

Introduction to AI, ML and Deep Learning



PG Level AP in AI&MLOps Cohort 3

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Let's meet a Virtual Assistant



- "Hey Siri. Show me a good breakfast restaurant near me."
- How does Siri process this query?
 - 1. Convert speech to text
 - 2. Understand the semantics of the question (e.g., understand keywords like breakfast, restaurant) and formulate a structured query (place type = "restaurant", good for="breakfast", rating 3-5, distance < 3 km)
 - It also needs your current location!
 - 3. Search for restaurants, filter by the structure above and rank based on ratings (or other metrics)
 - For rating, the system could use the star rating and the sentiments/points in the written review – another NLP task
 - 4. At the restaurant, it might translate the menu card from Kannada to English



Al System for Almond Processing Plant







The Learning Process



1

- AI/ML Fundamentals
- Ops Fundamentals
 - Programming

3

- Learning Tools
 - Using Tools



Image Courtesy: 123rf

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2

- ML Algorithms
- DL Algorithms
- Cloud MLOps
- MLOps at Scale

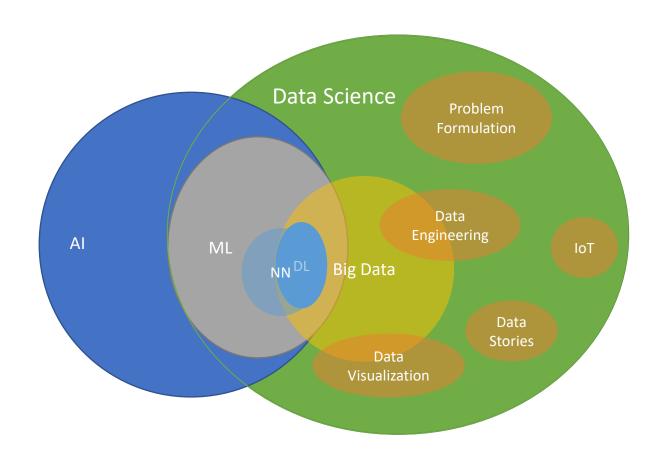
4

- Projects
- Portfolio



Data Science: ML/AI/DL – What is it?





- Data Science is an umbrella term
- It is the full building that we showed
- It has foundation, pillars, floors, walls, interiors, maintenance
- One can focus on a part of the building and develop deep expertise
- But should know the breadth as well

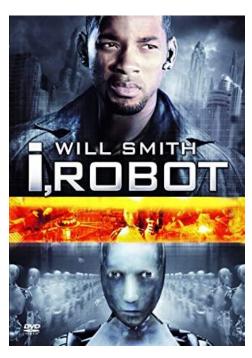


Artificial Intelligence



- AI: The ability of a digital computer or computer-controlled robot to perform tasks commonly associated with intelligent beings.
- Can be data-driven or model-driven (rule-based)
- Artificial General Intelligence is the ultimate goal in AI research







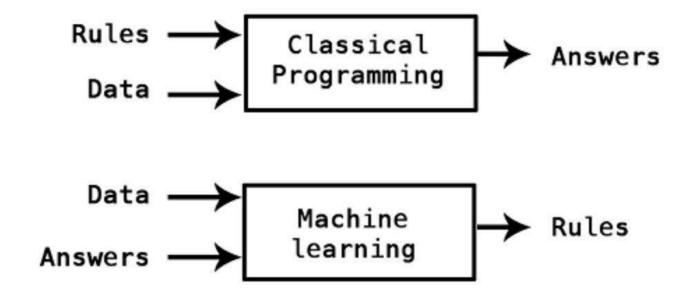


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Classical Programming vs ML







The AI/ML Workflow



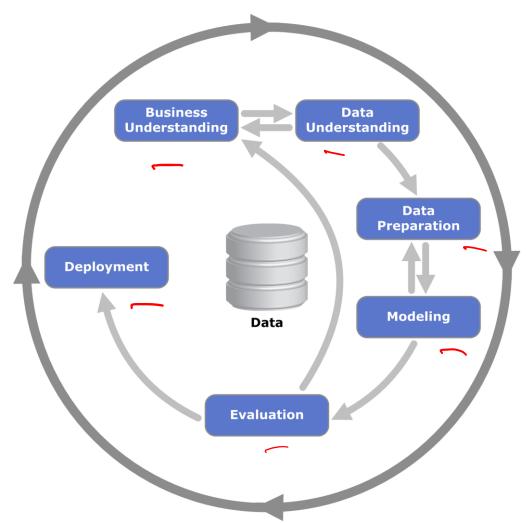
- 1. Frame the AI problem by looking at the business need
 - a. Identify subproblems (One/more of the 5 tasks a computer can do)
 - b. Establish a current baseline (What is currently done?)
 - c. Define success
- 2. Gather the data and do Data Munging/Wrangling + Baselines
 - a. Explore the data
 - b. Clean data and prepare for the downstream ML models
 - c. Establish a data, domain and SoTA baseline
- 3. Explore different models, improve them through Cross Validation and perhaps new model design
- 4. Form an ensemble of multiple models and solutions
- 5. Present your solution
 - a. Say a story with the data
- 6. Deploy



CRISP-DM



- Cross Industry Standard Process for Data Mining
- Initiative in the mid 90s by European Strategic Programme on Research in Information Technology (ESPRIT)
- The key ideas are in our 6step process as well





Types of Data



- Tabular Data
 - Most common form
 - Arises in almost all business use cases
 - Usually number of data points x features
- Timeseries Data
 - Tabular but at different times (a logical ordering in time)
- Image Data
 - Increasing in recent years
 - Usually number of data points x height x width x sensor channels
 - Time series of image data is video data
 - Vision Tasks
- Text Data
 - Language tasks
 - Usually text corpus Needs to be converted to number How?
- Speech Data
 - Language tasks
 - Usually recording corpus Signal Processing



Continuous vs Categorical Data



- Continuous Data mm of rainfall tomorrow
- Categorical data Will it rain or no?
- How to reason about categories?
- We will use the language of probability and statistics to answer these questions



ML: Mental Model



Data that can be collected

$$X \longrightarrow \text{Real World} \longrightarrow y$$

Quantity that must be predicted to make money

Data that can be collected

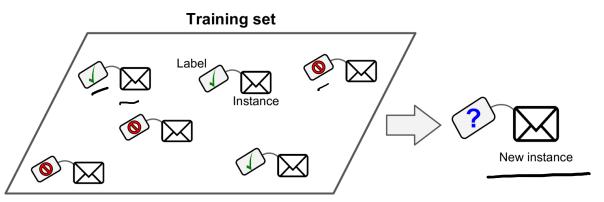
$$X \longrightarrow h_{\theta}(X;\theta) \longrightarrow \hat{y}$$

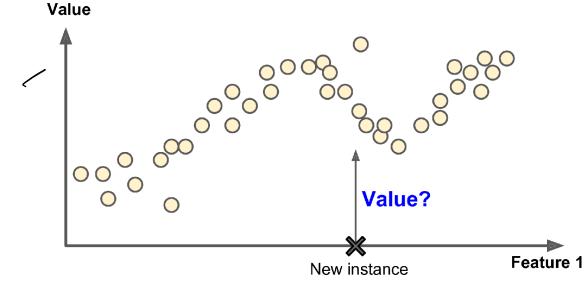
Machine's Prediction

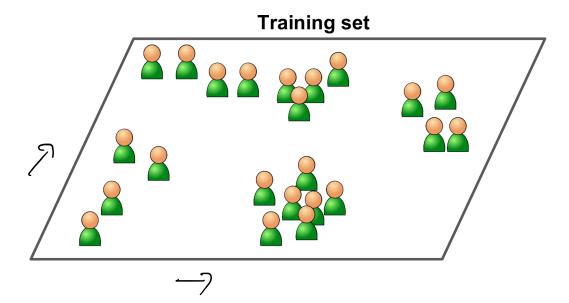


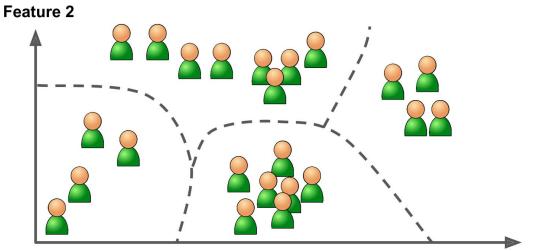
Tasks in ML/DS/AI: Visual Introduction







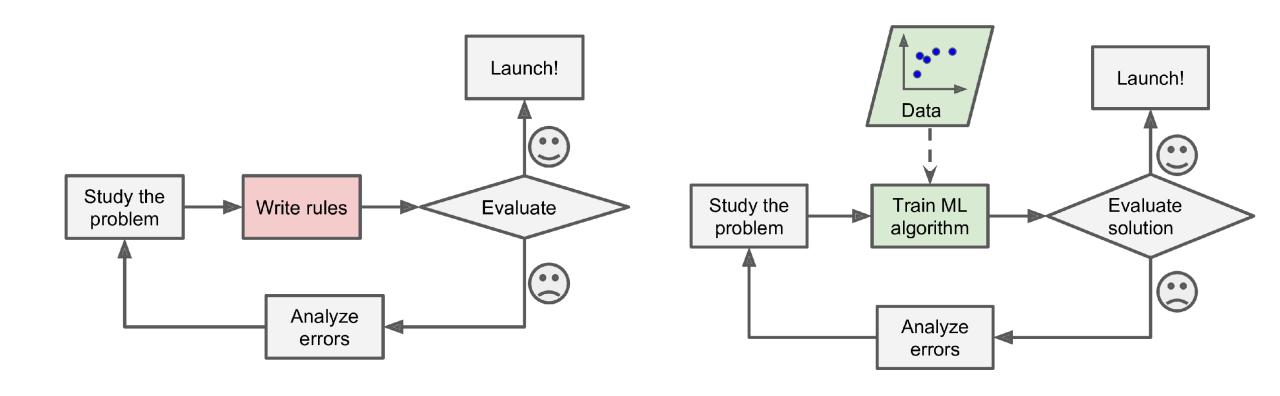






Traditional Approach vs ML

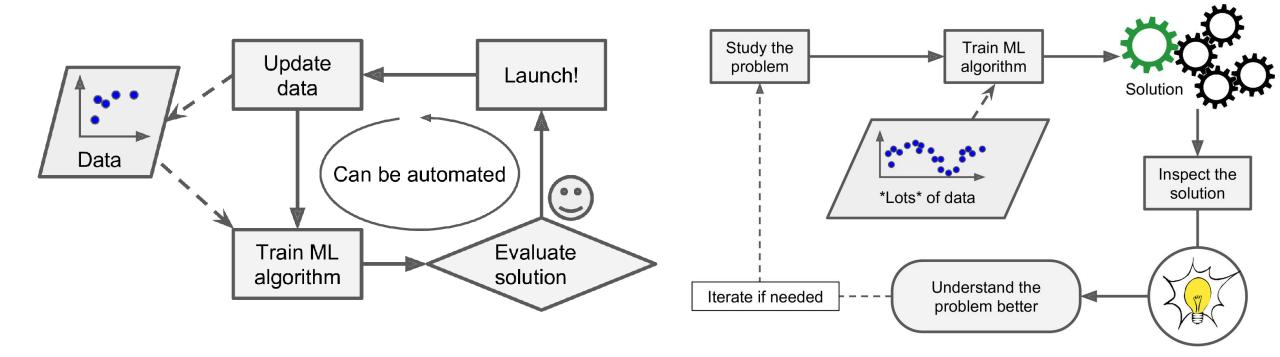






Uses of ML

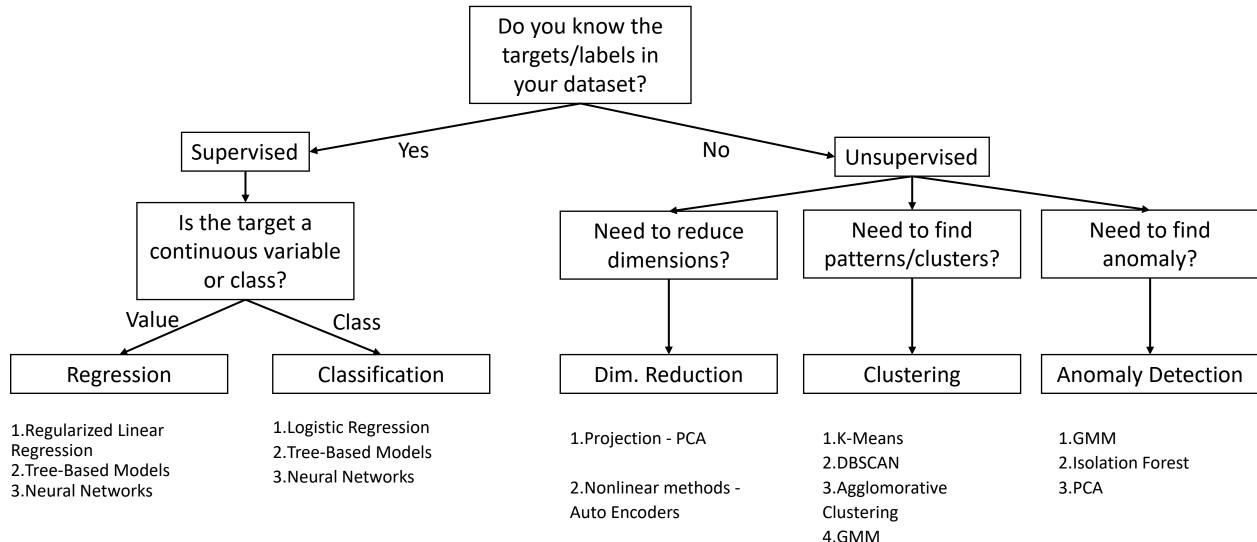






Summary of ML





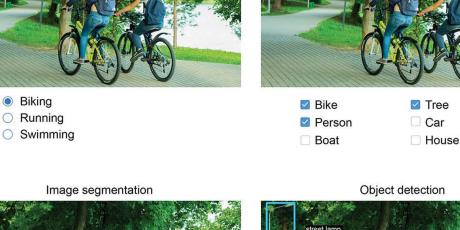


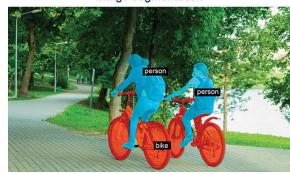
Three Essential Tasks in Computer Vision



- Image Classification
 - Single Label
 - Binary
 - Multiclass
 - Multi Label
- Image Segmentation
 - Pixel wise identify the class
 - Example: Zoom background replacement
- Object Detection
 - Bounding box around objects
 - Self-driving cars, face detection in cameras









Multi-label classification



NLP: Major Tasks



- Modern NLP Goal is not to understand language, but to ingest a piece of language as input and return useful quantities
- A collection of fundamental tasks repeatedly come in NLP
- Natural Language Understanding
 - "What is the topic of this text?" Topic Modelling
 - "Is this text inappropriate?" Content Filtering
 - "Is this text, positive, neutral or negative?" Sentiment Analysis
 - Named Entity Recognition, Part of Speech Tagging
 - Information retrieval (Keyword based)
- Natural Language Generation
 - "What is the next word or character?" Language Modeling, Sentence Completion
 - "What is "AI" in tamil?" Machine Translation
 - "What is the crux of this paragraph?" Text Summarization
 - Answer to "Where is the nearest hair salon?" Question Answering



Generative Al



- Text to Text
- Text to Image/Video
- Image/Video to Text
- Image/Video to Image/Video
- Text to Speech
- Speech to Text

• Input is the "Prompt"; Model is a Large Language/Vision Model; Output is Image/Video/Text/Speech



Explain your work to stakeholders and set expectations



- Success and customer trust are about consistently meeting or exceeding expectations
- The actual model is only half the picture; the level of expectation about system performance matters a lot
- Non-specialists expect AI to punch above its weight
 - They expect the system to "understand" and meet or exceed capability of a human doing the task
- Clearly setting the expectation is important
- Some guidelines
 - Don't talk in easily mis-understood terminology Accuracy is 98%
 - Show examples of what misclassification looks like
 - Understand if customer cares about False Positive or False Negative more
 - Discuss key parameters the probability above which a fraud has to be detected
 - Explain how many cases on average we expect the system to be falsely labelled as positive [False Positive, False Negative, Explain in simple language]



Sources of ML Data



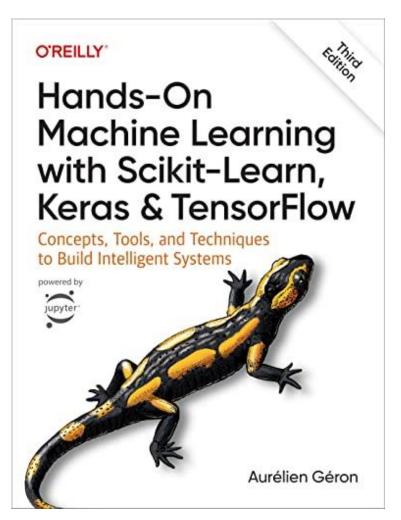
- There are several open data repositories for learning ML/DL
- UCI Repository https://archive.ics.uci.edu/ml/index.php
- Google Dataset Search https://datasetsearch.research.google.com/

 For our illustration and as a first case study, we will use the California Housing Prices dataset from Geron Textbook Chapter 2



Text Book 1



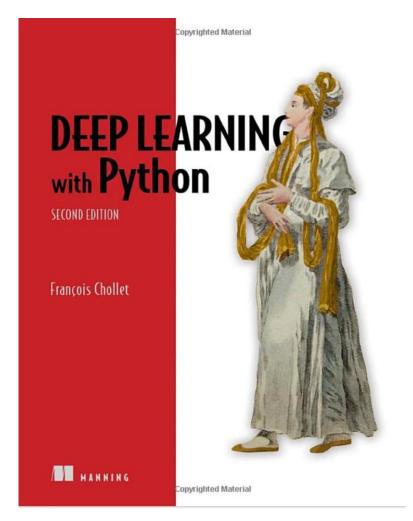


https://www.amazon.in/Hands-Machine-Learning-Scikit-Learn-TensorFlow-ebook/dp/B0BHCFNY9Q/



Text Book 2





https://www.amazon.in/Learning-Python-Second-Fran%C3%A7ois-Chollet-ebook/dp/B09K81XLN1/





