

```
import numpy as np
import pandas as pd
```

▼ Basic Data Cleaning

- Data cleaning means fixing bad data in your data set.
- Bad data could be:
 - Empty cells
 - Data in wrong format
 - Wrong data
 - Duplicates

1. Removing Duplicates :

- To discover duplicates, we can use the duplicated() method.
- The duplicated() method returns a Boolean values for each row.
- To remove duplicates, use the drop_duplicates() method.
- **Remember:** The (inplace = True) will make sure that the method does NOT return a new DataFrame, but it will remove all duplicates from the original DataFrame.

```
data = {
    'Id': [1,2,1,3,2,2,3],
    'Name': ['A','B','A','D','B','B','D'],
    'City': ['P','O','P','S','O','O','R'],
    'Year': [2013, 2023, 2013, 2025, 2023, 2023, 2017]
}
```

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

	Id	Name	City	Year
0	1	A	P	2013
1	2	B	O	2023
2	1	A	P	2013
3	3	D	S	2025
4	2	B	O	2023
5	2	B	O	2023
6	3	D	R	2017

```
# Find Duplicate rows in the DataFrame
# Returns True for duplicated rows and False otherwise.
```

```
print(df.duplicated())
```

0	False
1	False
2	True
3	False
4	True

```
5     True
6     False
dtype: bool
```

```
# Remove the duplicated rows
```

```
print(df.drop_duplicates())
```

```
# Returns DF with unique records only
```

```
   Id Name City  Year
0   1   A    P  2013
1   2   B    O  2023
3   3   D    S  2025
6   3   D    R  2017
```

2. Fixing Wrong Data:

- Wrong data may be data not in the default range.
- Or can also be an outlier.
- One way to fix wrong values is to replace them with something else.

```
# 1. Replacing the wrong data with some other value from the column.
```

```
data = {
    'Name': ['A', 'B', 'A', 'D', 'B', 'B', 'D'],
    'Duration': [30, 60, 45, 60, 30, 180, 30],
    'City': ['P', 'O', 'P', 'S', 'O', 'O', 'R'],
    'Year': [2013, 2023, 2013, 2025, 2023, 2023, 2017]
}
df = pd.DataFrame(data)
# In Duration, values are : 30, 45 OR 60, but 180 not lies in that.

# Replace 180 with other value
df.loc[5, 'Duration'] = 45

print(df)
```

```
   Name  Duration City  Year
0    A         30    P  2013
1    B         60    O  2023
2    A         45    P  2013
3    D         60    S  2025
4    B         30    O  2023
5    B         45    O  2023
6    D         30    R  2017
```

- For small number of wrong data, can replace them individually.
- But, for larger amount of wrong data, have to create some rules for replacing the wrong data.

```
# 2. Removing Rows
```

```
# So, no need to find what to replace with the wrong data.
```

```
# But, have to sure that, the wrong data must be very less.
```

```
data = {
    'Name': ['A', 'B', 'A', 'D', 'B', 'B', 'D'],
```

```

'Duration': [30, 60, 45, 60, 30, 180, 30],
'City': ['P','O','P','S','O','O','R'],
'Year': [2013, 2023, 2013, 2025, 2023, 2023, 2017]
}
df = pd.DataFrame(data)

# Remove rows with Duration >60
for i in df.index:
    if df.loc[i, 'Duration'] > 60:
        df.drop(i, inplace=True)

print(df)

```

	Name	Duration	City	Year
0	A	30	P	2013
1	B	60	O	2023
2	A	45	P	2013
3	D	60	S	2025
4	B	30	O	2023
6	D	30	R	2017

3. Cleaning Data of Wrong Format:

- Cells with data of wrong format can make it difficult, or even impossible, to analyze data.
- To fix it, you have two options:
 - remove the rows
 - convert all cells in the columns into the same format.

1. Convert to correct format

```

data = {
    'Name': ['A','B','A','D','B','B','D'],
    'Duration': [30, 60, 45, 60, 30, 180, 30],
    'City': ['P','O','P','S','O','O','R'],
    'Year': [2013, 2023, 2013, 2025, 2023, 2023, 2017],
    'Date': ['2020/12/01', '2020/12/02', '2020/12/03', '2020/12/04',
            '2020/12/05', np.nan, '2020/12/07']
}

```

Data column contains 1 NULL value, which does not match with date format

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

```
print(df)
```

	Name	Duration	City	Year	Date
0	A	30	P	2013	2020/12/01
1	B	60	O	2023	2020/12/02
2	A	45	P	2013	2020/12/03
3	D	60	S	2025	2020/12/04
4	B	30	O	2023	2020/12/05
5	B	180	O	2023	NaN
6	D	30	R	2017	2020/12/07

1. Convert to same format

```

# Pandas has a to_datetime() method,
# to convert all cells in the 'Date' column into dates.

```

```
# So, all values from other format, converted to DateTime.
```

```
df['Date'] = pd.to_datetime(df['Date'])
print(df)
# Here, NaT = Not a Date, for null values
```

	Name	Duration	City	Year	Date
0	A	30	P	2013	2020-12-01
1	B	60	O	2023	2020-12-02
2	A	45	P	2013	2020-12-03
3	D	60	S	2025	2020-12-04
4	B	30	O	2023	2020-12-05
5	B	180	O	2023	NaT
6	D	30	R	2017	2020-12-07

```
# 2. Remove rows with null values
```

```
df.dropna(subset=['Date'], inplace=True)
```

```
print(df)
# Thus, removes records with NULL values.
```

	Name	Duration	City	Year	Date
0	A	30	P	2013	2020-12-01
1	B	60	O	2023	2020-12-02
2	A	45	P	2013	2020-12-03
3	D	60	S	2025	2020-12-04
4	B	30	O	2023	2020-12-05
6	D	30	R	2017	2020-12-07

4. Cleaning Empty Cells:

- Empty cells can potentially give you a wrong result when you analyze data.

```
# 1. Remove Rows
```

```
# Removing some rows in very large dataset, does not affect much
```

```
data = {
    'Name': ['A', 'B', 'A', 'D', 'B', 'B', 'D', 'G', 'H'],
    'Duration': [30, 60, 45, 60, 30, 45, 30, 45, 60],
    'City': ['P', 'O', 'P', 'S', np.nan, 'O', 'R', 'S', 'R'],
    'Year': [2013, np.nan, 2013, 2025, 2023, 2023, 2017, 2013, 2018]
}
```

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

```
# Just remove the rows with null cell
```

```
df.dropna(inplace=True)
```

```
print(df)
```

	Name	Duration	City	Year
0	A	30	P	2013.0
2	A	45	P	2013.0
3	D	60	S	2025.0
5	B	45	O	2023.0
6	D	30	R	2017.0
7	G	45	S	2013.0
8	H	60	R	2018.0

```
# 2. Replace Empty Values
# Fill another value in the place of empty cells
data = {
    'Name': ['A','B','A','D','B','B','D','G','H'],
    'Duration': [30, 60, 45, 60, 30, 45, 30, 45, 60],
    'City': ['P','O','P','S',np.nan,'O','R','S','R'],
    'Year': [2013, np.nan, 2013, 2025, 2023, 2023, 2017, 2013, 2018]
}
```

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

```
# Can fill different values for each columns of the DF
df['City'].fillna('T', inplace=True)
df['Year'].fillna(2018, inplace=True)
```

```
print(df)
```

	Name	Duration	City	Year
0	A	30	P	2013.0
1	B	60	O	2018.0
2	A	45	P	2013.0
3	D	60	S	2025.0
4	B	30	T	2023.0
5	B	45	O	2023.0
6	D	30	R	2017.0
7	G	45	S	2013.0
8	H	60	R	2018.0

```
# 3. Replace Using Mean, Median, or Mode
```

```
data = {
    'Name': ['A','B','A','D','B','B','D','G','H'],
    'Age': [23, 25, 20, 19, 24, 27, np.nan, 30, 29],
    'Duration': [30, 60, 45, 60, 30, 45, 30, 45, 60],
    'City': ['P','O','P','S',np.nan,'O','R','S','R'],
    'Year': [2013, np.nan, 2013, 2025, 2023, 2023, 2017, 2013, 2018]
}
```

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

```
# Replace null in Age with Mean of the Age column
df['Age'].fillna(df['Age'].mean(), inplace=True)
```

```
# Replace null in City with Mode of the City column
df['City'].fillna(df['City'].mode(), inplace=True)
```

```
# Replace null in Year with Median of the Year column
df['Year'].fillna(df['Year'].median(), inplace=True)
```

```
print(df)
```

	Name	Age	Duration	City	Year
0	A	23.000	30	P	2013.0
1	B	25.000	60	O	2017.5
2	A	20.000	45	P	2013.0
3	D	19.000	60	S	2025.0
4	B	24.000	30	NaN	2023.0
5	B	27.000	45	O	2023.0
6	D	24.625	30	R	2017.0
7	G	30.000	45	S	2013.0
8	H	29.000	60	R	2018.0

▼ Data Cleaning and Preprocessing

Handling missing data with dropna(), fillna():

- Pandas dropna() method allows the user to analyze and drop Rows/Columns with Null values in different ways.
- **Syntax:**

```
DataFrameName.dropna(axis=0, how='any', thresh=None, subset=None, inplace=False)
```

- axis - Input can be 0 or 1 for Integer and 'index' or 'columns' for String.
- how - how takes string value of two kinds only ('any' or 'all'). 'any' drops the row/column if ANY value is Null and 'all' drops only if ALL values are null.
- thresh - thresh takes integer value which tells minimum amount of na values to drop.
- subset - It's an array which limits the dropping process to passed rows/columns through list.
- inplace - It is a boolean which makes the changes in data frame permanent if True.

```
data = {
    'Id': [1,2,3,4,5,6,7],
    'Name': ['A','B','C','D','E','F','G'],
    'City': ['P','O',np.nan,'Q','R','H',np.nan],
    'Year': [2013, 2023, np.nan, np.nan, 2018, 2014, 2017]
}

df = pd.DataFrame(data)

print(df.isna().sum()) # --> Return how many null present in each column

    Id      0
    Name    0
    City     2
    Year     2
    dtype: int64

# Drop all rows with null
# axis=0 : Drop rows having any null value
# axis=1 : Drop cols having any null value

# how='any': Drop even for only one null as well
# how='all': Drop only if all values are null
print(df.dropna(axis=0, how='all'))
```

	Id	Name	City	Year
0	1	A	P	2013.0
1	2	B	O	2023.0
2	3	C	NaN	NaN
3	4	D	Q	NaN
4	5	E	R	2018.0
5	6	F	H	2014.0
6	7	G	NaN	2017.0

```
print(df.dropna(axis=0,thresh=3))
# thresh=int : Determine how many non-null values must be present, otherwise drop.
# thresh=3: So, drop rows with less than 3 non-null values.
```

	Id	Name	City	Year
0	1	A	P	2013.0
1	2	B	O	2023.0
3	4	D	Q	NaN
4	5	E	R	2018.0
5	6	F	H	2014.0
6	7	G	NaN	2017.0

```
# Original DataFrame
print(df)
# So, can see that changes are not permanent to original DF
```

	Id	Name	City	Year
0	1	A	P	2013.0
1	2	B	O	2023.0
2	3	C	NaN	NaN
3	4	D	Q	NaN
4	5	E	R	2018.0
5	6	F	H	2014.0
6	7	G	NaN	2017.0

```
# inplace=True : Make changes permanent to the original DF
df.dropna(axis=0,inplace=True)
print(df)
```

	Id	Name	City	Year
0	1	A	P	2013.0
1	2	B	O	2023.0
4	5	E	R	2018.0
5	6	F	H	2014.0

fillna() :

- fillna() method replaces the NULL values with a specified value.
- **Syntax:**

```
dataframe.fillna(value, method, axis, inplace, limit, downcast)
```

- value - Required, Specifies the value to replace the NULL values with. This can also be values for the entire row or column. The value can be any data type: Number, String, Dictionary, Series, DataFrame
- method - Optional, default None. Specifies the method to use when replacing. 'backfill', 'bfill', 'pad', 'ffill', None
- axis - Optional, default 0. The axis to fill the NULL values along. 0, 1, 'index', 'columns'
- inplace - If True, make changes permanent to the original DF.
- limit - Optional, default None. Specifies the maximum number of NULL values to fill (if method is specified).
- downcast - Optional, a dictionary of values to fill for specific data types..

```
data = {
    'Id': [1,2,3,4,5,6,7],
    'Name': ['A','B','C','D','E','F','G'],
    'City': ['P','O',np.nan,'Q','R','H',np.nan],
    'Year': [2013, 2023, np.nan, np.nan, 2018, 2014, 2017]
}
```

```
df = pd.DataFrame(data)
df['City'].fillna('New_city', inplace=True)
# inplace=True : Makes changes permanent to the original df

print(df)
```

		Id	Name	City	Year
0	1	A		P	2013.0
1	2	B		O	2023.0
2	3	C	New_city		NaN
3	4	D		Q	NaN
4	5	E		R	2018.0
5	6	F		H	2014.0
6	7	G	New_city		2017.0

Removing duplicates with drop_duplicates()

- `inplace=True` : Make changes permanent to the original df.
- `ignore_index` : If True, starts indexing from 0 and original index otherwise on 0.

```
data = {
    'Id': [1,2,1,3,2,2,3],
    'Name': ['A','B','A','D','B','B','D'],
    'City': ['P','O','P','S','O','O','R'],
    'Year': [2013, 2023, 2013, 2025, 2023, 2023, 2017]
}
```

```
df = pd.DataFrame(data)
print(df)
# DF contains many duplicate entries
```

		Id	Name	City	Year
0	1	A		P	2013
1	2	B		O	2023
2	1	A		P	2013
3	3	D		S	2025
4	2	B		O	2023
5	2	B		O	2023
6	3	D		R	2017

```
# Subset : Take list of columns, from which to consider duplicates
print(df.drop_duplicates(subset=['Id']))
# print(df.drop_duplicates())
# Returns unique rows only.
```

		Id	Name	City	Year
0	1	A		P	2013
1	2	B		O	2023
3	3	D		S	2025

keep= keep is to control how to consider duplicate value. It has only three distinct value and default is 'first'.

- If 'first', it considers first value as unique and rest of the same values as duplicate.
- If 'last', it considers last value as unique and rest of the same values as duplicate.
- If False, it consider all of the same values as duplicates

```
print(df.drop_duplicates(keep=False))
```

	Id	Name	City	Year
3	3	D	S	2025
6	3	D	R	2017

▼ Data Transformation and Feature Engineering

Applying functions to DataFrame columns using apply() and map():

- Pandas.apply() allows to pass a function and apply it on every single value of the Pandas series. i.e. all values in the given column
- **Syntax:**

```
s.apply(func, convert_dtype=True, args=())
```

- func : Takes a function that needs to applied to all values in the Series.
- convert_dtype: If True, Convert dtype as per the function's operation.
- args=(): Additional arguments to pass to function instead of series.

```
data = {
    'Name': ['A', 'B', 'C', 'D', 'E', 'F'],
    'Marks': [56, 78, 90, 34, 62, 78],
    'City': ['P', 'O', 'P', 'S', 'O', 'O']
}
```

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

```
def give(num):
    if num>60:
        return 'Pass'
    else:
        return 'Fail'
```

```
print(df['Marks'].apply(give, convert_dtype=True))
```

0	Fail
1	Pass
2	Pass
3	Fail
4	Pass

```
5    Pass
   Name: Marks, dtype: object
```

map():

- `pandas.map()` is used to map values from two series having one column same.

```
s1 = pd.Series([0,1,2,3,4,5])
s2 = pd.Series([21,24,26,32,34,31])
```

```
print(s1.map(s2))
```

```
0    21
1    24
2    26
3    32
4    34
5    31
dtype: int64
```

▼ Working with Text Data

Handling text data in pandas (e.g., string methods):

- Series provides various methods, for operating on Strings, using the `str` attribute.

1. **`str.lower()`**: Method to convert a string's characters to lowercase.
2. **`str.upper()`**: Method to convert a string's characters to uppercase.
3. **`str.find()`**: Method is used to search a substring in each string present in a series.
4. **`str.rfind()`**: Method is used to search a substring in each string present in a series from the Right side.
5. **`str.findall()`**: Method is also used to find substrings or separators in each string in a series.
6. **`str.isalpha()`**: Method is used to check if all characters in each string in series are alphabetic(a-z/A-Z).
7. **`str.isdecimal()`**: Method is used to check whether all characters in a string are decimal.
8. **`str.title()`**: Method to capitalize the first letter of every word in a string.
9. **`str.len()`**: Method returns a count of the number of characters in a string.
10. **`str.replace()`**: Method replaces a substring within a string with another value that the user provides.
11. **`str.contains()`**: Method tests if pattern or regex is contained within a string of a Series or Index.
12. **`str.extract()`**: Extract groups from the first match of regular expression pattern.
13. **`str.startswith()`**: Method tests if the start of each string element matches a pattern.
14. **`str.endswith()`**: Method tests if the end of each string element matches a pattern.
15. **`str.isdigit()`**: Method is used to check if all characters in each string in series are digits.
16. **`str.lstrip()`**: Method removes whitespace from the left side (beginning) of a string.

17. **str.rstrip()**: Method removes whitespace from the right side (end) of a string.
18. **str.strip()**: Method to remove leading and trailing whitespace from string.
19. **str.split()**: Method splits a string value, based on an occurrence of a user-specified value.
20. **str.join()**: Method is used to join all elements in list present in a series with passed delimiter.
21. **str.cat()**: Method is used to concatenate strings to the passed caller series of string.
22. **str.repeat()**: Method is used to repeat string values in the same position of passed series itself.
23. **str.get()**: Method is used to get element at the passed position.
24. **str.partition()**: Method splits the string only at the first occurrence unlike str.split().
25. **str.rpartition()**: Method splits string only once and that too reversely. It works in a similar way like str.partition() and str.split()
26. **str.pad()**: Method to add padding (whitespaces or other characters) to every string element in a series.
27. **str.swapcase()**: Method to swap case of each string in a series.

```
data = {
    'Name': ['Abc', 'Bob', 'Cat', 'Dog', 'Ele', 'Fog'],
    'Marks': [56, 78, 90, 34, 62, 78],
    'City': ['Pa', 'OpD', 'PsT', 'SpY', 'On', 'Om']
}
```

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

```
print(df)
```

	Name	Marks	City
0	Abc	56	Pa
1	Bob	78	OpD
2	Cat	90	PsT
3	Dog	34	SpY
4	Ele	62	On
5	Fog	78	Om

```
# 1. str.lower() - Convert the string to lowercase chars
```

```
df['Name'] = df['Name'].str.lower()
```

```
print(df)
```

	Name	Marks	City
0	abc	56	Pa
1	bob	78	OpD
2	cat	90	PsT
3	dog	34	SpY
4	ele	62	On
5	fog	78	Om

```
# 2. str.upper() - Convert the string to uppercase chars
```

```
df['City'] = df['City'].str.upper()
```

```
print(df)
```

	Name	Marks	City
0	abc	56	PA
1	bob	78	OPD

2	cat	90	PST
3	dog	34	SPY
4	ele	62	ON
5	fog	78	OM

Regular expressions for pattern matching:

```
import re

data = {
    'Name': ['Abc', 'Bob', 'Cat', 'Dog', 'Ele', 'Fog'],
    'Marks': [56, 78, 90, 34, 62, 78],
    'City': ['Pa', 'OpD', 'PsT', 'SpY', 'On', 'Om']
}

df = pd.DataFrame(data)

pattern = '[0].*'

print(df[df.City.str.match(pattern)])
```

	Name	Marks	City
1	Bob	78	OpD
4	Ele	62	On
5	Fog	78	Om