### Master Data Cleaning with Python

Streamline your data preparation process



## Remove Duplicates

Duplicates in a dataset can skew analysis results. You can remove duplicates using Pandas' drop\_duplicates() method.



## Handle Missing Values

Missing values can impact the accuracy of analysis. You can identify and handle missing values using Pandas' isnull(), dropna(), or fillna() methods.



## **Convert Data Types**

Standardizing data formats ensures consistency across the dataset. You can use Pandas' to\_datetime() function or custom functions to convert data into a specific format.



### Handle Data Inconsistencies

Data inconsistencies such as typos or variations in formatting can be corrected using string methods or custom functions.



## Removing Outliers

Outliers can significantly affect statistical analysis. You can identify and remove outliers using statistical techniques like z-score or IQR method.



### Data Validation

Validating data against predefined rules or constraints helps ensure data integrity. You can use custom functions or libraries like pandasschema for data validation.



# Handling categorical variables

For machine learning models, categorical variables often need to be encoded. You can use Pandas' get\_dummies() method for one-hot encoding.





#### Scaling Numerical Features

Some machine learning algorithms require scaled features. You can use scikit-learn's preprocessing module for scaling.



#### In a nutshell

Clean data by removing duplicates, handling missing values, converting data types, and standardizing text.



# Ready to level up your data cleaning game?

Share your favorite data cleaning tips in the comments below!

