Data Science Project Lifecycle & Tools

1. Problem Definition and Planning:

▶ Define the problem statement and project goals.













2. Data Acquisition

▶ Identify relevant data sources and gather the required data.

TOOLS













3. Data Cleaning and Preprocessing

▶ Clean the data by handling missing values, outliers, and inconsistencies.

TOOLS

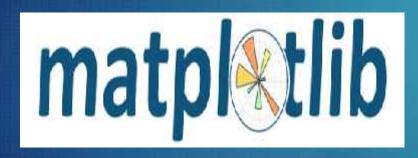




4. Exploratory Data Analysis (EDA)

- ▶ Understand the data through visualization and statistical analysis.
- ▶ Identify patterns, correlations, and outliers.

TOOLS









5. Model Selection and Development

- ➤ Select appropriate machine learning models based on the problem and data characteristics.
- ► Train and evaluate models using suitable metrics.

TOOLS







6. Model Tuning and Optimization

- ► Fine-tune model hyperparameters using techniques like grid search or random search.
- Optimize models for improved performance and generalization.

TOOLS







7. Model Evaluation and Validation

- ► Evaluate models using appropriate validation techniques such as cross-validation or train-test splits.
- ▶ Measure model performance using metrics specific to the problem.

TOOLS





8. Deployment and Integration

- ▶ Deploy the model into a production environment.
- ▶ Integrate the model with existing systems or applications.

TOOLS











9. Monitoring and Maintenance

- ► Continuously monitor model performance and data quality.
- ▶ Retrain or update models as needed.

TOOLS







10. Documentation and Communication

- Document the entire project, including data sources, methodology, and findings.
- ► Communicate results to stakeholders through reports or presentations.

TOOLS



















It's important to note that the specific tools used may vary depending on individual preferences, project requirements, and the data science ecosystem at the time. The list above provides a general overview of the stages and commonly used tools in a data science project life cycle.

THANK YOU