

# Fundamentals of Machine Learning (4341603) - Winter 2023 Solution

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## Question 1(a) [3 marks]

Define human learning and explain how machine learning is different from human learning?

### Solution

**Human Learning** is the process of acquiring knowledge through experience, observation, and reasoning.

**Table 1.** Human Learning vs Machine Learning

Aspect	Human Learning	Machine Learning
Method	Experience, trial and error	Data and algorithms
Speed	Slow, gradual	Fast processing
Data Requirement	Limited examples needed	Large datasets required

**Machine Learning** is the automated learning from data using algorithms to identify patterns without explicit programming.

### Mnemonic

“Humans Experience, Machines Analyze Data (HEMAD)”

## Question 1(b) [4 marks]

Describe the use of machine learning in finance and banking.

### Solution

**Table 2.** Applications in Finance and Banking

Application	Purpose	Benefit
Fraud Detection	Identify suspicious transactions	Reduce financial losses
Credit Scoring	Assess loan default risk	Better lending decisions
Algorithmic Trading	Automated trading decisions	Faster market responses

**Risk Assessment:** ML analyzes customer data to predict creditworthiness. **Customer Service:** Chatbots provide 24/7 support using NLP. **Regulatory Compliance:** Automated monitoring for suspicious activities.

### Mnemonic

“Finance Needs Smart Analysis (FNSA)”

## Question 1(c) [7 marks]

Give difference between Supervised Learning, Unsupervised Learning and Reinforcement Learning.

### Solution

**Table 3.** Comparison

Feature	Supervised	Unsupervised	Reinforcement
<b>Data Type</b>	Labeled data	Unlabeled data	Environment interaction
<b>Goal</b>	Predict output	Find patterns	Maximize rewards
<b>Examples</b>	Classification	Clustering	Game playing
<b>Feedback</b>	Immediate	None	Delayed rewards

**Supervised Learning:** Teacher-guided learning with correct answers provided.

**Unsupervised Learning:** Self-discovery of hidden patterns in data.

**Reinforcement Learning:** Learning through trial and error with rewards/penalties.

### Mnemonic

“Supervised Teachers, Unsupervised Explores, Reinforcement Rewards (STUER)”

## Question 1(c) OR [7 marks]

Explain different tools and technology used in machine learning.

### Solution

**Table 4.** ML Tools and Technologies

Category	Tools	Purpose
<b>Programming</b>	Python, R, Java	Algorithm implementation
<b>Libraries</b>	Scikit-learn, TensorFlow	Ready-made algorithms
<b>Visualization</b>	Matplotlib, Seaborn	Data visualization
<b>Data Processing</b>	Pandas, NumPy	Data manipulation

### Key Technologies:

- Cloud Platforms:** AWS, Google Cloud for scalable computing
- Development Environments:** Jupyter Notebook, Google Colab
- Big Data Tools:** Spark, Hadoop for large datasets

### Mnemonic

“Python Libraries Visualize Data Effectively (PLVDE)”

## Question 2(a) [3 marks]

Define outliers with one example.

### Solution

**Definition:** Outliers are data points that significantly differ from other observations in a dataset.

**Table 5.** Example: Student Heights

Student Heights (cm)	Classification
165, 170, 168, 172	Normal values
195	Outlier (too tall)
140	Outlier (too short)

**Detection:** Values beyond  $1.5 \times \text{IQR}$  from quartiles. **Impact:** Can skew statistical analysis and model performance.

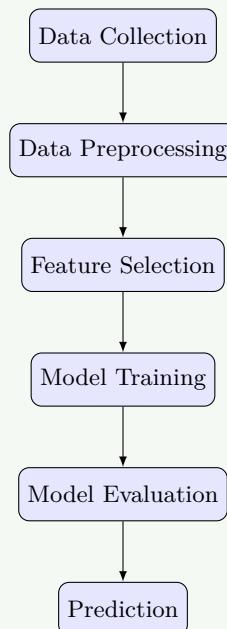
### Mnemonic

“Outliers Stand Apart (OSA)”

## Question 2(b) [4 marks]

Explain regression steps in detail.

### Solution



**Figure 1.** Regression Process Steps

### Detailed Steps:

- **Data Collection:** Gather relevant dataset with input-output pairs
- **Preprocessing:** Clean data, handle missing values, normalize features
- **Feature Selection:** Choose relevant variables that affect target
- **Model Training:** Fit regression line to minimize prediction errors

### Mnemonic

“Data Preprocessing Features Train Evaluation Predicts (DPFTEP)”

## Question 2(c) [7 marks]

Define Accuracy and for the following binary classifier's confusion matrix, find the various measurement parameters like 1. Accuracy 2. Precision.

### Solution

#### Confusion Matrix Analysis:

**Table 6.** Given Confusion Matrix

	Predicted No	Predicted Yes
Actual No	10 (TN)	3 (FP)
Actual Yes	2 (FN)	15 (TP)

#### Calculations:

- **Accuracy** formula:  $(TP + TN)/(TP + TN + FP + FN)$
- Calculation:  $(15 + 10)/(15 + 10 + 3 + 2) = 25/30 = 0.8333$
- **Accuracy Result: 83.33%**
- **Precision** formula:  $TP/(TP + FP)$
- Calculation:  $15/(15 + 3) = 15/18 = 0.8333$
- **Precision Result: 83.33%**

#### Definitions:

- **Accuracy:** Proportion of correct predictions out of total predictions.
- **Precision:** Proportion of true positive predictions out of all positive predictions.

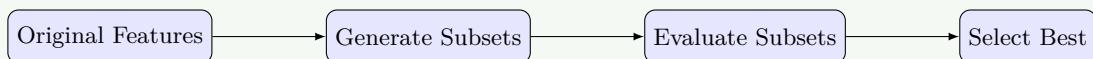
### Mnemonic

“Accuracy Counts All, Precision Picks Positives (ACAPP)”

## Question 2(a) OR [3 marks]

Identify basic steps of feature subset selection.

### Solution



**Figure 2.** Feature Subset Selection Process

#### Basic Steps:

1. **Generation:** Create different combinations of features
2. **Evaluation:** Test each subset using performance metrics
3. **Selection:** Choose optimal subset based on criteria

### Mnemonic

“Generate, Evaluate, Select (GES)”

## Question 2(b) OR [4 marks]

Discuss the strength and weakness of the KNN algorithm.

**Solution****Table 7.** KNN Algorithm Analysis

<b>Strengths</b>	<b>Weaknesses</b>
Simple to understand	Computationally expensive
No training required (Lazy)	Sensitive to irrelevant features
Works with non-linear data	Performance degrades with high dimensions
Effective for small datasets	Requires optimal K value selection

**Key Points:**

- **Lazy Learning:** No explicit training phase required.
- **Distance-Based:** Classification based on neighbor proximity.

**Mnemonic**

“Simple but Slow, Effective but Expensive (SBSEBE)”

**Question 2(c) OR [7 marks]**

Define Error-rate and for the following binary classifier's confusion matrix, find the various measurement parameters like 1. Error value 2. Recall.

**Solution****Confusion Matrix Analysis:****Table 8.** Given Confusion Matrix

	<b>Predicted No</b>	<b>Predicted Yes</b>
<b>Actual No</b>	20 (TN)	3 (FP)
<b>Actual Yes</b>	2 (FN)	15 (TP)

**Calculations:**

- **Error Rate formula:**  $(FP + FN)/(Total)$
- Calculation:  $(3 + 2)/(15 + 20 + 3 + 2) = 5/40 = 0.125$
- **Error Rate Result:** 12.5%
- **Recall formula:**  $TP/(TP + FN)$
- Calculation:  $15/(15 + 2) = 15/17 = 0.8824$
- **Recall Result:** 88.24%

**Definitions:**

- **Error Rate:** Proportion of incorrect predictions out of total predictions.
- **Recall:** Proportion of actual positives correctly identified.

**Mnemonic**

“Error Excludes, Recall Retrieves (EERR)”

**Question 3(a) [3 marks]**

Give any three examples of unsupervised learning.

**Solution****Table 9.** Unsupervised Learning Examples

Example	Description	Application
<b>Customer Segmentation</b>	Group customers by behavior	Marketing strategies
<b>Document Classification</b>	Organize documents by topics	Information retrieval
<b>Gene Sequencing</b>	Group similar DNA patterns	Medical research

**Additional Examples:**

- **Market Basket Analysis:** Finding product purchase patterns
- **Anomaly Detection:** Detecting unusual patterns in data

**Mnemonic**

“Customers, Documents, Genes Group Automatically (CDGGA)”

**Question 3(b) [4 marks]**

Find Mean and Median for the following data: 4,6,7,8,9,12,14,15,20

**Solution**

**Data:** 4, 6, 7, 8, 9, 12, 14, 15, 20 (Already sorted)

**Mean Calculation:**

- Sum =  $4 + 6 + 7 + 8 + 9 + 12 + 14 + 15 + 20 = 95$
- Count = 9
- Mean =  $95/9 = 10.56$

**Median Calculation:**

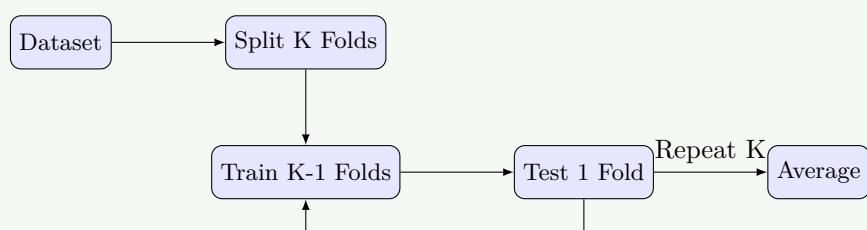
- N = 9 (Odd number)
- Position =  $(N + 1)/2 = 5$ th value
- 5th value in sorted list is 9
- Median = 9

**Mnemonic**

“Mean Averages All, Median Middle Value (MAAMV)”

**Question 3(c) [7 marks]**

Describe k-fold cross validation method in detail.

**Solution****Figure 3.** K-Fold Cross Validation**Process Steps:**

1. **Data Division:** Split data into K equal parts (folds).
2. **Iterative Training:** Use K-1 folds for training the model.
3. **Validation:** Test the model on the remaining fold.
4. **Averaging:** Repeat K times and average the performance metrics.

**Advantages:**

- **Unbiased Estimation:** Each data point used for both training and testing.
- **Reduced Overfitting:** Multiple validation rounds increase reliability.

#### Mnemonic

“K-fold Keeps Keen Knowledge (KKKK)”

## Question 3(a) OR [3 marks]

Give any three applications of multiple linear regression.

#### Solution

Table 10. Multiple Linear Regression Applications

Application	Variables	Purpose
House Price Prediction	Size, location, age	Real estate valuation
Sales Forecasting	Marketing spend, season	Business planning
Medical Diagnosis	Symptoms, age, history	Disease prediction

#### Mnemonic

“Houses, Sales, Medicine Predict Multiple Variables (HSMPV)”

## Question 3(b) OR [4 marks]

Find Standard Deviation for the following data: 4,15,20,28,35,45

#### Solution

**Data:** 4, 15, 20, 28, 35, 45 ( $N=6$ )

**Step 1: Calculate Mean**

- Sum =  $4 + 15 + 20 + 28 + 35 + 45 = 147$
- Mean ( $\bar{x}$ ) =  $147/6 = 24.5$

**Step 2: Calculate Squared Deviations**

- $(4 - 24.5)^2 = (-20.5)^2 = 420.25$
- $(15 - 24.5)^2 = (-9.5)^2 = 90.25$
- $(20 - 24.5)^2 = (-4.5)^2 = 20.25$
- $(28 - 24.5)^2 = (3.5)^2 = 12.25$
- $(35 - 24.5)^2 = (10.5)^2 = 110.25$
- $(45 - 24.5)^2 = (20.5)^2 = 420.25$

**Step 3: Calculate Variance and Std Dev**

- Sum of squared deviations = 1073.5
- Variance ( $\sigma^2$ ) =  $1073.5/6 = 178.92$
- Standard Deviation ( $\sigma$ ) =  $\sqrt{178.92} = 13.376$

**Mnemonic**

“Deviation Measures Data Spread (DMDS)”

**Question 3(c) OR [7 marks]**

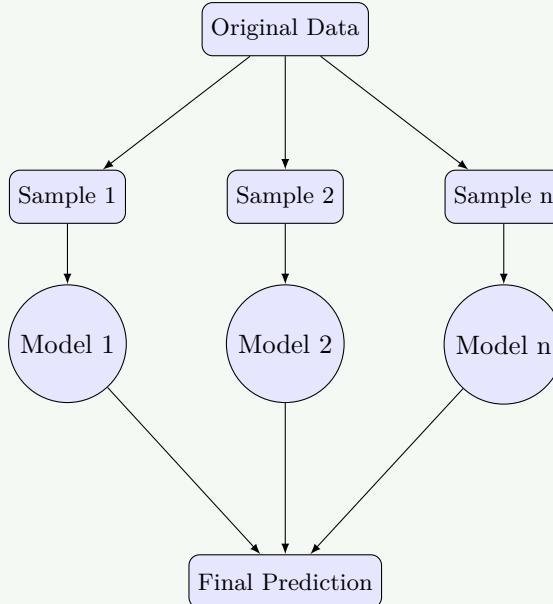
Explain Bagging, Boosting in detail.

**Solution**

**Table 11.** Bagging vs Boosting

Aspect	Bagging	Boosting
<b>Strategy</b>	Parallel training	Sequential training
<b>Data Sampling</b>	Random with replacement	Weighted sampling
<b>Goal</b>	Reduces variance	Reduces bias

**Bagging (Bootstrap Aggregating):** Trains multiple independent models in parallel using random subsets of data and averages their predictions.



**Figure 4.** Bagging Process

**Boosting:** Trains models sequentially, where each new model focuses on the errors made by previous models.

**Mnemonic**

“Bagging Builds Parallel, Boosting Builds Sequential (BBPBS)”

**Question 4(a) [3 marks]**

**Define:** Support, Confidence.

**Solution****Table 12.** Association Rule Metrics

Metric	Definition and Formula
<b>Support</b>	Frequency of itemset in transactions. $Support(A) = Count(A)/Total$
<b>Confidence</b>	Conditional probability of rule. $Confidence(A \rightarrow B) = Support(A \cup B)/Support(A)$

**Example:**

- **Support:** 60% of transactions have Bread.
- **Confidence:** 80% of people buying Bread also buy Butter.

**Mnemonic**

“Support Shows Frequency, Confidence Shows Connection (SSFC)”

**Question 4(b) [4 marks]**

Illustrate any two applications of logistic regression.

**Solution****Table 13.** Logistic Regression Applications

Application	Description	Outcome
Email Spam	Detect spam based on words	Spam/Not Spam
Medical Diagnosis	Predict disease from symptoms	Disease/Healthy
Credit Approval	Assess loan risk	Approve/Reject

**Key Features:**

- **Binary Classification:** Predicts probability (0 to 1).
- **Sigmoid Function:** Maps output to S-shaped curve.

**Mnemonic**

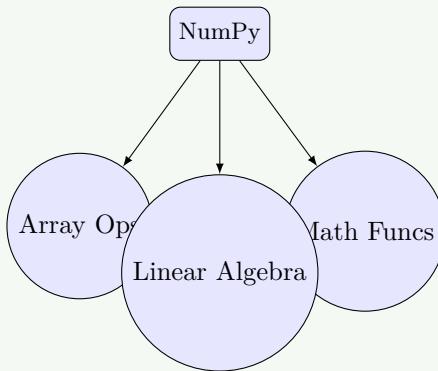
“Logistic Limits Linear Logic (LLLL)”

**Question 4(c) [7 marks]**

Discuss the main purpose of Numpy and Pandas in machine learning.

**Solution**

**NumPy** provides support for large, multi-dimensional arrays and matrices, along with mathematical functions.  
**Pandas** offers data structures and operations for manipulating numerical tables and time series.

**Figure 5.** NumPy Features**Table 14.** Comparison

Library	Primary Purpose	Key Features
NumPy	Numerical Computing	N-dim arrays, broadcasting
Pandas	Data Manipulation	DataFrames, cleaning, analysis

**Mnemonic**

“NumPy Numbers, Pandas Processes Data (NNPD)”

**Question 4(a) OR [3 marks]**

Give any three examples of Supervised Learning.

**Solution****Table 15.** Supervised Learning Examples

Example	Type	Input → Output
Email Classification	Classification	Email features → Spam/Not
House Price Prediction	Regression	House features → Price
Image Recognition	Classification	Pixels → Object Class

**Mnemonic**

“Emails, Houses, Images Learn Supervised (EHILS)”

**Question 4(b) OR [4 marks]**

Explain any two applications of the apriori algorithm.

**Solution****Table 16.** Apriori Applications

Application	Description
Market Basket Analysis	Finding products bought together (e.g., Bread & Butter) to optimize store layout.
Web Usage Mining	Discovering reliable navigation patterns to improve website UX.

Process:

1. Generate frequent itemsets.
2. Prune infrequent items based on support.
3. Generate association rules based on confidence.

#### Mnemonic

“Apriori Analyzes Associations Automatically (AAAA)”

### Question 4(c) OR [7 marks]

Explain the features and applications of Matplotlib.

#### Solution

Matplotlib is a comprehensive library for creating static, animated, and interactive visualizations in Python.

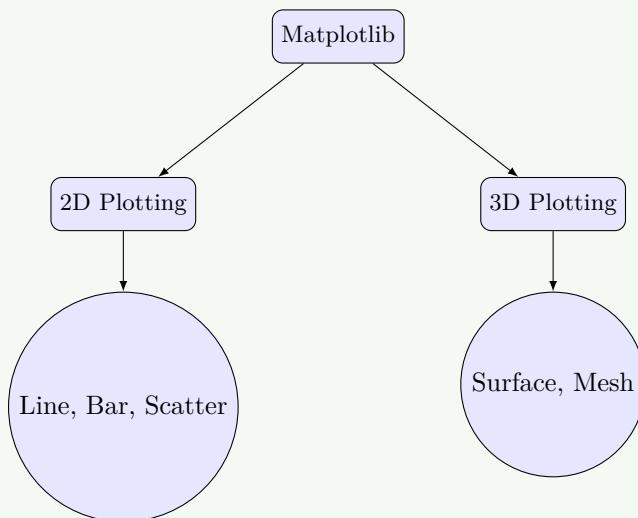


Figure 6. Matplotlib Capabilities

#### Applications:

- **Data Exploration:** Histograms, scatter plots to understand data.
- **Model Performance:** Plotting loss curves and accuracy.
- **Result Presentation:** Publication-quality figures.

#### Mnemonic

“Matplotlib Makes Meaningful Visual Displays (MMVD)”

### Question 5(a) [3 marks]

List out the major features of Numpy.

### Solution

#### Features of NumPy:

- **N-dimensional Arrays:** Fast and efficient multidimensional array object (`ndarray`).
- **Broadcasting:** Functions to perform operations on arrays of different shapes.
- **Linear Algebra:** Built-in support for matrix operations and Fourier transforms.
- **C/C++ Integration:** Tools for integrating C/C++ and Fortran code.

### Mnemonic

“NumPy Numbers Need Neat Operations (NNNNO)”

## Question 5(b) [4 marks]

How to load an iris dataset csv file in a Pandas Dataframe program? Explain with example.

### Solution

```

1 import pandas as pd
2
3 # Method 1: Load from local CSV file
4 df = pd.read_csv('iris.csv')
5
6 # Method 2: From sklearn (common in ML)
7 from sklearn.datasets import load_iris
8 iris = load_iris()
9 df_iris = pd.DataFrame(iris.data, columns=iris.feature_names)
10
11 # Display first 5 rows
12 print(df.head())

```

#### Explanation:

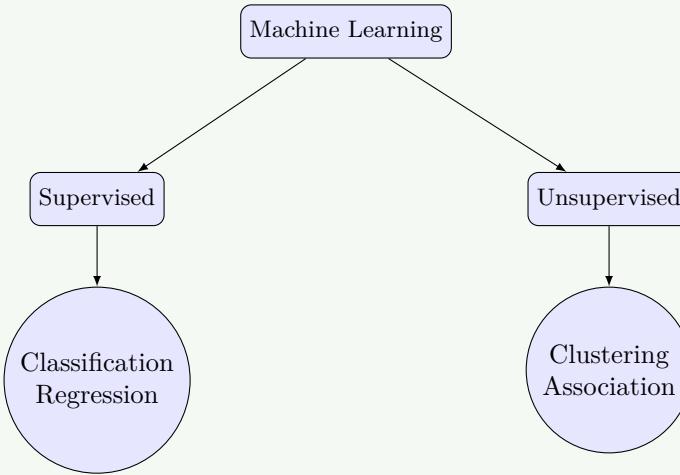
- `pd.read_csv()`: Function to read comma-separated values.
- `df.head()`: Returns the first n rows (default 5).

### Mnemonic

“Pandas Reads CSV Files Easily (PRCFE)”

## Question 5(c) [7 marks]

Compare and Contrast Supervised Learning and Unsupervised Learning.

**Solution****Figure 7.** ML Learning Types**Table 17.** Comparison

Aspect	Supervised	Unsupervised
<b>Data</b>	Labeled	Unlabeled
<b>Goal</b>	Predict output	Discover patterns
<b>Feedback</b>	Direct feedback	No feedback
<b>Complexity</b>	Validation is easier	Validation is harder

**Mnemonic**

“Supervised Seeks Specific Solutions, Unsupervised Uncovers Unknown (SSSUU)”

**Question 5(a) OR [3 marks]**

List out the applications of Pandas.

**Solution****Table 18.** Pandas Applications

Application	Description	Field
<b>Data Cleaning</b>	Handling missing data	General ML
<b>Financial Analysis</b>	Stock market trends	Finance
<b>Recommendation</b>	Analyzing user behavior	E-commerce

**Mnemonic**

“Pandas Processes Data Perfectly (PPDP)”

**Question 5(b) OR [4 marks]**

How to plot a vertical line and horizontal line in matplotlib? Explain with examples.

### Solution

```

1 import matplotlib.pyplot as plt
2
3 # Plot a simple line
4 plt.plot([1, 2, 3], [1, 4, 9])
5
6 # Vertical line at x = 2 (Red dashed)
7 plt.axvline(x=2, color='red', linestyle='--')
8
9 # Horizontal line at y = 4 (Green solid)
10 plt.axhline(y=4, color='green', linestyle='--')
11
12 plt.show()

```

#### Functions:

- `axvline(x)`: Adds a vertical line across the axes.
- `axhline(y)`: Adds a horizontal line across the axes.

## Question 5(c) OR [7 marks]

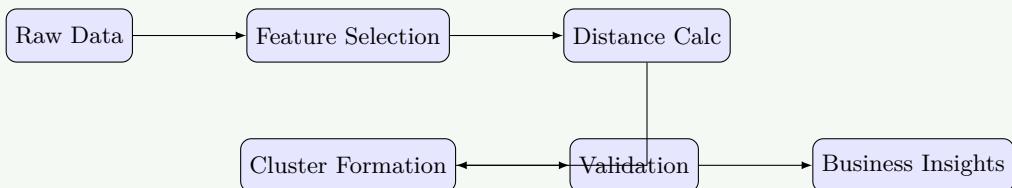
Describe the concept of clustering using appropriate real-world examples.

### Solution

**Clustering** is an unsupervised learning technique that groups similar data points such that points in the same group are more similar to each other than to those in other groups.

**Table 19.** Clustering Applications

Type	Example	Impact
Customer Seg.	Group by purchase behavior	Targeted marketing
Image Seg.	Tumor detection in MRI	Improved diagnosis
Gene Analysis	Group genes by expression	Drug discovery



**Figure 8.** Clustering Process

#### Real-World Examples:

1. **Customer Segmentation:** Identifying high-value customers vs. seasonal shoppers.
2. **Social Media Analysis:** Grouping users by interests (e.g., sports, tech).
3. **Market Research:** Finding segments with similar product needs.

### Mnemonic

“Clustering Creates Clear Categories (CCCC)”