

Data Transformtant Anids





### A Practical Guide for Data Cleaning in Python.

• As a Data Analyst Data cleaning is the first and most crucial step in any data analysis process. Python offers powerful functions to handle functions to handle missing values, convert data types, and prepare your data for deeper analysis.



### **Data Cleaning**

dropna()

Remove missing missing values

dataset

fillna()

Fill missing values with a specified value

astype()

Convert column data type

nan\_to\_nu m()

Replace NaN with numeric value

### reshape()

values from a

Reshape arrays without changing data

#### unique()

Get all unique values in a column



### dropna()



# Remove missing values from a dataset

Used to remove rows or columns with missing values (NaNs) in a DataFrame.



#### **Example:**

```
import pandas as pd

df =

pd.read_csv('data.csv')
```

df\_clean = df.dropna()

1	2	1.3	
1	30	133	200
2	30	188	234
4	30	239	238
5	29	238	237
0	38	238	238
4	30	274	235
6	50	295	239
7	38	238	230
18	42	238	280
17	38	235	246
14	28	233	286
15	38	288	280
10	36	238	257
10	65	353	250

## fillna()



#### **Purpose**

•Fill missing values in a dataset

Replaces missing values with a specified specified constant or method (like forward fill or backward fill).

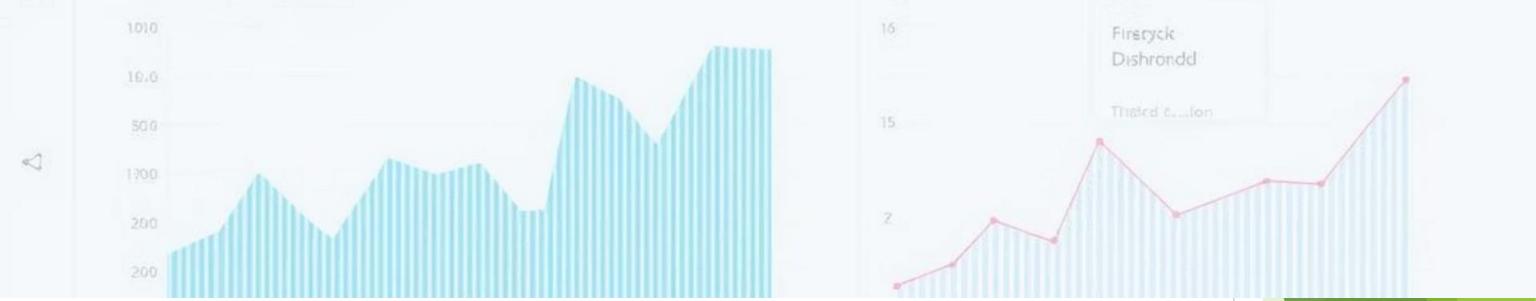


#### **Example**

df['column'].fillna(0, inplace=True)



	Tatr Oill	Pay Ont	Tate Oill	Tatr Oril	Patr Oill	Parcinil Fecestione	Peta Fecelo
	Polirence	Ecopmante	Represnate	Politocase	Pollenate	Popment	Pollen
ı	31270	13,757	17/20	13,7570	15,670	16,650	13
1	33700	11,420	12579	23/420	35,300	14,000	12
1	34570	13,750	13573	25,1319	14,999	25,730	12
1	42550	13,404	13544	263550	15,790	11,654	12
ı	35600	12,954	13744	223550	13,200	13,700	12
1	33800	22.790	15504	266570	19,820	13,844	1
1	35590	12,990	18083	227550	16,090	12,740	7
1	63170	13,740	14645	221252	16/355	11.597	3
1	33850	12,400	15564	225590	4,1150	13,487	
1	33555	14,200	16874	166752	3,550	13,747	2
1	34590	41,700	15148	286754	3.5545	13,847	3
1	31590	13,900	16574	245316	5/387	15,249	3
1	35590	13,630	12594	125538	9.1370	16,644	2
1	25560	11,790	13972	145508	3,1389	16.740	9
١	33062	15,990	13676	215560	34090	13,400	3
1	35770	15,990	12376	165390	41399	13,540	€
1	31650	12,700	18677	236440	35367	13,543	2
1	33380	15,797	15778	275108	93772	15,500	67
1	35120	15,800	14773	225550	43397	16,770	9
1	35760	13,500	18647	224550	83357	12,508	1
1	32350	12,450	19664	225858	45940	15,007	2
1	61140	13,950	15564	236990	35666	15,770	3
1	33590	11,750	15994	266559	45767	15,750	2
1	26750	13,280	16074	186550	93200	11,609	2
1	23140	13,780	15594	286550	34550	11,800	1
1	23780	11,782	16807	26,/525	11,900	13,500	2
1	23500	13,420	15000	227.360	12.070	12,320	2
1	35160	17,730	15979	23/024	13.500	15,477	2
1	31500	22,660	29907	235377	18,300	13,370	3
ı	33900	50,450	26677	283900	18,400	13,799	7
1	53500	45,720	26674	26,550	15,540	13,394	2
1	41730	32,920	15703	25/224	11,040	13,780	2
1	41350	33,790	16679	255530	13,157	13,799	2
1	35729	40,460	15077	286490	13,172	18,742	173
1	61252	62,010	17573	56/559	13)130	14,339	3



## astype()

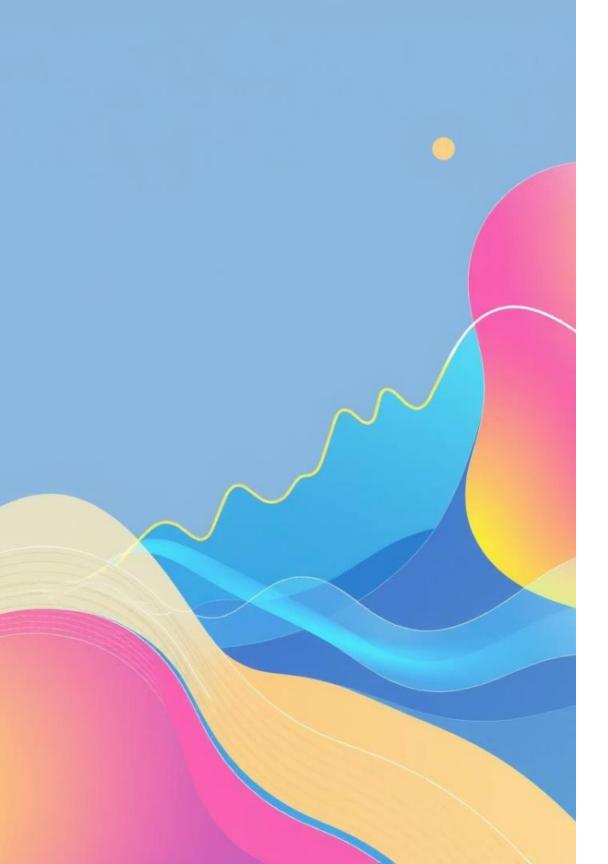
#### **Convert data types of columns**

Used to change the data type of a column (e.g., float to int, int, object to datetime).

#### **Example:**

df['column'] = df['column'].astype(int)





### nan\_to\_num()

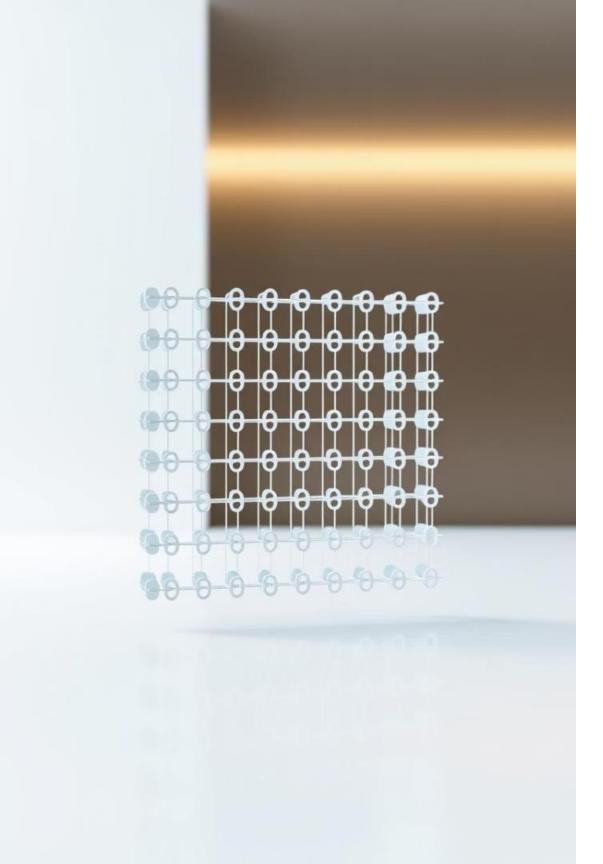
#### Replace NaN with numeric values (NumPy)

Useful for numerical computations where NaN values need to be replaced with 0 or other values.

#### **Example:**

```
import numpy as np
data = np.array([1, 2, np.nan])
clean_data = np.nan_to_num(data, nan=0)
```





### reshape()

1 Reshape arrays (NumPy)

Used to change the shape of arrays without modifying their data.

2 Example:

•array = np.array([[1, 2], [3, 4]])
reshaped = array.reshape((4,))



### unique()

#### Get unique values from a column

Returns unique values from a column or array, useful for understanding understanding category spreads.

#### **Example:**

unique\_values = df['column'].unique()



## Thank You!