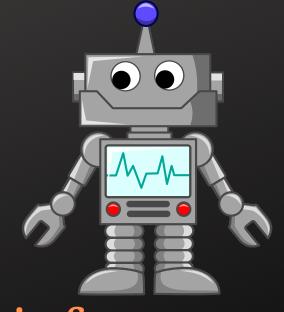
MACHINE LEARNING ENGINEERING

ML: The ability to Learn, Predict & Find Ways to Improve the Performance

Types of Machine Learning

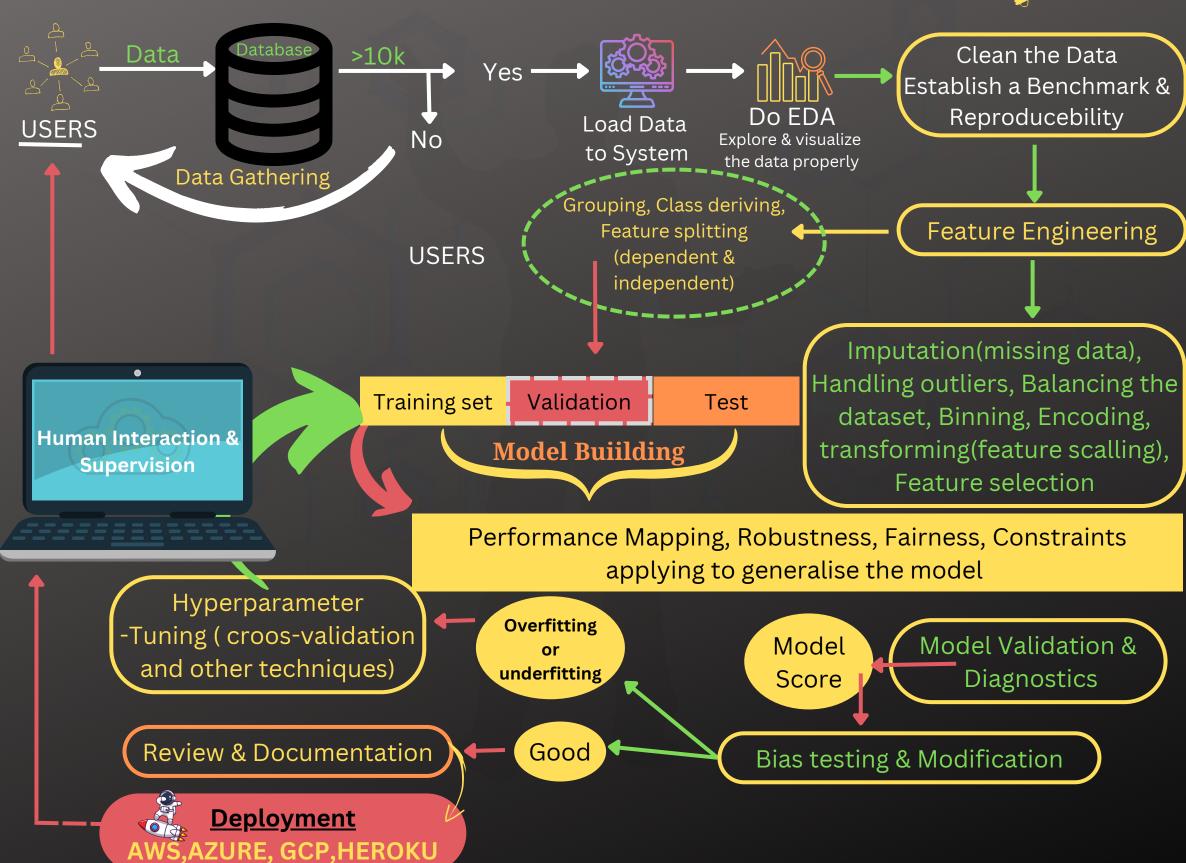






Unsupervised Reinforecement

MACHINE LEARNING WORK FLOW



BEST PRACTICES TO MAKE A MODEL

STAGE 1- ALL ABOUT DATA

| Process | Challenges to Overcome | Best Practice |
|--|---|--|
| Data collection (web scraping, from Database, User input & more) | Uncleaned Data(need to clean) Finding relevant data Curse of Dimensionality Baised Data Incomplete data, Sparsity Data Quality issue & many more | Enrich the Data Using data-related Key Performance Indicators (KPIs) to understand the Data Dimension-reduction techniques Data Modification Data representation |
| Untidy Data | Matching proper Rows and columnMultiple variables in same column | Restructure the data to be fit & tidy by using cast techniques |
| Missing Data | Information LossBias | Central tendency Imputation Tree-Based Modeling techniques Apply best missing technique accoring to dataset(more than 20+ ways there) |
| Outliers | Undue influence on Squared loss function Unknown categorical data | Winsorizing Capping Robust techniques and 10+ effective techniques availabe |
| EDA | Unstable parameter estimationNon-determinism on colinearlty | Use Profiling process Use statical approcah and graphical representation |

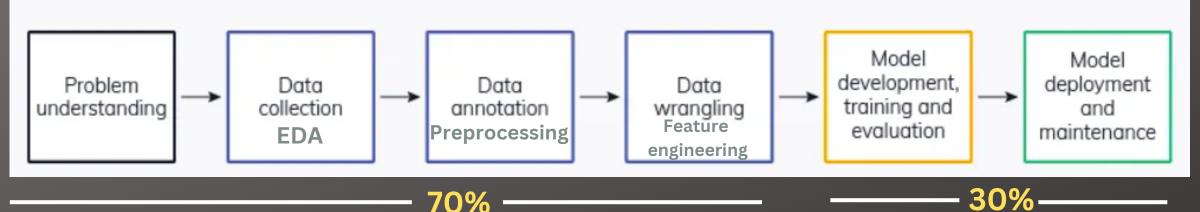
BEST PRACTICES TO MAKE A MODEL

STAGE 2- TRAING & TESTING (MODEL BUILDING

| Process | Challenges to Overcome | Best Practice |
|---|---|---|
| Overfitting & Underfitting | High variance and Low Bias (Not a Generalize model) | Regularization Noise reduction Cross validation Removing features Ensembling & many more techniques |
| Hyperparameter Tuning | Combinatorial explosion of hyperparameters in convetional algorithms | Parameter estimationGridsearchCvRandomsearchcv & many more |
| Ensemble models | Single models that fail to provide adequate High-variance and low-bias models that fail to generalize well | Powerful ensemble model like (bagging & boosting) Custom combinations of predictions |
| Model Interpretation & Validation | Imperfections in the Algorithm When Data Grows Nonrepresentative training data. | Variable selection by regularization (e.g. L1) Surrogate Models Partial dependency plots & variable dependency measures |
| Model deployment & Decay | Trained modellogic must be transferred from a development environment to an operational computing system to assit | Database, web scoring |

decission making process

MACHINE LEARNING LIFE CYCLE



On an Average a Data scientist spends Time

