

```
import pandas as pd
import numpy as np

names = ['Arun',"Varun",'Ram',"Mohon"]
age=[29,21,32,51]

df = pd.DataFrame(list(zip(names,age)))
df
```

	0	1
0	Arun	29
1	Varun	21
2	Ram	32
3	Mohon	51

```
df = pd.DataFrame(list(zip(names,age)),columns=["Name",'Val'])
df
```

	Name	Val
0	Arun	29
1	Varun	21
2	Ram	32
3	Mohon	51

```
lst =[
    {'name':"Arun",'age':29,"gender":"M"},
    {'name':"Varun",'age':21,"gender":"M"},
    {'name':"Ram",'age':32,"gender":"M"}
]
```

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

	name	age	gender
0	Arun	29	M
1	Varun	21	M
2	Ram	32	M

```
df =pd.read_csv('harry.csv')
df
```

	Unnamed: 0.3	Unnamed: 0.2	Unnamed: 0.1	Unnamed: 0	Train No
Name \					
0	0	0	0	0	12655
Nazmul					
1	1	1	1	1	46645
Siam					
2	2	2	2	2	54646
Sajid					

	Marks	City
0	90	Madaripur

```
1      80   Barishal
2      95   Jheniadh
```

```
df = pd.DataFrame #Blank DataFrame
df
```

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

```
df = pd.DataFrame(columns=['name', 'age']) #Blank DataFrame
df
```

```
Empty DataFrame
Columns: [name, age]
Index: []
```

```
df = pd.read_csv('name_age.csv')
df
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m
4	Ram	32	7/11/1988	m
5	Radha	23	6/9/1997	f
6	Mohan	51	3/3/1969	m
7	Devi	20	5/1/2000	f
8	Nidhi	29	10/1/1991	f

```
df.shape #output is row by column ,Output is a tuple
```

```
(9, 4)
```

```
rows,column =df.shape
```

```
rows
```

```
9
```

```
column
```

```
4
```

```
df.head()
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m
4	Ram	32	7/11/1988	m

```
df.head(3)
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f

```
df.tail()
```

	name	age	dob	gender
4	Ram	32	7/11/1988	m
5	Radha	23	6/9/1997	f
6	Mohan	51	3/3/1969	m
7	Devi	20	5/1/2000	f
8	Nidhi	29	10/1/1991	f

```
df[5:8] # include row #5 and exclude row #8
```

	name	age	dob	gender
5	Radha	23	6/9/1997	f
6	Mohan	51	3/3/1969	m
7	Devi	20	5/1/2000	f

```
df[:] #All rows
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m
4	Ram	32	7/11/1988	m
5	Radha	23	6/9/1997	f
6	Mohan	51	3/3/1969	m
7	Devi	20	5/1/2000	f
8	Nidhi	29	10/1/1991	f

```
df.columns
```

```
Index(['name', 'age', 'dob', 'gender'], dtype='object')
```

```
# Read the CSV file
```

```
df = pd.read_csv('name_age.csv') # Replace with your file path
df
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m
4	Ram	32	7/11/1988	m
5	Radha	23	6/9/1997	f
6	Mohan	51	3/3/1969	m

```
7  Devi    20    5/1/2000    f
8  Nidhi   29    10/1/1991   f
```

```
df.columns
```

```
Index(['name', 'age', 'dob', 'gender'], dtype='object')
```

```
df
```

```
   name  age    dob gender
0  Rita   23  20/02/97     f
1  Arun   29   1/7/1991     m
2  Sita   14   7/7/2006     f
3  Varun  21   1/5/1999     m
4   Ram   32   7/11/1988     m
5  Radha  23   6/9/1997     f
6  Mohan  51   3/3/1969     m
7  Devi   20   5/1/2000     f
8  Nidhi  29   10/1/1991     f
```

```
df["name"]
```

```
0    Rita
1    Arun
2    Sita
3  Varun
4     Ram
5  Radha
6  Mohan
7    Devi
8  Nidhi
```

```
Name: name, dtype: object
```

```
df["age"]
```

```
0    23
1    29
2    14
3    21
4    32
5    23
6    51
7    20
8    29
```

```
Name: age, dtype: int64
```

```
df.name
```

```
0    Rita
1    Arun
2    Sita
```

```
3    Varun
4      Ram
5    Radha
6    Mohan
7     Devi
8    Nidhi
Name: name, dtype: object
```

```
df.age
```

```
0    23
1    29
2    14
3    21
4    32
5    23
6    51
7    20
8    29
Name: age, dtype: int64
```

```
df[['name', 'age']].head() #Column name for accessing column
                             #Range for accessing row
```

	name	age
0	Rita	23
1	Arun	29
2	Sita	14
3	Varun	21
4	Ram	32

```
type(df)
```

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

```
type(df["age"])
```

```
pandas.core.series.Series
```

```
df['age'].max()
```

```
51
```

```
df['age'].min()
```

```
14
```

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

```
26.88888888888889
```

```
df["age"].std()
```

```
10.576441325470071
```

```
df.describe()
```

	age
count	9.000000
mean	26.888889
std	10.576441
min	14.000000
25%	21.000000
50%	23.000000
75%	29.000000
max	51.000000

```
df[df['age']>30]
```

	name	age	dob	gender
4	Ram	32	7/11/1988	m
6	Mohan	51	3/3/1969	m

```
df[df['age']==df['age'].min()]
```

	name	age	dob	gender
2	Sita	14	7/7/2006	f

```
df['name'][df['age'] > 30]
```

```
4      Ram
6      Mohan
```

```
Name: name, dtype: object
```

```
df[['name','dob','age']][df['age'] < 30] #multiple bracket for multiple column
```

#single bracket for accessing

single column

	name	dob	age
0	Rita	20/02/97	23
1	Arun	1/7/1991	29
2	Sita	7/7/2006	14
3	Varun	1/5/1999	21
5	Radha	6/9/1997	23
7	Devi	5/1/2000	20
8	Nidhi	10/1/1991	29

```
# accessing row with loc and iloc
```

```
# Loc takes the index value and gives the result for that index, whereas iloc
```

```
# takes the position of an index and gives the row.
```

```
df.loc[0]
```

```
name      Rita
age       23
dob      20/02/97
gender    f
Name: 0, dtype: object
```

```
df.iloc[1]
```

```
name      Arun
age       29
dob      1/7/1991
gender    m
Name: 1, dtype: object
```

```
df.index
```

```
RangeIndex(start=0, stop=9, step=1)
```

```
#df=df.set_index('name')
```

```
df.set_index('name',inplace=True)
```

```
df.head()
```

	age	dob	gender
name			
Rita	23	20/02/97	f
Arun	29	1/7/1991	m
Sita	14	7/7/2006	f
Varun	21	1/5/1999	m
Ram	32	7/11/1988	m

```
df.loc["Arun"]
```

```
age       29
dob      1/7/1991
gender    m
Name: Arun, dtype: object
```

```
df.iloc[0]
```

```
age       23
dob      20/02/97
gender    f
Name: Rita, dtype: object
```

```
df=pd.read_csv('name_age.csv',header=None)
df
```

	0	1	2	3
0	name	age	dob	gender
1	Rita	23	20/02/97	f

2	Arun	29	1/7/1991	m
3	Sita	14	7/7/2006	f
4	Varun	21	1/5/1999	m
5	Ram	32	7/11/1988	m
6	Radha	23	6/9/1997	f
7	Mohan	51	3/3/1969	m
8	Devi	20	5/1/2000	f
9	Nidhi	29	10/1/1991	f

Adding custom header

```
df=pd.read_csv('name_age.csv',
names=['Name','Age','dateOfBirth','GENDER'])
df.head()
```

	Name	Age	dateOfBirth	GENDER
0	name	age	dob	gender
1	Rita	23	20/02/97	f
2	Arun	29	1/7/1991	m
3	Sita	14	7/7/2006	f
4	Varun	21	1/5/1999	m

'nrows' will be the number of rows we want to see

```
df = pd.read_csv('name_age.csv',nrows=4)
df.head()
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m

```
df= pd.read_csv('prac2.csv')
df.head()
```

	name	Unnamed: 1	Unnamed: 2	Unnamed: 3
0	NaN	NaN	NaN	NaN
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m
4	Ram	32	7/11/1988	not avaialable

```
df = pd.read_csv('prac2.csv',skiprows=2) #skip first two rows
df
```

	Arun	29	1/7/1991	m
0	Sita	14	7/7/2006	f
1	Varun	21	1/5/1999	m
2	Ram	32	7/11/1988	not avaialable
3	Radha	na	6/9/1997	f
4	Mohan	51	3/3/1969	m


```
5   Devi   20   5/1/2000   f
6   Nidhi  29   10/1/1991  f
```

```
df = pd.read_csv("name_age.csv")
df
```

```
   name  age   dob gender
0   Rita   23  20/02/97     f
1   Arun   29  1/7/1991     m
2   Sita   14  7/7/2006     f
3  Varun   21  1/5/1999     m
4    Ram   32  7/11/1988     m
5  Radha   23  6/9/1997     f
6  Mohan   51  3/3/1969     m
7   Devi   20  5/1/2000     f
8  Nidhi   29  10/1/1991     f
```

```
df = pd.read_csv("name_age.csv", skiprows=2) #Skip two rows
df
```

```
   Arun   29  1/7/1991  m
0   Sita   14  7/7/2006  f
1  Varun   21  1/5/1999  m
2    Ram   32  7/11/1988  m
3  Radha   23  6/9/1997  f
4  Mohan   51  3/3/1969  m
5   Devi   20  5/1/2000  f
6  Nidhi   29  10/1/1991  f
```

```
df = pd.read_csv("name_age.csv", header = 2) #Skip two rows
df
```

```
   Arun   29  1/7/1991  m
0   Sita   14  7/7/2006  f
1  Varun   21  1/5/1999  m
2    Ram   32  7/11/1988  m
3  Radha   23  6/9/1997  f
4  Mohan   51  3/3/1969  m
5   Devi   20  5/1/2000  f
6  Nidhi   29  10/1/1991  f
```

```
df = pd.read_csv('prac2.csv')
df
```

```
   name  Unnamed: 1  Unnamed: 2  Unnamed: 3
0   NaN          NaN          NaN          NaN
1  Arun           29  1/7/1991           m
2  Sita           14  7/7/2006           f
3  Varun           21  1/5/1999           m
4   Ram           32  7/11/1988  not avaialable
5  Radha           na  6/9/1997           f
```

6	Mohan	51	3/3/1969	m
7	Devi	20	5/1/2000	f
8	Nidhi	29	10/1/1991	f

```
df = pd.read_csv('prac2.csv',na_values=['na','not avaialable'])
df
```

	name	Unnamed: 1	Unnamed: 2	Unnamed: 3
0	NaN	NaN	NaN	NaN
1	Arun	29.0	1/7/1991	m
2	Sita	14.0	7/7/2006	f
3	Varun	21.0	1/5/1999	m
4	Ram	32.0	7/11/1988	NaN
5	Radha	NaN	6/9/1997	f
6	Mohan	51.0	3/3/1969	m
7	Devi	20.0	5/1/2000	f
8	Nidhi	29.0	10/1/1991	f

```
df = pd.read_csv('name_age.csv')
df
```

cleaning has been applied to a specific column

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m
2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m
4	Ram	32	7/11/1988	m
5	Radha	23	6/9/1997	f
6	Mohan	51	3/3/1969	m
7	Devi	20	5/1/2000	f
8	Nidhi	29	10/1/1991	f

```
df.to_csv('newCSV.csv') #df is created a new file newCSV
```

```
!dir "newCSV.csv"
```

```
Volume in drive C is Acer
Volume Serial Number is D6B1-BC2D
```

```
Directory of C:\Users\USER
```

```
05/14/2025  08:18 PM                225 newCSV.csv
               1 File(s)                  225 bytes
               0 Dir(s)  76,663,754,752 bytes free
```

```
df
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1/7/1991	m

2	Sita	14	7/7/2006	f
3	Varun	21	1/5/1999	m
4	Ram	32	7/11/1988	m
5	Radha	23	6/9/1997	f
6	Mohan	51	3/3/1969	m
7	Devi	20	5/1/2000	f
8	Nidhi	29	10/1/1991	f

```
df.to_csv('newCSV.csv', index=False)
df.to_csv()
```

```
',name,age,dob,gender\r\n0,Rita,23,20/02/97,f\r\n1,Arun,29,1/7/1991,m\r\n2,Sita,14,7/7/2006,f\r\n3,Varun,21,1/5/1999,m\r\n4,Ram,32,7/11/1988,m\r\n5,Radha,23,6/9/1997,f\r\n6,Mohan,51,3/3/1969,m\r\n7,Devi,20,5/1/2000,f\r\n8,Nidhi,29,10/1/1991,f\r\n'
```

```
df = pd.read_csv('newCSV.csv')
print(df.to_csv(index=False))
```

```
name,age,dob,gender
Rita,23,20/02/97,f
Arun,29,1/7/1991,m
Sita,14,7/7/2006,f
Varun,21,1/5/1999,m
Ram,32,7/11/1988,m
Radha,23,6/9/1997,f
Mohan,51,3/3/1969,m
Devi,20,5/1/2000,f
Nidhi,29,10/1/1991,f
```

```
df.to_csv("newCSV.csv", columns=["name", "age", 'dob'], index=False)
df_new = pd.read_csv('newCSV.csv')
df_new.head()
```

	name	age	dob
0	Rita	23	20/02/97
1	Arun	29	1/7/1991
2	Sita	14	7/7/2006
3	Varun	21	1/5/1999
4	Ram	32	7/11/1988

```
df.to_csv("newCSV.csv", header=False, index=False)
df_new = pd.read_csv('newCSV.csv')
df_new
```

	Rita	23	20/02/97	f
0	Arun	29	1/7/1991	m

1	Sita	14	7/7/2006	f
2	Varun	21	1/5/1999	m
3	Ram	32	7/11/1988	m
4	Radha	23	6/9/1997	f
5	Mohan	51	3/3/1969	m
6	Devi	20	5/1/2000	f
7	Nidhi	29	10/1/1991	f

```
import pandas as pd
```

```
# Sample DataFrame
```

```
data = {'Name': ['Alice', 'Bob', 'Charlie'],
        'Age': [25, 30, 35],
        'City': ['NY', 'LA', 'Chicago']}
```

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

```
# Export WITHOUT headers
```

```
df.to_csv("no_headers.csv", index=False, header=False)
```

```
# Show file content
```

```
print("CSV WITHOUT headers:")
with open("no_headers.csv") as f:
    print(f.read())
```

```
CSV WITHOUT headers:
```

```
Alice,25,NY
```

```
Bob,30,LA
```

```
Charlie,35,Chicago
```

```
# Export WITH headers (default behavior)
```

```
df.to_csv("with_headers.csv", index=False) # header=True is default
```

```
# Show file content
```

```
print("CSV WITH headers:")
with open("with_headers.csv") as f:
    print(f.read())
```

```
CSV WITH headers:
```

```
Name,Age,City
```

```
Alice,25,NY
```

```
Bob,30,LA
```

```
Charlie,35,Chicago
```

```
df_new = pd.read_excel('prac3.xlsx')
```

```
df_new.head()
```

	name	age	dob	gender
0	Rita	23	2015-01-01 00:00:00	f
1	Arun	29	2017-02-14 00:00:00	m

2	Sita	14	1999-07-07 00:00:00	f
3	Varun	21	4//5/8	m
4	Ram	32	1997-09-09 00:00:00	m

```
df_new = pd.read_excel('prac3.xlsx', 'Sheet2')
df_new
```

	Name	dob	class
0	AB	2025-08-07	8
1	CD	2025-04-07	9

```
def changeClass(cell):
    if cell==8:
        return 'Eight'
    elif cell==9:
        return 'Nine'
    else:
        return 'NA'
```

```
df_new = pd.read_excel('prac3.xlsx', sheet_name='Sheet2', converters ={
    'class' : changeClass
})
df_new.head()
```

	Name	dob	class
0	AB	2025-08-07	Eight
1	CD	2025-04-07	Nine

```
df_new= pd.read_excel("newExcel.xlsx", "Sheet2")
df_new
```

```
Empty DataFrame
Columns: []
Index: []
```

```
df_new.to_excel("newExcel.xlsx", sheet_name='Sheet2', index=False, header
=False)
```

```
df_new
```

```
#overwrite the entire file
```

```
Empty DataFrame
Columns: []
Index: []
```

```
df1=pd.read_excel('prac4.xlsx', sheet_name='Sheet1')
df1.head()
```

	name	age	dob	gender
0	Rita	23	20/02/97	f
1	Arun	29	1991-01-07 00:00:00	m
2	Sita	14	2006-07-07 00:00:00	f

3	Varun	21	1999-01-05 00:00:00	m
4	Ram	32	1988-07-11 00:00:00	m

```
df2=pd.read_excel('prac4.xlsx',sheet_name='Sheet2')
df2.head()
```

	name	age		dob	gender
0	Rita	23		20/02/97	f
1	Arun	29	1991-01-07 00:00:00		m
2	Sita	14	2006-07-07 00:00:00		f

```
with pd.ExcelWriter("multipleSheet.xlsx") as writer:
    df1.to_excel(writer,sheet_name='Sheet1',index=False)
    df2.to_excel(writer,sheet_name='Sheet2',index=False)
```

```
print(df1.to_csv(index=False))
```

```
name,age,dob,gender
Rita,23,20/02/97,f
Arun,29,1991-01-07 00:00:00,m
Sita,14,2006-07-07 00:00:00,f
Varun,21,1999-01-05 00:00:00,m
Ram,32,1988-07-11 00:00:00,m
Radha,23,1997-06-09 00:00:00,f
Mohan,51,1969-03-03 00:00:00,m
Devi,20,2000-05-01 00:00:00,f
Nidhi,29,1991-10-01 00:00:00,f
```

```
df = pd.read_excel('multipleSheet.xlsx',sheet_name='Sheet1')
df
```

	name	age		dob	gender
0	Rita	23		20/02/97	f
1	Arun	29	1991-01-07 00:00:00		m
2	Sita	14	2006-07-07 00:00:00		f
3	Varun	21	1999-01-05 00:00:00		m
4	Ram	32	1988-07-11 00:00:00		m
5	Radha	23	1997-06-09 00:00:00		f
6	Mohan	51	1969-03-03 00:00:00		m
7	Devi	20	2000-05-01 00:00:00		f
8	Nidhi	29	1991-10-01 00:00:00		f

```
df.to_csv('test1.txt',sep='\t',index=False)
```

```
df=pd.read_csv('test1.txt')
df
```

	name\tage\tdob\tgender
0	Rita\t23\t20/02/97\tf

```

1  Arun\t29\t1991-01-07 00:00:00\tm
2  Sita\t14\t2006-07-07 00:00:00\tf
3  Varun\t21\t1999-01-05 00:00:00\tm
4   Ram\t32\t1988-07-11 00:00:00\tm
5  Radha\t23\t1997-06-09 00:00:00\tf
6  Mohan\t51\t1969-03-03 00:00:00\tm
7   Devi\t20\t2000-05-01 00:00:00\tf
8  Nidhi\t29\t1991-10-01 00:00:00\tf

```

```
df = pd.read_csv('weather.csv', parse_dates=["date"])
```

```
df
```

	date	temperature	windSpeed	status	Unnamed: 4
0	2020-05-06	35.6582	10.788378	sunny	NaN
1	2019-01-30	NaN	NaN	NaN	NaN
2	2023-10-27	30.9343	NaN	rainy	NaN
3	2024-11-29	NaN	6.889682	cloudy	NaN
4	2025-08-11	13.9082	19.012990	rainy	NaN
5	2024-09-09	23.9382	NaN	sunny	NaN

```
df.set_index('date', inplace=True)
```

```
df
```

	temperature	windSpeed	status	Unnamed: 4
date				
2020-05-06	35.6582	10.788378	sunny	NaN
2019-01-30	NaN	NaN	NaN	NaN
2023-10-27	30.9343	NaN	rainy	NaN
2024-11-29	NaN	6.889682	cloudy	NaN
2025-08-11	13.9082	19.012990	rainy	NaN
2024-09-09	23.9382	NaN	sunny	NaN

```
pd.isna(df['temperature'])
```

```
date
```

2020-05-06	False
2019-01-30	True
2023-10-27	False
2024-11-29	True
2025-08-11	False
2024-09-09	False

```
Name: temperature, dtype: bool
```

```
df['temperature'].notna()
```

```
date
```

2020-05-06	True
2019-01-30	False
2023-10-27	True
2024-11-29	False

```
2025-08-11      True
2024-09-09      True
Name: temperature, dtype: bool
```

```
pd.notna(df['windSpeed'])
```

```
date
2020-05-06      True
2019-01-30     False
2023-10-27     False
2024-11-29      True
2025-08-11      True
2024-09-09     False
Name: windSpeed, dtype: bool
```

```
df.isna()
```

	temperature	windSpeed	status	Unnamed: 4
date				
2020-05-06	False	False	False	True
2019-01-30	True	True	True	True
2023-10-27	False	True	False	True
2024-11-29	True	False	False	True
2025-08-11	False	False	False	True
2024-09-09	False	True	False	True

```
df.notna()
```

	temperature	windSpeed	status	Unnamed: 4
date				
2020-05-06	True	True	True	False
2019-01-30	False	False	False	False
2023-10-27	True	False	True	False
2024-11-29	False	True	True	False
2025-08-11	True	True	True	False
2024-09-09	True	False	True	False

```
None==None
```

```
True
```

```
np.nan == np.nan
```

```
False
```

```
np.nan != np.nan
```

```
True
```

```
df['temperature']==np.nan
```



```

date
2020-05-06    False
2019-01-30    False
2023-10-27    False
2024-11-29    False
2025-08-11    False
2024-09-09    False
Name: temperature, dtype: bool

```

```
import pandas as pd
```

```

data = {
    'date': ["20200508", "20200509", "20200510", "20200511",
             "20200512", "20200513"], # Filled missing dates with adjacent days
    'temperature': [35.6582, 30.0, 30.9343, 25.0, np.nan, 23.9382], #
    'windSpeed': [10.788378, 5.0, 8.0, np.nan, 19.012990, 7.0], #
    'status': ["sunny", np.nan, "rainy", "cloudy", np.nan, np.nan] #
}

```

```

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

```

	date	temperature	windSpeed	status
0	20200508	35.6582	10.788378	sunny
1	20200509	30.0000	5.000000	NaN
2	20200510	30.9343	8.000000	rainy
3	20200511	25.0000	NaN	cloudy
4	20200512	NaN	19.012990	NaN
5	20200513	23.9382	7.000000	NaN

```

df['date'] = pd.to_datetime(df['date'], format='%Y%m%d')
df

```

	date	temperature	windSpeed	status
0	2020-05-08	35.6582	10.788378	sunny
1	2020-05-09	30.0000	5.000000	NaN
2	2020-05-10	30.9343	8.000000	rainy
3	2020-05-11	25.0000	NaN	cloudy
4	2020-05-12	NaN	19.012990	NaN
5	2020-05-13	23.9382	7.000000	NaN

```

df.loc[[0,4,5],['date']] = np.nan
df

```

	date	temperature	windSpeed	status
0	NaT	35.6582	10.788378	sunny
1	2020-05-09	30.0000	5.000000	NaN
2	2020-05-10	30.9343	8.000000	rainy

3	2020-05-11	25.0000	NaN	cloudy
4	NaT	NaN	19.012990	NaN
5	NaT	23.9382	7.000000	NaN

```
s =pd.Series(['aa','bb','cc'])
s.loc[0]=None
s.loc[1]=np.nan
s
```

```
0    None
1     NaN
2      cc
dtype: object
```

```
s =pd.Series([11,22,33])
s.loc[0]=None      # missing number always show NaN
s.loc[1]=np.nan    #missing datetime always show NaT
s
```

```
0     NaN
1     NaN
2    33.0
dtype: float64
```

```
df.set_index('date',inplace=True)
df
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	NaN
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	NaN	cloudy
NaT	NaN	19.012990	NaN
NaT	23.9382	7.000000	NaN

```
df.fillna(0)
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	0
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	0.000000	cloudy
NaT	0.0000	19.012990	0
NaT	23.9382	7.000000	0

```
df.fillna({
    'temperature':0,
    'windSpeed':0,
    'status':'sunny'
})
```

```
})
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	sunny
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	0.000000	cloudy
NaT	0.0000	19.012990	sunny
NaT	23.9382	7.000000	sunny

```
df['status'].fillna('new status')
```

date	
NaT	sunny
2020-05-09	new status
2020-05-10	rainy
2020-05-11	cloudy
NaT	new status
NaT	new status

Name: status, dtype: object

```
df.fillna(method='ffill')
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1193302488.py:1:
FutureWarning: DataFrame.fillna with 'method' is deprecated and will
raise in a future version. Use obj.ffill() or obj.bfill() instead.
df.fillna(method='ffill')

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	sunny
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	8.000000	cloudy
NaT	25.0000	19.012990	cloudy
NaT	23.9382	7.000000	cloudy

```
df
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	NaN
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	NaN	cloudy
NaT	NaN	19.012990	NaN
NaT	23.9382	7.000000	NaN

```
df.fillna(method='bfill')
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\2831856154.py:1:
FutureWarning: DataFrame.fillna with 'method' is deprecated and will
raise in a future version. Use obj.ffill() or obj.bfill() instead.
df.fillna(method='bfill')
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	rainy
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	19.012990	cloudy
NaT	23.9382	19.012990	NaN
NaT	23.9382	7.000000	NaN

```
df.fillna(method='ffill',axis='columns')
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1140326825.py:1:
FutureWarning: DataFrame.fillna with 'method' is deprecated and will
raise in a future version. Use obj.ffill() or obj.bfill() instead.
df.fillna(method='ffill',axis='columns')
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1140326825.py:1:
FutureWarning: Downcasting object dtype arrays
on .fillna, .ffill, .bfill is deprecated and will change in a future
version. Call result.infer_objects(copy=False) instead. To opt-in to
the future behavior, set
`pd.set_option('future.no_silent_downcasting', True)`
df.fillna(method='ffill',axis='columns')
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0	5.0	5.0
2020-05-10	30.9343	8.0	rainy
2020-05-11	25.0	25.0	cloudy
NaT	NaN	19.01299	19.01299
NaT	23.9382	7.0	7.0

```
df.fillna(method='bfill',axis='columns')
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3576732574.py:1:
FutureWarning: DataFrame.fillna with 'method' is deprecated and will
raise in a future version. Use obj.ffill() or obj.bfill() instead.
df.fillna(method='bfill',axis='columns')
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3576732574.py:1:
FutureWarning: Downcasting object dtype arrays
on .fillna, .ffill, .bfill is deprecated and will change in a future
version. Call result.infer_objects(copy=False) instead. To opt-in to
the future behavior, set
`pd.set_option('future.no_silent_downcasting', True)`
df.fillna(method='bfill',axis='columns')
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0	5.0	NaN
2020-05-10	30.9343	8.0	rainy
2020-05-11	25.0	cloudy	cloudy
NaT	19.01299	19.01299	NaN
NaT	23.9382	7.0	NaN

df

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	NaN
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	NaN	cloudy
NaT	NaN	19.012990	NaN
NaT	23.9382	7.000000	NaN

```
df.fillna(method='ffill',limit=1)
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\393035141.py:1:
FutureWarning: DataFrame.fillna with 'method' is deprecated and will
raise in a future version. Use obj.ffill() or obj.bfill() instead.

```
df.fillna(method='ffill',limit=1)
```

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	sunny
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	8.000000	cloudy
NaT	25.0000	19.012990	cloudy
NaT	23.9382	7.000000	NaN

df

	temperature	windSpeed	status
date			
NaT	35.6582	10.788378	sunny
2020-05-09	30.0000	5.000000	NaN
2020-05-10	30.9343	8.000000	rainy
2020-05-11	25.0000	NaN	cloudy
NaT	NaN	19.012990	NaN
NaT	23.9382	7.000000	NaN

```
cols = ['temperature', 'windSpeed']
df[cols] = df[cols].fillna(df[cols].mean()) #multiple columns mean at
a time
df
```

	temperature	windSpeed	status
date			
NaT	35.65820	10.788378	sunny
2020-05-09	30.00000	5.000000	NaN
2020-05-10	30.93430	8.000000	rainy
2020-05-11	25.00000	9.960274	cloudy
NaT	29.10614	19.012990	NaN
NaT	23.93820	7.000000	NaN

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

```
9.9602736
```

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

```
29.10614
```

```
df
```

	temperature	windSpeed	status
date			
NaT	35.65820	10.788378	sunny
2020-05-09	30.00000	5.000000	NaN
2020-05-10	30.93430	8.000000	rainy
2020-05-11	25.00000	9.960274	cloudy
NaT	29.10614	19.012990	NaN
NaT	23.93820	7.000000	NaN

```
import pandas as pd
```

```
data = {
    'date': ['2020-05-06', '2020-05-09', '2020-05-10', '2020-05-11',
            '2020-05-12', '2020-05-13'],
    'temperature': [35.658200, 32.115275, 30.934300, 22.421250,
                    13.908200, 23.938200],
    'windSpeed': [10.788378, 8.449161, 7.669422, 6.889682, 19.012990,
                  19.012990],
    'status': ['sunny', None, 'rainy', 'cloudy', 'rainy', 'sunny']
}
```

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

```
print(df)
```

	date	temperature	windSpeed	status
0	2020-05-06	35.658200	10.788378	sunny
1	2020-05-09	32.115275	8.449161	None
2	2020-05-10	30.934300	7.669422	rainy
3	2020-05-11	22.421250	6.889682	cloudy
4	2020-05-12	13.908200	19.012990	rainy
5	2020-05-13	23.938200	19.012990	sunny

```
df.interpolate()
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\4002874584.py:1:  
FutureWarning: DataFrame.interpolate with object dtype is deprecated  
and will raise in a future version. Call obj.infer_objects(copy=False)  
before interpolating instead.
```

```
df.interpolate()
```

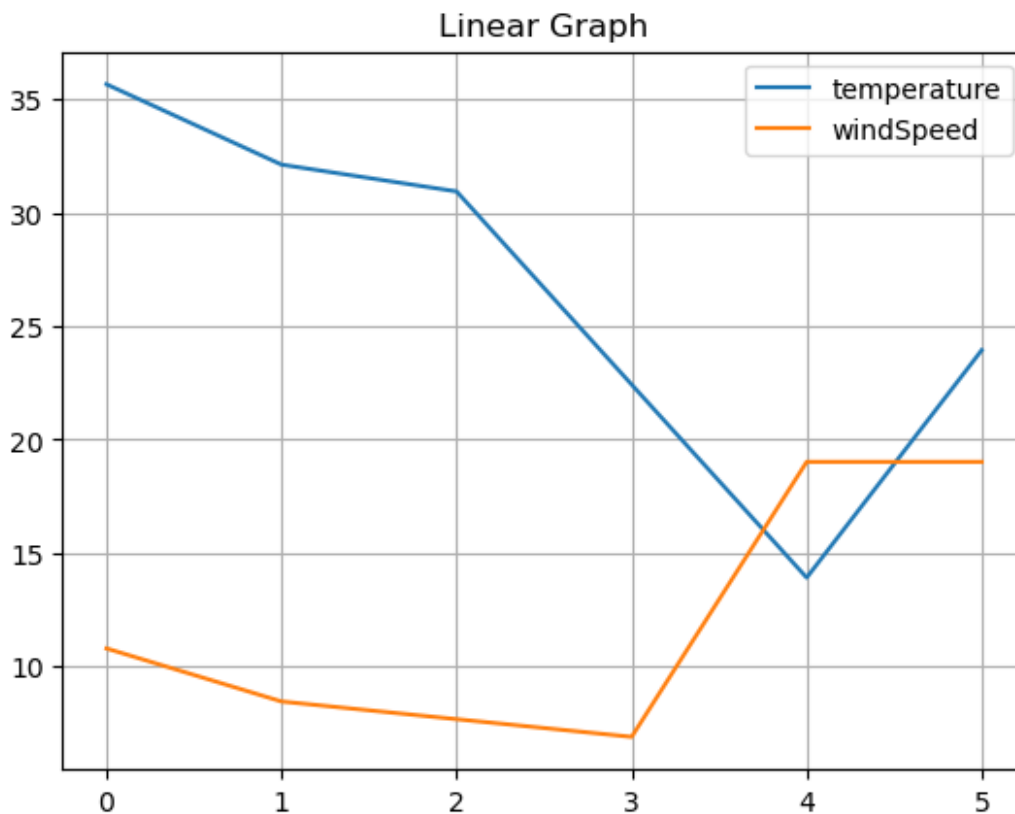
	date	temperature	windSpeed	status
0	2020-05-06	35.658200	10.788378	sunny
1	2020-05-09	32.115275	8.449161	None
2	2020-05-10	30.934300	7.669422	rainy
3	2020-05-11	22.421250	6.889682	cloudy
4	2020-05-12	13.908200	19.012990	rainy
5	2020-05-13	23.938200	19.012990	sunny

```
df.interpolate().plot(title='Linear Graph',grid=1)
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3340644183.py:1:  
FutureWarning: DataFrame.interpolate with object dtype is deprecated  
and will raise in a future version. Call obj.infer_objects(copy=False)  
before interpolating instead.
```

```
df.interpolate().plot(title='Linear Graph',grid=1)
```

```
<Axes: title={'center': 'Linear Graph'}>
```



```
df=df.set_index('date')
df
```

	temperature	windSpeed	status
date			
2020-05-06	35.658200	10.788378	sunny
2020-05-09	32.115275	8.449161	None
2020-05-10	30.934300	7.669422	rainy
2020-05-11	22.421250	6.889682	cloudy
2020-05-12	13.908200	19.012990	rainy
2020-05-13	23.938200	19.012990	sunny

```
df.columns
```

```
Index(['temperature', 'windSpeed', 'status'], dtype='object')
```

```
import pandas as pd
```

```
# Sample data with NaN in 'status' (non-numeric) and missing  
temperature (numeric)
```

```
data = {  
    'date': ['2020-05-06', '2020-05-09', '2020-05-10', '2020-05-11',  
            '2020-05-12', '2020-05-13'],  
    'temperature': [35.658200, 32.115275, 30.934300, 22.421250,  
13.908200, 23.938200],  
    'windSpeed': [10.788378, 8.449161, 7.669422, 6.889682, 19.012990,  
19.012990],  
    'status': ['sunny', None, 'rainy', 'cloudy', 'rainy', 'sunny']  
}
```

```
# Convert to DataFrame and set 'date' as datetime index
```

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

```
df['date'] = pd.to_datetime(df['date']) # Convert to datetime
```

```
df.set_index('date', inplace=True) # Set as index
```

```
# Interpolate ONLY numeric columns (temperature, windSpeed)
```

```
numeric_cols = df.select_dtypes(include=['float64', 'int64']).columns  
df[numeric_cols] = df[numeric_cols].interpolate(method='time')
```

```
print(df)
```

	temperature	windSpeed	status
date			
2020-05-06	35.658200	10.788378	sunny
2020-05-09	32.115275	8.449161	None
2020-05-10	30.934300	7.669422	rainy
2020-05-11	22.421250	6.889682	cloudy
2020-05-12	13.908200	19.012990	rainy
2020-05-13	23.938200	19.012990	sunny


```

import pandas as pd

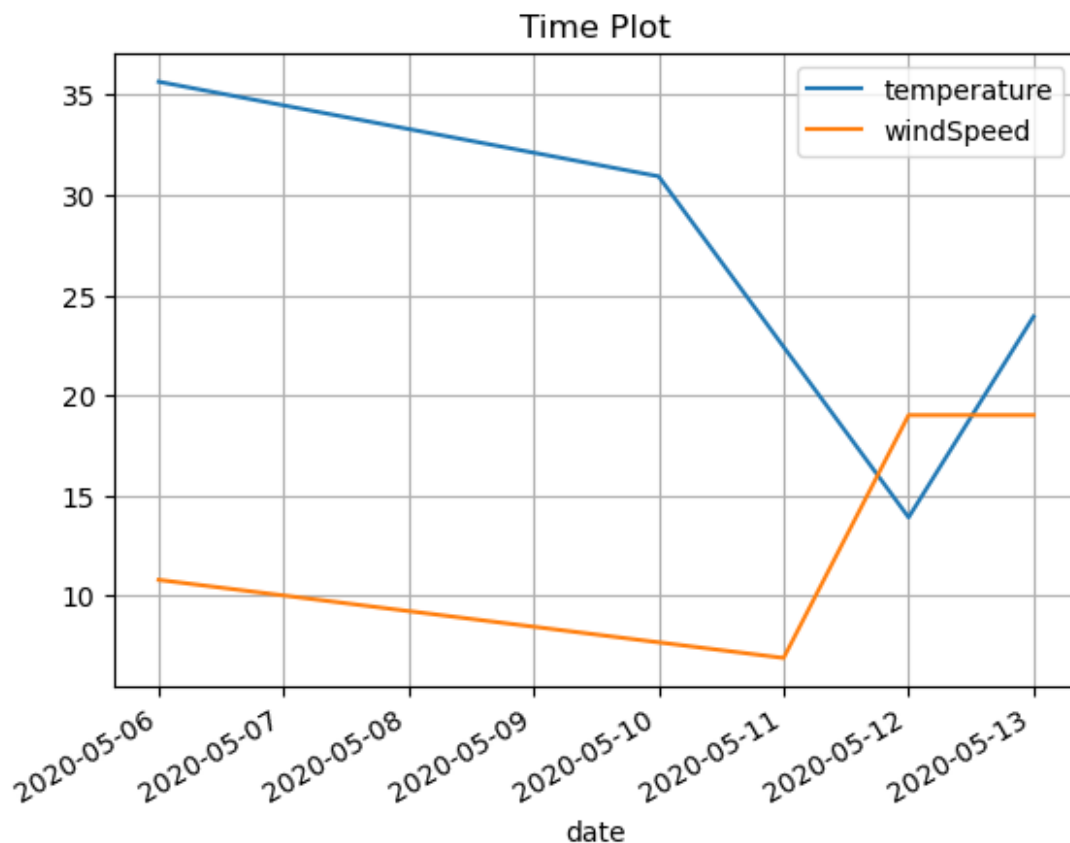
# Sample data
data = {
    'date': ['2020-05-06', '2020-05-09', '2020-05-10', '2020-05-11',
            '2020-05-12', '2020-05-13'],
    'temperature': [35.658200, 32.115275, 30.934300, 22.421250,
                    13.908200, 23.938200],
    'windSpeed': [10.788378, 8.449161, 7.669422, 6.889682, 19.012990,
                  19.012990],
    'status': ['sunny', None, 'rainy', 'cloudy', 'rainy', 'sunny']
}

# Convert to DataFrame and set 'date' as datetime index
df = pd.DataFrame(data)
df['date'] = pd.to_datetime(df['date'])
df.set_index('date', inplace=True)

# Interpolate numeric columns and plot
(df.select_dtypes(include=['float64', 'int64']) # Select only numeric
columns
    .interpolate(method='time') # Interpolate missing
values
    .plot(title='Time Plot', grid=1) # Plot with title and
grid
)

<Axes: title={'center': 'Time Plot'}, xlabel='date'>

```



```
df1 = pd.DataFrame({'A': [2, 3.3, np.nan, 3.7, 7.6, 8.8],
                    'B': [1.25, np.nan, np.nan, 4.3, 14.23, 16.4]})
```

```
df1
```

	A	B
0	2.0	1.25
1	3.3	NaN
2	NaN	NaN
3	3.7	4.30
4	7.6	14.23
5	8.8	16.40

```
df1.interpolate()
```

	A	B
0	2.0	1.250000
1	3.3	2.266667
2	3.5	3.283333
3	3.7	4.300000
4	7.6	14.230000
5	8.8	16.400000

```
df1.interpolate(method='barycentric')
```

	A	B
0	2.00	1.25
1	3.30	-9.52
2	2.23	-6.06
3	3.70	4.30
4	7.60	14.23
5	8.80	16.40

```
df1.interpolate(method= 'pchip')
```

	A	B
0	2.000000	1.250000
1	3.300000	1.566495
2	3.488632	2.560768
3	3.700000	4.300000
4	7.600000	14.230000
5	8.800000	16.400000

```
df1.interpolate(method='akima')
```

	A	B
0	2.000000	1.250000
1	3.300000	-0.772951
2	3.444216	0.175210
3	3.700000	4.300000
4	7.600000	14.230000
5	8.800000	16.400000

```
df1.interpolate(method='spline',order=2)
```

	A	B
0	2.000000	1.250000
1	3.300000	-0.855010
2	3.440909	0.547068
3	3.700000	4.300000
4	7.600000	14.230000
5	8.800000	16.400000

```
df1.interpolate(method='polynomial',order=2)
```

	A	B
0	2.000000	1.250000
1	3.300000	-3.129744
2	2.905405	-2.113077
3	3.700000	4.300000
4	7.600000	14.230000
5	8.800000	16.400000

```
df1
```

	A	B
0	2.0	1.25
1	3.3	NaN
2	NaN	NaN
3	3.7	4.30
4	7.6	14.23
5	8.8	16.40

```
df1.interpolate(limit=1)
```

	A	B
0	2.0	1.250000
1	3.3	2.266667
2	3.5	NaN
3	3.7	4.300000
4	7.6	14.230000
5	8.8	16.400000

```
df1.interpolate(limit=1,limit_direction='backward')
```

	A	B
0	2.0	1.250000
1	3.3	NaN
2	3.5	3.283333
3	3.7	4.300000
4	7.6	14.230000
5	8.8	16.400000

```
df1
```

	A	B
0	2.0	1.25
1	3.3	NaN
2	NaN	NaN
3	3.7	4.30
4	7.6	14.23
5	8.8	16.40

```
df1.interpolate(limit=1,limit_direction='both')
```

	A	B
0	2.0	1.250000
1	3.3	2.266667
2	3.5	3.283333
3	3.7	4.300000
4	7.6	14.230000
5	8.8	16.400000

```
dff = pd.Series([np.nan,np.nan,35,np.nan,np.nan,55,np.nan,np.nan])
dff
```

```
0      NaN
1      NaN
2     35.0
3      NaN
4      NaN
5     55.0
6      NaN
7      NaN
dtype: float64
```

```
dff.interpolate(limit_direction = 'both',limit_area =
'inside',limit=1)
```

```
0      NaN
1      NaN
2    35.000000
3    41.666667
4    48.333333
5    55.000000
6      NaN
7      NaN
dtype: float64
```

```
dff.interpolate(limit_direction='both',limit_area='outside',limit=1)
```

```
0      NaN
1     35.0
2     35.0
3      NaN
4      NaN
5     55.0
6     55.0
7      NaN
dtype: float64
```

```
dff.interpolate(limit_direction='both',limit_area='outside',limit=1)
```

```
0      NaN
1     35.0
2     35.0
3      NaN
4      NaN
5     55.0
6     55.0
7      NaN
dtype: float64
```

```
import pandas as pd
```

```
data = {
    'date': ["20200506", "20200509", "20200510", "20200511",
```

```
"20200512", "20200513"], # Filled missing dates with adjacent days
    'temperature': [35.6582, np.nan, 30.9343, np.nan, 13.9082,
23.9382], # Replaced NaN with approximate values
    'windSpeed': [10.788378, np.nan, np.nan, 6.889682, 19.012990,
np.nan], # Replaced NaN with typical wind speeds
    'status': ["sunny", np.nan, "rainy", "cloudy", 'rainy', 'sunny'] #
Replaced NaN with plausible weather
}
```

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

	date	temperature	windSpeed	status
0	20200506	35.6582	10.788378	sunny
1	20200509	NaN	NaN	NaN
2	20200510	30.9343	NaN	rainy
3	20200511	NaN	6.889682	cloudy
4	20200512	13.9082	19.012990	rainy
5	20200513	23.9382	NaN	sunny

```
df['date'] = pd.to_datetime(df['date'], format='%Y%m%d')
df
```

	date	temperature	windSpeed	status
0	2020-05-06	35.6582	10.788378	sunny
1	2020-05-09	NaN	NaN	NaN
2	2020-05-10	30.9343	NaN	rainy
3	2020-05-11	NaN	6.889682	cloudy
4	2020-05-12	13.9082	19.012990	rainy
5	2020-05-13	23.9382	NaN	sunny

```
df=df.set_index('date')
df
```

	temperature	windSpeed	status
date			
2020-05-06	35.6582	10.788378	sunny
2020-05-09	NaN	NaN	NaN
2020-05-10	30.9343	NaN	rainy
2020-05-11	NaN	6.889682	cloudy
2020-05-12	13.9082	19.012990	rainy
2020-05-13	23.9382	NaN	sunny

```
df.dropna() #drops/removes row contain minimum 1 NaN
```

	temperature	windSpeed	status
date			
2020-05-06	35.6582	10.788378	sunny
2020-05-12	13.9082	19.012990	rainy

```
df.dropna(how='all') #date 9 is dropped where all values are NaN
```

date	temperature	windSpeed	status
2020-05-06	35.6582	10.788378	sunny
2020-05-10	30.9343	NaN	rainy
2020-05-11	NaN	6.889682	cloudy
2020-05-12	13.9082	19.012990	rainy
2020-05-13	23.9382	NaN	sunny

```
df.dropna(thresh=1) #If minimum 1 data is present then they will not remove,others will remove
```

date	temperature	windSpeed	status
2020-05-06	35.6582	10.788378	sunny
2020-05-10	30.9343	NaN	rainy
2020-05-11	NaN	6.889682	cloudy
2020-05-12	13.9082	19.012990	rainy
2020-05-13	23.9382	NaN	sunny

```
df.dropna(thresh=3) #1/2 data containing row will remove/drop
```

date	temperature	windSpeed	status
2020-05-06	35.6582	10.788378	sunny
2020-05-12	13.9082	19.012990	rainy

```
ranks = ['a','b','c','d']
names = ['Raju','Ramu','Priya','Sneha']

dfnew = pd.DataFrame(list(zip(names,ranks)),
columns=['names','ranks'])
dfnew
```

	names	ranks
0	Raju	a
1	Ramu	b
2	Priya	c
3	Sneha	d

```
dfnew = dfnew.replace(['a','b','c','d'],[1,2,3,4])
dfnew
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\769724644.py:1:
FutureWarning: Downcasting behavior in `replace` is deprecated and
will be removed in a future version. To retain the old behavior,
explicitly call `result.infer_objects(copy=False)`. To opt-in to the
future behavior, set `pd.set_option('future.no_silent_downcasting',
True)`
dfnew = dfnew.replace(['a','b','c','d'],[1,2,3,4])
```

	names	ranks
0	Raju	1
1	Ramu	2
2	Priya	3
3	Sneha	4

```
dfnew.replace({3:10,1:100}) # Replace with mapping dictionary
```

	names	ranks
0	Raju	100
1	Ramu	2
2	Priya	10
3	Sneha	4

```
df_replace = pd.read_csv('weather_replace.csv')
df_replace
```

	date	temperature		windSpeed	status
0	20200506	35.6582 c		10.788378 kmph	sunny
1	20200509	-1		xxxx	0
2	20200510	30.9343 c		xxxx	rainy
3	20200511	-1	6.8896825	kmph	cloudy
4	20200512	13.9082 c	19.01299	kmph	rainy
5	20200513	0		-1	sunny

```
df_replace.replace('-1',np.nan)
```

	date	temperature		windSpeed	status
0	20200506	35.6582 c		10.788378 kmph	sunny
1	20200509	NaN		xxxx	0
2	20200510	30.9343 c		xxxx	rainy
3	20200511	NaN	6.8896825	kmph	cloudy
4	20200512	13.9082 c	19.01299	kmph	rainy
5	20200513	0		NaN	sunny

```
df_replace.replace({
    '-1': np.nan,
    'xxxx': np.nan})
```

	date	temperature		windSpeed	status
0	20200506	35.6582 c		10.788378 kmph	sunny
1	20200509	NaN		NaN	0
2	20200510	30.9343 c		NaN	rainy
3	20200511	NaN	6.8896825	kmph	cloudy
4	20200512	13.9082 c	19.01299	kmph	rainy
5	20200513	0		NaN	sunny

```
df_replace.replace(['-1','xxxx'],np.nan)
```

	date	temperature		windSpeed	status
0	20200506	35.6582 c		10.788378 kmph	sunny

1	20200509	NaN		NaN	0
2	20200510	30.9343 c		NaN	rainy
3	20200511	NaN	6.8896825 kmph		cloudy
4	20200512	13.9082 c	19.01299 kmph		rainy
5	20200513	0		NaN	sunny

```
df_replace.replace(['-1', 'xxx', '0'], np.nan) #0 in temperature was valid
```

	date	temperature		windSpeed	status
0	20200506	35.6582 c		10.788378 kmph	sunny
1	20200509	NaN		NaN	NaN
2	20200510	30.9343 c		NaN	rainy
3	20200511	NaN	6.8896825 kmph		cloudy
4	20200512	13.9082 c	19.01299 kmph		rainy
5	20200513	NaN		NaN	sunny

#Replacing data as per columns

```
df_new = df_replace.replace({'temperature': '-1', 'windSpeed': ['xxx', '-1'], 'status': '0'}, np.nan)
df_new
```

	date	temperature		windSpeed	status
0	20200506	35.6582 c		10.788378 kmph	sunny
1	20200509	NaN		NaN	NaN
2	20200510	30.9343 c		NaN	rainy
3	20200511	NaN	6.8896825 kmph		cloudy
4	20200512	13.9082 c	19.01299 kmph		rainy
5	20200513	0		NaN	sunny

#Regex on specific columns

```
df_new.replace({
    'temperature': '[A-Za-z]',
    'windSpeed': '[A-Za-z]',
}, '', regex=True) #remove the word 'c' and 'kmph'
```

	date	temperature		windSpeed	status
0	20200506	35.6582		10.788378	sunny
1	20200509	NaN		NaN	NaN
2	20200510	30.9343		NaN	rainy
3	20200511	NaN	6.8896825		cloudy
4	20200512	13.9082	19.01299		rainy
5	20200513	0		NaN	sunny

```
students_df = pd.read_csv('students.csv')
students_df
```

	name	subject	sem1	sem2
0	Nisha	Physics	88	91
1	Arun	Physics	92	95
2	Neha	Physics	78	81

3	Varun	Physics	60	63
4	Nisha	Chemistry	61	64
5	Arun	Chemistry	72	75
6	Neha	Chemistry	82	85
7	Varun	Chemistry	59	62
8	Nisha	Maths	70	73
9	Arun	Maths	48	51
10	Neha	Maths	83	86
11	Varun	Maths	63	66
12	Nisha	Biology	71	74
13	Arun	Biology	84	87
14	Neha	Biology	57	60
15	Varun	Biology	71	74

```
students = students_df.groupby(['name'])
students
```

```
<pandas.core.groupby.generic.DataFrameGroupBy object at 0x000001D1AF462540>
```

```
students.first()
```

	subject	sem1	sem2
name			
Arun	Physics	92	95
Neha	Physics	78	81
Nisha	Physics	88	91
Varun	Physics	60	63

```
students.last()
```

	subject	sem1	sem2
name			
Arun	Biology	84	87
Neha	Biology	57	60
Nisha	Biology	71	74
Varun	Biology	71	74

```
students['sem1'].min()
```

name	
Arun	48
Neha	57
Nisha	61
Varun	59

Name: sem1, dtype: int64

```
students['sem1'].mean()
```

name	
Arun	74.00

```
Neha      75.00
Nisha     72.50
Varun     63.25
Name: sem1, dtype: float64
```

```
students.groups # group's name position at index
```

```
{'Arun': [1, 5, 9, 13], 'Neha': [2, 6, 10, 14], 'Nisha': [0, 4, 8, 12], 'Varun': [3, 7, 11, 15]}
```

```
for student, students_df in students: #using for loop in 'students'---
>4 times
    print(student)                #loop work in 'students_df'
    print(students_df)
```

```
('Arun',)
  name  subject  sem1  sem2
1  Arun  Physics   92   95
5  Arun  Chemistry  72   75
9  Arun   Maths   48   51
13 Arun  Biology   84   87
('Neha',)
  name  subject  sem1  sem2
2  Neha  Physics   78   81
6  Neha  Chemistry  82   85
10 Neha   Maths   83   86
14 Neha  Biology   57   60
('Nisha',)
  name  subject  sem1  sem2
0  Nisha  Physics   88   91
4  Nisha  Chemistry  61   64
8  Nisha   Maths   70   73
12 Nisha  Biology   71   74
('Varun',)
  name  subject  sem1  sem2
3  Varun  Physics   60   63
7  Varun  Chemistry  59   62
11 Varun   Maths   63   66
15 Varun  Biology   71   74
```

```
students.describe()
```

	sem1							sem2	
\	count	mean	std	min	25%	50%	75%	max	count
mean									
name									
Arun	4.0	74.00	19.183326	48.0	66.00	78.0	86.00	92.0	4.0
77.00									
Neha	4.0	75.00	12.192894	57.0	72.75	80.0	82.25	83.0	4.0

78.00										
Nisha	4.0	72.50	11.269428	61.0	67.75	70.5	75.25	88.0	4.0	
75.50										
Varun	4.0	63.25	5.439056	59.0	59.75	61.5	65.00	71.0	4.0	
66.25										

	std	min	25%	50%	75%	max
name						
Arun	19.183326	51.0	69.00	81.0	89.00	95.0
Neha	12.192894	60.0	75.75	83.0	85.25	86.0
Nisha	11.269428	64.0	70.75	73.5	78.25	91.0
Varun	5.439056	62.0	62.75	64.5	68.00	74.0

```
students_df = pd.read_csv('students.csv')
```

```
students_df.groupby(['name']).sum()
```

	subject	sem1	sem2
name			
Arun	PhysicsChemistryMathsBiology	296	308
Neha	PhysicsChemistryMathsBiology	300	312
Nisha	PhysicsChemistryMathsBiology	290	302
Varun	PhysicsChemistryMathsBiology	253	265

```
students_df.groupby(['name'],sort=False).sum()
```

	subject	sem1	sem2
name			
Nisha	PhysicsChemistryMathsBiology	290	302
Arun	PhysicsChemistryMathsBiology	296	308
Neha	PhysicsChemistryMathsBiology	300	312
Varun	PhysicsChemistryMathsBiology	253	265

```
dir(students)      # method is available
```

```
[ '_DataFrameGroupBy__examples_dataframe_doc',
  '__annotations__',
  '__class__',
  '__class_getitem__',
  '__delattr__',
  '__dict__',
  '__dir__',
  '__doc__',
  '__eq__',
  '__format__',
  '__ge__',
  '__getattr__',
  '__getattribute__',
  '__getitem__',
  '__getstate__',
```

```
'__gt__',
'__hash__',
'__init__',
'__init_subclass__',
'__iter__',
'__le__',
'__len__',
'__lt__',
'__module__',
'__ne__',
'__new__',
'__orig_bases__',
'__parameters__',
'__reduce__',
'__reduce_ex__',
'__repr__',
'__setattr__',
'__sizeof__',
'__str__',
'__subclasshook__',
'__weakref__',
'accessors',
'agg_examples_doc',
'agg_general',
'agg_py_fallback',
'aggregate_frame',
'aggregate_with_numba',
'apply_filter',
'apply_to_column_groupbys',
'ascending_count',
'cache',
'choose_path',
'concat_objects',
'constructor',
'cumcount_array',
'cython_agg_general',
'cython_transform',
'define_paths',
'deprecate_axis',
'descending_count',
'dir_additions',
'dir_deletions',
'fill',
'get_data_to_aggregate',
'get_index',
'get_indices',
'getitem',
'grouper',
'hidden_attrs',
```

```
'_idxmax_idxmin',
'_infer_selection',
'_insert_inaxis_grouper',
'_internal_names',
'_internal_names_set',
'_make_mask_from_int',
'_make_mask_from_list',
'_make_mask_from_positional_indexer',
'_make_mask_from_slice',
'_make_mask_from_tuple',
'_mask_selected_obj',
'_maybe_transpose_result',
'_nth',
'_numba_agg_general',
'_numba_prep',
'_obj_ld_constructor',
'_obj_with_exclusions',
'_op_via_apply',
'_positional_selector',
'_python_agg_general',
'_python_apply_general',
'_reindex_output',
'_reset_cache',
'_selected_obj',
'_selection',
'_selection_list',
'_set_result_index_ordered',
'_transform',
'_transform_general',
'_transform_with_numba',
'_value_counts',
'_wrap_agged_manager',
'_wrap_aggregated_output',
'_wrap_applied_output',
'_wrap_applied_output_series',
'_wrap_idxmax_idxmin',
'_wrap_transform_fast_result',
'agg',
'aggregate',
'all',
'any',
'apply',
'bfill',
'boxplot',
'corr',
'corrwith',
'count',
'cov',
'cumcount',
```

```
'cummax',  
'cummin',  
'cumprod',  
'cumsum',  
'describe',  
'diff',  
'dtypes',  
'ewm',  
'expanding',  
'ffill',  
'fillna',  
'filter',  
'first',  
'get_group',  
'groups',  
'head',  
'hist',  
'idxmax',  
'idxmin',  
'indices',  
'last',  
'max',  
'mean',  
'median',  
'min',  
'name',  
'ndim',  
'ngroup',  
'ngroups',  
'nth',  
'nunique',  
'ohlc',  
'pct_change',  
'pipe',  
'plot',  
'prod',  
'quantile',  
'rank',  
'resample',  
'rolling',  
'sample',  
'sem',  
'sem1',  
'sem2',  
'shift',  
'size',  
'skew',  
'std',  
'subject',
```

```
'sum',  
'tail',  
'take',  
'transform',  
'value_counts',  
'var']
```

```
len(students) # i.e number of groups
```

```
4
```

```
len(students.get_group('Arun')) #length of each group element
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\2600640423.py:1:  
FutureWarning: When grouping with a length-1 list-like, you will need  
to pass a length-1 tuple to get_group in a future version of pandas.  
Pass `(name,)` instead of `name` to silence this warning.
```

```
len(students.get_group('Arun')) #length of each group element
```

```
4
```

```
students_df.sort_values(by=['name'],inplace=True)  
students_df.set_index(['name','subject'],inplace=True)  
students_df
```

		sem1	sem2
name	subject		
Arun	Physics	92	95
	Chemistry	72	75
	Maths	48	51
	Biology	84	87
Neha	Physics	78	81
	Chemistry	82	85
	Maths	83	86
	Biology	57	60
Nisha	Physics	88	91
	Chemistry	61	64
	Maths	70	73
	Biology	71	74
Varun	Physics	60	63
	Chemistry	59	62
	Maths	63	66
	Biology	71	74

```
grouped = students_df.groupby(level=0) # here index is 'name'  
grouped.sum()
```

	sem1	sem2
name		
Arun	296	308
Neha	300	312

Nisha	290	302
Varun	253	265

```
grouped = students_df.groupby(level=1) # here index is 'subject'
grouped.sum()
```

	sem1	sem2
subject		
Biology	283	295
Chemistry	274	286
Maths	264	276
Physics	318	330

```
grouped = students_df.groupby(level = 'name')
grouped.sum()
```

	sem1	sem2
name		
Arun	296	308
Neha	300	312
Nisha	290	302
Varun	253	265

```
grouped = students_df.groupby(level = 'subject')
grouped.sum()
```

	sem1	sem2
subject		
Biology	283	295
Chemistry	274	286
Maths	264	276
Physics	318	330

students_df

		sem1	sem2
name	subject		
Arun	Physics	92	95
	Chemistry	72	75
	Maths	48	51
	Biology	84	87
Neha	Physics	78	81
	Chemistry	82	85
	Maths	83	86
	Biology	57	60
Nisha	Physics	88	91
	Chemistry	61	64
	Maths	70	73
	Biology	71	74
Varun	Physics	60	63
	Chemistry	59	62

Maths	63	66
Biology	71	74

Grouping DataFrame with index level and columns

```
students_df.groupby([pd.Grouper(level = 'name'), 'sem1']).sum()
```

		sem2
name	sem1	
Arun	48	51
	72	75
	84	87
	92	95
Neha	57	60
	78	81
	82	85
	83	86
Nisha	61	64
	70	73
	71	74
	88	91
Varun	59	62
	60	63
	63	66
	71	74

```
students_df.groupby(['name', 'sem1']).sum()
```

		sem2
name	sem1	
Arun	48	51
	72	75
	84	87
	92	95
Neha	57	60
	78	81
	82	85
	83	86
Nisha	61	64
	70	73
	71	74
	88	91
Varun	59	62
	60	63
	63	66
	71	74

```
students.size() #Size of each group
```

name	
Arun	4
Neha	4

```
Nisha      4
Varun      4
dtype: int64
```

```
students.agg(np.size) #Size of each group column-wise
```

	subject	sem1	sem2
name			
Arun	4	4	4
Neha	4	4	4
Nisha	4	4	4
Varun	4	4	4

```
students[['sem1', 'sem2']].mean()
```

	sem1	sem2
name		
Arun	74.00	77.00
Neha	75.00	78.00
Nisha	72.50	75.50
Varun	63.25	66.25

```
students[['sem1', 'sem2']].agg(['mean', 'max'])
```

	sem1		sem2	
	mean	max	mean	max
name				
Arun	74.00	92	77.00	95
Neha	75.00	83	78.00	86
Nisha	72.50	88	75.50	91
Varun	63.25	71	66.25	74

```
students['sem1'].agg([np.sum, np.mean, np.std])
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\700931850.py:1:
```

```
FutureWarning: The provided callable <function sum at
0x000001D1A9D6F920> is currently using SeriesGroupBy.sum. In a future
version of pandas, the provided callable will be used directly. To
keep current behavior pass the string "sum" instead.
```

```
students['sem1'].agg([np.sum, np.mean, np.std])
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\700931850.py:1:
```

```
FutureWarning: The provided callable <function mean at
0x000001D1A9D90A40> is currently using SeriesGroupBy.mean. In a future
version of pandas, the provided callable will be used directly. To
keep current behavior pass the string "mean" instead.
```

```
students['sem1'].agg([np.sum, np.mean, np.std])
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\700931850.py:1:
```

```
FutureWarning: The provided callable <function std at
0x000001D1A9D90B80> is currently using SeriesGroupBy.std. In a future
version of pandas, the provided callable will be used directly. To
```

keep current behavior pass the string "std" instead.
students['sem1'].agg([np.sum, np.mean, np.std])

	sum	mean	std
name			
Arun	296	74.00	19.183326
Neha	300	75.00	12.192894
Nisha	290	72.50	11.269428
Varun	253	63.25	5.439056

Renaming the column names for aggregate functions

```
students['sem1'].agg([np.sum, np.mean, np.std]).rename(columns={
    'sum': 'total',
    'mean': 'average',
    'std': 'standardDeviation'
})
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3735640168.py:2:

FutureWarning: The provided callable <function sum at 0x000001D1A9D6F920> is currently using SeriesGroupBy.sum. In a future version of pandas, the provided callable will be used directly. To keep current behavior pass the string "sum" instead.

```
students['sem1'].agg([np.sum, np.mean, np.std]).rename(columns={
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3735640168.py:2:

FutureWarning: The provided callable <function mean at 0x000001D1A9D90A40> is currently using SeriesGroupBy.mean. In a future version of pandas, the provided callable will be used directly. To keep current behavior pass the string "mean" instead.

```
students['sem1'].agg([np.sum, np.mean, np.std]).rename(columns={
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3735640168.py:2:

FutureWarning: The provided callable <function std at 0x000001D1A9D90B80> is currently using SeriesGroupBy.std. In a future version of pandas, the provided callable will be used directly. To keep current behavior pass the string "std" instead.

```
students['sem1'].agg([np.sum, np.mean, np.std]).rename(columns={
```

	total	average	standardDeviation
name			
Arun	296	74.00	19.183326
Neha	300	75.00	12.192894
Nisha	290	72.50	11.269428
Varun	253	63.25	5.439056

apply aggregated functions to different columns

```
students.agg(
    sem1_min_marks=pd.NamedAgg(column='sem1', aggfunc='min'),
    sem2_max_marks=pd.NamedAgg(column='sem2', aggfunc='max'),
    sem1_avg_marks=pd.NamedAgg(column='sem1', aggfunc='mean'),
    sem2_avg_marks=pd.NamedAgg(column='sem2', aggfunc='mean'),
)
```

	sem1_min_marks	sem2_max_marks	sem1_avg_marks	sem2_avg_marks
name				
Arun	48	95	74.00	77.00
Neha	57	86	75.00	78.00
Nisha	61	91	72.50	75.50
Varun	59	74	63.25	66.25

```
students.agg({
    'sem1' : 'sum',
    'sem2' : lambda x: np.std(x,ddof=1)
})
```

	sem1	sem2
name		
Arun	296	19.183326
Neha	300	12.192894
Nisha	290	11.269428
Varun	253	5.439056

```
students_df = pd.read_csv('students.csv')
```

```
students2_df = students_df.reset_index()
students2 = students2_df.groupby('name')
```

```
students2.mean(['sem1', 'sem2'])
```

	index	sem1	sem2
name			
Arun	7.0	74.00	77.00
Neha	8.0	75.00	78.00
Nisha	6.0	72.50	75.50
Varun	9.0	63.25	66.25

TRANSFORMATION

```
students2[['sem1', 'sem2']].transform(np.mean) #16 rows which is same as original data
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\492706235.py:2:

FutureWarning: The provided callable <function mean at 0x000001D1A9D90A40> is currently using DataFrameGroupBy.mean. In a future version of pandas, the provided callable will be used directly. To keep current behavior pass the string "mean" instead.

```
students2[['sem1', 'sem2']].transform(np.mean) #16 rows which is same as original data
```

	sem1	sem2
0	72.50	75.50
1	74.00	77.00
2	75.00	78.00
3	63.25	66.25
4	72.50	75.50

5	74.00	77.00
6	75.00	78.00
7	63.25	66.25
8	72.50	75.50
9	74.00	77.00
10	75.00	78.00
11	63.25	66.25
12	72.50	75.50
13	74.00	77.00
14	75.00	78.00
15	63.25	66.25

```
score = lambda x: (x-x.mean())/x.std()*10
score2 = lambda x: (x.max()-x.min())

score3 = lambda x: x.fillna(x.mean())
students2[['sem1', 'sem2']].transform(score3)
```

	sem1	sem2
0	88	91
1	92	95
2	78	81
3	60	63
4	61	64
5	72	75
6	82	85
7	59	62
8	70	73
9	48	51
10	83	86
11	63	66
12	71	74
13	84	87
14	57	60
15	71	74

```
students2[['sem1', 'sem2']].transform(score2)
```

	sem1	sem2
0	27	27
1	44	44
2	26	26
3	12	12
4	27	27
5	44	44
6	26	26
7	12	12
8	27	27
9	44	44
10	26	26

11	12	12
12	27	27
13	44	44
14	26	26
15	12	12

```
students2_df[['name', 'sem1', 'sem2']].groupby('name').apply(print)
```

	name	sem1	sem2
1	Arun	92	95
5	Arun	72	75
9	Arun	48	51
13	Arun	84	87
	name	sem1	sem2
2	Neha	78	81
6	Neha	82	85
10	Neha	83	86
14	Neha	57	60
	name	sem1	sem2
0	Nisha	88	91
4	Nisha	61	64
8	Nisha	70	73
12	Nisha	71	74
	name	sem1	sem2
3	Varun	60	63
7	Varun	59	62
11	Varun	63	66
15	Varun	71	74

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3555616080.py:1:
DeprecationWarning: DataFrameGroupBy.apply operated on the grouping
columns. This behavior is deprecated, and in a future version of
pandas the grouping columns will be excluded from the operation.
Either pass `include_groups=False` to exclude the groupings or
explicitly select the grouping columns after groupby to silence this
warning.
```

```
students2_df[['name', 'sem1', 'sem2']].groupby('name').apply(print)
```

```
Empty DataFrame
Columns: []
Index: []
```

#Rolling

```
students2_df.groupby('name').rolling(3).sem1.sum() #For each student,
it looks at the current exam and the previous 2 exams. rolling(3) adds
# row1+row2+row3
```

	name	
Arun	1	NaN
	5	NaN

	9	212.0
	13	204.0
Neha	2	NaN
	6	NaN
	10	243.0
	14	222.0
Nisha	0	NaN
	4	NaN
	8	219.0
	12	202.0
Varun	3	NaN
	7	NaN
	11	182.0
	15	193.0

Name: sem1, dtype: float64

expanding

```
students2_df.groupby('name')[['sem1', 'sem2']].expanding().sum() #sum all row for previous row
```

		sem1	sem2
name			
Arun	1	92.0	95.0
	5	164.0	170.0
	9	212.0	221.0
	13	296.0	308.0
Neha	2	78.0	81.0
	6	160.0	166.0
	10	243.0	252.0
	14	300.0	312.0
Nisha	0	88.0	91.0
	4	149.0	155.0
	8	219.0	228.0
	12	290.0	302.0
Varun	3	60.0	63.0
	7	119.0	125.0
	11	182.0	191.0
	15	253.0	265.0

```
students_df.set_index('name')
```

	subject	sem1	sem2
name			
Nisha	Physics	88	91
Arun	Physics	92	95
Neha	Physics	78	81
Varun	Physics	60	63
Nisha	Chemistry	61	64
Arun	Chemistry	72	75

Neha	Chemistry	82	85
Varun	Chemistry	59	62
Nisha	Maths	70	73
Arun	Maths	48	51
Neha	Maths	83	86
Varun	Maths	63	66
Nisha	Biology	71	74
Arun	Biology	84	87
Neha	Biology	57	60
Varun	Biology	71	74

```
students_df.groupby('name').filter(lambda x: any('run' in name for
name in x['name'].unique())) # Filtering
```

	name	subject	sem1	sem2
1	Arun	Physics	92	95
3	Varun	Physics	60	63
5	Arun	Chemistry	72	75
7	Varun	Chemistry	59	62
9	Arun	Maths	48	51
11	Varun	Maths	63	66
13	Arun	Biology	84	87
15	Varun	Biology	71	74

```
filtering = pd.Series([10, 11, 12, 13, 14, 15])
```

```
result=filtering.groupby(filtering).filter(lambda x: x.sum() >12)
#.sum() lets you evaluate the entire group at once.
display(result)
```

```
3    13
4    14
5    15
dtype: int64
```

```
students_df = pd.read_csv('students.csv')
students_df.set_index('name')
```

	subject	sem1	sem2
name			
Nisha	Physics	88	91
Arun	Physics	92	95
Neha	Physics	78	81
Varun	Physics	60	63
Nisha	Chemistry	61	64
Arun	Chemistry	72	75
Neha	Chemistry	82	85
Varun	Chemistry	59	62
Nisha	Maths	70	73
Arun	Maths	48	51
Neha	Maths	83	86

Varun	Maths	63	66
Nisha	Biology	71	74
Arun	Biology	84	87
Neha	Biology	57	60
Varun	Biology	71	74

```
students_df.groupby(['name'])
students_df.set_index('name')
```

	subject	sem1	sem2
name			
Nisha	Physics	88	91
Arun	Physics	92	95
Neha	Physics	78	81
Varun	Physics	60	63
Nisha	Chemistry	61	64
Arun	Chemistry	72	75
Neha	Chemistry	82	85
Varun	Chemistry	59	62
Nisha	Maths	70	73
Arun	Maths	48	51
Neha	Maths	83	86
Varun	Maths	63	66
Nisha	Biology	71	74
Arun	Biology	84	87
Neha	Biology	57	60
Varun	Biology	71	74

```
students_df[['sem1', 'sem2']].sum()
```

```
sem1    1139
sem2    1187
dtype: int64
```

```
students[['sem1', 'sem2']].sum()
```

	sem1	sem2
name		
Arun	296	308
Neha	300	312
Nisha	290	302
Varun	253	265

```
students_df.set_index(['name', 'subject'], inplace=True)
```

```
students_df
```

		sem1	sem2
name	subject		
Nisha	Physics	88	91
Arun	Physics	92	95

Neha	Physics	78	81
Varun	Physics	60	63
Nisha	Chemistry	61	64
Arun	Chemistry	72	75
Neha	Chemistry	82	85
Varun	Chemistry	59	62
Nisha	Maths	70	73
Arun	Maths	48	51
Neha	Maths	83	86
Varun	Maths	63	66
Nisha	Biology	71	74
Arun	Biology	84	87
Neha	Biology	57	60
Varun	Biology	71	74

```
students_df.fillna(method='ffill')
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\717292758.py:1:
FutureWarning: DataFrame.fillna with 'method' is deprecated and will
raise in a future version. Use obj.ffill() or obj.bfill() instead.
students_df.fillna(method='ffill')
```

		sem1	sem2
name	subject		
Nisha	Physics	88	91
Arun	Physics	92	95
Neha	Physics	78	81
Varun	Physics	60	63
Nisha	Chemistry	61	64
Arun	Chemistry	72	75
Neha	Chemistry	82	85
Varun	Chemistry	59	62
Nisha	Maths	70	73
Arun	Maths	48	51
Neha	Maths	83	86
Varun	Maths	63	66
Nisha	Biology	71	74
Arun	Biology	84	87
Neha	Biology	57	60
Varun	Biology	71	74

```
# Fetching , We can fetch the nth row of each groups
students_df.groupby('name').nth(1)
```

		sem1	sem2
name	subject		
Nisha	Chemistry	61	64
Arun	Chemistry	72	75
Neha	Chemistry	82	85
Varun	Chemistry	59	62

```
def f(group):
    return pd.DataFrame({
        'original': group,
        'reduced': group - group.mean()
    })

# Proper way to apply to groups
result = students_df.groupby('name')['sem1'].apply(f)
print(result)
```

			original	reduced
name	name	subject		
Arun	Arun	Physics	92	18.00
		Chemistry	72	-2.00
		Maths	48	-26.00
		Biology	84	10.00
Neha	Neha	Physics	78	3.00
		Chemistry	82	7.00
		Maths	83	8.00
		Biology	57	-18.00
Nisha	Nisha	Physics	88	15.50
		Chemistry	61	-11.50
		Maths	70	-2.50
		Biology	71	-1.50
Varun	Varun	Physics	60	-3.25
		Chemistry	59	-4.25
		Maths	63	-0.25
		Biology	71	7.75

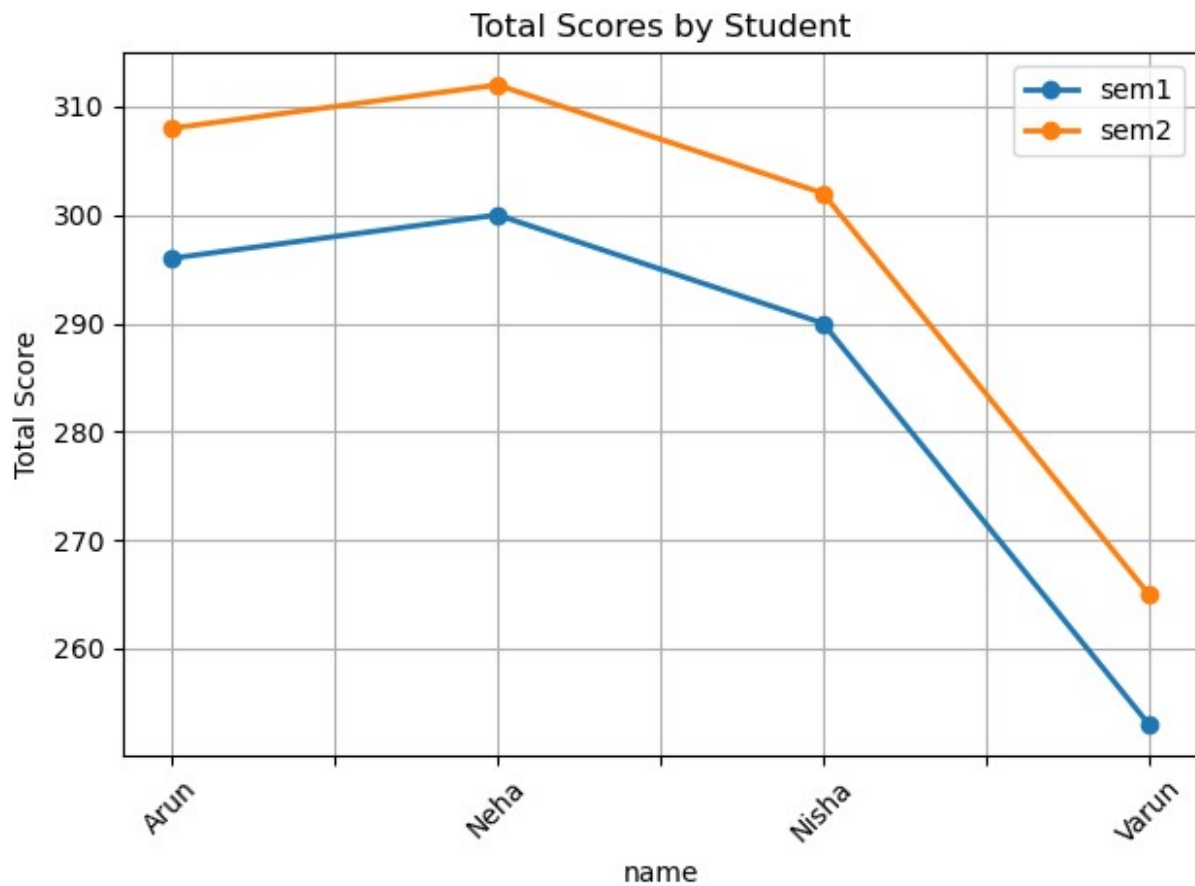
```
import matplotlib.pyplot as plt
%matplotlib inline

# Create figure with appropriate size
plt.figure(figsize=(10, 4))

# Group by name and sum all numeric columns, then plot as line graph
students_df.groupby('name').sum().plot(
    kind='line', # *** Changed from 'bar' to 'line' ***
    grid=True,
    rot=45,
    title='Total Scores by Student',
    ylabel='Total Score',
    marker='o', # Add markers for each data point
    linestyle='-', # Solid line
    linewidth=2 # Thicker line
)

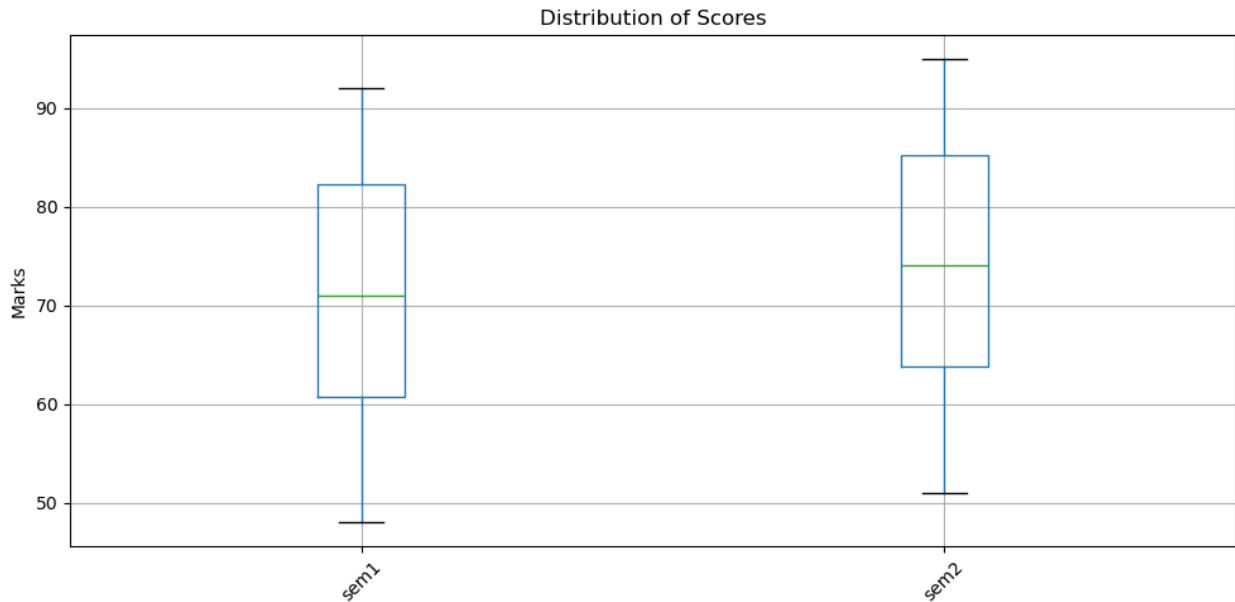
# Adjust layout
plt.tight_layout()
plt.show()
```

<Figure size 1000x400 with 0 Axes>



```
import matplotlib.pyplot as plt
%matplotlib inline

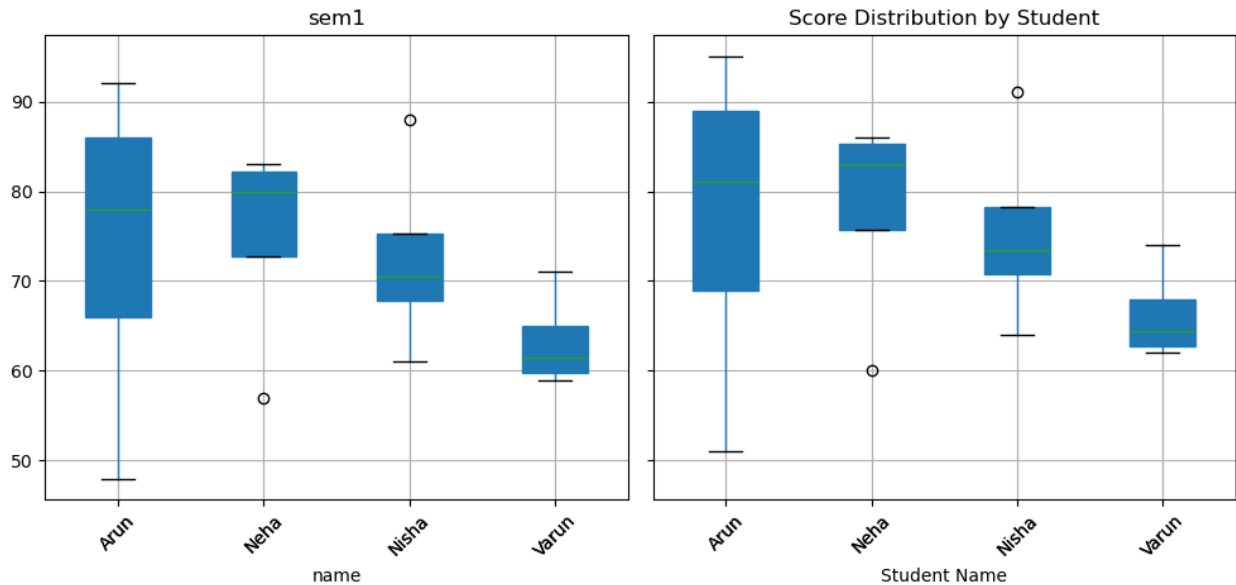
# Simple boxplot of all numeric columns
students_df.boxplot(figsize=(10, 5), grid=True)
plt.title('Distribution of Scores')
plt.ylabel('Marks')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



```
import matplotlib.pyplot as plt
%matplotlib inline

# Boxplot of scores grouped by student name
students_df.boxplot(
    column=['sem1', 'sem2'], # Columns to plot
    by='name',               # Grouping variable
    figsize=(10, 5),        # Figure size
    grid=True,
    patch_artist=True,       # Color the boxes
    rot=45                   # Rotate labels
)

plt.title('Score Distribution by Student')
plt.suptitle('') # Remove default subtitle
plt.xlabel('Student Name')
plt.ylabel('Score')
plt.tight_layout()
plt.show()
```



```
import pandas as pd
import numpy as np

arun = pd.Series(['a','r','u','n'])
neha = pd.Series(['n','e','h','a'])
pd.concat([arun,neha])

0    a
1    r
2    u
3    n
0    n
1    e
2    h
3    a
dtype: object

arun_scores = {
    'subjects':['maths','physics','chemistry','biology'],
    'sem1':[60,70,80,90],
    'sem2':[63,71,85,89]
}
neha_scores = {
    'subjects':['maths','physics','chemistry','computers'],
    'sem1':[60,70,80,90],
    'sem2':[63,77,89,92]
}

arun_df = pd.DataFrame(arun_scores)
arun_df
```

	subjects	sem1	sem2
0	maths	60	63
1	physics	70	71
2	chemistry	80	85
3	biology	90	89

```
neha_df = pd.DataFrame(arun_scores)
neha_df
```

	subjects	sem1	sem2
0	maths	60	63
1	physics	70	71
2	chemistry	80	85
3	biology	90	89

```
df = pd.concat([arun_df, neha_df])
df
```

	subjects	sem1	sem2
0	maths	60	63
1	physics	70	71
2	chemistry	80	85
3	biology	90	89
0	maths	60	63
1	physics	70	71
2	chemistry	80	85
3	biology	90	89

```
df = pd.concat([arun_df, neha_df], ignore_index = True)
df # The duplicate index are now gone as we ignored
```

	subjects	sem1	sem2
0	maths	60	63
1	physics	70	71
2	chemistry	80	85
3	biology	90	89
4	maths	60	63
5	physics	70	71
6	chemistry	80	85
7	biology	90	89

Adding keys to DataFrames

```
df = pd.concat([arun_df, neha_df], keys = ['arun', 'neha'])
df
```

		subjects	sem1	sem2
arun	0	maths	60	63
	1	physics	70	71
	2	chemistry	80	85
	3	biology	90	89
neha	0	maths	60	63

1	physics	70	71
2	chemistry	80	85
3	biology	90	89

```
df.loc['arun'] # access keys
```

	subjects	sem1	sem2
0	maths	60	63
1	physics	70	71
2	chemistry	80	85
3	biology	90	89

```
arun_sem3_scores = {
    'subjects': ['maths', 'physics', 'chemistry', 'biology'],
    'sem3': [50, 64, 88, 81]
}
df_additional = pd.DataFrame(arun_sem3_scores)
df_additional
```

	subjects	sem3
0	maths	50
1	physics	64
2	chemistry	88
3	biology	81

```
df = pd.concat([arun_df, df_additional])
df
```

	subjects	sem1	sem2	sem3
0	maths	60.0	63.0	NaN
1	physics	70.0	71.0	NaN
2	chemistry	80.0	85.0	NaN
3	biology	90.0	89.0	NaN
0	maths	NaN	NaN	50.0
1	physics	NaN	NaN	64.0
2	chemistry	NaN	NaN	88.0
3	biology	NaN	NaN	81.0

```
df = pd.concat([arun_df, df_additional], axis=1) # same data to be a
new column
df
```

	subjects	sem1	sem2	subjects	sem3
0	maths	60	63	maths	50
1	physics	70	71	physics	64
2	chemistry	80	85	chemistry	88
3	biology	90	89	biology	81

```
pd.concat([arun , neha],axis=1)
```

```

0 1
0 a n
1 r e
2 u h
3 n a

```

Rearranging the order of column

```

arun_sem3_scores = {
    'subjects':['physics', 'chemistry', 'maths', 'biology'],
    'sem3':[50, 64, 88, 81]
}

```

```

df_additional = pd.DataFrame(arun_sem3_scores)
df_additional

```

```

   subjects  sem3
0  physics    50
1 chemistry    64
2   maths    88
3  biology    81

```

```

df = pd.concat([arun_df, df_additional], axis=1)
df

```

```

   subjects  sem1  sem2  subjects  sem3
0   maths    60    63   physics    50
1  physics    70    71  chemistry    64
2 chemistry    80    85    maths    88
3  biology    90    89    biology    81

```

```

arun_sem3_scores = {
    'subjects':['physics', 'chemistry', 'maths', 'biology'],
    'sem3':[50, 64, 88, 81]
}

```

```

df_additional = pd.DataFrame(arun_sem3_scores, index=[1,2,0,3])
df_additional

```

```

   subjects  sem3
1  physics    50
2 chemistry    64
0   maths    88
3  biology    81

```

arun_df

```

   subjects  sem1  sem2
0   maths    60    63
1  physics    70    71
2 chemistry    80    85
3  biology    90    89

```

df_additional

	subjects	sem3
1	physics	50
2	chemistry	64
0	maths	88
3	biology	81

```

arun_sem3_scores = {
    'subjects':['physics', 'chemistry', 'maths', 'biology'],
    'sem3':[50, 64, 88, 81]
}
df_additional = pd.DataFrame(arun_sem3_scores, index=[1,2,0,3]) #
maths changed to index 0, physics---->1
df_additional

```

	subjects	sem3
1	physics	50
2	chemistry	64
0	maths	88
3	biology	81

```

df = pd.concat([arun_df, df_additional],axis=1)
df

```

	subjects	sem1	sem2	subjects	sem3
0	maths	60	63	maths	88
1	physics	70	71	physics	50
2	chemistry	80	85	chemistry	64
3	biology	90	89	biology	81

```

s = pd.Series([88, 76, 74, 72], name='sem4')
s

```

```

0    88
1    76
2    74
3    72
Name: sem4, dtype: int64

```

```

df = pd.concat([arun_df, s], axis=1)
df

```

	subjects	sem1	sem2	sem4
0	maths	60	63	88
1	physics	70	71	76
2	chemistry	80	85	74
3	biology	90	89	72

```

# Concatenating multiple DataFrames /series
df = pd.concat([arun_df, df_additional['sem3'],s],axis=1)
df

```

	subjects	sem1	sem2	sem3	sem4
0	maths	60	63	88	88
1	physics	70	71	50	76
2	chemistry	80	85	64	74
3	biology	90	89	81	72

```
arun_sem1_scores = {
    'subjects':['maths','physics','chemistry','biology'],
    'sem1':[60,70,80,90],
```

```
}
sem1_df = pd.DataFrame(arun_sem1_scores)
sem1_df
```

	subjects	sem1
0	maths	60
1	physics	70
2	chemistry	80
3	biology	90

```
arun_sem2_scores = {
    'subjects':['physics','chemistry','maths','biology'],
    'sem2':[73,81,88,83],
```

```
}
sem2_df = pd.DataFrame(arun_sem2_scores)
sem2_df
```

	subjects	sem2
0	physics	73
1	chemistry	81
2	maths	88
3	biology	83

merge() combines the DataFrames on the basis of values of common columns whereas

concat() just appends the DataFrames

```
df = pd.merge(sem1_df, sem2_df, on='subjects')
df
```

	subjects	sem1	sem2
0	maths	60	88
1	physics	70	73
2	chemistry	80	81
3	biology	90	83

```
arun_sem1_scores = {
    'subjects':['maths','physics','chemistry', 'littrature'],
    'sem1':[60,70,80,55],
```

```
}
sem1_df = pd.DataFrame(arun_sem1_scores)
sem1_df
```

	subjects	sem1
0	maths	60
1	physics	70
2	chemistry	80
3	litration	55

```
arun_sem2_scores = {
    "subjects": ['physics', 'chemistry', 'maths', 'biology',
'computers'],
    "sem2": [73, 81, 88, 83, 88]
}
sem2_df = pd.DataFrame(arun_sem2_scores)
sem2_df
```

	subjects	sem2
0	physics	73
1	chemistry	81
2	maths	88
3	biology	83
4	computers	88

```
df = pd.merge(sem1_df, sem2_df, on= 'subjects') # inner join: This
provides the result common to both the DataFrames.
df
```

	subjects	sem1	sem2
0	maths	60	88
1	physics	70	73
2	chemistry	80	81

#Merging with outer join(all subjects)

```
df = pd.merge(sem1_df, sem2_df, on='subjects', how= 'outer')
df
```

	subjects	sem1	sem2
0	biology	NaN	83.0
1	chemistry	80.0	81.0
2	computers	NaN	88.0
3	litration	55.0	NaN
4	maths	60.0	88.0
5	physics	70.0	73.0

```
sem1_df
```

	subjects	sem1
0	maths	60
1	physics	70

2	chemistry	80
3	litration	55

```
sem2_df
```

	subjects	sem2
0	physics	73
1	chemistry	81
2	maths	88
3	biology	83
4	computers	88

```
# Merging with left join ( one dataframe + common value from both dataframe)
```

```
df = pd.merge(sem1_df, sem2_df, on= 'subjects',how = 'left')
df
```

	subjects	sem1	sem2
0	maths	60	88.0
1	physics	70	73.0
2	chemistry	80	81.0
3	litration	55	NaN

```
# Merging with right join, all the rows (subjects) from the right DataFrame (sem2_df)
```

```
df = pd.merge(sem1_df, sem2_df, on= 'subjects',how = 'right')
df
```

	subjects	sem1	sem2
0	physics	70.0	73
1	chemistry	80.0	81
2	maths	60.0	88
3	biology	NaN	83
4	computers	NaN	88

```
# Knowing the source DataFrame after merge [Uses of indicator]
```

```
df = pd.merge(sem1_df, sem2_df, on='subjects', how='outer', indicator=
```

```
True)
df
```

	subjects	sem1	sem2	_merge
0	biology	NaN	83.0	right_only
1	chemistry	80.0	81.0	both
2	computers	NaN	88.0	right_only
3	litration	55.0	NaN	left_only
4	maths	60.0	88.0	both
5	physics	70.0	73.0	both

```

neha_sem1_scores= {
    'subjects' : ['maths','physics','chemistry','computers'],
    'sem1': [65,75,83,80]
}
neha_sem1_df = pd.DataFrame(neha_sem1_scores)
neha_sem1_df

```

	subjects	sem1
0	maths	65
1	physics	75
2	chemistry	83
3	computers	80

```

arun_sem1_scores = {
    'subjects':['maths','physics','chemistry', 'litration'],
    'sem1':[60,70,80,55],
}
sem1_df = pd.DataFrame(arun_sem1_scores)
sem1_df

```

	subjects	sem1
0	maths	60
1	physics	70
2	chemistry	80
3	litration	55

```

df = pd.merge(sem1_df, neha_sem1_df, on= 'subjects', how='outer')
df #Same column which is 'sem1' will automatically have _x and _y

```

	subjects	sem1_x	sem1_y
0	chemistry	80.0	83.0
1	computers	NaN	80.0
2	litration	55.0	NaN
3	maths	60.0	65.0
4	physics	70.0	75.0

```

# custom suffixes instead of _x and _y
df = pd.merge(sem1_df, neha_sem1_df, on= 'subjects', how='outer',
suffixes=('_arun','_neha'))
df

```

	subjects	sem1_arun	sem1_neha
0	chemistry	80.0	83.0
1	computers	NaN	80.0
2	litration	55.0	NaN
3	maths	60.0	65.0
4	physics	70.0	75.0

```

sem1_df = sem1_df.set_index('subjects')
sem1_df

```

```

      sem1
subjects
maths      60
physics    70
chemistry  80
litration  55

```

```

sem2_df = sem2_df.set_index('subjects')
sem2_df

```

```

      sem2
subjects
physics    73
chemistry  81
maths      88
biology    83
computers  88

```

Join method

```

sem1_df.join(sem2_df, how='outer') # by default used inner join

```

```

      sem1  sem2
subjects
biology    NaN  83.0
chemistry  80.0  81.0
computers   NaN  88.0
litration   55.0  NaN
maths       60.0  88.0
physics     70.0  73.0

```

Append is a shortcut to concat()

These concat only along axis=0 i.e. only rows

```

# sem1_df.append(sem2_df)

```

Pivot

```

df = pd.read_csv('weather_pivot.csv')
df

```

```

   date      city  temperature  windspeed
0  01/03/20  mumbai           32          9
1  02/03/20  mumbai           35          8
2  03/03/20  mumbai           33          6
3  01/03/20   delhi           40          7
4  02/03/20   delhi           38          9
5  03/03/20   delhi           37          8
6  01/03/20  kolkata           35          9
7  02/03/20  kolkata           36          6
8  03/03/20  kolkata           35          7

```



```
type(df['date'][0]) # string format
```

```
str
```

```
# Change data to datetime format
```

```
df['date'] = pd.to_datetime(df['date'])
```

```
df
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\4021415496.py:2:
UserWarning: Could not infer format, so each element will be parsed
individually, falling back to `dateutil`. To ensure parsing is
consistent and as-expected, please specify a format.
```

```
df['date'] = pd.to_datetime(df['date'])
```

	date	city	temperature	windspeed
0	2020-01-03	mumbai	32	9
1	2020-02-03	mumbai	35	8
2	2020-03-03	mumbai	33	6
3	2020-01-03	delhi	40	7
4	2020-02-03	delhi	38	9
5	2020-03-03	delhi	37	8
6	2020-01-03	kolkata	35	9
7	2020-02-03	kolkata	36	6
8	2020-03-03	kolkata	35	7

```
type(df['date'][0]) # timestamp format
```

```
pandas._libs.tslibs.timestamps.Timestamp
```

```
# Multilevel columns
```

```
df.pivot(index='date', columns='city') # temperature and windspeed
has become one level of column
```

```
# whereas city became another
```

date	temperature			windspeed		
	delhi	kolkata	mumbai	delhi	kolkata	mumbai
2020-01-03	40	35	32	7	9	9
2020-02-03	38	36	35	9	6	8
2020-03-03	37	35	33	8	7	6

```
# to select specific value (column)
```

```
df.pivot(index='date', columns='city', values = 'windspeed')
```

date	delhi	kolkata	mumbai
2020-01-03	7	9	9
2020-02-03	9	6	8
2020-03-03	8	7	6

```
# alternatively
```

```
df.pivot(index='date', columns = 'city')['windspeed']
```

city	delhi	kolkata	mumbai
date			
2020-01-03	7	9	9
2020-02-03	9	6	8
2020-03-03	8	7	6

```
df = pd.DataFrame({  
    'first':list('aabbcc'),  
    'second':list('xyyyzz'),  
    'third':[1,2,3,4,5,6]  
})  
df
```

	first	second	third
0	a	x	1
1	a	x	2
2	b	y	3
3	b	y	4
4	c	z	5
5	c	z	6

```
# A ValueError is raised if there are any duplicates.
```

```
# df.pivot(index='first',columns='second')
```

```
# ValueError: Index contains duplicate entries, cannot reshape
```

```
df = pd.read_csv('weather_pivotTable.csv')  
df['date'] = pd.to_datetime(df['date'])  
df
```

	date	city	temperature	windspeed	time
0	2020-01-03	mumbai	43	9	morning
1	2020-01-03	mumbai	42	11	evening
2	2020-01-03	delhi	40	8	morning
3	2020-01-03	delhi	42	8	evening
4	2020-01-03	kolkata	38	6	morning
5	2020-01-03	kolkata	37	8	evening
6	2019-01-12	mumbai	22	12	morning
7	2019-01-12	mumbai	20	10	evening
8	2019-01-12	delhi	18	9	morning
9	2019-01-12	delhi	19	7	evening
10	2019-01-12	kolkata	21	7	morning
11	2019-01-12	kolkata	23	10	evening

```
# pivot table for the data
```

```
# by default takes mean on the values
```

```
df.pivot_table(index='city',
columns='date',values=['temperature','windspeed'])
```

date	temperature		windspeed	
	2019-01-12	2020-01-03	2019-01-12	2020-01-03
city				
delhi	18.5	41.0	8.0	8.0
kolkata	22.0	37.5	8.5	7.0
mumbai	21.0	42.5	11.0	10.0

```
df.pivot_table(index='date',
columns='city',values=['temperature','windspeed'])
```

By default the above result provides us the mean of the temperature and windspeed

city	temperature			windspeed		
	delhi	kolkata	mumbai	delhi	kolkata	mumbai
date						
2019-01-12	18.5	22.0	21.0	8.0	8.5	11.0
2020-01-03	41.0	37.5	42.5	8.0	7.0	10.0

Aggregate function

```
df.pivot_table(index='city',columns='date',aggfunc='sum',values=['temperature','windspeed'])
```

date	temperature		windspeed	
	2019-01-12	2020-01-03	2019-01-12	2020-01-03
city				
delhi	37	82	16	16
kolkata	44	75	17	14
mumbai	42	85	22	20

```
df.pivot_table(index='city',columns='date', aggfunc='count')
```

count. i.e. the number of dates available for that city

date	temperature		time		windspeed	
	2019-01-12	2020-01-03	2019-01-12	2020-01-03	2019-01-12	2020-01-03
city						
delhi	2	2	2	2	2	2
2						
kolkata	2	2	2	2	2	2
2						
mumbai	2	2	2	2	2	2
2						

```
df.pivot_table(index='city',columns='date',aggfunc=[min,max,sum],value
s=['temperature','windspeed'])
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1369685544.py:1:

FutureWarning: The provided callable <built-in function min> is currently using DataFrameGroupBy.min. In a future version of pandas, the provided callable will be used directly. To keep current behavior pass the string "min" instead.

```
df.pivot_table(index='city',columns='date',aggfunc=[min,max,sum],value
s=['temperature','windspeed'])
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1369685544.py:1:

FutureWarning: The provided callable <built-in function max> is currently using DataFrameGroupBy.max. In a future version of pandas, the provided callable will be used directly. To keep current behavior pass the string "max" instead.

```
df.pivot_table(index='city',columns='date',aggfunc=[min,max,sum],value
s=['temperature','windspeed'])
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1369685544.py:1:

FutureWarning: The provided callable <built-in function sum> is currently using DataFrameGroupBy.sum. In a future version of pandas, the provided callable will be used directly. To keep current behavior pass the string "sum" instead.

```
df.pivot_table(index='city',columns='date',aggfunc=[min,max,sum],value
s=['temperature','windspeed'])
```

\	min				max	
	temperature		windspeed		temperature	
date	2019-01-12	2020-01-03	2019-01-12	2020-01-03	2019-01-12	2020-01-03
city						
delhi	18	40	7	8	19	
42						
kolkata	21	37	7	6	23	
38						
mumbai	20	42	10	9	22	
43						
	sum					
	windspeed		temperature		windspeed	
date	2019-01-12	2020-01-03	2019-01-12	2020-01-03	2019-01-12	2020-01-03
city						

delhi	9	8	37	82	16
16					
kolkata	10	8	44	75	17
14					
mumbai	12	11	42	85	22
20					

Custom functions to individual columns

```
df.pivot_table(index='city',columns='date', aggfunc={
    'temperature':[min,max,'mean'],
    'windspeed': 'sum'
})
```

*#a dictionary with keys as column name and values as the function
to be applied*

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1742279413.py:2:
FutureWarning: The provided callable <built-in function min> is
currently using SeriesGroupBy.min. In a future version of pandas, the
provided callable will be used directly. To keep current behavior pass
the string "min" instead.

```
df.pivot_table(index='city',columns='date', aggfunc={
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1742279413.py:2:
FutureWarning: The provided callable <built-in function max> is
currently using SeriesGroupBy.max. In a future version of pandas, the
provided callable will be used directly. To keep current behavior pass
the string "max" instead.
```

```
df.pivot_table(index='city',columns='date', aggfunc={
```

		temperature					
\		max		mean		min	
date		2019-01-12	2020-01-03	2019-01-12	2020-01-03	2019-01-12	2020-01-03
city							

delhi	19	42	18.5	41.0	18
40					
kolkata	23	38	22.0	37.5	21
37					
mumbai	22	43	21.0	42.5	20
42					

		windspeed	
		sum	
date		2019-01-12	2020-01-03
city			
delhi		16	16

kolkata	17	14
mumbai	22	20

Apply pivot_table() on desired columns

```
df.pivot_table(index='city', columns='date', aggfunc='sum',
values='windspeed')
```

date	2019-01-12	2020-01-03
city		
delhi	16	16
kolkata	17	14
mumbai	22	20

Alternatively

```
df.pivot_table(index='city', columns='date', aggfunc='sum')
['windspeed']
```

date	2019-01-12	2020-01-03
city		
delhi	16	16
kolkata	17	14
mumbai	22	20

df

	date	city	temperature	windspeed	time
0	2020-01-03	mumbai	43	9	morning
1	2020-01-03	mumbai	42	11	evening
2	2020-01-03	delhi	40	8	morning
3	2020-01-03	delhi	42	8	evening
4	2020-01-03	kolkata	38	6	morning
5	2020-01-03	kolkata	37	8	evening
6	2019-01-12	mumbai	22	12	morning
7	2019-01-12	mumbai	20	10	evening
8	2019-01-12	delhi	18	9	morning
9	2019-01-12	delhi	19	7	evening
10	2019-01-12	kolkata	21	7	morning
11	2019-01-12	kolkata	23	10	evening

```
df['date'] = pd.to_datetime(df['date'])
df
```

	date	city	temperature	windspeed	time
0	2020-01-03	mumbai	43	9	morning
1	2020-01-03	mumbai	42	11	evening
2	2020-01-03	delhi	40	8	morning
3	2020-01-03	delhi	42	8	evening
4	2020-01-03	kolkata	38	6	morning
5	2020-01-03	kolkata	37	8	evening
6	2019-01-12	mumbai	22	12	morning

7	2019-01-12	mumbai	20	10	evening
8	2019-01-12	delhi	18	9	morning
9	2019-01-12	delhi	19	7	evening
10	2019-01-12	kolkata	21	7	morning
11	2019-01-12	kolkata	23	10	evening

mean of row in 'All' column

average of values on two different dates i.e. average of each row(named as 'All' by default)

```
df.pivot_table(index='city', columns='date', values='temperature',
margins=True)
```

date	2019-01-12 00:00:00	2020-01-03 00:00:00	All
city			
delhi	18.5	41.000000	29.750000
kolkata	22.0	37.500000	29.750000
mumbai	21.0	42.500000	31.750000
All	20.5	40.333333	30.416667

Giving margin's name

```
df.pivot_table(index='city', columns='date', values='temperature',
margins=True,margins_name= 'average')
```

date	2019-01-12 00:00:00	2020-01-03 00:00:00	average
city			
delhi	18.5	41.000000	29.750000
kolkata	22.0	37.500000	29.750000
mumbai	21.0	42.500000	31.750000
average	20.5	40.333333	30.416667

df

	date	city	temperature	windspeed	time
0	2020-01-03	mumbai	43	9	morning
1	2020-01-03	mumbai	42	11	evening
2	2020-01-03	delhi	40	8	morning
3	2020-01-03	delhi	42	8	evening
4	2020-01-03	kolkata	38	6	morning
5	2020-01-03	kolkata	37	8	evening
6	2019-01-12	mumbai	22	12	morning
7	2019-01-12	mumbai	20	10	evening
8	2019-01-12	delhi	18	9	morning
9	2019-01-12	delhi	19	7	evening
10	2019-01-12	kolkata	21	7	morning
11	2019-01-12	kolkata	23	10	evening

[[[[[[Grouper]]]]]]

```
import pandas as pd
```

Convert 'date' to datetime if not already

```
df['date'] = pd.to_datetime(df['date'])

# Create pivot table with yearly aggregation
pivot = df.pivot_table(
    index=pd.Grouper(freq='Y', key='date'), # Yearly grouping
    columns='city',
    values=['temperature', 'windspeed'], # Values to aggregate
    aggfunc='mean' # Default is mean, but
    explicit here
)
```

```
print(pivot)
```

	temperature			windspeed		
city	delhi	kolkata	mumbai	delhi	kolkata	mumbai
date						
2019-12-31	18.5	22.0	21.0	8.0	8.5	11.0
2020-12-31	41.0	37.5	42.5	8.0	7.0	10.0

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\4142982490.py:10:
FutureWarning: 'Y' is deprecated and will be removed in a future
version, please use 'YE' instead.

```
index=pd.Grouper(freq='Y', key='date'), # Yearly grouping
```

```
df = pd.DataFrame({
    "first":list("aaabbbccd"),
    "second":list("xyzxyzxyy"),
    "third":[1,2,3,4,5,6,7,8,9]
})
df
```

	first	second	third
0	a	x	1
1	a	y	2
2	a	z	3
3	b	x	4
4	b	y	5
5	b	z	6
6	c	x	7
7	c	y	8
8	d	y	9

```
df.pivot_table(index='first', columns='second')
```

	third		
second	x	y	z
first			
a	1.0	2.0	3.0
b	4.0	5.0	6.0
c	7.0	8.0	NaN
d	NaN	9.0	NaN


```
df.pivot_table(index= 'first' , columns = 'second', fill_value="NILL Data")
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1873935781.py:1:

FutureWarning: Downcasting object dtype arrays

on .fillna, .ffill, .bfill is deprecated and will change in a future version. Call result.infer_objects(copy=False) instead. To opt-in to the future behavior, set

```
`pd.set_option('future.no_silent_downcasting', True)`
```

```
df.pivot_table(index= 'first' , columns = 'second', fill_value="NILL Data")
```

second	third		
	x	y	z
first			
a	1.0	2.0	3.0
b	4.0	5.0	6.0
c	7.0	8.0	NILL Data
d	NILL Data	9.0	NILL Data

Reshape DataFrame using melt

```
df = pd.read_csv('subject_melt.csv')  
df
```

	subjects	arun	varun	neha
0	maths	72	88	87
1	physics	92	74	81
2	chemistry	55	69	78
3	biology	82	77	89
4	computers	68	71	76

```
df1 = pd.melt(df, id_vars=['subjects'])  
df1
```

	subjects	variable	value
0	maths	arun	72
1	physics	arun	92
2	chemistry	arun	55
3	biology	arun	82
4	computers	arun	68
5	maths	varun	88
6	physics	varun	74
7	chemistry	varun	69
8	biology	varun	77
9	computers	varun	71
10	maths	neha	87
11	physics	neha	81
12	chemistry	neha	78
13	biology	neha	89
14	computers	neha	76

```
df1[df1['subjects']=='maths']
```

	subjects	variable	value
0	maths	arun	72
5	maths	varun	88
10	maths	neha	87

```
# Melt for only one column
```

```
df1 = pd.melt(df, id_vars=['subjects'], value_vars=['arun'])  
df1
```

	subjects	variable	value
0	maths	arun	72
1	physics	arun	92
2	chemistry	arun	55
3	biology	arun	82
4	computers	arun	68

```
# Melt multiple columns
```

```
df1 = pd.melt(df, id_vars=['subjects'], value_vars=['arun','neha'])  
df1
```

	subjects	variable	value
0	maths	arun	72
1	physics	arun	92
2	chemistry	arun	55
3	biology	arun	82
4	computers	arun	68
5	maths	neha	87
6	physics	neha	81
7	chemistry	neha	78
8	biology	neha	89
9	computers	neha	76

```
df1 = pd.melt(df, id_vars=['subjects'], var_name='nAMe',  
value_name='Marks')  
df1
```

	subjects	nAMe	Marks
0	maths	arun	72
1	physics	arun	92
2	chemistry	arun	55
3	biology	arun	82
4	computers	arun	68
5	maths	varun	88
6	physics	varun	74
7	chemistry	varun	69
8	biology	varun	77
9	computers	varun	71
10	maths	neha	87
11	physics	neha	81

```
12 chemistry neha 78
13 biology neha 89
14 computers neha 76
```

Reshaping using stack and unstack
converting columns to rows and rows to columns by stack and unstack respectively

```
df
```

	subjects	arun	varun	neha
0	maths	72	88	87
1	physics	92	74	81
2	chemistry	55	69	78
3	biology	82	77	89
4	computers	68	71	76

```
df = pd.read_excel('students_stack.xlsx', header=[0,1,2], index_col=0)
df
```

	name					
	arun			varun		
	maths	physics	chemistry	maths	physics	chemistry
sem1	60	63	62	58	66	65
sem2	58	61	60	56	64	63
sem3	62	65	64	60	68	67
sem4	67	70	69	65	73	72

df.stack() # by default stacks the last level of header , here last header is row 2

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1380500203.py:1:
FutureWarning: The previous implementation of stack is deprecated and
will be removed in a future version of pandas. See the What's New
notes for pandas 2.1.0 for details. Specify future_stack=True to adopt
the new implementation and silence this warning.
```

df.stack() # by default stacks the last level of header , here last header is row 2

	name		
	arun varun		
sem1	chemistry	62	65
	maths	60	58
	physics	63	66
sem2	chemistry	60	63
	maths	58	56
	physics	61	64
sem3	chemistry	64	67
	maths	62	60
	physics	65	68
sem4	chemistry	69	72

maths	67	65
physics	70	73

```
df_stacked = df.stack(level=1)
df_stacked
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\2980282631.py:1:
FutureWarning: The previous implementation of stack is deprecated and will be removed in a future version of pandas. See the What's New notes for pandas 2.1.0 for details. Specify future_stack=True to adopt the new implementation and silence this warning.

```
df_stacked = df.stack(level=1)
```

	name	maths	physics	chemistry
sem1	arun	60	63	62
	varun	58	66	65
sem2	arun	58	61	60
	varun	56	64	63
sem3	arun	62	65	64
	varun	60	68	67
sem4	arun	67	70	69
	varun	65	73	72

Alternatively

```
df_stacked = df.stack(1)
df_stacked
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\457864059.py:2:
FutureWarning: The previous implementation of stack is deprecated and will be removed in a future version of pandas. See the What's New notes for pandas 2.1.0 for details. Specify future_stack=True to adopt the new implementation and silence this warning.

```
df_stacked = df.stack(1)
```

	name	maths	physics	chemistry
sem1	arun	60	63	62
	varun	58	66	65
sem2	arun	58	61	60
	varun	56	64	63
sem3	arun	62	65	64
	varun	60	68	67
sem4	arun	67	70	69
	varun	65	73	72

Stack on multiple levels of column

```
df_stacked = df.stack(level=[1,2]) # Or, df_stacked([1,2])
df_stacked                          # df_stacked
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\155454708.py:2:
FutureWarning: The previous implementation of stack is deprecated and
will be removed in a future version of pandas. See the What's New
notes for pandas 2.1.0 for details. Specify future_stack=True to adopt
the new implementation and silence this warning.
```

```
df_stacked = df.stack(level=[1,2]) # Or, df_stacked([1,2])
```

		name	
sem1	arun	chemistry	62
		maths	60
		physics	63
	varun	chemistry	65
		maths	58
		physics	66
sem2	arun	chemistry	60
		maths	58
		physics	61
	varun	chemistry	63
		maths	56
		physics	64
sem3	arun	chemistry	64
		maths	62
		physics	65
	varun	chemistry	67
		maths	60
		physics	68
sem4	arun	chemistry	69
		maths	67
		physics	70
	varun	chemistry	72
		maths	65
		physics	73

```
# Stack by default removes the row with all missing values
# but still we can control the behaviour by parameter called 'dropna'
```

```
columns2 = pd.MultiIndex.from_tuples([('weight', 'kilogram'),
                                      ('height', 'meter')]) # First
column: Level 0='weight', Level 1='kilogram'
animals_df = pd.DataFrame([[1.3, None], [8, 2.8]], # Second
column: Level 0='height', Level 1='meter'
index=['rat', 'dog'],
columns= columns2
) #columns=columns2 sets
```

```
the column headers
```

```
animals_df
```

	weight	height
	kilogram	meter

rat	1.3	NaN
dog	8.0	2.8

```
animals_df.stack() #Header is used i.e, kilogram and meter
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\34287367.py:1:
FutureWarning: The previous implementation of stack is deprecated and will be removed in a future version of pandas. See the What's New notes for pandas 2.1.0 for details. Specify future_stack=True to adopt the new implementation and silence this warning.

```
animals_df.stack() #Header is used i.e, kilogram and meter
```

		weight	height
rat	kilogram	1.3	NaN
dog	kilogram	8.0	NaN
	meter	NaN	2.8

```
animals_df.stack(dropna = False) # rat [height and meter both are none]
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1013946429.py:1:
FutureWarning: The previous implementation of stack is deprecated and will be removed in a future version of pandas. See the What's New notes for pandas 2.1.0 for details. Specify future_stack=True to adopt the new implementation and silence this warning.

```
animals_df.stack(dropna = False) # rat [height and meter both are none]
```

		weight	height
rat	kilogram	1.3	NaN
	meter	NaN	NaN
dog	kilogram	8.0	NaN
	meter	NaN	2.8

Unstack the stacked DataFrame

```
df = pd.read_excel('students_stack.xlsx', header=[0,1,2], index_col=0)  
df
```

	name			varun		
	arun			maths	physics	chemistry
	maths	physics	chemistry	maths	physics	chemistry
sem1	60	63	62	58	66	65
sem2	58	61	60	56	64	63
sem3	62	65	64	60	68	67
sem4	67	70	69	65	73	72

```
df_stacked = df.stack(2)  
df_stacked
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1602624969.py:1:
FutureWarning: The previous implementation of stack is deprecated and

will be removed in a future version of pandas. See the What's New notes for pandas 2.1.0 for details. Specify `future_stack=True` to adopt the new implementation and silence this warning.

```
df_stacked = df.stack(2)
```

		name	
		arun	varun
sem1	chemistry	62	65
	maths	60	58
	physics	63	66
sem2	chemistry	60	63
	maths	58	56
	physics	61	64
sem3	chemistry	64	67
	maths	62	60
	physics	65	68
sem4	chemistry	69	72
	maths	67	65
	physics	70	73

```
df_stacked.unstack()
```

	name					
	arun			varun		
	chemistry	maths	physics	chemistry	maths	physics
sem1	62	60	63	65	58	66
sem2	60	58	61	63	56	64
sem3	64	62	65	67	60	68
sem4	69	67	70	72	65	73

```
df = pd.read_excel('students_stack.xlsx', header=[0,1,2], index_col=0)
df
```

	name					
	arun			varun		
	maths	physics	chemistry	maths	physics	chemistry
sem1	60	63	62	58	66	65
sem2	58	61	60	56	64	63
sem3	62	65	64	60	68	67
sem4	67	70	69	65	73	72

```
df_stacked = df.stack(level = [1,2])
df_stacked
```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\425497088.py:1:
FutureWarning: The previous implementation of stack is deprecated and will be removed in a future version of pandas. See the What's New notes for pandas 2.1.0 for details. Specify `future_stack=True` to adopt the new implementation and silence this warning.

```
df_stacked = df.stack(level = [1,2])
```

			name
sem1	arun	chemistry	62
		maths	60
		physics	63
	varun	chemistry	65
		maths	58
		physics	66
sem2	arun	chemistry	60
		maths	58
		physics	61
	varun	chemistry	63
		maths	56
		physics	64
sem3	arun	chemistry	64
		maths	62
		physics	65
	varun	chemistry	67
		maths	60
		physics	68
sem4	arun	chemistry	69
		maths	67
		physics	70
	varun	chemistry	72
		maths	65
		physics	73

```
df_stacked.unstack(level=0) # sem index will be column i.e, 0
```

		name	sem1	sem2	sem3	sem4
arun	chemistry	62	60	64	69	
	maths	60	58	62	67	
	physics	63	61	65	70	
varun	chemistry	65	63	67	72	
	maths	58	56	60	65	
	physics	66	64	68	73	

```
df_stacked.unstack(1)
```

		name	arun	varun
sem1	chemistry	62	65	
	maths	60	58	
	physics	63	66	
sem2	chemistry	60	63	
	maths	58	56	
	physics	61	64	
sem3	chemistry	64	67	
	maths	62	60	
	physics	65	68	

sem4	chemistry	69	72
	maths	67	65
	physics	70	73

Unstack multiple indexes

```
df_stacked.unstack(level=[1,2])
```

	name					
	arun			varun		
	chemistry	maths	physics	chemistry	maths	physics
sem1	62	60	63	65	58	66
sem2	60	58	61	63	56	64
sem3	64	62	65	67	60	68
sem4	69	67	70	72	65	73

Frequency distribution of DataFrame column

interrelation between two variables and can help find interactions between them

```
df = pd.read_csv('hair_color.csv')
df
```

	name	country	gender	age	hair_color
0	Ram	India	M	23	black
1	Mathew	UK	M	27	brown
2	Gillian	UK	F	43	brown
3	Tom	USA	M	33	brown
4	Anna	USA	F	25	blonde
5	Sophia	USA	F	27	blonde
6	Emma	UK	F	52	blonde
7	Sweta	India	F	23	black
8	Mohan	India	M	44	black
9	Amelia	UK	F	24	blonde

```
pd.crosstab(df.country, df.hair_color)
```

The 1st parameter to crosstab function is the 'index' and the 2nd parameter is

the 'column' on which we want to apply the frequency distribution

hair_color	black	blonde	brown
country			
India	3	0	0
UK	0	2	2
USA	0	2	1

```
pd.crosstab(df.gender, df.hair_color) # the parameter passed are the
DataFrame column (as df.gender etc.)
```

hair_color	black	blonde	brown
gender			
F	1	4	1
M	2	0	2

```
pd.crosstab(df.gender, df.hair_color, margins = True) # Get total of rows/columns
```

hair_color	black	blonde	brown	All
gender				
F	1	4	1	6
M	2	0	2	4
All	3	4	3	10

Multilevel columns

```
pd.crosstab(df.country, [df.gender, df.hair_color]) # M has no blonde color
```

gender	F			M	
	black	blonde	brown	black	brown
country					
India	1	0	0	2	0
UK	0	2	1	0	1
USA	0	2	0	0	1

Multilevel indexes

```
pd.crosstab([df.gender, df.country], df.hair_color) #Index are df.gender, df.country
```

hair_color		black	blonde	brown
gender	country			
F	India	1	0	0
	UK	0	2	1
	USA	0	2	0
M	India	2	0	0
	UK	0	0	1
	USA	0	0	1

```
pd.crosstab([df.gender, df.country],
             df.hair_color,
             rownames = ['GENDER', 'COUNTRY'],
             colnames = ['HAIRcolor'])
```

We have provided two names to rownames as we have two indexes and 1 to

colnames as we have only one column

HAIRcolor		black	blonde	brown
GENDER	COUNTRY			
F	India	1	0	0

	UK	0	2	1
	USA	0	2	0
M	India	2	0	0
	UK	0	0	1
	USA	0	0	1

Normalize (percentage) of the frequency

```
pd.crosstab([df.gender], df.hair_color, normalize='index')
```

hair_color	black	blonde	brown
gender			
F	0.166667	0.666667	0.166667
M	0.500000	0.000000	0.500000

```
pd.crosstab(df.gender,
            df.hair_color,
            values = df.age,
            aggfunc = 'mean')
```

hair_color	black	blonde	brown
gender			
F	23.0	32.0	43.0
M	33.5	NaN	30.0

Drop unwanted rows/columns

```
df = pd.read_csv('hair_color.csv')
df
```

	name	country	gender	age	hair_color
0	Ram	India	M	23	black
1	Mathew	UK	M	27	brown
2	Gillian	UK	F	43	brown
3	Tom	USA	M	33	brown
4	Anna	USA	F	25	blonde
5	Sophia	USA	F	27	blonde
6	Emma	UK	F	52	blonde
7	Sweta	India	F	23	black
8	Mohan	India	M	44	black
9	Amelia	UK	F	24	blonde

```
df_drop = pd.crosstab([df.gender, df.country], df.hair_color)
df_drop
```

hair_color	black	blonde	brown
gender country			
F India	1	0	0
UK	0	2	1
USA	0	2	0
M India	2	0	0

UK	0	0	1
USA	0	0	1

```
df_drop.drop('M') # by default axis = 0 [row]
```

hair_color		black	blonde	brown
gender	country			
F	India	1	0	0
	UK	0	2	1
	USA	0	2	0

```
#Delete rows of custom index level
```

```
df_drop.drop(index='India', level= 1)
```

```
# we have deleted the row with index 'India' at level 1
```

hair_color		black	blonde	brown
gender	country			
F	UK	0	2	1
	USA	0	2	0
M	UK	0	0	1
	USA	0	0	1

```
# Delete multiple rows
```

```
df_drop.drop(index= ['India','UK'], level=1)
```

hair_color		black	blonde	brown
gender	country			
F	USA	0	2	0
M	USA	0	0	1

```
# Drop column
```

```
# These operation are related to crosstab
```

```
df_drop.drop(columns= 'brown')
```

hair_color		black	blonde
gender	country		
F	India	1	0
	UK	0	2
	USA	0	2
M	India	2	0
	UK	0	0
	USA	0	0

```
# Alternatively
```

```
df_drop.drop('brown', axis=1)
```

hair_color		black	blonde
gender	country		

F	India	1	0
	UK	0	2
	USA	0	2
M	India	2	0
	UK	0	0
	USA	0	0

Delete multilevel columns

```
df_drop = pd.crosstab(df.country,
                      [df.gender, df.hair_color])
```

df_drop

gender	F			M	
hair_color	black	blonde	brown	black	brown
country					
India	1	0	0	2	0
UK	0	2	1	0	1
USA	0	2	0	0	1

```
df_drop.drop(columns = ['black', 'blonde'] , level=1 ) #all column
with 'black' and 'brown' name were removed
```

gender	F	M
hair_color	brown	brown
country		
India	0	0
UK	1	1
USA	0	1

```
df_drop = pd.crosstab([df.country], [df.gender,df.hair_color])
```

df_drop

gender	F			M	
hair_color	black	blonde	brown	black	brown
country					
India	1	0	0	2	0
UK	0	2	1	0	1
USA	0	2	0	0	1

Delete both rows & columns

pass both 'index' (rows) and 'columns' as parameter
to delete both rows and columns together

```
df_drop.drop(index= 'UK' , columns='M')
```

gender	F		
hair_color	black	blonde	brown
country			
India	1	0	0
USA	0	2	0

Remove duplicate values

```
df = pd.DataFrame({
    "name": ['arun', 'varun', 'neha', 'varun', 'varun', 'arun'],
    'instruments': ['violin', 'drum', 'flute', 'guitar', 'bongo', 'tabla'],
    'start_date': ['Jan 10, 2020', 'Mar 3, 2003', 'Feb 6, 2005', 'Dec 8,
2008',
    'Nov 5, 2011', 'Mar 10, 2011']
})
df.start_date= pd.to_datetime(df.start_date)
df
```

	name	instruments	start_date
0	arun	violin	2020-01-10
1	varun	drum	2003-03-03
2	neha	flute	2005-02-06
3	varun	guitar	2008-12-08
4	varun	bongo	2011-11-05
5	arun	tabla	2011-03-10

Remove duplicate

```
df.drop_duplicates(subset = 'name') # remaining rows are the rows
with their first occurrence
```

	name	instruments	start_date
0	arun	violin	2020-01-10
1	varun	drum	2003-03-03
2	neha	flute	2005-02-06

ALternatively

```
df.drop_duplicates('name')
```

	name	instruments	start_date
0	arun	violin	2020-01-10
1	varun	drum	2003-03-03
2	neha	flute	2005-02-06

Fetch custom occurrence of data

```
df.drop_duplicates('name', keep = 'first') #by default keeps first
occurence
```

	name	instruments	start_date
0	arun	violin	2020-01-10
1	varun	drum	2003-03-03
2	neha	flute	2005-02-06

```
df
```

	name	instruments	start_date
0	arun	violin	2020-01-10

```
1  varun      drum  2003-03-03
2   neha      flute 2005-02-06
3  varun      guitar 2008-12-08
4  varun      bongo  2011-11-05
5   arun      tabla  2011-03-10
```

```
df.drop_duplicates('name', keep= 'last') #Index has no sequence,which
problem will solve by ignore_index
```

```
   name instruments start_date
2  neha      flute  2005-02-06
4  varun      bongo  2011-11-05
5   arun      tabla  2011-03-10
```

Remove all duplicates

```
df.drop_duplicates('name' , keep = False)
```

```
   name instruments start_date
2  neha      flute  2005-02-06
```

while deleting the duplicates the indexes becomes nonsequential.This can be

handled by a parameter 'ignore_index'.

non-sequential : not following a particular order, or not following one after the other in order

#Ignore Index

```
df.drop_duplicates('name',
                  keep='last',
                  ignore_index=True) #index 0,1,2
```

```
   name instruments start_date
0  neha      flute  2005-02-06
1  varun      bongo  2011-11-05
2   arun      tabla  2011-03-10
```

Sort the data

```
df = pd.DataFrame({
    'alphabet': list('dpbtkbc'),
    'num1': [1,2,np.nan,4,3,7,2],
    'num2': [3,4,3,4,7,5,4]
})
```

df #In book 'num2':[3,4,3,4,2,5,4] which is changed for understanding if 2nd column has no order with 1st columns

```
   alphabet  num1  num2
0         d    1.0     3
1         p    2.0     4
2         b    NaN     3
```

3	t	4.0	4
4	b	3.0	7
5	k	7.0	5
6	c	2.0	4

```
#Sort columns
# columns: axis=0 [by default]
# rows: axis=1 [if we need]
```

```
df.sort_values(by='alphabet')
```

```
# OR, df.sort_values(by=['alphabet'])
# OR, df.sort_values('alphabet')
```

	alphabet	num1	num2
2	b	NaN	3
4	b	3.0	7
6	c	2.0	4
0	d	1.0	3
5	k	7.0	5
1	p	2.0	4
3	t	4.0	4

```
# Sorting multiple columns
df.sort_values(by=['alphabet', 'num2'])
```

```
# OR, df.sort_values(by=['alphabet', 'num2']) ; 1st sort alphabet
column then if
# possible num2 is sorted(maybe all values will not be sorted)
```

	alphabet	num1	num2
2	b	NaN	3
4	b	3.0	7
6	c	2.0	4
0	d	1.0	3
5	k	7.0	5
1	p	2.0	4
3	t	4.0	4

```
# Sorting order
```

```
df.sort_values(by='alphabet', ascending=False) # descending
```

	alphabet	num1	num2
3	t	4.0	4
1	p	2.0	4
5	k	7.0	5
0	d	1.0	3
6	c	2.0	4
2	b	NaN	3
4	b	3.0	7


```
# by default missing value is at the end
```

```
df.sort_values('num1')
```

	alphabet	num1	num2
0	d	1.0	3
1	p	2.0	4
6	c	2.0	4
4	b	3.0	7
3	t	4.0	4
5	k	7.0	5
2	b	NaN	3

```
# Positioning missing value
```

```
df.sort_values('num1',na_position = 'first')
```

	alphabet	num1	num2
2	b	NaN	3
0	d	1.0	3
1	p	2.0	4
6	c	2.0	4
4	b	3.0	7
3	t	4.0	4
5	k	7.0	5

```
# Working with date and time
```

```
df = pd.read_csv('TCS_data.csv')  
df
```

	Date	Open	High	Low	Close	Volume
0	2019/04/01	2010.00	2039.95	2008.25	2031.65	2095740
1	2019/04/02	2037.10	2086.00	2037.00	2079.30	3719663
2	2019/04/03	2085.00	2089.60	2058.10	2079.30	2939886
3	2019/04/04	2078.15	2079.70	2007.40	2014.50	4397518
4	2019/04/05	2028.65	2054.40	2018.80	2048.30	3152103
...
240	2020/03/24	1653.05	1770.00	1632.85	1703.15	6354209
241	2020/03/25	1700.00	1810.00	1680.00	1750.30	2765527
242	2020/03/26	1831.60	1832.05	1722.55	1790.95	4556071
243	2020/03/27	1820.00	1850.00	1750.40	1824.50	4331310
244	2020/03/30	1766.00	1905.00	1763.55	1778.50	8513608

```
[245 rows x 6 columns]
```

```
type(df['Date'][0])
```

```
str
```

```
# Converting date to timestamp and set as index
```

```
df = pd.read_csv('TCS_data.csv', parse_dates=['Date'],
```

```
index_col='Date')
df.head()
```

	Open	High	Low	Close	Volume
Date					
2019-04-01	2010.00	2039.95	2008.25	2031.65	2095740
2019-04-02	2037.10	2086.00	2037.00	2079.30	3719663
2019-04-03	2085.00	2089.60	2058.10	2079.30	2939886
2019-04-04	2078.15	2079.70	2007.40	2014.50	4397518
2019-04-05	2028.65	2054.40	2018.80	2048.30	3152103

```
df.index
```

```
DatetimeIndex(['2019-04-01', '2019-04-02', '2019-04-03', '2019-04-04',
                '2019-04-05', '2019-04-08', '2019-04-09', '2019-04-10',
                '2019-04-11', '2019-04-12',
                ...,
                '2020-03-17', '2020-03-18', '2020-03-19', '2020-03-20',
                '2020-03-23', '2020-03-24', '2020-03-25', '2020-03-26',
                '2020-03-27', '2020-03-30'],
              dtype='datetime64[ns]', name='Date', length=245,
```

```
freq=None)
```

```
# Access data for particular year
```

```
df.loc['2020']
```

	Open	High	Low	Close	Volume
Date					
2020-01-01	2168.00	2183.90	2154.00	2167.60	1354908
2020-01-02	2179.95	2179.95	2149.20	2157.65	2380752
2020-01-03	2164.00	2223.00	2164.00	2200.65	4655761
2020-01-06	2205.00	2225.95	2187.90	2200.45	3023209
2020-01-07	2200.50	2214.65	2183.80	2205.85	2429317
...
2020-03-24	1653.05	1770.00	1632.85	1703.15	6354209
2020-03-25	1700.00	1810.00	1680.00	1750.30	2765527
2020-03-26	1831.60	1832.05	1722.55	1790.95	4556071
2020-03-27	1820.00	1850.00	1750.40	1824.50	4331310
2020-03-30	1766.00	1905.00	1763.55	1778.50	8513608

```
[63 rows x 5 columns]
```

```
# Access data for particular month
```

```
df.loc['2020-1'].head()
```

	Open	High	Low	Close	Volume
Date					
2020-01-01	2168.00	2183.90	2154.0	2167.60	1354908
2020-01-02	2179.95	2179.95	2149.2	2157.65	2380752
2020-01-03	2164.00	2223.00	2164.0	2200.65	4655761

```
2020-01-06  2205.00  2225.95  2187.9  2200.45  3023209
2020-01-07  2200.50  2214.65  2183.8  2205.85  2429317
```

```
df.loc['2020-01'].Close.mean() # Calculating average closing price for
any month
```

```
2188.8934782608694
```

```
# Access a date range
```

```
df.loc['2020-01-01':'2020-01-07']
```

	Open	High	Low	Close	Volume
Date					
2020-01-01	2168.00	2183.90	2154.0	2167.60	1354908
2020-01-02	2179.95	2179.95	2149.2	2157.65	2380752
2020-01-03	2164.00	2223.00	2164.0	2200.65	4655761
2020-01-06	2205.00	2225.95	2187.9	2200.45	3023209
2020-01-07	2200.50	2214.65	2183.8	2205.85	2429317

```
# Resampling the data
```

```
df.Close.resample('M').mean() # 'M'
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1914949084.py:2:
FutureWarning: 'M' is deprecated and will be removed in a future
version, please use 'ME' instead.
```

```
df.Close.resample('M').mean() # 'M'
```

Date	
2019-04-30	2109.397368
2019-05-31	2120.722727
2019-06-30	2241.686842
2019-07-31	2145.432609
2019-08-31	2219.017500
2019-09-30	2131.200000
2019-10-31	2080.752500
2019-11-30	2131.205000
2019-12-31	2134.290000
2020-01-31	2188.893478
2020-02-29	2146.592500
2020-03-31	1841.205000

```
Freq: ME, Name: Close, dtype: float64
```

```
#Business days are the weekdays when most companies operate, typically
Monday through Friday, excluding weekends
# and public holidays
```

```
#'WOM' The X-th Day of the Y-th Week of Each Month
```

```
# Y = The week number (1st, 2nd, 3rd, 4th, or last week).
```

```
# X = The day of the week (Monday=0, Tuesday=1, ..., Sunday=6)
```

```
# 'M' calendar month end
```

```
# 'M': "calendar month end" refers to the last day of each month

#'BM' stands for "Business Month End", which refers to the last
business day (weekday) of each month

#'SM' always generates two timestamps per month:
# 15th day (mid-month).
# Last calendar day (e.g., 2019-03-31)

# 'Q' stands for calendar quarter-end, which marks the last day of
each financial quarter (regardless of
# weekends/holidays)
```

```
df.Close.resample('SM').mean()
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3815591759.py:1:
FutureWarning: 'SM' is deprecated and will be removed in a future
version, please use 'SME' instead.
  df.Close.resample('SM').mean()
```

```
Date
2019-03-31    2048.955000
2019-04-15    2166.081250
2019-04-30    2159.985000
2019-05-15    2093.320833
2019-05-31    2222.360000
2019-06-15    2256.500000
2019-06-30    2176.215000
2019-07-15    2114.758333
2019-07-31    2217.620000
2019-08-15    2219.077273
2019-08-31    2190.275000
2019-09-15    2087.130000
2019-09-30    2042.105556
2019-10-15    2096.886364
2019-10-31    2184.650000
2019-11-15    2095.204545
2019-11-30    2064.638889
2019-12-15    2194.235000
2019-12-31    2197.663636
2020-01-15    2187.741667
2020-01-31    2144.833333
2020-02-15    2141.433333
2020-02-29    1984.933333
2020-03-15    1723.609091
Freq: SME-15, Name: Close, dtype: float64
```

```
# Quarterly frequency
df.Close.resample('Q').mean() # Jan,Feb,Mar=1 quarter
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\365181218.py:2:
FutureWarning: 'Q' is deprecated and will be removed in a future
version, please use 'QE' instead.
```

```
df.Close.resample('Q').mean() # Jan,Feb,Mar=1 quarter
```

```
Date
```

```
2019-06-30    2155.441667
```

```
2019-09-30    2164.808065
```

```
2019-12-31    2115.415833
```

```
2020-03-31    2065.087302
```

```
Freq: QE-DEC, Name: Close, dtype: float64
```

```
df.Close.resample('A').mean()
```

```
#'A' stands for calendar year-end, which refers to December 31 of each
year (regardless of weekends/holidays)
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\36632666.py:1:
```

```
FutureWarning: 'A' is deprecated and will be removed in a future
version, please use 'YE' instead.
```

```
df.Close.resample('A').mean()
```

```
Date
```

```
2019-12-31    2145.437088
```

```
2020-12-31    2065.087302
```

```
Freq: YE-DEC, Name: Close, dtype: float64
```

```
# Plotting the resampled data
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
# Assuming df is your DataFrame with a datetime index and 'Close'
column
```

```
monthly_avg = df['Close'].resample('M').mean() # Calculate monthly
averages
```

```
# Create the plot with proper formatting
```

```
ax = monthly_avg.plot(
    figsize=(12, 6),           # Set figure size
    title='Monthly Average Closing Price', # Add title
    grid=True,                # Show grid lines
    color='blue',             # Line color
    linewidth=2,              # Line thickness
    style='--o',              # Line with circle markers
    markersize=5,             # Marker size
    alpha=0.8,                # Slight transparency
)
```

```
# Customize the plot further
```

```
ax.set_xlabel('Date', fontsize=12)
```

```
ax.set_ylabel('Closing Price', fontsize=12)
ax.tick_params(axis='both', which='major', labelsize=10)
```

```
# Rotate x-axis labels for better readability
plt.xticks(rotation=45)
```

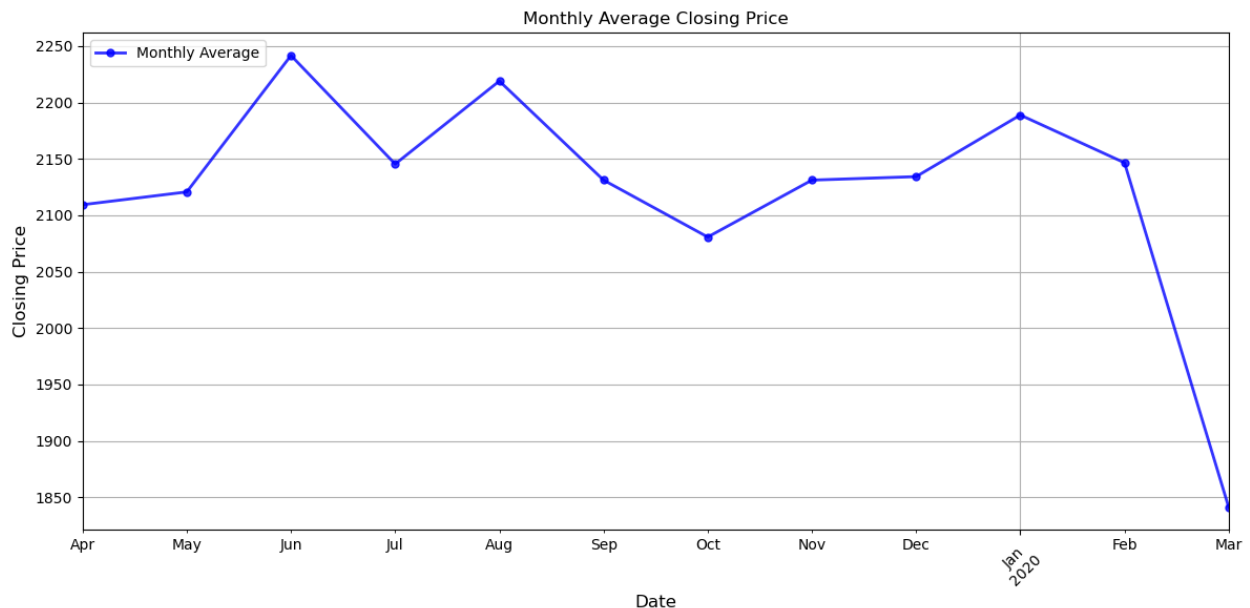
```
# Add a legend
ax.legend(['Monthly Average'], loc='upper left')
```

```
# Adjust layout to prevent label cutoff
plt.tight_layout()
```

```
# Show the plot
plt.show()
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1964578811.py:7:
FutureWarning: 'M' is deprecated and will be removed in a future
version, please use 'ME' instead.
```

```
monthly_avg = df['Close'].resample('M').mean() # Calculate monthly
averages
```



```
df.Close.resample('Q').mean()
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3015364846.py:1:
FutureWarning: 'Q' is deprecated and will be removed in a future
version, please use 'QE' instead.
```

```
df.Close.resample('Q').mean()
```

Date

2019-06-30	2155.441667
2019-09-30	2164.808065

```
2019-12-31    2115.415833
2020-03-31    2065.087302
Freq: QE-DEC, Name: Close, dtype: float64
```

```
# df.Close.resample('Q').mean().plot(kind='bar')
```

```
df = pd.read_csv('TCS_data_withoutdate.csv')
df.head()
```

	Open	High	Low	Close	Volume
0	2168.00	2183.90	2154.0	2167.60	1354908
1	2179.95	2179.95	2149.2	2157.65	2380752
2	2164.00	2223.00	2164.0	2200.65	4655761
3	2205.00	2225.95	2187.9	2200.45	3023209
4	2200.50	2214.65	2183.8	2205.85	2429317

```
rng = pd.date_range(start='01/01/2020', end='01/31/2020', freq='B')
rng
```

```
DatetimeIndex(['2020-01-01', '2020-01-02', '2020-01-03', '2020-01-06',
               '2020-01-07', '2020-01-08', '2020-01-09', '2020-01-10',
               '2020-01-13', '2020-01-14', '2020-01-15', '2020-01-16',
               '2020-01-17', '2020-01-20', '2020-01-21', '2020-01-22',
               '2020-01-23', '2020-01-24', '2020-01-27', '2020-01-28',
               '2020-01-29', '2020-01-30', '2020-01-31'],
              dtype='datetime64[ns]', freq='B')
```

```
# Apply the above date range to our data
```

```
df.set_index(rng, inplace=True)
df.head(10)
```

	Open	High	Low	Close	Volume
2020-01-01	2168.00	2183.90	2154.00	2167.60	1354908
2020-01-02	2179.95	2179.95	2149.20	2157.65	2380752
2020-01-03	2164.00	2223.00	2164.00	2200.65	4655761
2020-01-06	2205.00	2225.95	2187.90	2200.45	3023209
2020-01-07	2200.50	2214.65	2183.80	2205.85	2429317
2020-01-08	2205.00	2260.00	2202.05	2255.25	5197454
2020-01-09	2248.75	2251.95	2210.00	2214.35	3734173
2020-01-10	2228.00	2234.00	2208.00	2213.55	1915807
2020-01-13	2217.85	2218.95	2184.70	2190.35	2843893
2020-01-14	2195.00	2229.80	2195.00	2206.90	2948452

```
rng = pd.date_range(start="01/01/2020", end="01/31/2020", freq="D")
rng
```

```
# This would have given all the dates, but this can't be applied
because we
# have data for only working days. By above method we would have
```

generated

the dates but would have not able to map it with the data.

```
DatetimeIndex(['2020-01-01', '2020-01-02', '2020-01-03', '2020-01-04',  
              '2020-01-05', '2020-01-06', '2020-01-07', '2020-01-08',  
              '2020-01-09', '2020-01-10', '2020-01-11', '2020-01-12',  
              '2020-01-13', '2020-01-14', '2020-01-15', '2020-01-16',  
              '2020-01-17', '2020-01-18', '2020-01-19', '2020-01-20',  
              '2020-01-21', '2020-01-22', '2020-01-23', '2020-01-24',  
              '2020-01-25', '2020-01-26', '2020-01-27', '2020-01-28',  
              '2020-01-29', '2020-01-30', '2020-01-31'],  
              dtype='datetime64[ns]', freq='D')
```

Generate the missing data with missing dates by 'asfreq()' function

```
df.asfreq('D').head(7)
```

	Open	High	Low	Close	Volume
2020-01-01	2168.00	2183.90	2154.0	2167.60	1354908.0
2020-01-02	2179.95	2179.95	2149.2	2157.65	2380752.0
2020-01-03	2164.00	2223.00	2164.0	2200.65	4655761.0
2020-01-04	NaN	NaN	NaN	NaN	NaN
2020-01-05	NaN	NaN	NaN	NaN	NaN
2020-01-06	2205.00	2225.95	2187.9	2200.45	3023209.0
2020-01-07	2200.50	2214.65	2183.8	2205.85	2429317.0

To fill the data we can do various methods as we learned in previous chapters

```
df.asfreq('D', method='pad') # 'pad' or 'ffill'
```

	Open	High	Low	Close	Volume
2020-01-01	2168.00	2183.90	2154.00	2167.60	1354908
2020-01-02	2179.95	2179.95	2149.20	2157.65	2380752
2020-01-03	2164.00	2223.00	2164.00	2200.65	4655761
2020-01-04	2164.00	2223.00	2164.00	2200.65	4655761
2020-01-05	2164.00	2223.00	2164.00	2200.65	4655761
2020-01-06	2205.00	2225.95	2187.90	2200.45	3023209
2020-01-07	2200.50	2214.65	2183.80	2205.85	2429317
2020-01-08	2205.00	2260.00	2202.05	2255.25	5197454
2020-01-09	2248.75	2251.95	2210.00	2214.35	3734173
2020-01-10	2228.00	2234.00	2208.00	2213.55	1915807
2020-01-11	2228.00	2234.00	2208.00	2213.55	1915807
2020-01-12	2228.00	2234.00	2208.00	2213.55	1915807
2020-01-13	2217.85	2218.95	2184.70	2190.35	2843893
2020-01-14	2195.00	2229.80	2195.00	2206.90	2948452
2020-01-15	2213.00	2231.00	2194.20	2226.90	2620681
2020-01-16	2226.95	2249.00	2215.00	2238.80	3117214
2020-01-17	2240.75	2253.55	2213.00	2219.10	3281059
2020-01-18	2240.75	2253.55	2213.00	2219.10	3281059

2020-01-19	2240.75	2253.55	2213.00	2219.10	3281059
2020-01-20	2194.90	2242.20	2156.20	2170.35	5817599
2020-01-21	2169.95	2186.55	2158.05	2171.05	1902980
2020-01-22	2181.00	2210.00	2173.70	2206.90	1773686
2020-01-23	2209.80	2217.75	2183.70	2190.95	2069866
2020-01-24	2190.95	2190.95	2170.00	2183.40	1319430
2020-01-25	2190.95	2190.95	2170.00	2183.40	1319430
2020-01-26	2190.95	2190.95	2170.00	2183.40	1319430
2020-01-27	2189.70	2193.45	2165.00	2169.25	1549101
2020-01-28	2174.00	2187.80	2152.00	2183.75	1743024
2020-01-29	2185.00	2186.95	2150.00	2154.60	2306761
2020-01-30	2160.00	2165.00	2125.00	2137.85	2098567
2020-01-31	2139.40	2144.35	2071.60	2079.05	3287223

```
df.asfreq('D', method='ffill').head(7)
```

	Open	High	Low	Close	Volume
2020-01-01	2168.00	2183.90	2154.0	2167.60	1354908
2020-01-02	2179.95	2179.95	2149.2	2157.65	2380752
2020-01-03	2164.00	2223.00	2164.0	2200.65	4655761
2020-01-04	2164.00	2223.00	2164.0	2200.65	4655761
2020-01-05	2164.00	2223.00	2164.0	2200.65	4655761
2020-01-06	2205.00	2225.95	2187.9	2200.45	3023209
2020-01-07	2200.50	2214.65	2183.8	2205.85	2429317

We can get the weekly prices too

```
df.asfreq('W', method='pad') # Sunday
```

	Open	High	Low	Close	Volume
2020-01-05	2164.00	2223.00	2164.0	2200.65	4655761
2020-01-12	2228.00	2234.00	2208.0	2213.55	1915807
2020-01-19	2240.75	2253.55	2213.0	2219.10	3281059
2020-01-26	2190.95	2190.95	2170.0	2183.40	1319430

Date range with periods (one more way to generate the date range with date_range() function)

```
rng = pd.date_range(start='01/01/2020', periods=23, freq='B') #pass 'periods' parameter instead of 'end' parameter
```

```
rng
```

```
DatetimeIndex(['2020-01-01', '2020-01-02', '2020-01-03', '2020-01-06',
                '2020-01-07', '2020-01-08', '2020-01-09', '2020-01-10',
                '2020-01-13', '2020-01-14', '2020-01-15', '2020-01-16',
                '2020-01-17', '2020-01-20', '2020-01-21', '2020-01-22',
                '2020-01-23', '2020-01-24', '2020-01-27', '2020-01-28',
                '2020-01-29', '2020-01-30', '2020-01-31'],
              dtype='datetime64[ns]', freq='B')
```

```
df.set_index(rng, inplace=True)
df.head(8)
```

	Open	High	Low	Close	Volume
2020-01-01	2168.00	2183.90	2154.00	2167.60	1354908
2020-01-02	2179.95	2179.95	2149.20	2157.65	2380752
2020-01-03	2164.00	2223.00	2164.00	2200.65	4655761
2020-01-06	2205.00	2225.95	2187.90	2200.45	3023209
2020-01-07	2200.50	2214.65	2183.80	2205.85	2429317
2020-01-08	2205.00	2260.00	2202.05	2255.25	5197454
2020-01-09	2248.75	2251.95	2210.00	2214.35	3734173
2020-01-10	2228.00	2234.00	2208.00	2213.55	1915807

Working with custom holidays

```
pd.date_range(start='12-01-2019', end='12-31-2019', freq='B') #month-
date-year
```

#output:

working date

```
DatetimeIndex(['2019-12-02', '2019-12-03', '2019-12-04', '2019-12-05',
                '2019-12-06', '2019-12-09', '2019-12-10', '2019-12-11',
                '2019-12-12', '2019-12-13', '2019-12-16', '2019-12-17',
                '2019-12-18', '2019-12-19', '2019-12-20', '2019-12-23',
                '2019-12-24', '2019-12-25', '2019-12-26', '2019-12-27',
                '2019-12-30', '2019-12-31'],
              dtype='datetime64[ns]', freq='B')
```

Adding US holidays

```
from pandas.tseries.holiday import USFederalHolidayCalendar
from pandas.tseries.offsets import CustomBusinessDay
```

```
usb = CustomBusinessDay(calendar = USFederalHolidayCalendar())
usb
```

```
<CustomBusinessDay>
```

The 25th has been removed as it was Christmas in US

```
rng = pd.date_range(start='12/01/2019', end='12/31/2019', freq=usb)
rng
```

```
DatetimeIndex(['2019-12-02', '2019-12-03', '2019-12-04', '2019-12-05',
                '2019-12-06', '2019-12-09', '2019-12-10', '2019-12-11',
                '2019-12-12', '2019-12-13', '2019-12-16', '2019-12-17',
                '2019-12-18', '2019-12-19', '2019-12-20', '2019-12-23',
                '2019-12-24', '2019-12-26', '2019-12-27', '2019-12-30',
                '2019-12-31'],
              dtype='datetime64[ns]', freq='C')
```

Creating custom calendar

```

from pandas.tseries.holiday import Holiday, \
AbstractHolidayCalendar, \
nearest_workday

class myHolidayCalendar(AbstractHolidayCalendar):
    rules = [
        Holiday('ExampleHoliday1', month=1, day=2),
        Holiday('ExampleHoliday1', month=1, day=7)
    ]
myc = CustomBusinessDay(calendar= myHolidayCalendar())
myc

<CustomBusinessDay>

pd.date_range(start='1-1-2020',end='1/11/2020', freq=myc)

# we have provided as holidays in myHolidayCalendar (2nd and 7th) have
been removed from the date range

DatetimeIndex(['2020-01-01', '2020-01-03', '2020-01-06', '2020-01-08',
               '2020-01-09', '2020-01-10'],
              dtype='datetime64[ns]', freq='C')

# Observance rule
from pandas.tseries.holiday import Holiday, AbstractHolidayCalendar,
nearest_workday
from pandas.tseries.offsets import CustomBusinessDay

class myHolidayCalendar(AbstractHolidayCalendar):
    rules = [
        Holiday('ExampleHoliday1', month=1, day=5,
observance=nearest_workday),
        Holiday('ExampleHoliday2', month=1, day=9)
    ]

myc = CustomBusinessDay(calendar=myHolidayCalendar())
myc

<CustomBusinessDay>

# The 5th was sunday thus the nearest weekday was 6th(monday)
# If we would have provided holiday on 4th then 3rd would have been
nearest holiday

pd.date_range(start='1-1-2020',end='1-11-2020', freq= myc)

DatetimeIndex(['2020-01-01', '2020-01-02', '2020-01-03', '2020-01-07',
               '2020-01-08', '2020-01-10'],
              dtype='datetime64[ns]', freq='C')

# Custom week days

```

```

custom_weekdays = 'Sun Mon Tue Wed Thu'
custom_businessDays = CustomBusinessDay(weekmask=custom_weekdays)
pd.date_range(start='1/1/2020', end='1/11/2020', freq=
custom_businessDays)

# In This example Friday(3rd) and Saturday(4th) are the weekends

DatetimeIndex(['2020-01-01', '2020-01-02', '2020-01-05', '2020-01-06',
               '2020-01-07', '2020-01-08', '2020-01-09'],
              dtype='datetime64[ns]', freq='C')

# Custom holiday

custom_weekdays = 'Sun Mon Tue Wed Thu'
custom_businessDays = CustomBusinessDay(weekmask=custom_weekdays ,
holidays=['2020-01-06'])
pd.date_range(start='1/1/2020', end='1/11/2020', freq=
custom_businessDays)

# passed 6th as a holiday to 'holiday'

DatetimeIndex(['2020-01-01', '2020-01-02', '2020-01-05', '2020-01-07',
               '2020-01-08', '2020-01-09'],
              dtype='datetime64[ns]', freq='C')

# Working with date formats

#Converting to a common format
import pandas as pd

dates = ['2020-03-10', 'Mar 10, 2020', '03/10/2020', '2020.03.10',
         '2020/03/10', '20200310']
pd.to_datetime(dates, format='mixed')

DatetimeIndex(['2020-03-10', '2020-03-10', '2020-03-10', '2020-03-10',
               '2020-03-10', '2020-03-10'],
              dtype='datetime64[ns]', freq=None)

# Time conversion

dates = ['2020-03-10 04:30:00 PM',
         'Mar 10, 2020 16:30:00',
         '03/10/2020',
         '2020.03.10',
         '2020/03/10',
         '20200310']
pd.to_datetime(dates, format='mixed') #both the different time
formats (04:30 PM and 16:30) have become a single format (16:30)

DatetimeIndex(['2020-03-10 16:30:00', '2020-03-10 16:30:00',
               '2020-03-10 00:00:00', '2020-03-10 00:00:00',

```

```

                '2020-03-10 00:00:00', '2020-03-10 00:00:00'],
                dtype='datetime64[ns]', freq=None)

#10th March, 2020
date = '10/03/2020' # month/date/year
pd.to_datetime(date) # Normal form

Timestamp('2020-10-03 00:00:00')

# Dayfirst formats
pd.to_datetime(date, dayfirst=True)

Timestamp('2020-03-10 00:00:00')

# Remove custom delimiter in date

date = '10@03@2020'
pd.to_datetime(date, format='%d@%m@%Y')

Timestamp('2020-03-10 00:00:00')

date = '10@03@2020'
pd.to_datetime(date, format='%d@%m@%Y')

Timestamp('2020-03-10 00:00:00')

# Remove custom delimiter in time

import pandas as pd

date = '10@03@2020 04&30'
pd.to_datetime(date, format='%d@%m@%Y %H&%M')

Timestamp('2020-03-10 04:30:00')

# Handling errors in datetime

dates = ['Mar 10, 2020', None]
pd.to_datetime(dates, format='%b %d,%Y', errors='coerce')

##b Abbreviated month name (3 letters)

DatetimeIndex(['2020-03-10', 'NaT'], dtype='datetime64[ns]',
freq=None)

# by default this will raise an error
# to avoid we can pass errors=ignore
# but none of the other conversions will be done

dates = ['Mar 10, 2020', 'xyz']
pd.to_datetime(dates, format='%b %d,%Y', errors='ignore')

```

```

C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3481787600.py:6:
FutureWarning: errors='ignore' is deprecated and will raise in a
future version. Use to_datetime without passing `errors` and catch
exceptions explicitly instead
    pd.to_datetime(dates, format='%b %d,%Y',errors='ignore')

Index(['Mar 10, 2020', 'xyz'], dtype='object')

# Coerce the error
dates = ['Mar 10, 2020', 'xyz']
pd.to_datetime(dates, errors='coerce') # We can see that the
conversion to the valid dates have been done but the
                                         # garbage one has been
converted to NaT

DatetimeIndex(['2020-03-10', 'NaT'], dtype='datetime64[ns]',
freq=None)

# Epoch to datetime [ number to date]

t=1
pd.to_datetime(t)

Timestamp('1970-01-01 00:00:00.000000001')

# by default in ns(nano second)
t = 1583861574
pd.to_datetime(t)

Timestamp('1970-01-01 00:00:01.583861574')

# unit = 's' for sec
# unit = 'ms' for milli sec

dt = pd.to_datetime([t], unit='s')
dt

# nix timestamp, representing the number of seconds since January 1,
1970, 00:00:00 UTC (the Unix epoch)

DatetimeIndex(['2020-03-10 17:32:54'], dtype='datetime64[ns]',
freq=None)

# datetime to Epoch [date to number]

dt.view('int64') # This gives back, the epoch time

array([1583861574000000000], dtype=int64)

# Annual Period

y_2020 = pd.Period('2020') # Period covers a range
y_2020

```

```

# output: 'Y-DEC': The frequency is yearly (Y)
Period('2020', 'Y-DEC')
y_2020.start_time # Operations in annual period
Timestamp('2020-01-01 00:00:00')
y_2020.end_time # Operations in annual period
Timestamp('2020-12-31 23:59:59.999999999')

# Addition/Subtraction to annual period:
y_2019 = pd.Period('2019')
y_2019 + 1
Period('2020', 'Y-DEC')

# Monthly period
m_2020 = pd.Period('2020-01')
m_2020
Period('2020-01', 'M')
m_2020 = pd.Period('2020-2', freq='3M') # February, March and April
m_2020
Period('2020-02', '3M')
m_2020.start_time
Timestamp('2020-02-01 00:00:00')
m_2020.end_time
Timestamp('2020-04-30 23:59:59.999999999')

# Addition/Subtraction on monthly period:
m_2020 = pd.Period('2020-01')
m_2020 - 1
Period('2019-12', 'M')

# Daily period
d_2020 = pd.Period('2020-03-01')
d_2020
Period('2020-03-01', 'D')

```

```

d_2020 = pd.Period('2020-03-01', freq='3D')
d_2020
Period('2020-03-01', '3D')
d_2020.start_time
Timestamp('2020-03-01 00:00:00')
d_2020.end_time
Timestamp('2020-03-03 23:59:59.999999999')
d_2020
Period('2020-03-01', '3D')
# Adding/Subtracting days:
d_2020+2 # date: 1 + 3*2 ( freq='3D')
Period('2020-03-07', '3D')
# Hourly period
h_2020 = pd.Period('2020-03-01 04:00')
h_2020
Period('2020-03-01 04:00', 'min')
h_2020 = pd.Period('2020-03-01 04:00', freq='H')
h_2020
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\3817598763.py:1:
FutureWarning: 'H' is deprecated and will be removed in a future
version, please use 'h' instead.
  h_2020 = pd.Period('2020-03-01 04:00', freq='H')
Period('2020-03-01 04:00', 'h')
# Adding/Subtracting hours
h_2020 - 1
Period('2020-03-01 03:00', 'h')
# Quarterly period
# A year has four quarters (Jan to Mar, Apr to Jun, Jul to Sept, Oct
to Dec by default)
# Only 1st Quarter
q1_2020 = pd.Period('2020', freq='Q')
q1_2020

```



```

Period('2020Q1', 'Q-DEC')
q1_2020.start_time
Timestamp('2020-01-01 00:00:00')
q1_2020.end_time
Timestamp('2020-03-31 23:59:59.999999999')

# Multiple Quarters
q_2020 = pd.Period('2020', freq='3Q')
q_2020
Period('2020Q1', '3Q-DEC')
q_2020.start_time
Timestamp('2020-01-01 00:00:00')
q_2020.end_time #'3Q' 1st 9 months of a year [1 quarter = 3 month]
Timestamp('2020-09-30 23:59:59.999999999')

# Defined Quarter
q3_2020 = pd.Period('2020Q3', freq='Q') # 3rd quarter=July to September
q3_2020
Period('2020Q3', 'Q-DEC')
q3_2020.start_time
Timestamp('2020-07-01 00:00:00')
q3_2020.end_time
Timestamp('2020-09-30 23:59:59.999999999')

# Custom quarters
#To get a period where the quarter starts with April we have to pass the 'freq' parameter as 'Q-MAR'.

q1_2020 = pd.Period('2020Q1', freq='Q-MAR') # Fiscal year 2020 (ending March 2020) [*** began April 2019]
q1_2020
Period('2020Q1', 'Q-MAR')
q1_2020.start_time

```

```

Timestamp('2019-04-01 00:00:00')
q1_2020.end_time
Timestamp('2019-06-30 23:59:59.999999999')
q1_2020
Period('2020Q1', 'Q-MAR')

# Converting one frequency to another

# We'll take the quarterly frequency we generated in above topic and
convert that to monthly.

q1_2020.asfreq('M') # output shows the end of the period.
Period('2019-06', 'M')
q1_2020.asfreq('M', how='start')
Period('2019-04', 'M')
q1_2020.asfreq('M', how="end")
# OR, q1_2020.asfreq('M', 'e')
Period('2019-06', 'M')

# Arithmetic between two periods
q1_2020 = pd.Period('2020Q1', freq='Q-MAR')
q_new = pd.Period('2018Q3', freq='Q-MAR')
q1_2020 - q_new # This shows the difference between is of
6 quarters.

<6 * QuarterEnds: startingMonth=3>

q1_2020.start_time # start April 2019
Timestamp('2019-04-01 00:00:00')
q1_2020.end_time # end at June 2019
Timestamp('2019-06-30 23:59:59.999999999')
q_new.start_time # start at October 2017
Timestamp('2017-10-01 00:00:00')
q_new.end_time # end at December 2017
Timestamp('2017-12-31 23:59:59.999999999')

```

```

# October 2017 to June 2019 have 6 quarter
# Oct-Dec 2017 (2018Q3) → Jan-Mar 2018 (2018Q4) → Apr-Jun 2018 (2019Q1) →
# Jul-Sep 2018 (2019Q2) → Oct-Dec 2018 (2019Q3) → Jan-Mar 2019 (2019Q4) →
# Apr-Jun 2019 (2020Q1)

# Period Index

idx = pd.period_range('2015', '2020', freq='Q')
idx

PeriodIndex(['2015Q1', '2015Q2', '2015Q3', '2015Q4', '2016Q1',
            '2016Q2',
            '2016Q3', '2016Q4', '2017Q1', '2017Q2', '2017Q3',
            '2017Q4',
            '2018Q1', '2018Q2', '2018Q3', '2018Q4', '2019Q1',
            '2019Q2',
            '2019Q3', '2019Q4', '2020Q1'],
            dtype='period[Q-DEC]')

# Getting given number of periods

idx = pd.period_range('2015', periods=5, freq='Q')
idx

# This provides us 5 periods, starting from 2015.

PeriodIndex(['2015Q1', '2015Q2', '2015Q3', '2015Q4', '2016Q1'],
            dtype='period[Q-DEC]')

# Period index to DataFrame

ps = pd.Series(np.random.randn(len(idx)), idx)
ps

# We have created a DataFrame with random numbers and added the
period index to that.

2015Q1    -0.283528
2015Q2    -1.045122
2015Q3     0.419786
2015Q4     0.403627
2016Q1     0.279300
Freq: Q-DEC, dtype: float64

# Extract annual data

ps['2015']

2015Q1    -0.283528
2015Q2    -1.045122

```

```

2015Q3      0.419786
2015Q4      0.403627
Freq: Q-DEC, dtype: float64

ps['2016']

2016Q1      0.2793
Freq: Q-DEC, dtype: float64

# Extract a range of periods data

ps['2015Q3' : '2016'] # range from 2015 Q3 to 2016 Q1

2015Q3      0.419786
2015Q4      0.403627
2016Q1      0.279300
Freq: Q-DEC, dtype: float64

# Convert DatetimeIndex to PeriodIndex

import pandas as pd

# Create a DateTimeIndex
dti = pd.date_range('2023-01-01', periods=5, freq='D')
print("DateTimeIndex:", dti)

# Convert to PeriodIndex with daily frequency
period_index = dti.to_period('D')
print("PeriodIndex:", period_index)

DateTimeIndex: DatetimeIndex(['2023-01-01', '2023-01-02', '2023-01-03',
                               '2023-01-04',
                               '2023-01-05'],
                               dtype='datetime64[ns]', freq='D')
PeriodIndex: PeriodIndex(['2023-01-01', '2023-01-02', '2023-01-03',
                           '2023-01-04',
                           '2023-01-05'],
                           dtype='period[D]')

# Working with time zones
# Two types of times: 1,time zone aware...2,time zone unaware

df = pd.read_csv('timezone.csv',
                  index_col='date',
                  parse_dates=(['date']))

df

```

date	price
2020-10-03 01:00:00	57
2020-10-03 02:00:00	58
2020-10-03 03:00:00	59

```

2020-10-03 04:00:00      60
2020-10-03 05:00:00      61
2020-10-03 06:00:00      62
2020-10-03 07:00:00      63
2020-10-03 08:00:00      64
2020-10-03 09:00:00      65
2020-10-03 10:00:00      66

# Make naïve time to time zone aware

df1 = df.tz_localize(tz='US/Eastern')
df1.index

# EST (Eastern Standard Time) = UTC-5 (used in winter, no DST).
# EDT (Eastern Daylight Time) = UTC-4 (used in summer, DST active).

DatetimeIndex(['2020-10-03 01:00:00-04:00', '2020-10-03 02:00:00-
04:00',
              '2020-10-03 03:00:00-04:00', '2020-10-03 04:00:00-
04:00',
              '2020-10-03 05:00:00-04:00', '2020-10-03 06:00:00-
04:00',
              '2020-10-03 07:00:00-04:00', '2020-10-03 08:00:00-
04:00',
              '2020-10-03 09:00:00-04:00', '2020-10-03 10:00:00-
04:00'],
              dtype='datetime64[ns, US/Eastern]', name='date',
              freq=None)

# Available timezones

import pytz
pytz.all_timezones[::30]

['Africa/Abidjan',
 'Africa/Kinshasa',
 'America/Argentina/Catamarca',
 'America/Catamarca',
 'America/Guadeloupe',
 'America/Maceio',
 'America/Paramaribo',
 'America/Tegucigalpa',
 'Asia/Aqtobe',
 'Asia/Hovd',
 'Asia/Pontianak',
 'Asia/Yakutsk',
 'Australia/North',
 'Eire',
 'Etc/GMT0',
 'Europe/Kyiv',

```

```

'Europe/Uzhgorod',
'Indian/Reunion',
'Pacific/Gambier',
'Pacific/Wallis']

# For INDIA:
pytz.country_timezones('IN')

['Asia/Kolkata']

# For Bangladesh:
pytz.country_timezones('BD')

['Asia/Dhaka']

df1.index

DatetimeIndex(['2020-10-03 01:00:00-04:00', '2020-10-03 02:00:00-
04:00',
              '2020-10-03 03:00:00-04:00', '2020-10-03 04:00:00-
04:00',
              '2020-10-03 05:00:00-04:00', '2020-10-03 06:00:00-
04:00',
              '2020-10-03 07:00:00-04:00', '2020-10-03 08:00:00-
04:00',
              '2020-10-03 09:00:00-04:00', '2020-10-03 10:00:00-
04:00'],
              dtype='datetime64[ns, US/Eastern]', name='date',
              freq=None)

# Convert on time zone to other

df = df1.tz_convert(tz='Europe/Guernsey') # df1 was in
'US/Eastern' time zone and is getting converted to
# 'Europe/Guernsey' time
zone
df.index

DatetimeIndex(['2020-10-03 06:00:00+01:00', '2020-10-03
07:00:00+01:00',
              '2020-10-03 08:00:00+01:00', '2020-10-03
09:00:00+01:00',
              '2020-10-03 10:00:00+01:00', '2020-10-03
11:00:00+01:00',
              '2020-10-03 12:00:00+01:00', '2020-10-03
13:00:00+01:00',
              '2020-10-03 14:00:00+01:00', '2020-10-03
15:00:00+01:00'],
              dtype='datetime64[ns, Europe/Guernsey]', name='date',
              freq=None)

```

```
# Time zone in a date range
```

```
rng = pd.date_range(start='10/3/2020', periods=10, freq='H',  
tz='dateutil/Asia/Kolkata')  
rng    # [ 1 to 8 hour ] + 5: 30,  
       # IST is UTC+5:30 (always, no DST).
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_9660\1899864123.py:3:  
FutureWarning: 'H' is deprecated and will be removed in a future  
version, please use 'h' instead.
```

```
    rng = pd.date_range(start='10/3/2020', periods=10, freq='H',  
tz='dateutil/Asia/Kolkata')
```

```
DatetimeIndex(['2020-10-03 00:00:00+05:30', '2020-10-03  
01:00:00+05:30',  
              '2020-10-03 02:00:00+05:30', '2020-10-03  
03:00:00+05:30',  
              '2020-10-03 04:00:00+05:30', '2020-10-03  
05:00:00+05:30',  
              '2020-10-03 06:00:00+05:30', '2020-10-03  
07:00:00+05:30',  
              '2020-10-03 08:00:00+05:30', '2020-10-03  
09:00:00+05:30'],  
              dtype='datetime64[ns, tzfile('Asia/Calcutta')]',  
              freq='h')
```

```
# Data shifts in DataFrame
```

```
df = pd.read_csv('shifting.csv', index_col='date', parse_dates =  
(['date']))  
df
```

	price
date	
2020-03-10	43
2020-03-11	54
2020-03-12	73
2020-03-13	85
2020-03-14	53
2020-03-15	74
2020-03-16	76
2020-03-17	44
2020-03-18	62
2020-03-19	84

```
# Shifting the price down
```

```
df.shift()
```

```
# the price for date 10 March, 2020 has been shifted to 11 march, 2020
```

date	price
2020-03-10	NaN
2020-03-11	43.0
2020-03-12	54.0
2020-03-13	73.0
2020-03-14	85.0
2020-03-15	53.0
2020-03-16	74.0
2020-03-17	76.0
2020-03-18	44.0
2020-03-19	62.0

Shifting by multiple rows

df.shift(3)

date	price
2020-03-10	NaN
2020-03-11	NaN
2020-03-12	NaN
2020-03-13	43.0
2020-03-14	54.0
2020-03-15	73.0
2020-03-16	85.0
2020-03-17	53.0
2020-03-18	74.0
2020-03-19	76.0

Reverse shifting

df.shift(-3)

the price which was for 19 is now for 17 and so on

date	price
2020-03-10	85.0
2020-03-11	53.0
2020-03-12	74.0
2020-03-13	76.0
2020-03-14	44.0
2020-03-15	62.0
2020-03-16	84.0
2020-03-17	NaN
2020-03-18	NaN
2020-03-19	NaN

```
df = pd.read_csv('shifting.csv', index_col='date', parse_dates =
(['date']))
df
```


date	price
2020-03-10	43
2020-03-11	54
2020-03-12	73
2020-03-13	85
2020-03-14	53
2020-03-15	74
2020-03-16	76
2020-03-17	44
2020-03-18	62
2020-03-19	84

New column with last day price:

```
df['previous_day_prices'] = df.shift()
df
```

date	price	previous_day_prices
2020-03-10	43	NaN
2020-03-11	54	43.0
2020-03-12	73	54.0
2020-03-13	85	73.0
2020-03-14	53	85.0
2020-03-15	74	53.0
2020-03-16	76	74.0
2020-03-17	44	76.0
2020-03-18	62	44.0
2020-03-19	84	62.0

New column with change in the price from last day:

```
df['changed_price'] = df['price'] - df['previous_day_prices']
df
```

date	price	previous_day_prices	changed_price
2020-03-10	43	NaN	NaN
2020-03-11	54	43.0	11.0
2020-03-12	73	54.0	19.0
2020-03-13	85	73.0	12.0
2020-03-14	53	85.0	-32.0
2020-03-15	74	53.0	21.0
2020-03-16	76	74.0	2.0
2020-03-17	44	76.0	-32.0
2020-03-18	62	44.0	18.0
2020-03-19	84	62.0	22.0

```
df['return_3days'] = (df['price'] -
df['price'].shift(3)*100/df['price'].shift(3))
```

```
df
```

```
# (85 - 43) * 100 / 43 ≈ 97.67%
```

```
# (53 - 54) * 100 / 54 ≈ -1.85%
```

	price	previous_day_prices	changed_price	return_3days
date				
2020-03-10	43	NaN	NaN	NaN
2020-03-11	54	43.0	11.0	NaN
2020-03-12	73	54.0	19.0	NaN
2020-03-13	85	73.0	12.0	-15.0
2020-03-14	53	85.0	-32.0	-47.0
2020-03-15	74	53.0	21.0	-26.0
2020-03-16	76	74.0	2.0	-24.0
2020-03-17	44	76.0	-32.0	-56.0
2020-03-18	62	44.0	18.0	-38.0
2020-03-19	84	62.0	22.0	-16.0

```
df
```

	price	previous_day_prices	changed_price	return_3days
date				
2020-03-10	43	NaN	NaN	NaN
2020-03-11	54	43.0	11.0	NaN
2020-03-12	73	54.0	19.0	NaN
2020-03-13	85	73.0	12.0	-15.0
2020-03-14	53	85.0	-32.0	-47.0
2020-03-15	74	53.0	21.0	-26.0
2020-03-16	76	74.0	2.0	-24.0
2020-03-17	44	76.0	-32.0	-56.0
2020-03-18	62	44.0	18.0	-38.0
2020-03-19	84	62.0	22.0	-16.0

```
# DatetimeIndex shift
```

```
df.shift(freq='D')
```

```
# the index has been shifted by one row down and day 10 has gone and  
taken over by 11
```

	price	previous_day_prices	changed_price	return_3days
date				
2020-03-11	43	NaN	NaN	NaN
2020-03-12	54	43.0	11.0	NaN
2020-03-13	73	54.0	19.0	NaN
2020-03-14	85	73.0	12.0	-15.0
2020-03-15	53	85.0	-32.0	-47.0
2020-03-16	74	53.0	21.0	-26.0
2020-03-17	76	74.0	2.0	-24.0
2020-03-18	44	76.0	-32.0	-56.0

2020-03-19	62	44.0	18.0	-38.0
2020-03-20	84	62.0	22.0	-16.0

Reverse DatetimeIndex shift

`df.shift(freq='-1D')` *# not working: '-D'*

vanished 19 from the index.

	price	previous_day_prices	changed_price	return_3days
date				
2020-03-09	43	NaN	NaN	NaN
2020-03-10	54	43.0	11.0	NaN
2020-03-11	73	54.0	19.0	NaN
2020-03-12	85	73.0	12.0	-15.0
2020-03-13	53	85.0	-32.0	-47.0
2020-03-14	74	53.0	21.0	-26.0
2020-03-15	76	74.0	2.0	-24.0
2020-03-16	44	76.0	-32.0	-56.0
2020-03-17	62	44.0	18.0	-38.0
2020-03-18	84	62.0	22.0	-16.0

Working with MySQL

```
import pandas as pd
import sqlalchemy
```

Create connection

```
engine =
sqlalchemy.create_engine('mysql+pymysql://root:NAZMULhasan11#@localhost:3306/students_db')
engine
```

```
Engine(mysql+pymysql://root:***@localhost:3306/students_db)
```

Read table data

```
df_students = pd.read_sql_table("students", engine)
df_students
```

	id	student_name	age
0	1	arun	29
1	2	varun	20
2	3	neha	22
3	4	nisha	35
4	5	robert	38
5	6	michael	27
6	7	gillian	52
7	8	graeme	49
8	9	rohan	22

9	10	robert	38
10	11	michael	27
11	12	gillian	52
12	13	graeme	49
13	14	rohan	22
14	15	robert	38
15	16	michael	27
16	17	gillian	52
17	18	graeme	49
18	19	rohan	22

Fetching specific columns from table

```
df_students = pd.read_sql_table('students',
                                engine,
                                columns = ['student_name'])
```

df_students

	student_name
0	arun
1	varun
2	neha
3	nisha
4	robert
5	michael
6	gillian
7	graeme
8	rohan
9	robert
10	michael
11	gillian
12	graeme
13	rohan
14	robert
15	michael
16	gillian
17	graeme
18	rohan

Execute a query

```
query = '''
select student_name, age from students where student_name='arun'
'''
```

```
df_query = pd.read_sql(query, engine)
df_query
```

	student_name	age
0	arun	29

Insert data to table

	id	student_name	age
0	1	arun	29
1	2	varun	20
2	3	neha	22
3	4	nisha	35
4	5	robert	38

```
# Common function to read table and query
# 'read_sql_table' or, 'read_sql_query' =read_sql()
```

```
# read_sql() to fetch table data
```

```
df_students_table = pd.read_sql('students', engine)
df_students_table.head()
```

	id	student_name	age
0	1	arun	29
1	2	varun	20
2	3	neha	22
3	4	nisha	35
4	5	robert	38

```
# read_sql() to fetch query data
```

```
query = '''
select student_name, age
from students
where student_name ='arun'
'''
```

```
df_students_query = pd.read_sql(query, engine)
df_students_query
```

	student_name	age
0	arun	29

```
# Working with MongoDB
# !pip install pymongo
```

```
# Create connection
from pymongo import MongoClient
```

```
client = MongoClient()
db = client.students_db
collection = db.students
```

```
from pymongo import MongoClient
```

```
# Connect to MongoDB
client = MongoClient()
db = client.students_db # Database name
collection = db.students # Collection name
```

```

# Insert a sample document (creates the collection automatically)
insert_result = collection.insert_one({"student_name": "Alice", "age":
24, "grade": "A"})

# Simply return/print the collection object
print(collection)

# Or to verify the collection exists (without updating anything)
print("Collections in 'students_db':", db.list_collection_names())

Collection(Database(MongoClient(host=['localhost:27017'],
document_class=dict, tz_aware=False, connect=True), 'students_db'),
'students')
Collections in 'students_db': ['students']

import pandas as pd
from pymongo import MongoClient

# Connect and query with limit
collection = MongoClient().students_db.students
data = pd.DataFrame(list(collection.find().limit(1))) # ← Only get 1
document

print(data.head()) # Shows first 5 rows

```

	_id	student_name	age	grade
0	6824a5e70428e16ab6d6c92f	Alice	24	A

```

# Fetching specific columns

data = pd.DataFrame(list(
    collection.find({}, {'age': 1, '_id': 0}).limit(2) # ← Limits to
2 docs at DB level
))
print(data)

```

	age
0	24
1	24

```

# Insert records

students = [
    {"student_name": "robert", "age": 38},
    {"student_name": "michael", "age": 27},
    {"student_name": "gillian", "age": 52},
    {"student_name": "graeme", "age": 49},
    {"student_name": "rohan", "age": 22},
]
collection.insert_many(students)

```

```
InsertManyResult([ObjectId('6824a64672eda7ac2bb4752d'),
ObjectId('6824a64672eda7ac2bb4752e'),
ObjectId('6824a64672eda7ac2bb4752f'),
ObjectId('6824a64672eda7ac2bb47530'),
ObjectId('6824a64672eda7ac2bb47531')], acknowledged=True)
```

```
collection.delete_many({
    "$or": [
        {"student_name": {"$ne": "Alice"}},
        {"student_name": {"$exists": False}}
    ]
})
```

```
DeleteResult({'n': 5, 'ok': 1.0}, acknowledged=True)
```

```
# Delete records
```

```
data = pd.DataFrame(list(collection.find().limit(1)))
display(data)
```

	_id	student_name	age	grade
0	6824a5e70428e16ab6d6c92f	Alice	24	A