In [2]:

import numpy as np
import pandas as pd
#import os
#import matplotlib.pyplot as plt
#%matplotlib inline

In [5]:

	c_train = pd c_train.head	_	sv("I	∃:/Machine L	earnin.	g/Mach	nine-Lear	ning	g-Masters	-master,	/Machi	ne-L	.e
4												1	•
4	5	0	3	William Henry	male	35.0	0	0	373450	8.0500	NaN		•
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583	NaN		
6	7	0	1	McCarthy, Mr. Timothy J	male	54.0	0	0	17463	51.8625	E46		
7	8	0	3	Palsson, Master. Gosta Leonard	male	2.0	3	1	349909	21.0750	NaN		
8	9	1	3	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	female	27.0	0	2	347742	11.1333	NaN		
9	10	1	2	Nasser, Mrs. Nicholas (Adele Achem)	female	14.0	1	0	237736	30.0708	NaN	,	•
4												•	

In [10]:

titanic_train

Out[10]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500

891 rows × 12 columns

In [17]:

titanic_train.tail(10)

Out[17]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Faı
881	882	0	3	Markun, Mr. Johann	male	33.0	0	0	349257	7.895
882	883	0	3	Dahlberg, Miss. Gerda Ulrika	female	22.0	0	0	7552	10.516
883	884	0	2	Banfield, Mr. Frederick James	male	28.0	0	0	C.A./SOTON 34068	10.500
884	885	0	3	Sutehall, Mr. Henry Jr	male	25.0	0	0	SOTON/OQ 392076	7.050
885	886	0	3	Rice, Mrs. William (Margaret Norton)	female	39.0	0	5	382652	29.125
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.000
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.000
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.450
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.000
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.750
4										•

In [19]:

```
titanic_train.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	PassengerId	891 non-null	int64
1	Survived	891 non-null	int64
2	Pclass	891 non-null	int64
3	Name	891 non-null	object
4	Sex	891 non-null	object
5	Age	714 non-null	float64
6	SibSp	891 non-null	int64
7	Parch	891 non-null	int64
8	Ticket	891 non-null	object
9	Fare	891 non-null	float64
10	Cabin	204 non-null	object
11	Embarked	889 non-null	object
dtyp	es: float64(2), int64(5), obj	ect(5)

memory usage: 83.7+ KB

In [13]:

```
sum(titanic_train.isnull().sum())
```

Out[13]:

866

In [6]:

```
titanic_train.describe() #only of numerical data
```

Out[6]:

	Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare
count	156.000000	156.000000	156.000000	126.000000	156.000000	156.000000	156.000000
mean	78.500000	0.346154	2.423077	28.141508	0.615385	0.397436	28.109587
std	45.177428	0.477275	0.795459	14.613880	1.056235	0.870146	39.401047
min	1.000000	0.000000	1.000000	0.830000	0.000000	0.000000	6.750000
25%	39.750000	0.000000	2.000000	19.000000	0.000000	0.000000	8.003150
50%	78.500000	0.000000	3.000000	26.000000	0.000000	0.000000	14.454200
75%	117.250000	1.000000	3.000000	35.000000	1.000000	0.000000	30.371850
max	156.000000	1.000000	3.000000	71.000000	5.000000	5.000000	263.000000

```
In [17]:
```

```
titanic_train['Sex'].value_counts()
```

Out[17]:

male 577 female 314

Name: Sex, dtype: int64

In [6]:

```
titanic_train.dtypes[titanic_train.dtypes == "object"]
```

Out[6]:

Name object
Sex object
Ticket object
Cabin object
Embarked object
dtype: object

In [15]:

```
titanic_train.dtypes == "object"
```

Out[15]:

False PassengerId Survived False Pclass False Name True Sex True False Age SibSp False Parch False Ticket True Fare False Cabin True **Embarked** True dtype: bool

In [18]:

```
a=titanic_train.dtypes[titanic_train.dtypes == "object"]
print(a)
```

Name object
Sex object
Ticket object
Cabin object
Embarked object
dtype: object

In [18]:

```
a.describe()
```

Out[18]:

count 5
unique 1
top object
freq 5
dtype: object

In [101]:

```
titanic_train.dtypes[titanic_train.dtypes == "object"].index
```

Out[101]:

Index(['Name', 'Sex', 'Ticket', 'Cabin', 'Embarked'], dtype='object')

In [103]:

```
list_col = titanic_train.dtypes[titanic_train.dtypes == "object"].index
titanic_train[list_col].describe()
```

Out[103]:

	Name	Sex	licket	Cabin	Embarked
count	156	156	156	31	155
unique	156	2	145	28	3
top	Williams, Mr. Charles Eugene	male	113803	C123	S
freq	1	100	2	2	110

In [163]:

```
#now for categorical column
titanic_train.dtypes == "object"
a = titanic_train.dtypes[titanic_train.dtypes == "object"].index
a
titanic_train[a].describe()
```

Out[163]:

	Name	Sex	Ticket	Cabin	Embarked
count	156	156	156	31	155
unique	156	2	145	28	3
top	Williams, Mr. Charles Eugene	male	113803	C123	S
freq	1	100	2	2	110

In [4]:

```
titanic_train.info()
```

```
RangeIndex: 156 entries, 0 to 155
Data columns (total 12 columns):
    Column
                 Non-Null Count Dtype
                  -----
0
    PassengerId 156 non-null
                                  int64
 1
    Survived
                 156 non-null
                                 int64
 2
    Pclass
                 156 non-null
                                 int64
 3
    Name
                 156 non-null
                                 object
 4
                 156 non-null
    Sex
                                 object
 5
                 126 non-null
                                 float64
    Age
 6
    SibSp
                 156 non-null
                                 int64
 7
                 156 non-null
                                 int64
    Parch
 8
    Ticket
                 156 non-null
                                 object
 9
    Fare
                 156 non-null
                                 float64
 10 Cabin
                 31 non-null
                                 object
                 155 non-null
 11 Embarked
                                 object
dtypes: float64(2), int64(5), object(5)
memory usage: 14.8+ KB
```

<class 'pandas.core.frame.DataFrame'>

In [6]:

```
titanic_train.columns
```

Out[6]:

In [110]:

```
titanic_train[['Name','Sex','Age']]
```

Out[110]:

	Name	Sex	Age
0	Braund, Mr. Owen Harris	male	22.0
1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0
2	Heikkinen, Miss. Laina	female	26.0
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0
4	Allen, Mr. William Henry	male	35.0
5	Moran, Mr. James	male	NaN
6	McCarthy, Mr. Timothy J	male	54.0
7	Palsson, Master. Gosta Leonard	male	2.0
8	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	female	27.0
9	Nasser, Mrs. Nicholas (Adele Achem)	female	14.0
10	Sandstrom, Miss. Marguerite Rut	female	4.0
11	Bonnell, Miss. Elizabeth	female	58.0
12	Saundercock, Mr. William Henry	male	20.0
13	Andersson, Mr. Anders Johan	male	39.0
14	Vestrom, Miss. Hulda Amanda Adolfina	female	14.0
15	Hewlett, Mrs. (Mary D Kingcome)	female	55.0
16	Rice, Master. Eugene	male	2.0
17	Williams, Mr. Charles Eugene	male	NaN
18	Vander Planke, Mrs. Julius (Emelia Maria Vande	female	31.0
19	Masselmani, Mrs. Fatima	female	NaN
20	Fynney, Mr. Joseph J	male	35.0
21	Beesley, Mr. Lawrence	male	34.0
22	McGowan, Miss. Anna "Annie"	female	15.0
23	Sloper, Mr. William Thompson	male	28.0
24	Palsson, Miss. Torborg Danira	female	8.0
25	Asplund, Mrs. Carl Oscar (Selma Augusta Emilia	female	38.0
26	Emir, Mr. Farred Chehab	male	NaN
27	Fortune, Mr. Charles Alexander	male	19.0
28	O'Dwyer, Miss. Ellen "Nellie"	female	NaN
29	Todoroff, Mr. Lalio	male	NaN
126	McMahon, Mr. Martin	male	NaN
127	Madsen, Mr. Fridtjof Arne	male	24.0
128	Peter, Miss. Anna	female	NaN
129	Ekstrom, Mr. Johan	male	45.0

	Name	Sex	Age
130	Drazenoic, Mr. Jozef	male	33.0
131	Coelho, Mr. Domingos Fernandeo	male	20.0
132	Robins, Mrs. Alexander A (Grace Charity Laury)	female	47.0
133	Weisz, Mrs. Leopold (Mathilde Francoise Pede)	female	29.0
134	Sobey, Mr. Samuel James Hayden	male	25.0
135	Richard, Mr. Emile	male	23.0
136	Newsom, Miss. Helen Monypeny	female	19.0
137	Futrelle, Mr. Jacques Heath	male	37.0
138	Osen, Mr. Olaf Elon	male	16.0
139	Giglio, Mr. Victor	male	24.0
140	Boulos, Mrs. Joseph (Sultana)	female	NaN
141	Nysten, Miss. Anna Sofia	female	22.0
142	Hakkarainen, Mrs. Pekka Pietari (Elin Matilda	female	24.0
143	Burke, Mr. Jeremiah	male	19.0
144	Andrew, Mr. Edgardo Samuel	male	18.0
145	Nicholls, Mr. Joseph Charles	male	19.0
146	Andersson, Mr. August Edvard ("Wennerstrom")	male	27.0
147	Ford, Miss. Robina Maggie "Ruby"	female	9.0
148	Navratil, Mr. Michel ("Louis M Hoffman")	male	36.5
149	Byles, Rev. Thomas Roussel Davids	male	42.0
150	Bateman, Rev. Robert James	male	51.0
151	Pears, Mrs. Thomas (Edith Wearne)	female	22.0
152	Meo, Mr. Alfonzo	male	55.5
153	van Billiard, Mr. Austin Blyler	male	40.5
154	Olsen, Mr. Ole Martin	male	NaN
155	Williams, Mr. Charles Duane	male	51.0

156 rows × 3 columns

In [166]:

```
sorted(titanic_train["Name"])[5 :10:2]# Check the first 15 sorted names
```

Out[166]:

```
['Andersson, Mr. August Edvard ("Wennerstrom")',
'Andrew, Mr. Edgardo Samuel',
'Asplund, Mrs. Carl Oscar (Selma Augusta Emilia Johansson)']
```

```
In [167]:
```

```
titanic_train["Name"].describe()
Out[167]:
count
                                    156
unique
                                    156
          Williams, Mr. Charles Eugene
top
freq
Name: Name, dtype: object
In [168]:
titanic_train[["Ticket"]][4:9] #operation on df as [['ticket']]
Out[168]:
    Ticket
  373450
5 330877
    17463
7 349909
8 347742
In [169]:
titanic_train["Ticket"].describe()
titanic_train.columns
Out[169]:
```

```
dtype='object')
```

In [171]:

```
titanic_train["sudh"]="sdffs"
titanic_train.head()
```

Out[171]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	(
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	_
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	
4										•	

In [172]:

titanic_train["Cabin"][0:15] # Check the first 15 tickets

Out[172]:

NaN 1 C85 2 NaN 3 C123 4 NaN 5 NaN 6 E46 7 NaN 8 NaN 9 NaN 10 G6 11 C103 12 NaN 13 NaN 14 NaN

Name: Cabin, dtype: object

In [118]:

titanic_train.dtypes# Check number of unique cabins

Out[118]:

PassengerId int64 Survived int64 Pclass int64 Name object object Sex Age float64 int64 SibSp int64 Parch object Ticket Fare float64 object Cabin Embarked object int64 test

dtype: object

In [18]:

```
titanic_train["Survived"]
Out[18]:
0
        0
1
        1
2
        1
3
        1
4
        0
5
        0
6
        0
7
        0
        1
8
9
        1
        1
10
11
        1
12
        0
13
        0
14
        0
15
        1
        0
16
17
        1
18
        0
19
        1
20
        0
        1
21
22
        1
        1
23
24
        0
25
        1
26
        0
27
        0
        1
28
29
        0
126
        0
127
        1
128
        1
129
        0
        0
130
        0
131
        0
132
        1
133
134
        0
135
        0
        1
136
137
        0
        0
138
139
        0
        0
140
141
        1
142
        1
        0
143
144
        0
145
        0
146
        1
147
        0
```

```
12/7/21, 11:56 AM
```

150 0

151 1

152 0

153 0

0

155 0

154

Name: Survived, Length: 156, dtype: int64

In [119]:

```
titanic_train.columns
```

Out[119]:

In [120]:

```
titanic_train.describe()
```

Out[120]:

	Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare	
count	156.000000	156.000000	156.000000	126.000000	156.000000	156.000000	156.000000	1
mean	78.500000	0.346154	2.423077	28.141508	0.615385	0.397436	28.109587	
std	45.177428	0.477275	0.795459	14.613880	1.056235	0.870146	39.401047	
min	1.000000	0.000000	1.000000	0.830000	0.000000	0.000000	6.750000	
25%	39.750000	0.000000	2.000000	19.000000	0.000000	0.000000	8.003150	
50%	78.500000	0.000000	3.000000	26.000000	0.000000	0.000000	14.454200	
75%	117.250000	1.000000	3.000000	35.000000	1.000000	0.000000	30.371850	
max	156.000000	1.000000	3.000000	71.000000	5.000000	5.000000	263.000000	
4								

In [179]:

```
new_Pclass = pd.Categorical(titanic_train["Pclass"])
new_Pclass
```

Out[179]:

```
[3, 1, 3, 1, 3, ..., 1, 3, 3, 3, 1]
Length: 156
Categories (3, int64): [1, 2, 3]
```

```
In [193]:
```

```
new_Pclass = pd.Categorical(titanic_train["Pclass"])
new_Pclass
```

Out[193]:

```
[90, dfsf, 90, dfsf, 90, ..., dfsf, 90, 90, 90, dfsf]
Length: 156
Categories (3, object): [dfsf, 9, 90]
```

In [181]:

```
titanic_train["Cabin"].unique() # Check unique cabins
```

Out[181]:

In [7]:

```
# to create a new column to 1st leter of cabin
import numpy as np
char_cabin = titanic_train["Cabin"].astype(str) # Convert data to str

# for i in char_cabin:
# new_cabin.append(i[0])

new_Cabin = [cabin[0] for cabin in char_cabin] # Take first letter in list

new_Cabin = pd.Categorical(new_Cabin)
#new_Cabin
new_Cabin
```

Out[7]:

```
[n, C, n, C, n, ..., C, n, n, n, n]
Length: 156
Categories (8, object): [A, B, C, D, E, F, G, n]
```

In [8]:

```
titanic_train["Cabin1"] = new_Cabin
titanic_train.head()
```

Out[8]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	(
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	_
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	



In [7]:

titanic_train["Age"].isnull() #all index whereage is null

Out[7]:

```
0
       False
       False
1
2
       False
3
       False
4
       False
       . . .
151
       False
152
       False
153
       False
154
        True
155
       False
```

Name: Age, Length: 156, dtype: bool

In [8]:

titanic_train[titanic_train["Age"].isnull()==True]

Out[8]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.
17	18	1	2	Williams, Mr. Charles Eugene	male	NaN	0	0	244373	13.
19	20	1	3	Masselmani, Mrs. Fatima	female	NaN	0	0	2649	7.
26	27	0	3	Emir, Mr. Farred Chehab	male	NaN	0	0	2631	7.
28	29	1	3	O'Dwyer, Miss. Ellen "Nellie"	female	NaN	0	0	330959	7.
29	30	0	3	Todoroff, Mr. Lalio	male	NaN	0	0	349216	7.
31	32	1	1	Spencer, Mrs. William Augustus (Marie Eugenie)	female	NaN	1	0	PC 17569	146.
32	33	1	3	Glynn, Miss. Mary Agatha	female	NaN	0	0	335677	7.
36	37	1	3	Mamee, Mr. Hanna	male	NaN	0	0	2677	7.
42	43	0	3	Kraeff, Mr. Theodor	male	NaN	0	0	349253	7.
45	46	0	3	Rogers, Mr. William John	male	NaN	0	0	S.C./A.4. 23567	8.
46	47	0	3	Lennon, Mr. Denis	male	NaN	1	0	370371	15.
47	48	1	3	O'Driscoll, Miss. Bridget	female	NaN	0	0	14311	7.
48	49	0	3	Samaan, Mr. Youssef	male	NaN	2	0	2662	21.
55	56	1	1	Woolner, Mr. Hugh	male	NaN	0	0	19947	35.
64	65	0	1	Stewart, Mr. Albert A	male	NaN	0	0	PC 17605	27.
65	66	1	3	Moubarek, Master. Gerios	male	NaN	1	1	2661	15.
76	77	0	3	Staneff, Mr. Ivan	male	NaN	0	0	349208	7.
77	78	0	3	Moutal, Mr. Rahamin Haim	male	NaN	0	0	374746	8.

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	
82	83	1	3	McDermott, Miss. Brigdet Delia	female	NaN	0	0	330932	7.
87	88	0	3	Slocovski, Mr. Selman Francis	male	NaN	0	0	SOTON/OQ 392086	8.
95	96	0	3	Shorney, Mr. Charles Joseph	male	NaN	0	0	374910	8.
101	102	0	3	Petroff, Mr. Pastcho ("Pentcho")	male	NaN	0	0	349215	7.
107	108	1	3	Moss, Mr. Albert Johan	male	NaN	0	0	312991	7.
109	110	1	3	Moran, Miss. Bertha	female	NaN	1	0	371110	24.
121	122	0	3	Moore, Mr. Leonard Charles	male	NaN	0	0	A4. 54510	8.
126	127	0	3	McMahon, Mr. Martin	male	NaN	0	0	370372	7.
128	129	1	3	Peter, Miss. Anna	female	NaN	1	1	2668	22.
140	141	0	3	Boulos, Mrs. Joseph (Sultana)	female	NaN	0	2	2678	15.
154	155	0	3	Olsen, Mr. Ole Martin	male	NaN	0	0	Fa 265302	7.
4										•

In []:

In [9]:

```
missing = np.where(titanic_train["Age"].isnull() == True)
missing
```

Out[9]:

```
(array([ 5, 17, 19, 26, 28, 29, 31, 32, 36, 42, 45, 46, 47, 48, 55, 64, 65, 76, 77, 82, 87, 95, 101, 107, 109, 121, 126, 128, 140, 154], dtype=int64),)
```

In [19]:

```
max(titanic_train["Fare"])
```

Out[19]:

263.0

```
In [20]:
```

```
titanic_train["Fare"]==max(titanic_train["Fare"])
Out[20]:
```

_ _

- 0 False
- 1 False
- 2 False
- 3 False
- 4 False
- . .
- 151 False
- 152 False
- 153 False
- 154 False
- 155 False

Name: Fare, Length: 156, dtype: bool

In []:

```
In [132]:
```

np.where(titanic_train["Fare"]==max(titanic_train["Fare"]))

Out[132]:

(array([27, 88]),)

In [191]:

titanic_train.iloc[np.where(titanic_train["Fare"]==max(titanic_train["Fare"]))]

Out[191]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
27	28	0	dfsf	Fortune, Mr. Charles Alexander	male	19.0	3	2	19950	263.0	С
88	89	1	dfsf	Fortune, Miss. Mabel Helen	female	23.0	3	2	19950	263.0	С
4											•

In [30]:

```
np.where(titanic_train["Fare"]==max(titanic_train["Fare"]))
```

Out[30]:

(array([27, 88]),)

```
In [31]:
```

```
len(missing[0])
```

Out[31]:

30

In [194]:

```
titanic_train["Family"] = titanic_train["SibSp"] + titanic_train["Parch"]
titanic_train["Family"]
most_family = np.where(titanic_train["Family"] == max(titanic_train["Family"]))
most_family
```

Out[194]:

(array([59, 71]),)

In [135]:

```
titanic_train["Family"] = titanic_train["SibSp"] + titanic_train["Parch"]
most_family = np.where(titanic_train["Family"] == max(titanic_train["Family"]))
most_family
titanic_train.iloc[most_family]
```

Out[135]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
59	60	0	3	Goodwin, Master. William Frederick	male	11.0	5	2	CA 2144	46.9	NaN
71	72	0	3	Goodwin, Miss. Lillian Amy	female	16.0	5	2	CA 2144	46.9	NaN
4											•

In [196]:

```
import numpy as np
import pandas as pd
labels = ['a','b','c']
my_data = [10,20,30]
arr = np.array(my_data)
d = {'a':10,'b':20,'c':30}
print ("Labels:", labels)
print("My data:", my_data)
print("Dictionary:", d)
pd.Series(my_data, index=labels)
Labels: ['a', 'b', 'c']
My data: [10, 20, 30]
Dictionary: {'a': 10, 'b': 20, 'c': 30}
Out[196]:
     10
а
     20
     30
C
dtype: int64
In [35]:
pd.Series(my_data, index=labels)
Out[35]:
     10
а
     20
b
     30
dtype: int64
In [199]:
print ("\nHolding numerical data\n",'-'*25, sep='')
print(pd.Series(arr)[1])
Holding numerical data
20
```

```
In [198]:
```

```
print ("\nHolding text labels\n",'-'*20, sep='')
print(pd.Series(labels))
```

```
Holding text labels

0 a
1 b
2 c
dtype: object
```

In [201]:

```
d = {'a':"kjhk",'b':20,'c':30}
d.items
pd.DataFrame(d,index = ['a','b','c'])
```

Out[201]:

```
        a
        b
        c

        a
        kjhk
        20
        30

        b
        kjhk
        20
        30

        c
        kjhk
        20
        30
```

In [203]:

```
print ("\nHolding objects from a dictionary\n",'-'*40, sep='')
print(pd.Series([type, sum, max]))
```

```
Holding objects from a dictionary
```

```
-----
```

In [216]:

```
ser1 = pd.Series([1,2,3,4],index = [2,4,6,8])
ser2 = pd.Series([1,2,5,4],['CA', 'OR', 'NV', 'AZ'])
ser1
```

Out[216]:

```
2 1
4 2
6 3
8 4
dtype: int64
```

```
In [217]:
ser1[0:3:2]
Out[217]:
     1
2
     3
dtype: int64
In [42]:
print ("\nIndexing by number (positional value in the series)\n",'-'*52, sep='')
print("Value for CA in ser1:", ser1[0])
print("Value for AZ in ser1:", ser1[3])
print("Value for NV in ser2:", ser2[2])
ser1
Indexing by number (positional value in the series)
Value for CA in ser1: 1
Value for AZ in ser1: 4
Value for NV in ser2: 5
Out[42]:
CA
     1
OR
      2
CO
      3
      4
ΑZ
dtype: int64
In [215]:
print ("\nIndexing by a range\n",'-'*25, sep='')
print ("Value for OR, CO, and AZ in ser1:\n", ser1[0:3:2], sep='')
Indexing by a range
-----
Value for OR, CO, and AZ in ser1:
100
       1
CO
dtype: int64
```

```
In [3]:
```

```
ser1 = pd.Series([1,2,3,4],['CA', 'OR', 'CO', 'CA'])
ser2 = pd.Series([1,2,5,4],['CA', 'NV', 'AZ','OR'])
ser3 = ser1+ser2
ser3
```

Out[3]:

AZ NaN
CA 2.0
CA 5.0
CO NaN
NV NaN
OR 6.0
dtype: float64

In [45]:

```
print ("\nAfter adding the two series, the result looks like this...\n",'-'*59, sep='')
print(ser3)
print("\nPython tries to add values where it finds common index name, and puts NaN where in
```

```
After adding the two series, the result looks like this...
```

AZ NaN
CA 2.0
CA 5.0
CO NaN
NV NaN
OR 6.0

dtype: float64

Python tries to add values where it finds common index name, and puts NaN where indices are missing

In [46]:

```
print ("\nThe idea works even for multiplication...\n",'-'*43, sep='')
print (ser1*ser2)
```

```
The idea works even for multiplication...
```

N.7. NoN

AZ NaN
CA 1.0
CA 4.0
CO NaN
NV NaN
OR 8.0
dtype: float64

In [4]:

```
from numpy.random import randn as rn
#np.random.seed(101)
matrix_data = rn(5,4)
row_labels = ['A','B','C','D','E']
column_headings = ['W','X','Y','Z']
df = pd.DataFrame(matrix_data,row_labels,column_headings)
#print("\nThe data frame looks like\n",'-'*45, sep='')
```

Out[4]:

	W	Х	Y	Z
Α	-0.964125	1.620393	0.862816	0.877327
В	-1.027721	-0.235968	0.395199	0.105760
С	0.638297	0.191965	-0.910558	-0.262456
D	0.361685	-0.391555	-0.986207	0.231495
Ε	-0.090619	-1.711412	0.552947	-1.348353

In [5]:

```
df.iloc[2] #index position
```

Out[5]:

```
0.638297
W
Χ
     0.191965
    -0.910558
Υ
Ζ
    -0.262456
```

Name: C, dtype: float64

In [8]:

```
df.loc['A'] # label of row
```

Out[8]:

```
-0.964125
W
     1.620393
Χ
Υ
     0.862816
     0.877327
```

Name: A, dtype: float64

```
In [6]:
```

```
print("\nType of the pair of columns: ", type(df[['X','Z']]), sep='')
print ("\nSo, for more than one column, the object turns into a DataFrame")
```

Type of the pair of columns: <class 'pandas.core.frame.DataFrame'>

So, for more than one column, the object turns into a DataFrame

In [7]:

```
print("\nThe 'X' column accessed by DOT method (NOT recommended)\n",'-'*55, sep='')
print(df["X"])
```

The 'X' column accessed by DOT method (NOT recommended)

```
A 1.620393
B -0.235968
```

C 0.191965

D -0.391555

E -1.711412

Name: X, dtype: float64

In [9]:

A column is created by assigning it in relation to an existing column

```
      W
      X
      Y
      Z
      New
      New
      (Sum of X and Z)

      A -0.964125
      1.620393
      0.862816
      0.877327
      2.497720
      2.497720

      B -1.027721
      -0.235968
      0.395199
      0.105760
      -0.130209
      -0.130209

      C 0.638297
      0.191965
      -0.910558
      -0.262456
      -0.070492
      -0.070492

      D 0.361685
      -0.391555
      -0.986207
      0.231495
      -0.160059
      -0.160059

      E -0.090619
      -1.711412
      0.552947
      -1.348353
      -3.059765
      -3.059765
```

In [11]:

```
df.iloc[2,1]
```

Out[11]:

0.19196476782656316

In [14]:

```
df.iloc[1:4 , 2:4]
```

Out[14]:

	Y	
В	0.395199	0.105760

- **C** -0.910558 -0.262456
- **D** -0.986207 0.231495

In [52]:

df

Out[52]:

	W	Х	Y	Z	New	New (Sum of X and Z)
-	0.150705	-0.446910	1.669238	-1.138071	-1.584981	-1.584981
E	0.462929	0.129871	0.128628	1.865870	1.995741	1.995741
C	0.783021	0.561695	-1.608215	-0.997146	-0.435451	-0.435451
	1.124639	0.626070	-0.129314	-0.075665	0.550405	0.550405
E	0.133687	1.988570	0.416920	-0.729497	1.259073	1.259073

In [17]:

print("\nA column is dropped by using df.drop() method\n",'-'*55, sep=' ') df.drop("C",inplace=True) # Notice the axis=1 option, axis = 0 is default, so one has to ch

A column is dropped by using df.drop() method

In [20]:

df

Out[20]:

	W	X	Υ	Z	New	New (Sum of X and Z)
Α	-0.964125	1.620393	0.862816	0.877327	2.497720	2.497720
В	-1.027721	-0.235968	0.395199	0.105760	-0.130209	-0.130209
D	0.361685	-0.391555	-0.986207	0.231495	-0.160059	-0.160059
F	-0 090619	-1 711412	0 552947	-1 348353	-3 059765	-3 059765

In [22]:

```
df1=df.drop('X',axis =1) #1 =coloum , \theta = row #print("\nA row (index) is dropped by using df.drop() method and axis=\theta\n",'-'*65, sep='') print(df1)
```

A row (index) is dropped by using df.drop() method and axis=0

 W
 Y
 Z
 New
 New (Sum of X and Z)

 A -0.964125
 0.862816
 0.877327
 2.497720
 2.497720

 B -1.027721
 0.395199
 0.105760
 -0.130209
 -0.130209

 D 0.361685
 -0.986207
 0.231495
 -0.160059

 E -0.090619
 0.552947
 -1.348353
 -3.059765

In [56]:

print("\nAn in-place change can be done by making inplace=True in the drop method\n",'-'*75 df.drop('New (Sum of X and Z)', axis=1, inplace=True) print(df)

An in-place change can be done by making inplace=True in the drop method

```
W X Y Z New
A 0.150705 -0.446910 1.669238 -1.138071 -1.584981
B 0.462929 0.129871 0.128628 1.865870 1.995741
D 1.124639 0.626070 -0.129314 -0.075665 0.550405
E 0.133687 1.988570 0.416920 -0.729497 1.259073
```

In [11]:

```
### Selecting/indexing Rows
#* Label-based 'loc' method
#* Index (numeric) 'iloc' method
df
```

Out[11]:

	w	X	Υ	Z
Α	0.463014	2.351775	0.355781	0.127966
В	0.783331	1.794840	-0.892002	-0.596769
С	0.119252	-0.020044	-2.072349	2.130845
D	0.128286	0.206469	-1.227162	1.602765
Е	-0.197037	1.957458	0.238325	1.542864

```
In [58]:
```

```
print("\nLabel-based 'loc' method can be used for selecting row(s)\n",'-'*60, sep='')
print("\nSingle row\n")
print(df.iloc[3])
```

Label-based 'loc' method can be used for selecting row(s)

Single row

```
W 0.133687
X 1.988570
Y 0.416920
Z -0.729497
New 1.259073
Name: E, dtype: float64
```

In [59]:

```
print("\nMultiple rows\n")
print(df.loc[['B','C']]) #index first and last row
```

Multiple rows

```
W X Y Z New
B 0.462929 0.129871 0.128628 1.86587 1.995741
C NaN NaN NaN NaN NaN
```

/Users/sudhanshukumar/anaconda3/lib/python3.7/site-packages/ipykernel_launch er.py:2: FutureWarning:

Passing list-likes to .loc or [] with any missing label will raise KeyError in the future, you can use .reindex() as an alternative.

See the documentation here:

https://pandas.pydata.org/pandas-docs/stable/indexing.html#deprecate-loc-rei ndex-listlike (https://pandas.pydata.org/pandas-docs/stable/indexing.html#deprecate-loc-reindex-listlike)

```
In [60]:
```

```
print("\nIndex position based 'iloc' method can be used for selecting row(s)\n",'-'*70, sep
print("\nSingle row\n")
print(df.iloc[2])
```

```
print("\nMultiple rows\n")
print(df.iloc[[1,2]])
```

Multiple rows

```
W X Y Z New
B 0.462929 0.129871 0.128628 1.865870 1.995741
D 1.124639 0.626070 -0.129314 -0.075665 0.550405
```

In [62]:

```
#### Subsetting DataFrame
```

In [63]:

```
matrix_data = rn(5,4)
row_labels = ['A','B','C','D','E']
column_headings = ['W','X','Y','Z']
df = pd.DataFrame(data=matrix_data, index=row_labels, columns=column_headings)
```

```
In [64]:
```

```
print("\nThe DatFrame\n",'-'*45, sep='')
print(df)
```

The DatFrame

```
W X Y Z
A -0.320697 1.586737 -0.962361 0.984044
B -0.236481 -0.520044 -0.137304 -1.719868
C -0.415743 -0.280577 1.293790 -0.495861
D -0.564725 -1.331318 -1.098884 0.385147
E 1.079266 1.101438 1.097766 0.452282
```

In [65]:

```
print("\nElement at row 'B' and column 'Y' is\n")
print(df.loc[['B','C'],['Y','W']])
```

Element at row 'B' and column 'Y' is

```
Y W
B -0.137304 -0.236481
C 1.293790 -0.415743
```

In [66]:

```
print("\nSubset comprising of rows B and D, and columns W and Y, is\n")
df.iloc[[1,2,3],[0,1]]
```

Subset comprising of rows B and D, and columns W and Y, is

Out[66]:

```
        W
        X

        B
        -0.236481
        -0.520044

        C
        -0.415743
        -0.280577

        D
        -0.564725
        -1.331318
```

In [67]:

```
print(df.loc[['B','C'],])
```

```
W X Y Z
B -0.236481 -0.520044 -0.137304 -1.719868
C -0.415743 -0.280577 1.293790 -0.495861
```

```
In [68]:
```

```
print("\nThe DataFrame\n",'-'*45, sep='')
print(df)
```

The DataFrame

```
W X Y Z
A -0.320697 1.586737 -0.962361 0.984044
B -0.236481 -0.520044 -0.137304 -1.719868
C -0.415743 -0.280577 1.293790 -0.495861
D -0.564725 -1.331318 -1.098884 0.385147
E 1.079266 1.101438 1.097766 0.452282
```

In [23]:

```
print("\nBoolean DataFrame(s) where we are checking if the values are greater than 0\n",'-' print(df>0)
```

Boolean DataFrame(s) where we are checking if the values are greater than 0

```
Z
                Υ
            Χ
                             New New (Sum of X and Z)
 False
         True
               True
                            True
                      True
                                               True
Α
  False False
               True
                      True False
                                               False
D
   True False False
                     True False
                                               False
 False False True False False
                                               False
```

In [70]:

```
print("\n")
print(df.loc[['A','B','C']]>0)
```

```
W X Y Z
A False True False True
B False False False
C False False True False
```

In [71]:

```
booldf = df>0
print("\nDataFrame indexed by boolean dataframe\n",'-'*45, sep='')
print()df[booldf]
```

DataFrame indexed by boolean dataframe

```
Χ
                            Υ
         W
                           NaN 0.984044
            1.586737
       NaN
В
       NaN
                 NaN
                           NaN
                                     NaN
C
                      1.293790
       NaN
                 NaN
                                     NaN
D
       NaN
                 NaN
                           NaN 0.385147
  1.079266 1.101438 1.097766 0.452282
```

In [72]:

```
import pandas as pd
import numpy as np
matrix_data = np.matrix('22,66,140;42,70,148;30,62,125;35,68,160;25,62,152')
row_labels = ['A','B','C','D','E']
column_headings = ['Age', 'Height', 'Weight']
matrix_data
```

Out[72]:

In [73]:

```
df = pd.DataFrame(data=matrix_data, index=row_labels, columns=column_headings)
print("\nA new DataFrame\n",'-'*25, sep='')
print(df)
```

A new DataFrame

Age Height Weight 22 66 140 Α 42 70 148 В C 30 62 125 D 35 68 160 25 62 152

In [74]:

```
df[df['Height']>65]
```

Out[74]:

	Age	Height	Weight
Α	22	66	140
В	42	70	148
D	35	68	160

```
12/7/21, 11:56 AM
                                            Pandas data manipulation - Jupyter Notebook
  In [75]:
  print("\nRows with Height > 65 inch\n",'-'*35, sep='')
  print(df[df['Height']>65])
  Rows with Height > 65 inch
     Age Height Weight
      22
             66
                      140
  Α
              70
  В
      42
                      148
  D
      35
              68
                      160
  In [76]:
 df['Height']>65
 Out[76]:
        True
  Α
  В
        True
  C
       False
  D
       True
  Ε
       False
  Name: Height, dtype: bool
  In [77]:
  booldf1 = df['Height']>65
 booldf2 = df['Weight']>145
  In [78]:
  print("\nRows with Height > 65 inch and Weight >145 lbs\n",'-'*55, sep='')
  print(df[(booldf1) & (booldf2)])
  Rows with Height > 65 inch and Weight >145 lbs
     Age Height Weight
              70
      42
                      148
  В
      35
  D
              68
                      160
```

```
In [79]:
```

```
print("\nDataFrame with only Age and Weight columns whose Height > 65 inch\n",'-'*68, sep='
print(df[booldf1][['Age', 'Weight']])
```

```
DataFrame with only Age and Weight columns whose Height > 65 inch
```

```
Weight
   Age
Δ
    22
            140
    42
            148
D
    35
            160
```

```
In [80]:
```

```
matrix_data = np.matrix('22,66,140;42,70,148;30,62,125;35,68,160;25,62,152')
row_labels = ['A','B','C','D','E']
column_headings = ['Age', 'Height', 'Weight']
```

In [81]:

```
df = pd.DataFrame(data=matrix_data, index=row_labels, columns=column_headings)
print("\nThe DataFrame\n",'-'*25, sep='')
print(df)
```

The DataFrame

	Age	Height	Weight
Α	22	66	140
В	42	70	148
C	30	62	125
D	35	68	160
Ε	25	62	152

In [82]:

```
print("\nAfter resetting index\n",'-'*35, sep='')
print(df.reset_index())
```

After resetting index

	index	Age	Height	Weight
0	Α	22	66	140
1	В	42	70	148
2	C	30	62	125
3	D	35	68	160
4	Е	25	62	152

In [83]:

```
print("\nAfter resetting index with 'drop' option TRUE\n",'-'*45, sep='')
print(df.reset_index(drop=True))
"Student Teacher Engineer Doctor Nurse".split()
```

After resetting index with 'drop' option TRUE

```
Age Height Weight
0
    22
             66
                     140
1
    42
             70
                    148
2
    30
             62
                    125
3
    35
             68
                     160
    25
             62
                     152
```

Out[83]:

```
['Student', 'Teacher', 'Engineer', 'Doctor', 'Nurse']
```

In [84]:

```
print("\nAdding a new column 'Profession'\n",'-'*45, sep='')
df['Profession'] = "Student Teacher Engineer Doctor Nurse".split()
print(df)
```

Adding a new column 'Profession'

	Age	Height	Weight	Profession	
Α	22	66	140	Student	
В	42	70	148	Teacher	
C	30	62	125	Engineer	
D	35	68	160	Doctor	
Ε	25	62	152	Nurse	

In [85]:

```
print("\nSetting 'Profession' column as index\n",'-'*45, sep='')
print (df.set_index('Profession'))
```

Setting 'Profession' column as index

Age	Height	Weight
22	66	140
42	70	148
30	62	125
35	68	160
25	62	152
	22 42 30 35	22 66 42 70 30 62 35 68

In [14]:

```
#multi-indexing
# Index Levels
outside = ['G1','G1','G2','G2','G2']
inside = [1,2,3,1,2,3]
hier_index = list(zip(outside,inside))
```

In [15]:

```
print("\nTuple pairs after the zip and list command\n",'-'*45, sep='')
print(hier_index)
```

```
In [24]:
```

```
hier_index = pd.MultiIndex.from_tuples(hier_index)
print("\nIndex hierarchy\n",'-'*25, sep='')
print(hier_index)
```

```
Index hierarchy
```

```
MultiIndex([('G1', 1), ('G1', 2), ('G1', 3), ('G2', 1), ('G2', 2), ('G2', 3)],
```

In [17]:

```
print("\nIndex hierarchy type\n",'-'*25, sep='')
print(type(hier_index))
```

```
Index hierarchy type
-----
<class 'pandas.core.indexes.multi.MultiIndex'>
```

In [18]:

```
print("\nCreating DataFrame with multi-index\n",'-'*37, sep='')
#np.random.seed(101)
df1 = pd.DataFrame(data=np.round(rn(6,3)), index= hier_index, columns= ['A','B','C'])
print(df1)
```

Creating DataFrame with multi-index

```
A B C
G1 1 -1.0 -0.0 1.0
2 2.0 1.0 1.0
3 1.0 -2.0 0.0
G2 1 -0.0 -0.0 -1.0
2 1.0 -0.0 -0.0
3 1.0 -1.0 1.0
```

In [91]:

```
#cross tabulation like pivot table
```

```
In [92]:
print("\nGrabbing a cross-section from outer level\n",'-'*45, sep='')
print(df1.xs('G1'))
Grabbing a cross-section from outer level
     Α
          В
1 -0.0 1.0 -0.0
2 2.0 -1.0 0.0
3 -0.0 -0.0 1.0
In [19]:
df1.loc['G1']
Out[19]:
    Α
            С
         В
1 -1.0 -0.0 1.0
  2.0 1.0 1.0
   1.0 -2.0 0.0
In [22]:
df1.loc['G1'].loc[3,['B','C']]
Out[22]:
    -2.0
В
C
     0.0
Name: 3, dtype: float64
In [26]:
#do it in 5
l1 = ['a','a','a','b','b','c','c','c']
12 = [1,2,3,1,2,3,1,2,3]
13 = [1,2,3,4,5,6,7,8,9]
1 = list(zip(11,12,13))
g = pd.MultiIndex.from tuples(1)
g
Out[26]:
MultiIndex([('a', 1, 1),
            ('a', 2, 2),
            ('a', 3, 3),
            ('b', 1, 4),
            ('b', 2, 5),
            ('b', 3, 6),
            ('c', 1, 7),
            ('c', 2, 8),
            ('c', 3, 9)],
```

In [27]:

```
df= pd.DataFrame(data=np.round(rn(9,4)),index=g)
df
```

Out[27]:

			0	1	2	3
а	1	1	-1.0	0.0	-0.0	1.0
	2	2	-1.0	3.0	2.0	1.0
	3	3	0.0	-0.0	0.0	-0.0
b	1	4	1.0	-1.0	-2.0	-0.0
	2	5	1.0	-0.0	1.0	-1.0
	3	6	-1.0	1.0	0.0	1.0
С	1	7	-0.0	1.0	1.0	-1.0
	2	8	1.0	-1.0	-0.0	-0.0
	3	9	0.0	-0.0	1.0	0.0