

```
In [3]: import pandas as pd
import numpy as np
```

```
In [4]: dict1={'Name': ['Ramesh', 'Suresh', 'Sathish', np.nan],
              'Age': [30, 33, np.nan, 36],
              'City': [np.nan, 'Hyd', 'Hyd', 'Blr']}
```

```
In [5]: data=pd.DataFrame(dict1)
data
```

```
Out[5]:
```

	Name	Age	City
0	Ramesh	30.0	NaN
1	Suresh	33.0	Hyd
2	Sathish	NaN	Hyd
3	NaN	36.0	Blr

```
In [11]: data.to_csv('data1.csv', index=False)
```

isnull

```
In [13]: data.isnull()
```

```
Out[13]:
```

	Name	Age	City
0	False	False	True
1	False	False	False
2	False	True	False
3	True	False	False

```
In [15]: data.isnull().sum()
```

```
Out[15]: Name      1
Age          1
City         1
dtype: int64
```

dropna

- drop the null values data based on rows and columns

```
In [17]: # I want to extract the data with out Null values
data.dropna()
```

```
Out[17]:
```

	Name	Age	City
1	Suresh	33.0	Hyd

```
In [10]: data
```

```
Out[10]:
```

	Name	Age	City
0	Ramesh	30.0	NaN
1	Suresh	33.0	Hyd
2	Sathish	NaN	Hyd
3	NaN	36.0	Blr

```
In [11]: data.dropna(axis=1)
```

```
Out[11]:
```

0
1
2
3

```
In [19]: dict1={'Name': ['Ramesh', 'Suresh', 'Sathish', np.nan],  
              'Age': [30, 33, np.nan, 36],  
              'City': [np.nan, 'Hyd', 'Hyd', 'Blr'],  
              'Company': ['Google', 'Microsoft', 'Apple', 'Meta']}  
data1=pd.DataFrame(dict1)  
data1
```

```
Out[19]:
```

	Name	Age	City	Company
0	Ramesh	30.0	NaN	Google
1	Suresh	33.0	Hyd	Microsoft
2	Sathish	NaN	Hyd	Apple
3	NaN	36.0	Blr	Meta

```
In [21]: data1.dropna(axis=1)
```

```
Out[21]:
```

	Company
0	Google
1	Microsoft
2	Apple
3	Meta

Fillna

```
In [25]: data1.fillna(30)
```

```
Out[25]:
```

	Name	Age	City	Company
0	Ramesh	30.0	30	Google
1	Suresh	33.0	Hyd	Microsoft
2	Sathish	30.0	Hyd	Apple
3	30	36.0	Blr	Meta

```
In [27]: # first select the column
# then apply Fill na
data1['Name'].fillna('Manish')
```

```
Out[27]: 0    Ramesh
1    Suresh
2    Sathish
3    Manish
Name: Name, dtype: object
```

Draw back: Filling with Random value is not a Good approach

- To avoid the we have some methods
 - backfill
 - bfill
 - ffill
 - pad

pad

```
In [39]: data1.fillna(method='pad')
```

```
Out[39]:
```

	Name	Age	City	Company
0	Ramesh	30.0	NaN	Google
1	Suresh	33.0	Hyd	Microsoft
2	Sathish	33.0	Hyd	Apple
3	Sathish	36.0	Blr	Meta

```
In [23]: import warnings
warnings.filterwarnings('ignore')
data1.fillna(method='pad')
```

```
Out[23]:
```

	Name	Age	City	Company
0	Ramesh	30.0	NaN	Google
1	Suresh	33.0	Hyd	Microsoft
2	Sathish	33.0	Hyd	Apple
3	Sathish	36.0	Blr	Meta

```
In [41]: import warnings
warnings.filterwarnings('ignore')
data1.fillna(method='pad',axis=1)
```

```
Out[41]:
```

	Name	Age	City	Company
0	Ramesh	30.0	30.0	Google
1	Suresh	33.0	Hyd	Microsoft
2	Sathish	Sathish	Hyd	Apple
3	NaN	36.0	Blr	Meta

```
In [34]: data1
```

```
Out[34]:
```

	Name	Age	City	Company
0	Ramesh	30.0	NaN	Google
1	Suresh	33.0	Hyd	Microsoft
2	Sathish	NaN	Hyd	Apple
3	NaN	36.0	Blr	Meta

```
In [ ]: idea-1: filling with random value = df.fillna(<random value>)
idea-2: filling with rv based on column=df[<col>].fillna(<rv>)
idea-3: filling with some pattern using a method: df.fillna(<method>)
idea-4: filling with m-m-m based on column=df[<col>].fillna(<m>or<m>or<m>)
idea-5: filling with avg value only selected neighbours
```

```
In [ ]:
```

Mean-Median-Mode

```
In [31]: # step-1: select the column
# step-2: get the mean
# step-3: apply fill na

age_mean=data1['Age'].mean()
print('age_mean:',age_mean)
data1['Age'].fillna(age_mean)
```

```
age_mean: 33.0
```

```
Out[31]: 0    30.0
         1    33.0
         2    33.0
         3    36.0
         Name: Age, dtype: float64
```

```
In [47]: age_mean=data1['Age'].mean()
         age_mean
```

```
Out[47]: 33.0
```

```
In [51]: city_mode=data1['City'].mode()
         city_mode.values[0]
```

```
Out[51]: 'Hyd'
```

```
In [33]: city_mode=data1['City'].mode()
         data1['City'].fillna(city_mode)
```

```
Out[33]: 0    Hyd
         1    Hyd
         2    Hyd
         3    Blr
         Name: City, dtype: object
```

```
In [35]: age_mean # fixed value
```

```
Out[35]: 33.0
```

```
In [37]: city_mode # series
```

```
Out[37]: 0    Hyd
         Name: City, dtype: object
```

```
In [41]: city_mode.values[0]
```

```
Out[41]: 'Hyd'
```

- in future sometime if we direct impute mode value we will get error
- mode is coming as series
- mean is coming as fixed value
- so first convert series to fixed value then apply it

```
In [53]: city_mode.values[0]
```

```
Out[53]: 'Hyd'
```

```
In [33]: city_mode=data1['City'].mode()
         data1['City'].fillna(city_mode.values[0])
```

```
Out[33]: 0    Hyd
          1    Hyd
          2    Hyd
          3    Blr
          Name: City, dtype: object
```

impute class

- under sklearn we have impute class is there

```
In [53]: from sklearn import impute
```

```
In [54]: dir(impute)
```

```
Out[54]: ['KNNImputer',
          'MissingIndicator',
          'SimpleImputer',
          '__all__',
          '__builtins__',
          '__cached__',
          '__doc__',
          '__file__',
          '__getattr__',
          '__loader__',
          '__name__',
          '__package__',
          '__path__',
          '__spec__',
          '_base',
          '_knn',
          'typing']
```

KNN Imputer

- K- Nearest Neighbours
- Where k = hyper parameter which means the user can change
- Idea: Instead of taking all the samples average
 - First fix the neighbours number i.e. k
 - Then find the neighbours using distance metric
 - Then take the average of those samepls to fill the missing value

```
In [43]: import numpy as np
          from sklearn.impute import KNNImputer
          X = [[1, 2, np.nan], [3, np.nan, 3], [np.nan, 60, 5], [8, 8, 7]]
          pd.DataFrame(X)
```

```
Out[43]:
```

	0	1	2
0	1.0	2.0	NaN
1	3.0	NaN	3.0
2	NaN	60.0	5.0
3	8.0	8.0	7.0

```
In [57]: import numpy as np
from sklearn.impute import KNNImputer
X = [[1, 2, np.nan], [3, np.nan, 3], [np.nan, 60, 5], [8, 8, 7]]
imputer = KNNImputer(n_neighbors=2, weights='uniform')
imputer.fit_transform(X)
```

```
Out[57]: array([[ 1. ,  2. ,  5. ],
 [ 3. , 31. ,  3. ],
 [ 5.5, 60. ,  5. ],
 [ 8. ,  8. ,  7. ]])
```

```
In [45]: import numpy as np
from sklearn.impute import KNNImputer
X = [[1, 2, np.nan], [3, np.nan, 3], [np.nan, 60, 5], [8, 8, 7]]
imputer = KNNImputer(n_neighbors=2, weights='distance')
imputer.fit_transform(X)
```

```
Out[45]: array([[ 1.          ,  2.          ,  3.93905505],
 [ 3.          , 31.          ,  3.          ],
 [ 3.25775362, 60.          ,  5.          ],
 [ 8.          ,  8.          ,  7.          ]])
```

```
In [ ]:
```

```
In [45]: import numpy as np
from sklearn.impute import KNNImputer
X = [[1, 2, np.nan], [3, np.nan, 3], [np.nan, 60, 5], [8, 8, 7]]
imputer = KNNImputer(n_neighbors=2)
imputer.fit_transform(X)
```

```
Out[45]: array([[ 1. ,  2. ,  5. ],
 [ 3. , 31. ,  3. ],
 [ 5.5, 60. ,  5. ],
 [ 8. ,  8. ,  7. ]])
```

```
In [ ]: idea-1: filling with random value = df.fillna(<random value>)
idea-2: filling with rv based on column=df[<col>].fillna(<rv>)
idea-3: filling with some pattern using a method: df.fillna(<method>)
idea-4: filling with m-m-m based on column=df[<col>].fillna(<m>or<m>or<m>)
idea-5: filling with avg value only selected neighbours
idea-6: filling with a value based on correlation with another columns
```