

## 1. Data Understanding and Preprocessing

- **Loading and Understanding Data**
    - Importing datasets (e.g., CSV, SQL, JSON) effectively.
    - Summarizing datasets: Checking the shape, columns, data types, missing values, and basic statistics.
    - Demonstrating domain understanding through exploratory data analysis (EDA).
  - **Handling Missing Data**
    - Identifying missing or incorrect data and applying appropriate imputation techniques (mean, median, mode, interpolation, etc.).
  - **Data Cleaning**
    - Removing duplicates, correcting data entry errors, and ensuring uniformity in categorical variables.
  - **Feature Encoding and Transformation**
    - Encoding categorical variables (e.g., one-hot encoding, label encoding).
    - Scaling and normalizing numerical features (e.g., MinMaxScaler, StandardScaler).
  - **Feature Engineering**
    - Creating new features from existing ones to improve model performance.
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## 2. Exploratory Data Analysis (EDA) and Visualization

- **Descriptive Statistics**
  - Summarizing data using measures like mean, median, variance, and correlation.
- **Visualizations**
  - Creating insightful plots to reveal trends, patterns, and anomalies:
    - Histograms, box plots, scatter plots, and pair plots.
    - Advanced plots: Heatmaps, time-series plots, or geospatial visualizations.

- Interactive visualizations using Plotly, Tableau, or Dash (optional).
  - **Insights**
    - Clearly communicating findings from EDA (e.g., trends, correlations, anomalies).
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### **3. Model Development and Implementation**

- **Model Selection**
    - Choosing the appropriate algorithm(s) based on the problem statement (e.g., regression, classification, clustering).
    - Justifying the choice of the model with reasoning.
  - **Implementation**
    - Writing clean, modular, and well-documented code for model training and evaluation.
    - Using frameworks such as Scikit-learn, TensorFlow, PyTorch, or XGBoost.
  - **Hyperparameter Tuning**
    - Demonstrating techniques like GridSearchCV, RandomizedSearchCV, or Bayesian Optimization for improving model performance.
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### **4. Model Evaluation and Validation**

- **Performance Metrics**
  - Properly calculating and interpreting relevant metrics, such as:
    - Regression: RMSE, MAE,  $R^2$ .
    - Classification: Accuracy, Precision, Recall, F1-score, ROC-AUC.
    - Clustering: Silhouette score, Davies-Bouldin index.
- **Cross-Validation**
  - Implementing K-Fold or other cross-validation techniques to ensure robustness.

- **Error Analysis**
    - Analyzing incorrect predictions and understanding areas of model weakness.
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## **5. Explainability and Interpretation**

- **Feature Importance**
    - Using techniques like SHAP, LIME, or built-in feature importance to explain model predictions.
  - **Model Transparency**
    - Providing insights into how the model arrives at predictions (especially for black-box models).
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## **6. Scalability and Deployment (Optional only for DA Student)**

For advanced students, additional marks can be awarded for implementing solutions that are scalable and deployable.

- **Deployment Readiness**
    - Building a simple web app using Flask, Django, or Streamlit for model predictions.
    - Creating APIs for the trained model.
  - **Scalability Features**
    - Writing efficient code for large datasets or parallelizing computations.
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## **7. Documentation and Presentation**

- **Code Documentation**
  - Writing meaningful comments and ensuring proper function and variable naming.
- **Report Submission**
  - Delivering a concise, well-structured report or presentation covering:
    - Problem understanding.

- Key steps taken (EDA, modeling, evaluation).
    - Final solution and recommendations.
  - **Presentation Skills**
    - Explaining the results and code logic effectively during a demo or Q&A session.
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## **8. Innovation and Creativity**

- **Novel Approaches**
  - Implementing innovative techniques or proposing unconventional solutions.
  - Effective use of advanced tools, libraries, or methods.
- **Real-World Applicability**
  - Connecting their solution to real-world use cases or suggesting practical implications.