

AI ML Projects

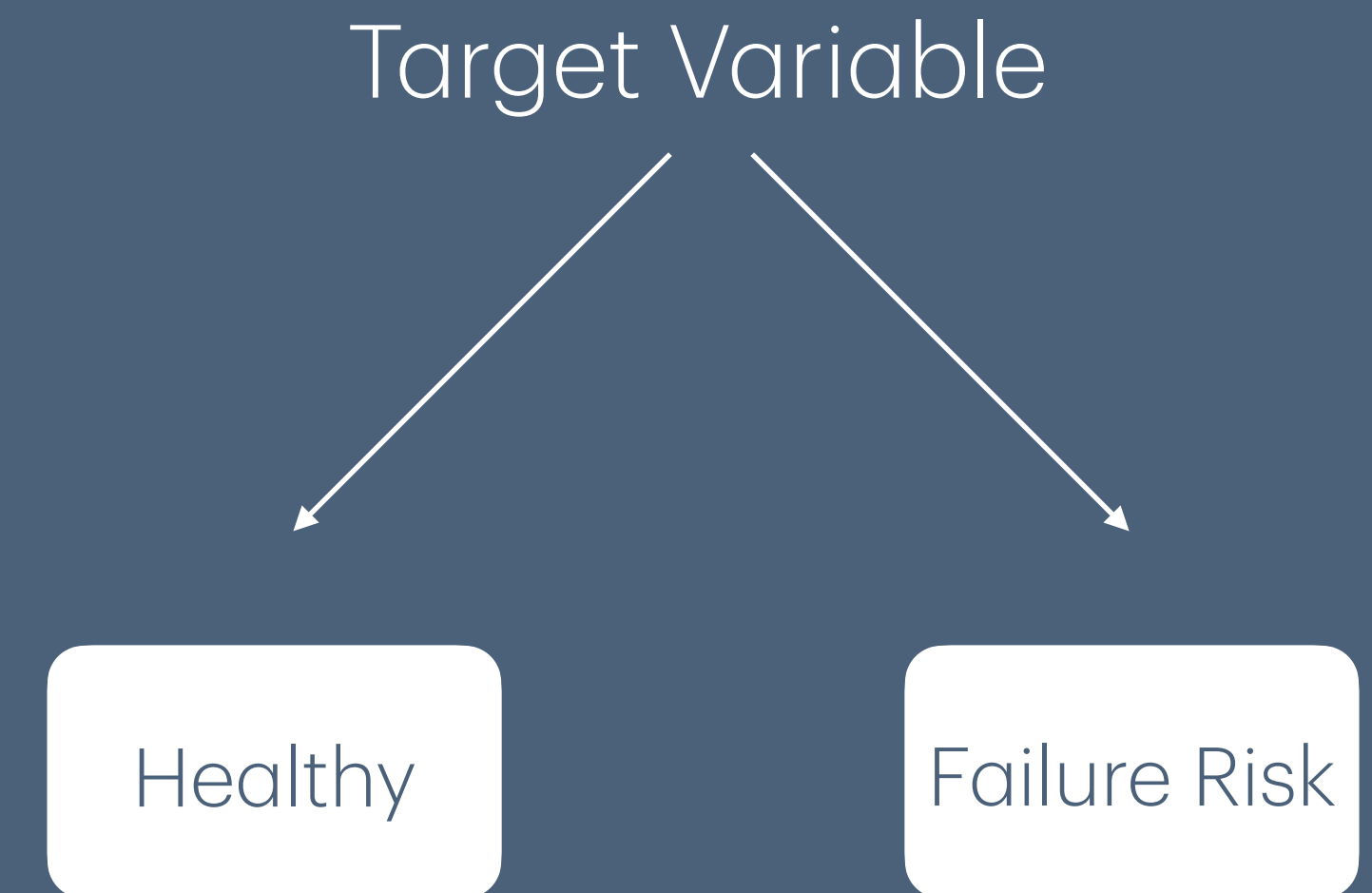
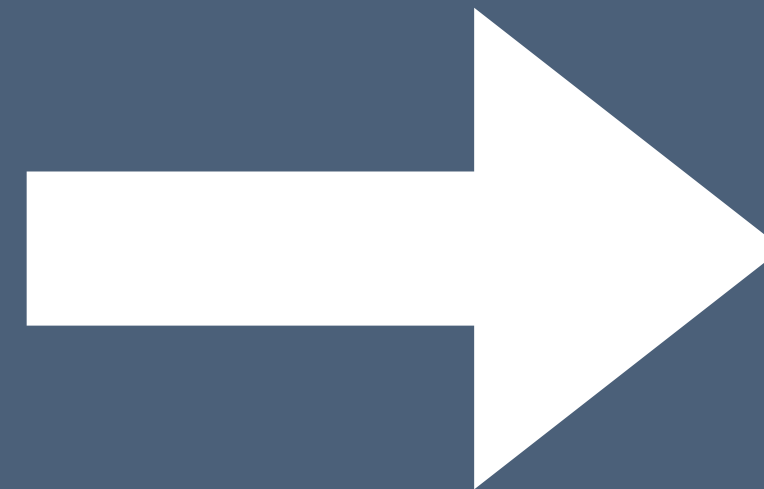
Supervised & Unsupervised

Project 1: Supervised Machine learning

Supervised Machine Learning

Problem Statement

Unexpected equipment failures lead to costly downtimes. Predicting failures before they occur allows proactive maintenance.

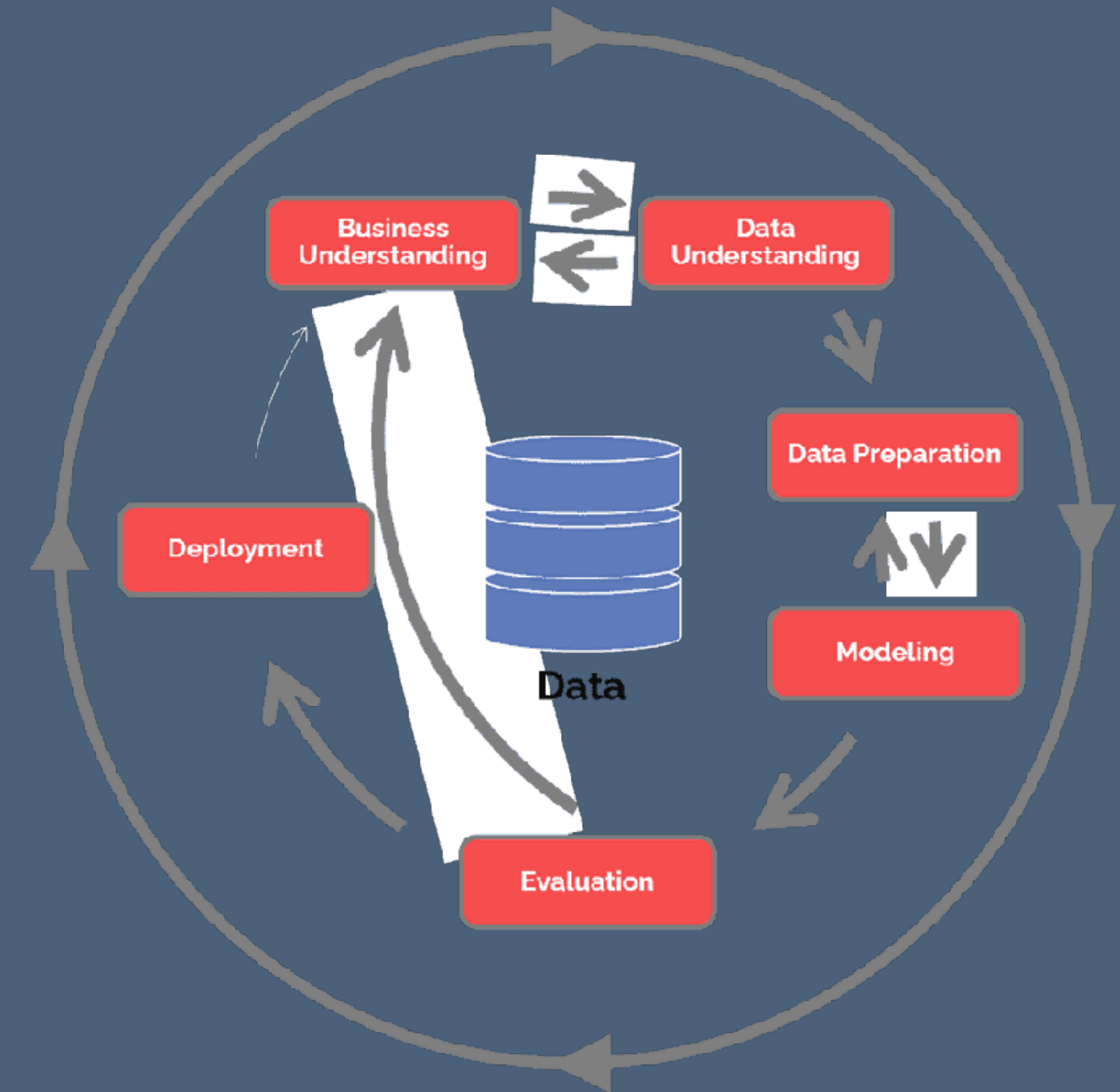
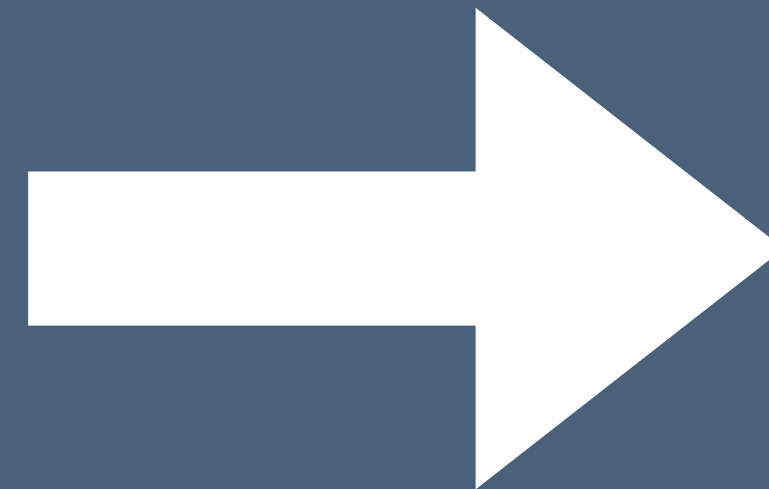


Binary Classification

Methodology Used

CRISP - DM : Cross Industry Standard Process for Data Mining

It Starts from Business
Understanding to Model
Deployment



Step By Step Process

Inspired by CRISP-DM

Data Extraction



Web Data, CSV, Excel Sheet etc...

Data Analysis



Pattern & Trend Analysis

Feature Engineering



Handling Missing Values, Unknowns etc ...

Model Building



Model Building based on use case

Model Evaluation



Model Performance Metrics like RMSE,MSE etc...

Python Use Cases



NumPy

A fundamental package to manipulate arrays

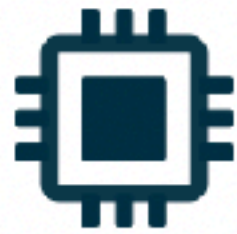
NumPy



The fundamental package for scientific computing with Python

LATEST RELEASE: NUMPY 2.3. [VIEW ALL RELEASES](#)

Quantum Computing



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[PyQuil](#)
[Qiskit](#)
[PennyLane](#)

Statistical Computing



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[statsmodels](#)
[Xarray](#)
[Seaborn](#)

Signal Processing



[SciPy](#)
[PyWavelets](#)
[python-control](#)
[HyperSpy](#)

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[OpenCV](#)
[Mahotas](#)

Graphs and Networks



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[graph-tool](#)
[igraph](#)
[PyGSP](#)

Astronomy



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[SunPy](#)
[SpacePy](#)

Cognitive Psychology



[PsychoPy](#)

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[Scikit-Bio](#)
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[ETE](#)

Bayesian Inference



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[ArviZ](#)
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[SymPy](#)
[cvxpy](#)
[FEniCS](#)

Chemistry



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Geoscience



[Pangeo](#)
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[ObsPy](#)
[Fatiando a Terra](#)

Geographic Processing

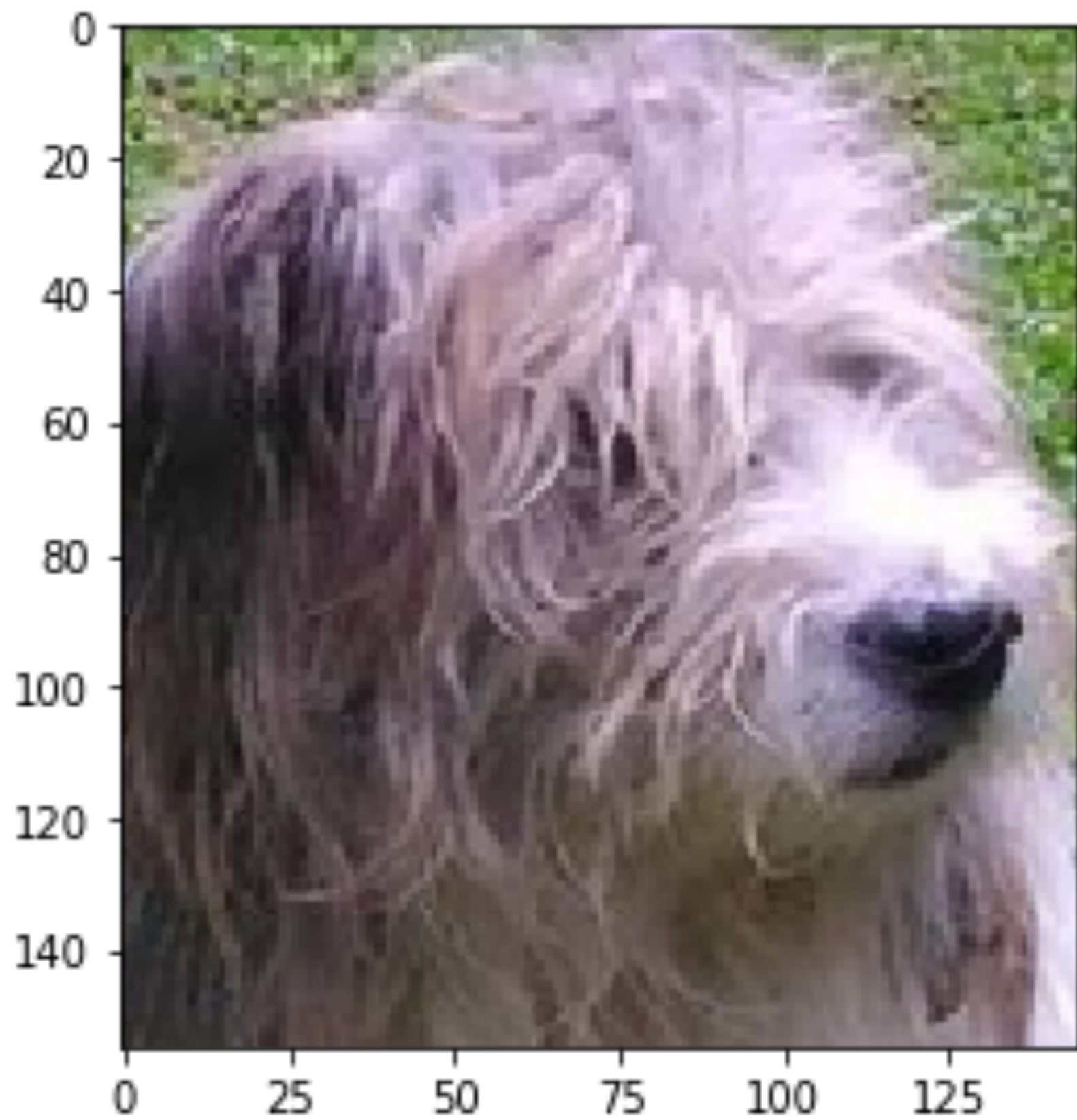


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Architecture & Engineering



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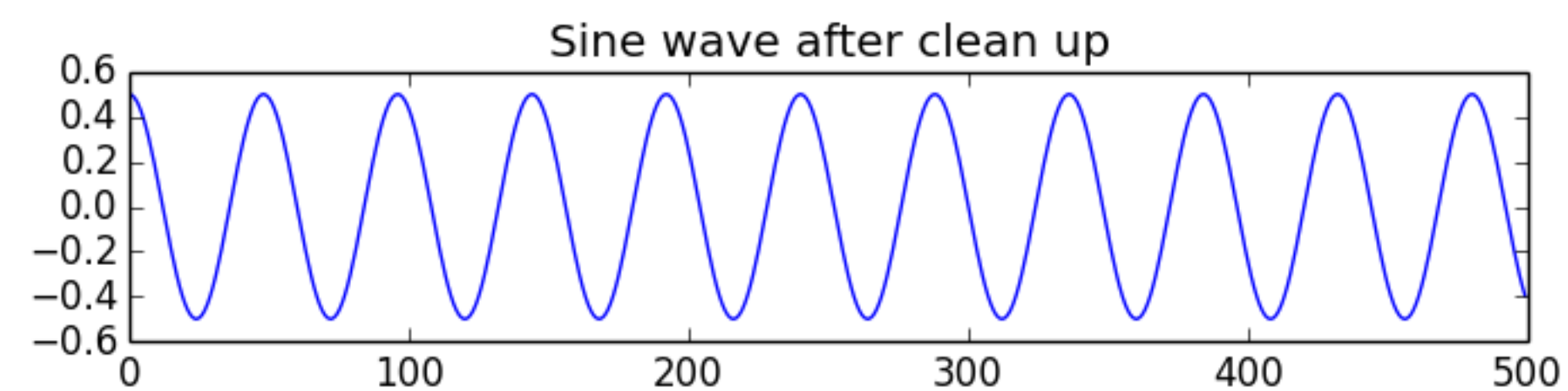
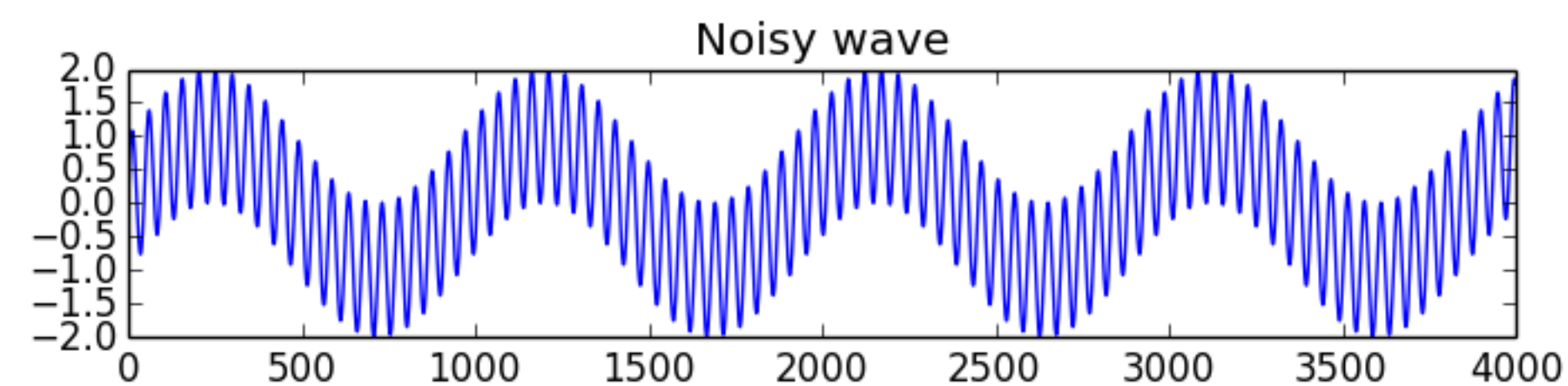
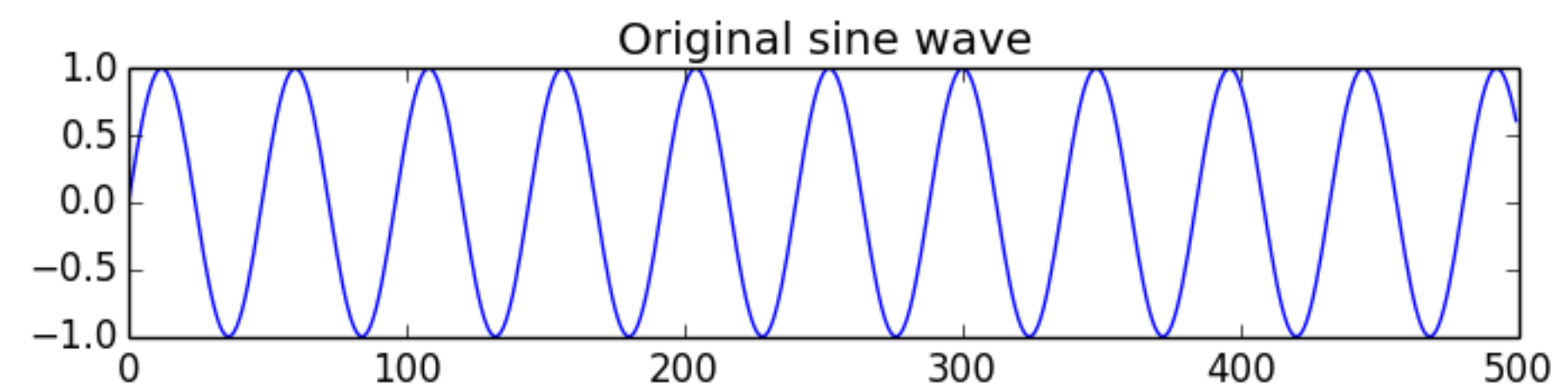
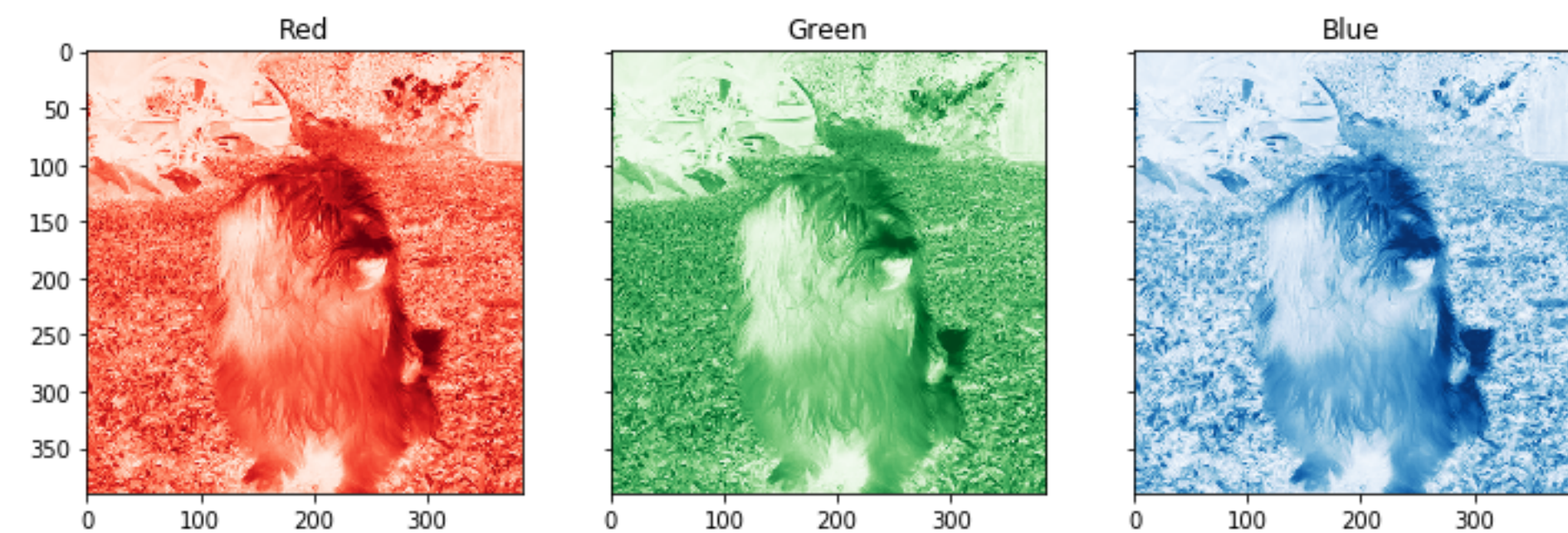
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         [ 43, 42, 41]],

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         [ 33, 36, 29],
         [ 32, 35, 28],
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         [ 37, 35, 36],
         [ 44, 43, 42]],

       [[ 34, 36, 30],
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         [ 34, 37, 30],
         ...,
         [ 40, 38, 41],
         [ 37, 35, 36],
         [ 40, 38, 38]],

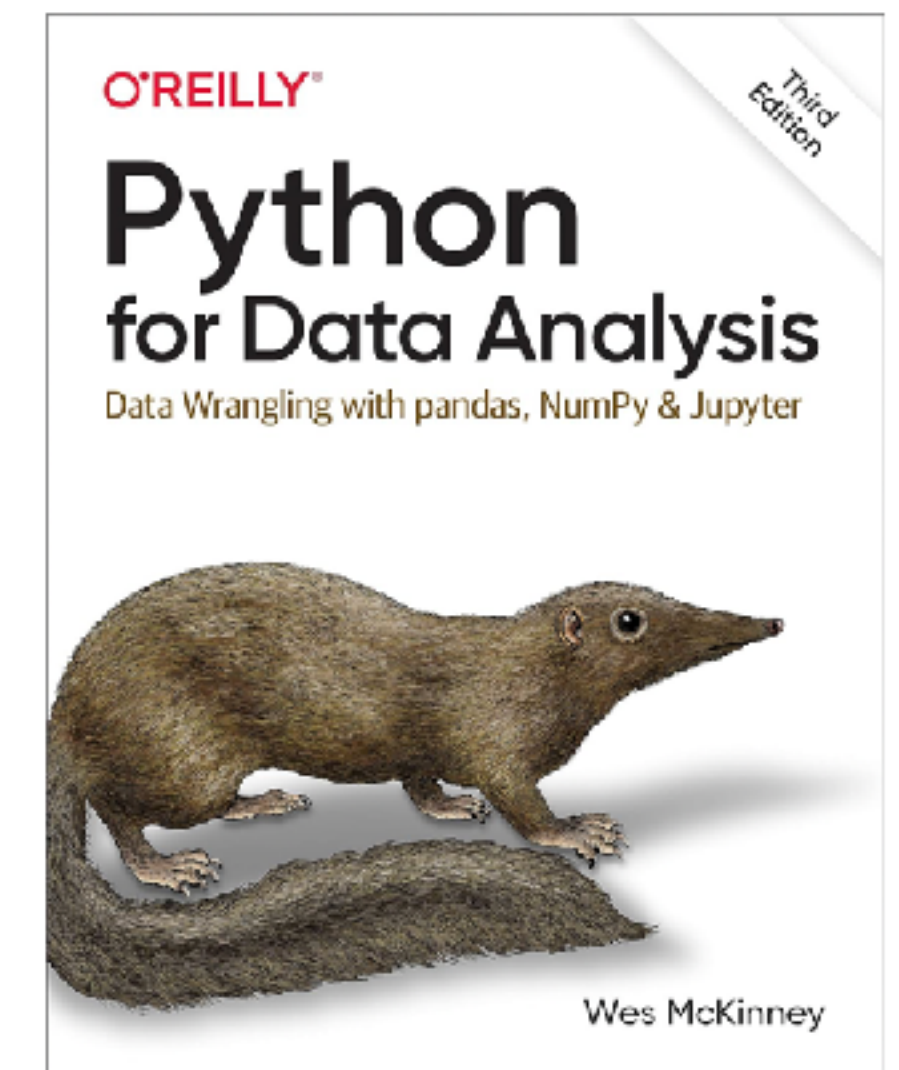
       ...,

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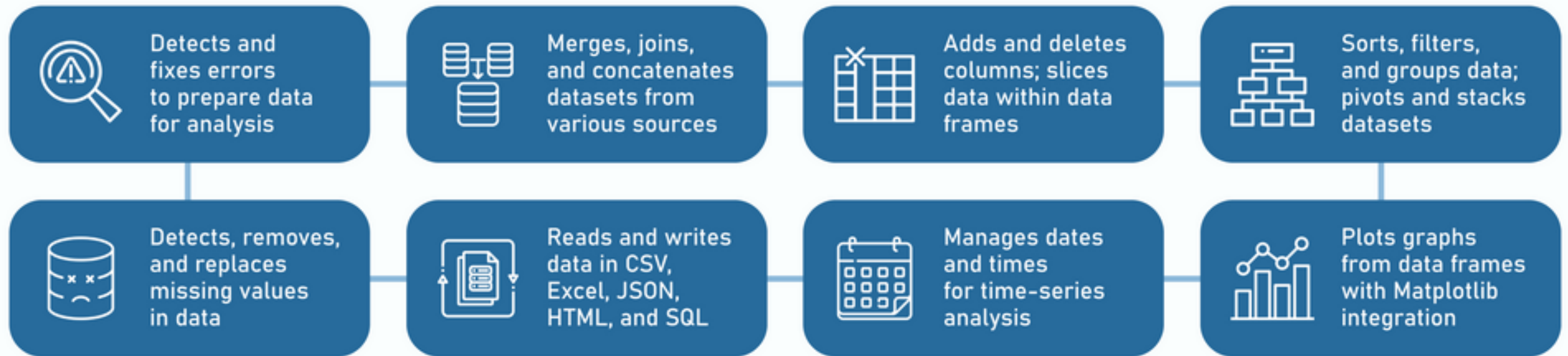


Pandas

Pandas is a fast, powerful, flexible and easy to use open source **data analysis and manipulation tool**, built on top of the [Python](#) programming language.



PANDAS MAIN CAPABILITIES



Pandas

Series

pd.Series

pd.Series(data, index)

Creation of Series

Arithmetic operations

Data Frame

pd.DataFrame

pd.DataFrame(data, index, columns)

pd.read_csv('file_path')

df.head()

df.tail()

df.describe()

df.info()

Create/drop the columns

iloc and loc

Conditional Filtering

pd.DataFrame(Conditional)

df[df['column_name'] <= 'value']

Multiple columns filtering -

And(&) Or (|)

df.isin()

GroupBy Operations

df.groupby()

Evaluation Metrics

$$\textit{Accuracy} = \frac{TP + TN}{TP + TN + FP + FN}$$

$$\textit{Precision} = \frac{TP}{TP + FP}$$

$$\textit{Recall} = \frac{TP}{TP + FN}$$

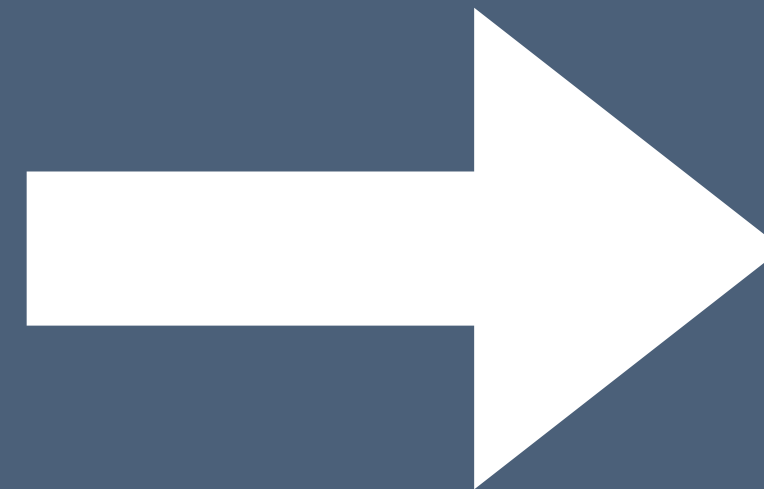
$$F_1 = 2 \cdot \frac{\textit{Precision} \cdot \textit{Recall}}{\textit{Precision} + \textit{Recall}}$$

Project 2: Un-supervised Machine learning

Un - supervised Machine Learning

Problem Statement and Objective

Retailers struggle with one-size-fits-all marketing. Segmenting customers helps tailor promotions and improve retention.



- Identify distinct customer groups based on purchase behavior.
- Provide actionable business strategies for each segment.

Methodology

Step By Step Approach

1. Data Cleaning: Remove null transactions, aggregate purchases per customer.
2. Feature Engineering:
 1. RFM (Recency, Frequency, Monetary Value).
 2. Derived features (avg basket size, category diversity).
3. Clustering:
 1. K-Means, Gaussian Mixture, DBSCAN.
 2. Dimensionality reduction with PCA/UMAP for visualization.
4. Cluster Validation:
 1. Silhouette score, Davies–Bouldin index.
5. Business Insight:
 1. Map clusters → marketing personas (e.g., “loyal high spenders,” “deal-seekers”).

