

Standard Model Roadmap

DOMS

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① Define the Problem

- Problem Statement
- Objectives
- Scope
- Data Requirements
- Project Plan

② Data Collection

- Data sources
- Data acquisition
- Data Quality
- Privacy
- Augmentation (if required)
→ increase dataset artificially.

③ Data Preprocessing

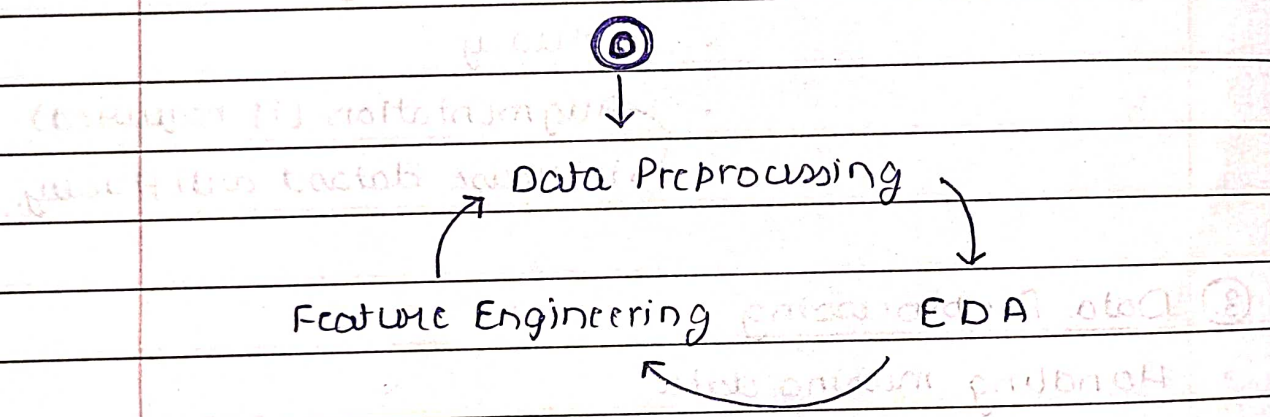
- Handling missing data
- Outliers and anomaly detection and Handling
- Cleaning
→ correcting typos, inconsistency entries.
- Data splitting into train, validate & test.
- Augmentation (if required.)

④ Exploratory Data Analysis [EDA]

- Summarization & visualization
- Correlation analysis
- Outlier & anomaly detection
- Feature interaction.
→ analyse features via different plots.
- Hypothesis Testing
- Report finding & Making

⑤. Feature Engineering

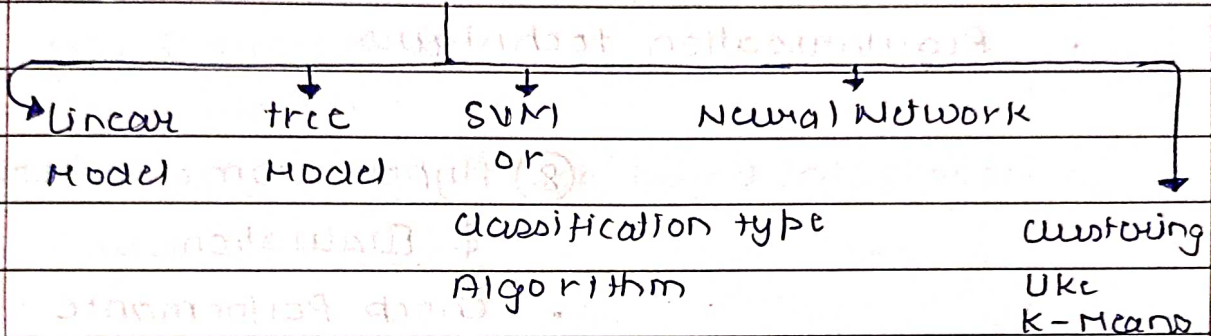
- New Feature creation
- Transformation
→ log transformation, square, square root, scale etc
- Encoding categorical variables
- Text Data Processing like TF-IDF, word2vec. etc
- Scaling.
- Normalisation & standardisation
- Dimensionality Reduction.
- ✓ • Feature selection.



★ All these comes in a loop and any no. of cycles can be performed.

⑥. Model Selection.

- Problem type identification.
→ Classification, regression, clustering, recommend etc.
- Model selection criteria
→ identify speed, scalability performance measurement
- Algorithm choice



- Choose between ML & DL
- Ensemble Methods (if applicable)
- Create a custom Model (if applicable)
- Hyperparameter tuning & Regularisation.
- Cross Validation
- Model Evaluation Metrics
- Bench Marking.

★ It is very vast, above mentioned are very generalised and therefore experience via projects.

⑦. DL Architecture

- choose Loss function
- optimization algorithm
- Hyperparameter tuning
- Epochs and Batches
- Early stopping & checkpointing
- Regularisation techniques

⑧. Hyperparameter tuning & Evaluation.

- check Performance metrics.
- Model comparison
- Model Robustness
- Visualisation of Evaluation Metrics
- Metrics Comparison

⑨. Deployment

- Environment Setup
→ cloud, infrastructure or hardware to host model.
- Model Serialisation.
→ save trained model in pickle or onnx format
- API Development
if applicable
Develop a API or web based integration of model to web.
- Model Containerisation. using docker, Kubernetes
- Integration with Real time data sources
- Model versioning via implementing version control system. for updates and roll backs.
- Monitoring and logging
- Model Inference for new and unseen data
- Feed back loop (if applicable)
→ mechanism to continuously improve model by collecting prediction.
- Performance optimization.
- Load Balancing (if applicable)
- A/B Testing (if applicable)
- Rollout, Backup, Recovery strategy
- Compliance and Governance.

any number of times depending upon Results and Objectives.

