

### 12. PANDAS – Handling missing or NaN values

#### Contents

|   |    |
|---|----|
| 1. NaN Value .....  | 2  |
| 2. Creating a DataFrame by loading csv file .....               | 3  |
| 3. isna() and isnull() method – Checking NaN values.....        | 4  |
| 4. notnull() method – Checking NaN values.....                  | 6  |
| 5. Counting NaN values in column wise .....                     | 7  |
| 6. dropna() method – Handling missing values.....               | 9  |
| 7. dropna(inplace = True) method – Handling missing values..... | 12 |
| 8. fillna() method – Handling missing values .....              | 13 |

### 12. PANDAS – Handling missing or NaN values

#### 1. NaN Value

- ✓ The full form of NaN is **Not a Number**
- ✓ The purpose of NaN is, to represent the missing values in data.
- ✓ The data type of NaN is **float**.
- ✓ While loading csv file, if file having missing values then it will be considered as NaN values.
- ✓ During data analysis we need to handle these NaN values.
  - For Example, suppose different users being surveyed may choose not to share their income, some user may choose not to share the address in this way many datasets went missing.

#### **None** and **NaN**

- ✓ **None** : None is a Python object which is holding nothing
- ✓ **NaN** : NaN is a pandas related object which represents missing data

### 2. Creating a DataFrame by loading csv file

- ✓ We can create DataFrame by loading csv file
- ✓ The given fruits.csv file having missing values.
- ✓ Kindly observe the missing/NaN values in DataFrame.

**Program** Loading fruits csv file  
**Name** demo1.py  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv("fruits1.csv")

print(df1)
```

#### Output

```
   Order Id  Apples  Orange  Banana
0      16723     9.0     NaN     6.0
1      16724    12.0    12.0     6.0
2      16725     9.0     NaN     NaN
3      16726    12.0     NaN    24.0
4      16727    30.0     8.0    10.0
..      ...     ...     ...     ...
145     16868    24.0    12.0     6.0
146     16869     8.0     8.0    12.0
147     16870     5.0    30.0    30.0
148     16871    24.0     NaN    10.0
149     16872     5.0    30.0    10.0

[150 rows x 4 columns]
```

### 3. isna() and isnull() method – Checking NaN values

- ✓ isna() and isnull() are a predefined methods in DataFrame
- ✓ We can access these methods by using DataFrame object.
- ✓ By using these methods we can check missing values exist in DataFrame or not.
- ✓ If missing values are available then it return as **True**, otherwise **False**

**Program** isna() method  
**Name** demo2.py  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv("fruits1.csv")
df2 = df1.isna()

print(df1.head())
print()
print(df2.head())
```

#### Output

|   | Order Id | Apples | Orange | Banana |
|---|----------|--------|--------|--------|
| 0 | 16723    | 9.0    | NaN    | 6.0    |
| 1 | 16724    | 12.0   | 12.0   | 6.0    |
| 2 | 16725    | 9.0    | NaN    | NaN    |
| 3 | 16726    | 12.0   | NaN    | 24.0   |
| 4 | 16727    | 30.0   | 8.0    | 10.0   |

  

|   | Order Id | Apples | Orange | Banana |
|---|----------|--------|--------|--------|
| 0 | False    | False  | True   | False  |
| 1 | False    | False  | False  | False  |
| 2 | False    | False  | True   | True   |
| 3 | False    | False  | True   | False  |
| 4 | False    | False  | False  | False  |

**Program** isnull() method

**Name** demo3.py

**Input file** fruits1.csv

```
import pandas as pd
```

```
df1 = pd.read_csv("fruits1.csv")
```

```
df2 = df1.isnull()
```

```
print(df1.head())
```

```
print()
```

```
print(df2.head())
```

**Output**

```
Order Id  Apples  Orange  Banana
0    16723    9.0    NaN    6.0
1    16724   12.0   12.0    6.0
2    16725    9.0    NaN    NaN
3    16726   12.0    NaN   24.0
4    16727   30.0    8.0   10.0

Order Id  Apples  Orange  Banana
0     False  False   True   False
1     False  False  False   False
2     False  False   True   True
3     False  False   True   False
4     False  False  False   False
```

### Make a note

- ✓ isnull() and isna() both methods works in same way

### 4. notnull() method – Checking NaN values

- ✓ notnull() is a predefined method in DataFrame
- ✓ We can access this method by using DataFrame object.
- ✓ By using this method we can check missing values exist in DataFrame or not.
- ✓ If missing values are available then it return as **False**, otherwise **True**

**Program** notnull() method  
**Name** demo4.py  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv("fruits1.csv")
df2 = df1.notnull()

print(df1.head())
print()
print(df2.head())
```

#### Output

|   | Order Id | Apples | Orange | Banana |
|---|----------|--------|--------|--------|
| 0 | 16723    | 9.0    | NaN    | 6.0    |
| 1 | 16724    | 12.0   | 12.0   | 6.0    |
| 2 | 16725    | 9.0    | NaN    | NaN    |
| 3 | 16726    | 12.0   | NaN    | 24.0   |
| 4 | 16727    | 30.0   | 8.0    | 10.0   |

  

|   | Order Id | Apples | Orange | Banana |
|---|----------|--------|--------|--------|
| 0 | True     | True   | False  | True   |
| 1 | True     | True   | True   | True   |
| 2 | True     | True   | False  | False  |
| 3 | True     | True   | False  | True   |
| 4 | True     | True   | True   | True   |

### 5. Counting NaN values in column wise

- ✓ We can count number of missing values in DataFrame
- ✓ By using `isna()` and `sum()` methods we can count the number of missing values in each column.

**Program** Counting the missing values in each column  
**Name** demo5.py  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv('fruits1.csv')
s = df1.isna().sum()

print(s)
```

**Output**

```
Order Id    0
Apples      16
Orange      16
Banana      21
dtype: int64
```

**Program Name** Counting the missing values in each column with percentage  
**demo6.py**  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv('fruits1.csv')
s = df1.isna().sum()
per = (s * 100) / len(df1)

print(per)
```

**Output**

```
Order Id      0.000000
Apples       10.666667
Orange       10.666667
Banana       14.000000
dtype: float64
```



### 6. dropna() method – Handling missing values

- ✓ dropna() is a predefined method in DataFrame
- ✓ We can access dropna() method by using DataFrame object.
- ✓ This method drops the rows where at least one value is missing.

**Program** Dropping rows where NaN values existing  
**Name** demo7.py  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv("fruits1.csv")
df2 = df1.dropna()

print(df2)
```

#### Output

```
   Order Id  Apples  Orange  Banana
1    16724    12.0    12.0     6.0
4    16727    30.0     8.0    10.0
5    16728     6.0     6.0    12.0
6    16729    12.0    24.0     9.0
7    16730    24.0     6.0     8.0
..      ...     ...     ...     ...
143   16866    12.0    10.0    10.0
145   16868    24.0    12.0     6.0
146   16869     8.0     8.0    12.0
147   16870     5.0    30.0    30.0
149   16872     5.0    30.0    10.0

[105 rows x 4 columns]
```

**Program Name**      Dropping rows where NaN values existing and counting  
**demo8.py**  
**Input file**        fruits1.csv

```
import pandas as pd

df1 = pd.read_csv("fruits1.csv")
df2 = df1.dropna()
s = df2.isna().sum()

print(s)
```

**Output**

```
Order Id    0
Apples      0
Orange      0
Banana      0
dtype: int64
```

**Program** Converting float column type into int data type  
**Name** demo9.py  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv('fruits1.csv')
df2 = df1.dropna()
df3 = df2.astype(int)

print(df2.head())
print()
print(df3.head())
```

### Output

```
Order Id Apples Orange Banana
1    16724    12.0    12.0     6.0
4    16727    30.0     8.0    10.0
5    16728     6.0     6.0    12.0
6    16729    12.0    24.0     9.0
7    16730    24.0     6.0     8.0

Order Id Apples Orange Banana
1    16724     12     12      6
4    16727     30      8     10
5    16728      6      6     12
6    16729     12     24      9
7    16730     24      6      8
```

### 7. dropna(inplace = True) method – Handling missing values

- ✓ dropna(inplace = True) is a predefined method in DataFrame
- ✓ We can access this method by using DataFrame object.
- ✓ This method drops the rows and perform changes on existing DataFrame.

**Program Name** Dropping NaN values by using inplace parameter  
**demo10.py**  
**Input file** fruits1.csv

```
import pandas as pd

df1 = pd.read_csv("fruits1.csv")
df1.dropna(inplace = True)

print(df1)
```

#### Output

```
   Order Id  Apples  Orange  Banana
1      16724    12.0    12.0     6.0
4      16727    30.0     8.0    10.0
5      16728     6.0     6.0    12.0
6      16729    12.0    24.0     9.0
7      16730    24.0     6.0     8.0
..      ...     ...     ...     ...
143    16866    12.0    10.0    10.0
145    16868    24.0    12.0     6.0
146    16869     8.0     8.0    12.0
147    16870     5.0    30.0    30.0
149    16872     5.0    30.0    10.0

[105 rows x 4 columns]
```

### 8. fillna() method – Handling missing values

- ✓ fillna() is a predefined method in DataFrame
- ✓ We can access this method by using DataFrame object.
- ✓ By using this method we can fill missing/NaN values with specific value.
  - fillna(0)                   ->     This method fill NaN with Zero values
  - fillna(number)           ->     This method fill NaN with number

**Program**     Filling NaN values with zero  
**Name**       demo11.py  
**Input file**   fruits1.csv

```
import pandas as pd

df1 = pd.read_csv("fruits1.csv")
df2 = df1.fillna(0)

print(df1.head())
print()
print(df2.head())
```

#### Output

|   | Order Id | Apples | Orange | Banana |
|---|----------|--------|--------|--------|
| 0 | 16723    | 9.0    | NaN    | 6.0    |
| 1 | 16724    | 12.0   | 12.0   | 6.0    |
| 2 | 16725    | 9.0    | NaN    | NaN    |
| 3 | 16726    | 12.0   | NaN    | 24.0   |
| 4 | 16727    | 30.0   | 8.0    | 10.0   |

  

|   | Order Id | Apples | Orange | Banana |
|---|----------|--------|--------|--------|
| 0 | 16723    | 9.0    | 0.0    | 6.0    |
| 1 | 16724    | 12.0   | 12.0   | 6.0    |
| 2 | 16725    | 9.0    | 0.0    | 0.0    |
| 3 | 16726    | 12.0   | 0.0    | 24.0   |
| 4 | 16727    | 30.0   | 8.0    | 10.0   |

**Program Name**      Filling NaN values with specific value  
demo12.py

```
import pandas as pd
import numpy as np

data = [
    ["Rajan", 26, 40000],
    ["Daniel", 16, 20000],
    ["Veeru", 45, 90000],
    ["Venkat", np.nan, 45000],
    ["Sumanth", 20, 95000],
    ["Shafi", np.nan, 97000]
]

df1 = pd.DataFrame(data, columns = ['Name', 'Age', 'Salary'])
df2 = df1.fillna(22)

print(df1)
print()
print(df2)
```

**Output**

```
   Name  Age  Salary
0  Rajan  26.0  40000
1  Daniel  16.0  20000
2  Veeru  45.0  90000
3  Venkat   NaN  45000
4  Sumanth 20.0  95000
5  Shafi   NaN  97000

   Name  Age  Salary
0  Rajan  26.0  40000
1  Daniel  16.0  20000
2  Veeru  45.0  90000
3  Venkat  22.0  45000
4  Sumanth 20.0  95000
5  Shafi  22.0  97000
```

**Program Name**      Filling NaN value with mean value  
demo13.py

```
import pandas as pd
import numpy as np

data = [
    ["Shahid", 26, 40000],
    ["Daniel", 16, 20000],
    ["Karteek", np.nan, 90000],
    ["Venkat", np.nan, 45000],
    ["Veeru", 24, 95000],
    ["Shafi", np.nan, 97000]
]

df1 = pd.DataFrame(data, columns = ['Name', 'Age', 'Salary'])

print(df1)
m = df1['Age'].mean()
df1['Age'] = df1['Age'].fillna(m)
print()
print(df1)
```

**Output**

|   | Name    | Age  | Salary |
|---|---------|------|--------|
| 0 | Shahid  | 26.0 | 40000  |
| 1 | Daniel  | 16.0 | 20000  |
| 2 | Karteek | NaN  | 90000  |
| 3 | Venkat  | NaN  | 45000  |
| 4 | Veeru   | 24.0 | 95000  |
| 5 | Shafi   | NaN  | 97000  |

  

|   | Name    | Age  | Salary |
|---|---------|------|--------|
| 0 | Shahid  | 26.0 | 40000  |
| 1 | Daniel  | 16.0 | 20000  |
| 2 | Karteek | 22.0 | 90000  |
| 3 | Venkat  | 22.0 | 45000  |
| 4 | Veeru   | 24.0 | 95000  |
| 5 | Shafi   | 22.0 | 97000  |

**Program Name**      Creating dataframe and replacing nan values with specific value  
demo14.py

```
import pandas as pd
import numpy as np

data = [
    ['Shahid', np.nan, 40000],
    ['Daniel', 16, 20000],
    ['Veeru', 45, 90000],
    ['Sumanth', 20, 95000]
]

df1 = pd.DataFrame(data, columns = ['Name', 'Age', 'Salary'])

print(df1)

df2 = df1.replace(np.nan, 0)

print()
print(df2)
```

**Output**

```
   Name  Age  Salary
0  Shahid  NaN   40000
1  Daniel  16.0  20000
2   Veeru  45.0  90000
3 Sumanth  20.0  95000

   Name  Age  Salary
0  Shahid   0.0   40000
1  Daniel  16.0  20000
2   Veeru  45.0  90000
3 Sumanth  20.0  95000
```