

MLOps Training

MACHINE LEARNING OPERATIONS



Exploratory Data Analysis (EDA)

- Day 1 Recap
- Day 1 Workshop Quiz
- Understanding datasets: nulls, outliers, distributions
- Visualization tools (Pandas, Seaborn)
- Feature engineering and transformation
- Day 2 Quiz
- Day 2 Workshop

Day 2 Agenda



Day 1 Recap

OVERVIEW OF MACHINE LEARNING



Q1: Fill in the blank to correctly import the Linear Regression model in Python:



Choose the correct option

A: LinearRegression

B: linear_regression

C: regression

D: lin_reg



Q1: Fill in the blank to correctly import the Linear Regression model in Python:



```
import pandas as pd
from sklearn.linear_model import ______
```

Choose the correct option

A: LinearRegression

B: linear_regression

C: regression

D: lin_reg



Q2: Fill in the blank to correctly import the Logistic Regression model in Python:



```
import pandas as pd
from sklearn.linear_model import ______
```

Choose the correct option

A: Regression

B: logistic_regression

C: LogisticRegression

D: log_reg



Q2: Fill in the blank to correctly import the Logistic Regression model in Python:



```
import pandas as pd
from sklearn.linear_model import ______
```

Choose the correct option

A: Regression

B: logistic_regression

C: LogisticRegression

D: log_reg



Q3:Which module should you import train_test_split from in Python?



```
import pandas as pd
from _____ import train_test_split
```

- A. sklearn.preprocessing
- B. sklearn.model_selection
- C. sklearn.metrics
- D. sklearn.linear_model



Q3:Which module should you import train_test_split from in Python?



```
import pandas as pd
from _____ import train_test_split
```

- A. sklearn.preprocessing
- B. sklearn.model_selection
- C. sklearn.metrics
- D. sklearn.linear_model



Q4: Which of the following is the correct function used to load a .csv file into a pandas DataFrame in Python?

```
import pandas as pd
df = pd._____("ice_cream.csv")
```

- A. pandas.load_csv('filename.csv')
- B. pandas.open_csv('filename.csv')
- C. pandas.import_csv('filename.csv')
- D. pd.read_csv('filename.csv')





Q4: Which of the following is the correct function used to load a .csv file into a pandas DataFrame in Python?

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- A. pandas.load_csv('filename.csv')
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- C. pandas.import_csv('filename.csv')
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Q5: What does the following line do in Python?

joblib.dump(model_lin_reg, 'linear_regression_model.pkl')

- A. It saves the model to a file
- B. It loads a model from a file
- C. It evaluates the model and prints results
- D. It deletes the model from memory





Q5: What does the following line do in Python?

joblib.dump(model_lin_reg, 'linear_regression_model.pkl')

- A. It saves the model to a file
- B. It loads a model from a file
- C. It evaluates the model and prints results
- D. It deletes the model from memory





Q6: What does test_size=0.2 indicate in the train_test_split() function?

```
X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.2,random_state=1)
```

- A. The data will not be split
- B. 20% of the data will be used for testing
- C. 80% of the data will be used for testing
- D. 20% of the data will be used for training





Q6: What does test_size=0.2 indicate in the train_test_split() function?

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- A. The data will not be split
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- C. 80% of the data will be used for testing
- D. 20% of the data will be used for training



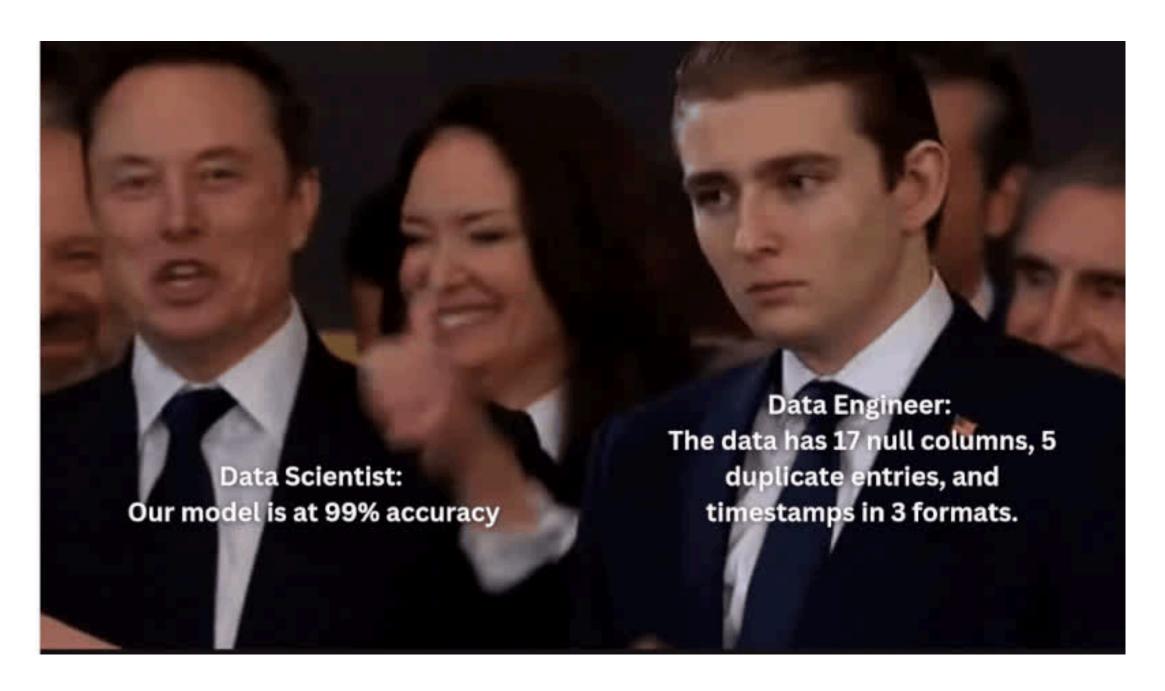




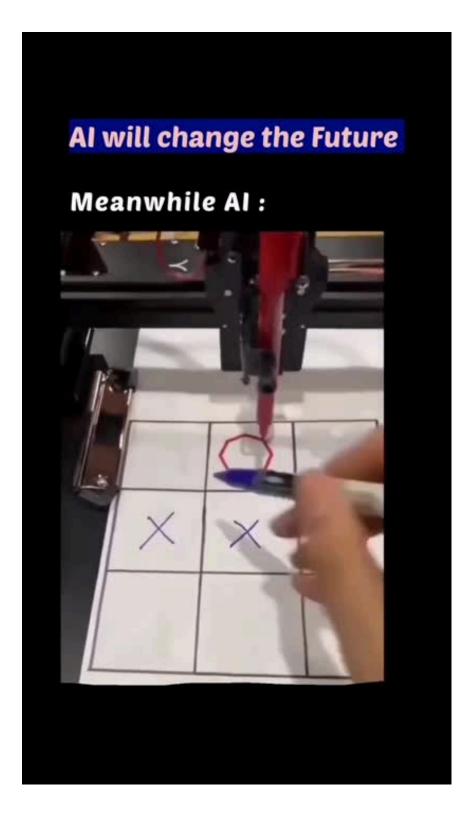






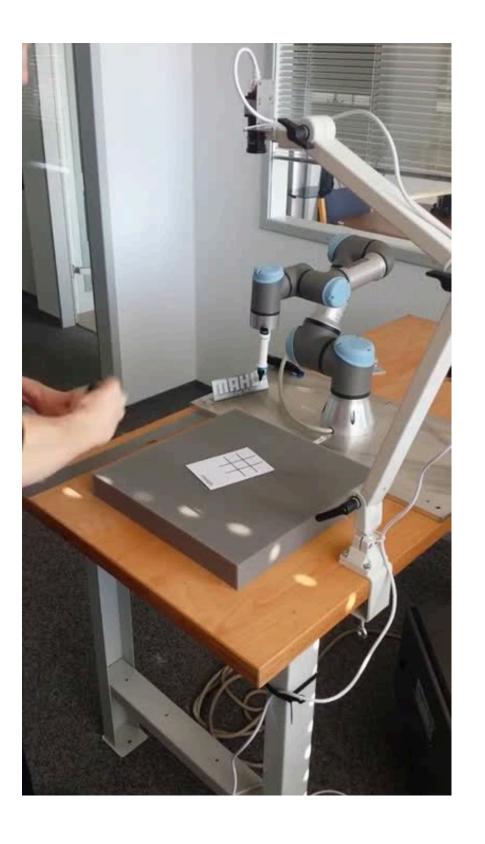


Before EDA



After EDA







EDA helps us explore and understand the structure, patterns, and issues in data before modeling.

- It's the diagnostic phase of data science.
- Helps detect missing values, outliers, and shape of distributions.
- Guides feature engineering and cleaning decisions



Can we split EDA?

- EDA Level O Initial Exploration of Raw Data
 - Gaining a basic understanding of the unprocessed dataset.
- EDA Level 1 Data Cleaning and Transformation
 - Applying preprocessing techniques to make the data usable.
- EDA Level 2 Analysis of Processed Data
 - Exploring patterns, trends, and relationships in the cleaned dataset.



Why is EDA Important?

Before you build models, you must understand the data. A model is only as good as the data you feed it. If your dataset contains inconsistencies, missing values, or outliers, your model may underperform or give incorrect results.

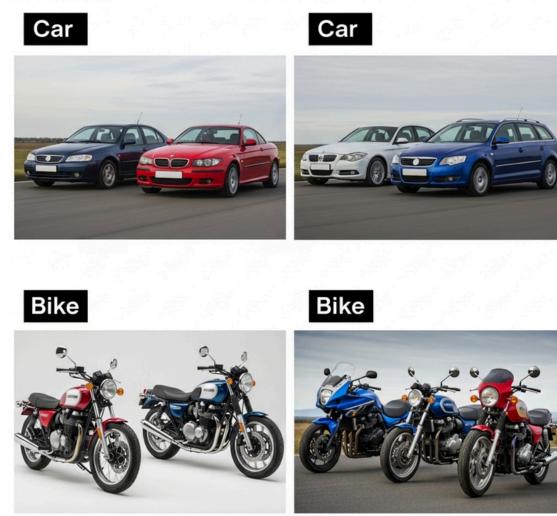
EDA enables you to:

- Detect data quality issues (nulls, duplicates, wrong types)
- Understand feature distributions and relationships
- Uncover hidden trends and patterns
- Guide feature selection and transformation



Labeled Data

Labeled data consists of data points, such as images, text, or sensor readings, that have been explicitly tagged with meaningful labels or categories.





Unlabeled data is raw data that doesn't have any pre-defined labels or categories. It's essentially the raw, unprocessed data.

Unlabeled Data







Data that conforms to a well-defined, pre-defined format, typically organized in rows and columns (like in a spreadsheet or database).

Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilities ... PoolArea PoolQC Fence MiscFeature AllPub 0 1 60 RL 65.0 8450 Pave NaN Reg Lvl NaN NaN NaN 1 2 20 RLAllPub 80.0 9600 Pave NaN Reg Lvl 0 NaN NaN NaN 2 3 60 RL 68.0 11250 Pave NaN IR1 Lvl AllPub NaN NaN 3 4 70 RL 60.0 9550 Pave NaN IR1 Lvl AllPub 0 NaN NaN NaN 4 5 60 RL 84.0 14260 Pave NaN IR1 Lvl AllPub NaN NaN AllPub **5** 6 50 RL 85.0 14115 Pave NaN IR1 Lvl NaN MnPrv Shed Reg 20 RL10084 Pave AllPub NaN 75.0 NaN Lvl NaN NaN 7 8 AllPub 60 RL 10382 Pave Lvl NaN NaN NaN Shed NaN 8 9 50 RM 6120 AllPub NaN 51.0 Pave NaN Lvl NaN 9 10 190 RL 50.0 AllPub 7420 Pave NaN Reg Lvl NaN NaN NaN

10 rows × 81 columns

Structured Data

Unstructured Data



Data that does not conform to a predefined format and lacks a clear structure.

Example:

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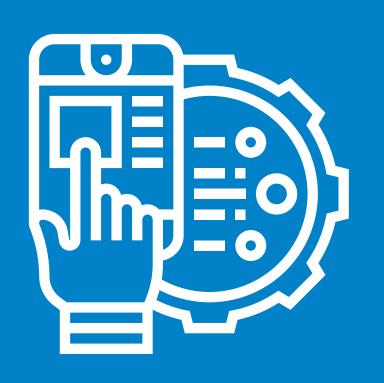
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.txt .jpg .mp3





Understanding Datasets



What is the Dataset?

Definition

- A Dataset is a set of data grouped into a collection with which engineers can work to meet their goals.
- In a structured dataset, the rows represent the number of data points and the columns represent the features of the Dataset.



Why are datasets used?

Datasets are used to train and test AI models, analyze trends, and gain insights from data. They provide the raw material for computers to learn patterns and make predictions.



Dataset Types

- Numerical Dataset
- Categorical Dataset
- Web Dataset
- Time series Dataset
- Image Dataset



Dataset Properties

- Center of data
- Skewness of data
- Correlation among the data
- Presence of outliers



Dataset Features

The features of a dataset are the most critical aspect of the dataset, as based on the features of each available data point, will there be any possibility of deploying models to find the output to predict the features of any new data point that may be added to the dataset.

Numerical Features:

Categorical Features:

Metadata:

Size of the Data:



Examples

	Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities	•••	PoolArea	PoolQC	Fence
0	1	60	RL	65.0	8450	Pave	NaN	Reg	Lvl	AllPub		0	NaN	NaN
1	2	20	RL	80.0	9600	Pave	NaN	Reg	Lvl	AllPub		0	NaN	NaN
2	3	60	RL	68.0	11250	Pave	NaN	IR1	Lvl	AllPub		0	NaN	NaN
3	4	70	RL	60.0	9550	Pave	NaN	IR1	Lvl	AllPub		0	NaN	NaN
4	5	60	RL	84.0	14260	Pave	NaN	IR1	Lvl	AllPub		0	NaN	NaN
5	6	50	RL	85.0	14115	Pave	NaN	IR1	Lvl	AllPub		0	NaN	MnPrv
6	7	20	RL	75.0	10084	Pave	NaN	Reg	Lvl	AllPub		0	NaN	NaN
7	8	60	RL	NaN	10382	Pave	NaN	IR1	Lvl	AllPub		0	NaN	NaN
8	9	50	RM	51.0	6120	Pave	NaN	Reg	Lvl	AllPub		0	NaN	NaN
9	10	190	RL	50.0	7420	Pave	NaN	Reg	Lvl	AllPub		0	NaN	NaN

10 rows × 81 columns



Dataset Understanding

This includes checking the number of rows and columns, data types, and summary statistics.

- Viewing top rows using .head()
- Understanding structure using .info() and .describe()
- Checking data types (numerical, categorical, datetime)
- Identifying duplicates or anomalies



Handling Missing Values

Detect → Drop or Impute → Validate

- Use isnull().sum()
- Imputation: Mean, Median, Mode, Domain logic
- Sometimes missing ≠ bad depends on context



Handling Missing Values

Detect → Drop or Impute → Validate

- Use isnull().sum()
- Imputation: Mean, Median, Mode, Domain logic
- Sometimes missing ≠ bad depends on context



Dealing with Outliers

Outliers are extreme values that differ significantly from other observations. They can skew analysis and impact model performance.

Ways to detect them:

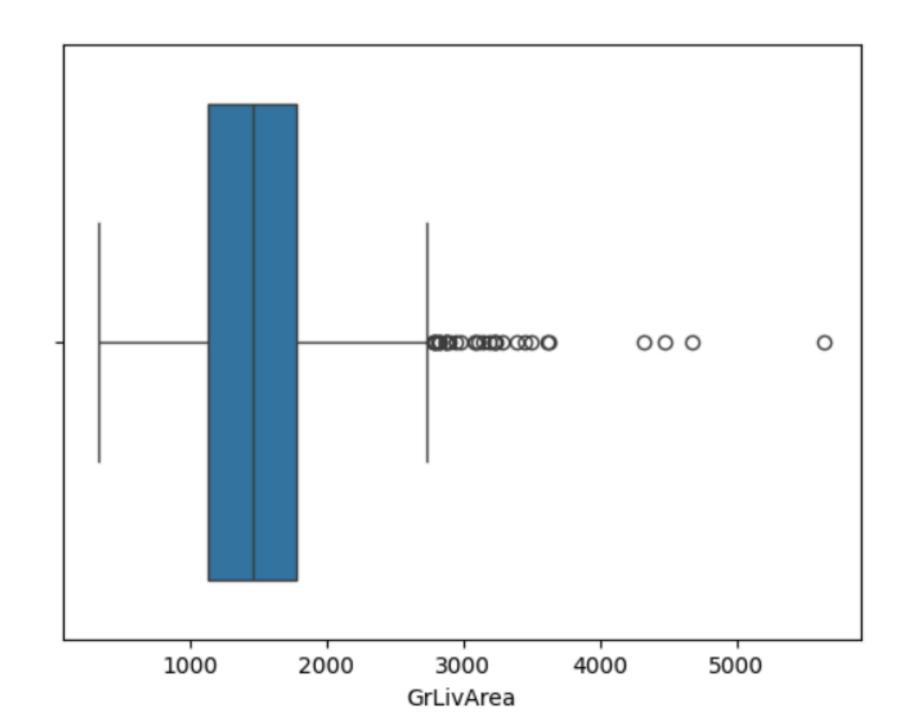
• Using boxplots, histograms

Once detected:

- You can drop them if they're errors
- Cap them if they are rare but legitimate
- Or transform the variable (e.g., apply log)



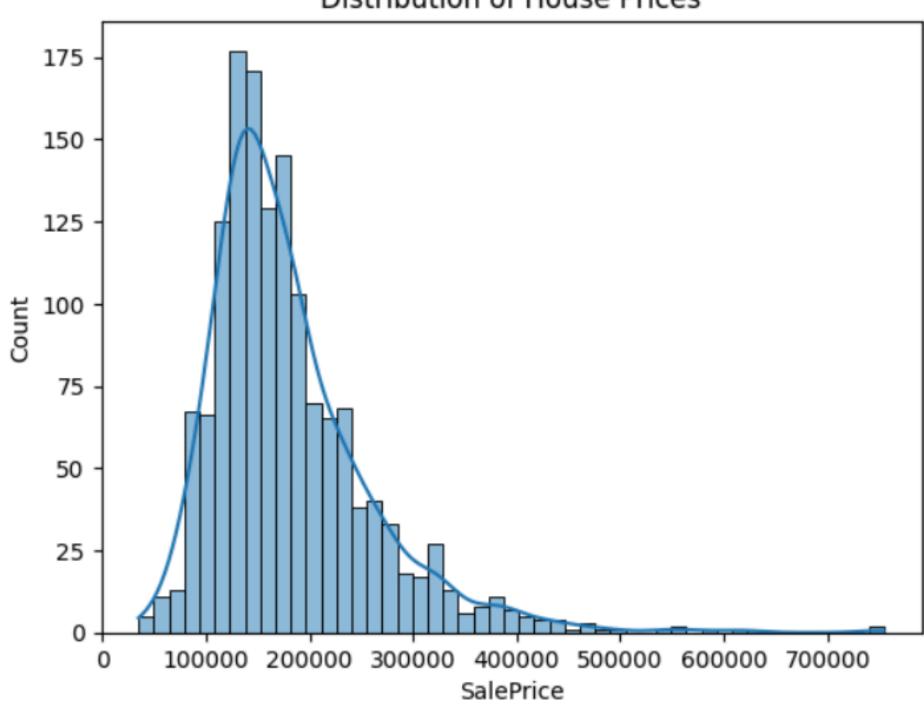
Boxplot Outliers



Hisplot Outliers









Understanding Distributions

A distribution shows how data is spread. Know your shape: Normal,

Skewed, Peaked.

Fix with log/sqrt transforms if needed.

Why It Matters

- Affects model assumptions
- May require transformation

Fixes

- Use log() or sqrt() on skewed features
- Tools: histplot, KDE, df.skew(), df.kurtosis()



Understanding Distributions

A distribution shows how data is spread. Knowing the shape helps in choosing transformations and ML algorithms.

Common Types:

- Normal: Bell-shaped; good for linear models
- Skewed: Left or right-tailed; may require log/sqrt transformation
- Bimodal: Two peaks; may indicate subgroups





Visualization Tools



Visualization Tools

Visualization is a core part of Exploratory Data Analysis. It helps you understand patterns, spot anomalies, and uncover relationships between features that are hard to detect from raw numbers alone.

Python offers two widely used libraries for data visualization in EDA:

- Pandas
- Seaborn