Frey (Jinglebell)
3      Aegon Targaryen
4      Adrack Humble
5      Aemon Costayne
Name: Name, dtype: object
```

loc

```
In [23]: df2.set_index('Name',inplace=True)
```

```
In [24]: print("\n\nUsing loc method:\t",df2.loc['Adrack Humble','Death Year'])
```

```
Using loc method:      300
```

## CREATING SUBSETS USING RELATIONAL OPERATORS (RETRIEVING THOSE WITH DEATH YEAR < 400)

### Code

```
df3=df2.copy(deep=True)
df3['Death Year']=df3['Death Year'].astype(float)
df3 = df3[df3['Death Year']<400]
print(df3)
```

### Output

```
In [25]: df3=df2.copy(deep=True)
df3['Death Year']=df3['Death Year'].astype(float)
```

```
In [26]: df3 = df3[df3['Death Year']<400]
print(df3)
```

	Allegiances	Death Year	BoD	\
0				
Name				
Aegon Frey (Jinglebell)	None	299.0	3	
Adrack Humble	House Greyjoy	300.0	5	
Aemon Targaryen (son of Maekar I)	Night's Watch	300.0	4	
Aenys Frey	None	300.0	5	
Aggar	House Greyjoy	299.0	2	
...	...	...	..	
Young Henly	Night's Watch	299.0	3	
Yurkhaz zo Yunzak	None	300.0	5	
Yezzan Zo Qaggaz	None	300.0	5	
Torwynd the Tame	Wildling	300.0	5	
Talbert Serry	Tyrell	300.0	4	

  

	Death	Chapter	Book	Intro	Chapter	Gender	\
0							
Name							
Aegon Frey (Jinglebell)	51				49	1	
Adrack Humble	20				20	1	
Aemon Targaryen (son of Maekar I)	35				21	1	
Aenys Frey	NaN				59	0	
Aggar	56				50	1	
...	...				...	...	
Young Henly	55				55	1	
Yurkhaz zo Yunzak	59				47	1	
Yezzan Zo Qaggaz	57				25	1	
Torwynd the Tame	73				73	1	
Talbert Serry	29				29	1	

  

	Nobility	GoT	CoK	SoS	FfC	DwD
0						
Name						
Aegon Frey (Jinglebell)	1	0	0	1	0	0
Adrack Humble	1	0	0	0	0	1
Aemon Targaryen (son of Maekar I)	1	1	0	1	1	0
Aenys Frey	1	1	1	1	0	1
Aggar	0	0	1	0	0	0
...	...	...	...	...	...	...
Young Henly	0	0	0	1	0	0
Yurkhaz zo Yunzak	0	0	0	0	0	1
Yezzan Zo Qaggaz	1	0	0	0	0	1
Torwynd the Tame	0	0	0	1	0	0
Talbert Serry	1	0	0	0	1	0

[305 rows x 12 columns]

## USING GROUPBY (GROUPING DATA BY 'DEATH YEAR')

### Code

```
import numpy as np
df5 = df2.groupby('Death Year').count()
df5
```

### Output

```
In [28]: import numpy as np
```

```
In [29]: df5 = df2.groupby('Death Year').count()
df5
```

```
Out[29]:
```

	Allegiances	BoD	Death Chapter	Book Intro Chapter	Gender	Nobility	GoT	CoK	SoS	FfC	DwD
Death Year											
297	3	3	3	3	3	3	3	3	3	3	3
298	46	46	46	46	46	46	46	46	46	46	46
299	156	156	154	154	156	156	156	156	156	156	156
300	100	100	92	97	100	100	100	100	100	100	100

## USING AGGREGATE WITH GROUPBY

### Code

```
df3=df2.copy(deep=True)
df3.fillna(0)
df3['Gender']=df3['Gender'].astype(int)
df3['Death Year']=df3['Death Year'].astype(float)
df5 = df3.groupby("Death Year").agg({"Death Year":np.sum,"Gender":np.sum})
df5
```

### Output

```
In [30]: df3=df2.copy(deep=True)
df3.fillna(0)
df3['Gender']=df3['Gender'].astype(int)
df3['Death Year']=df3['Death Year'].astype(float)
df5 = df3.groupby("Death Year").agg({"Death Year":np.sum,"Gender":np.sum})
df5
```

```
Out[30]:
```

	Death Year	Gender
Death Year		
297.0	891.0	3
298.0	13708.0	43
299.0	46644.0	141
300.0	30000.0	82

## VISUALIZING THE DATA

### Code

```
import matplotlib.pyplot as plt
df5.plot(x='Death Year',y='Gender',kind='scatter')
plt.show()
```

### Output

```
In [35]: import matplotlib.pyplot as plt
df5.plot(x='Death Year',y='Gender',kind='scatter')
```

```
Out[35]: <AxesSubplot:xlabel='Death Year', ylabel='Gender'>
```

```
In [36]: plt.show()
```

