

CSE 445 / L-1 / 21.05.2025 /

Absent

Attendance - 51.

Assignment - 201.

→ 3 out of 4

i. ML Project base Code

ii. Project Proposal

iii. VIVA on Project progress

iv. Video demonstration of project.

Midterm - 251.

Final - 251.

Project - 251.

L-2 / 26.05.2025 /

⊗ Mean :

— Arithmetic mean:

⇒ direct average of a data set, is found by adding all numbers in the data set and then dividing by the number of values in the set.



⇒ Harmonic Mean:

- defined as the reciprocal of the average of the reciprocals of the data values.

⊕ Median:

- is the middle value when a data set is ordered from least to greatest.

⊕ Mode:

- is the number that occurs most often in a data set.

⊕ Conditional Probability:

- is a measure of the probability of an event occurring, given that another event  $E$  has already occurred.

⊕ Tom Mitchell:

- a computer program is said to learn from experience  $E$  with respect to some class of tasks  $T$  and performance measure  $P$ , if its performance at task in  $T$ , as measured by  $P$ , improves with experience  $E$ .

⊕ Arthur Samuel:

- Machine learning is the field of study that gives the computer the ability to learn without being explicitly programmed.

\* The difference between traditional programming and the ML approach: ~~are~~

Slide - 9, 10

\* ML Approach:

- allow fine tuning and make long list of rules.
- can adapt in changing environment.
- provide insight from large amount of data.
- solve complex problem.

\* Deep learning:

- subset of ML
- automatic features extraction
- function like human brain.
- more data, better prediction.

+

$AI > ML > DL$

\* Narrow AI:

- train for specific task.

\* Parametric knowledge:

- learn from data.



L-3 / 28.05.2025 /

### ⊗ Supervised Learning:

- trained with labeled data.

⇒ - Linear Regression

- Logistic Regression

- SVM

- Decision Tree

- NN

### ⊗ Unsupervised Learning:

- data has no label.

- model try to make group based on familiar features.

⇒ Used in

- clustering

- anomaly detection

- association mining

- data preprocessing

⇒ - k-means

- SVD

- PCA

- ICA

### ⊗ Semisupervised Learning:

- partially labeled data

- model categorise the data based on familiar feature and then asked for the label to human.

### ⊗ Reinforcement Learning:

- first provide a output, based on the feedback of the output, it try to re-learn its state for the data

## ⊗ Batch Learning:

- learning is not possible after deployment.
- for new data, we need to train from scratch.

## ⊗ Online Learning:

- continue learning after deployment as new data comes.
- use parallel computing, no downtime.

## ⊗ Instance Based Learning:

- memorize known data and try to match with these.
- KNN

## ⊗ Model - Based Learning:

- more generalize, it divide the data area and build a model, then predict the output.

## ⊗ 3 V's of Big Data:

- Volume: amount of data
- Variety: data for each possibility.
- Velocity: how the data change over time

## ⊗ Overfitting:

- very good on training data
- very bad on test data.
- model need to be simple respect to the data.
- need to tune the hyperparameters to control the regularization.



## \* Data Split!

- Training
- Validation
- Test

⇒ Good model: 60-20-20

⇒ an okay model: 70-15-15

⇒ barely acceptable model: 80-10-10

allowed only when we have millions of instances in the dataset.

L-4 / 02.06.2025 /

## \* Problem with the housing data:

- price changes over time.
- influencing features also change

Might be a good research area.

## \* Classification problem:

- predict the class of the data

## \* Regression problem:

- predict continuous variable.

⇒ one value ⇒ univariate regression problem  
multiple value ⇒ multivariate regression problem.

## \* Performance Measurement:

- RMSE: Root mean squared error

$$= \sqrt{\frac{1}{m} \sum_{i=1}^m (h(x^{(i)}) - y^{(i)})^2}$$

- MAE: Mean absolute error

$$= \frac{1}{m} \sum_{i=1}^m |h(x^{(i)}) - y^{(i)}|$$

⇒

RMSE:

- straight line distance
- Euclidean norm or L2 norm

\* MAE:

- city block distance
- Manhattan norm or L1 norm

} differences available  
on google

⇒ Describe Code from colab:

## \* Correlation between attributes:

⇒ 1:

- if  $x$  goes up,  $y$  goes up too
- both are in same direction

⇒ -1:

- ~~both~~  $x$  and  $y$  in opposite direction.

⇒  $\approx 0$ :

- no relation between them.



L-5/04.06.2025/

## \* Explained Code from colab.

### \* Steps:

- (i) Download the data
- (ii) Quick look at Data structure
  - plot histogram
- (iii) Data visualization:
  - scatter plot
  - bar plot
- (iv) Preprocess the data
  - remove outlier
  - remove or manage ~~the~~ NaN value
  - Transform the data
    - use encoder
    - scaling
- (v) Split the data : stratified sampling
- (vi) Select model and train.
  - validate using validation set
- (vii) Test the model:
  - using test set.



L-06/16.06.2025/

⑦ What is one-hot encoder?

⑧ From similar correlation, we need to take only one, as the effect is same.

⑨ During encode, we need to maintain the order.

⑩ Remove outlier before apply normalization.

- or use standardization.

⑪ What the fit and fit-transform do in the custom transformer method?

L-07/18.06.2025/

⑫ Confusion Matrix:

TP  $\Rightarrow$  True Positive  $\Rightarrow$  Actual true & Predicted True.

TN  $\Rightarrow$  True Negative  $\Rightarrow$  Actual Negative & Predicted Negative

FP  $\Rightarrow$  False Positive  $\Rightarrow$  Actual Negative & Predicted Positive

FN  $\Rightarrow$  False Negative  $\Rightarrow$  Actual Positive & Predicted Negative

$$\text{Accuracy} = \frac{TP + TN}{TP + TN + FP + FN}$$

Precision =  $\frac{TP}{TP+FP}$  ; How precisely this model can identify the positive class.

$$\text{Recall} = \frac{TP}{TP+FN}$$

$$F1 \text{ Score} = \frac{2}{\frac{1}{\text{Precision}} + \frac{1}{\text{Recall}}} = \frac{2 \times \text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

~~$$\text{Precision} \times \frac{1}{\text{Recall}}$$~~