

**Public Policy  
& Opinion Cell**  
Indian Institute of Technology, Kanpur

**Summer Project'24**

# **FACT FILTER**

**End-Term Evaluation**



# CONTENTS

**01**

INTRODUCTION

**02**

PROJECT TIMELINE

**03**

WEEK-WISE ACCOMPLISHMENT

- WEEK 1
- WEEK 2
- WEEK 3
- WEEK 4
- WEEK 5

**04**

OUR TEAM



# INTRODUCTION

- With the proliferation of misinformation online, the ability to automatically identify fake news is crucial for maintaining the integrity of information dissemination.
- The principal aim of this project was to develop a fact filtering machine learning model that can reliably differentiate between authentic and fraudulent news stories.
- The model will be trained on datasets of both real and fake news articles, utilizing both machine learning and natural language processing (NLP) techniques for feature extraction and classification of news articles as fake or real.





# PROJECT TIMELINE



**WEEK  
1**

**WEEK  
2**

**WEEK  
3**

**WEEK  
4**

**WEEK  
5**

**Introduction  
to Machine  
learning and  
Python  
Libraries**

**Data collection  
and  
Preprocessing**

**Feature  
Extraction**

**Machine  
learning  
models and  
evaluation  
metrics**

**Final Model**



# WEEK-WISE ACCOMPLISHMENT

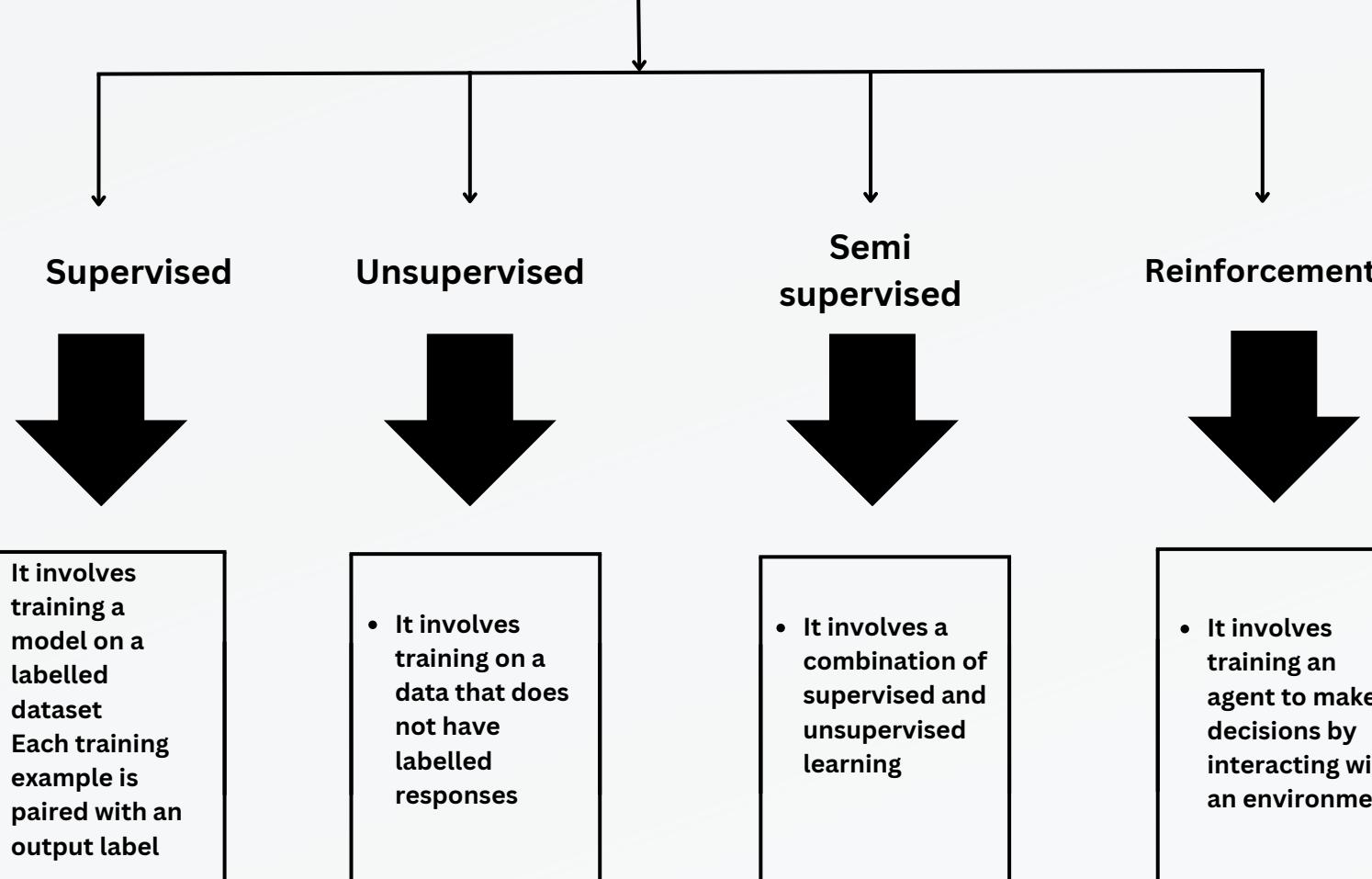


# WEEK-1

## Introduction to Machine Learning and Python Libraries

ML is basically a subset of AI that focuses on developing algorithms and statistical models by analyzing data and identifying patterns

### Machine Learning



## Used MACHINE Learning Techniques

### Logistic Regression

Use for binary classification problems  
It predicts the probability of a binary outcome

### Decision Trees

A decision tree in machine learning splits data into branches based on feature values to predict outcomes and make decisions. Used for both classification and regression tasks

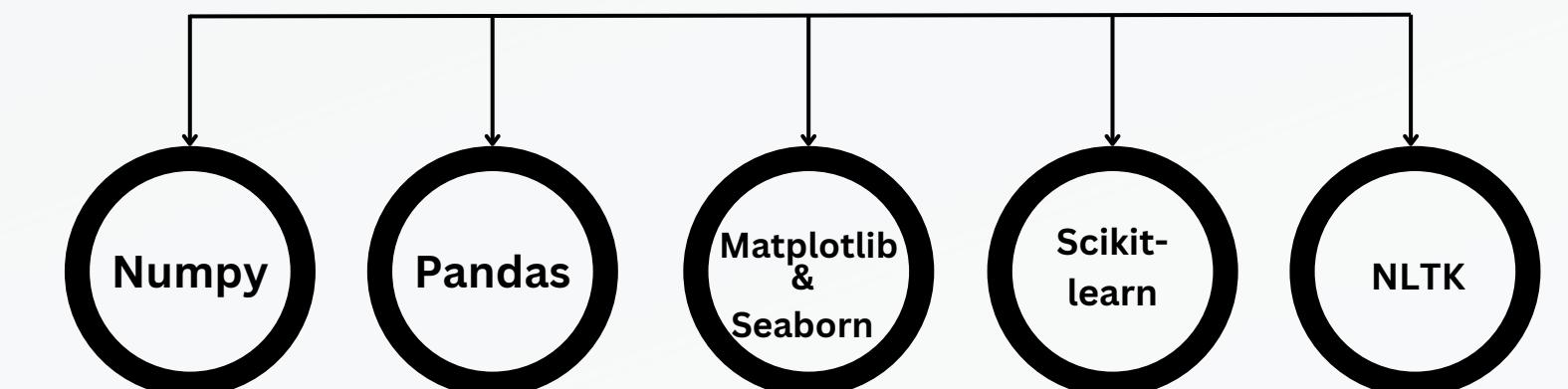
### SVMs

Support Vector Machines (SVMs) are supervised learning models that classify data by finding the optimal hyperplane that separates classes in feature space.

### Random forest

Random forest is an ensemble learning method that combines multiple decision trees to improve prediction accuracy and control overfitting.

### Python libraries



For numerical computations and array operations

For data manipulation and analysis using DataFrames

For data visualization

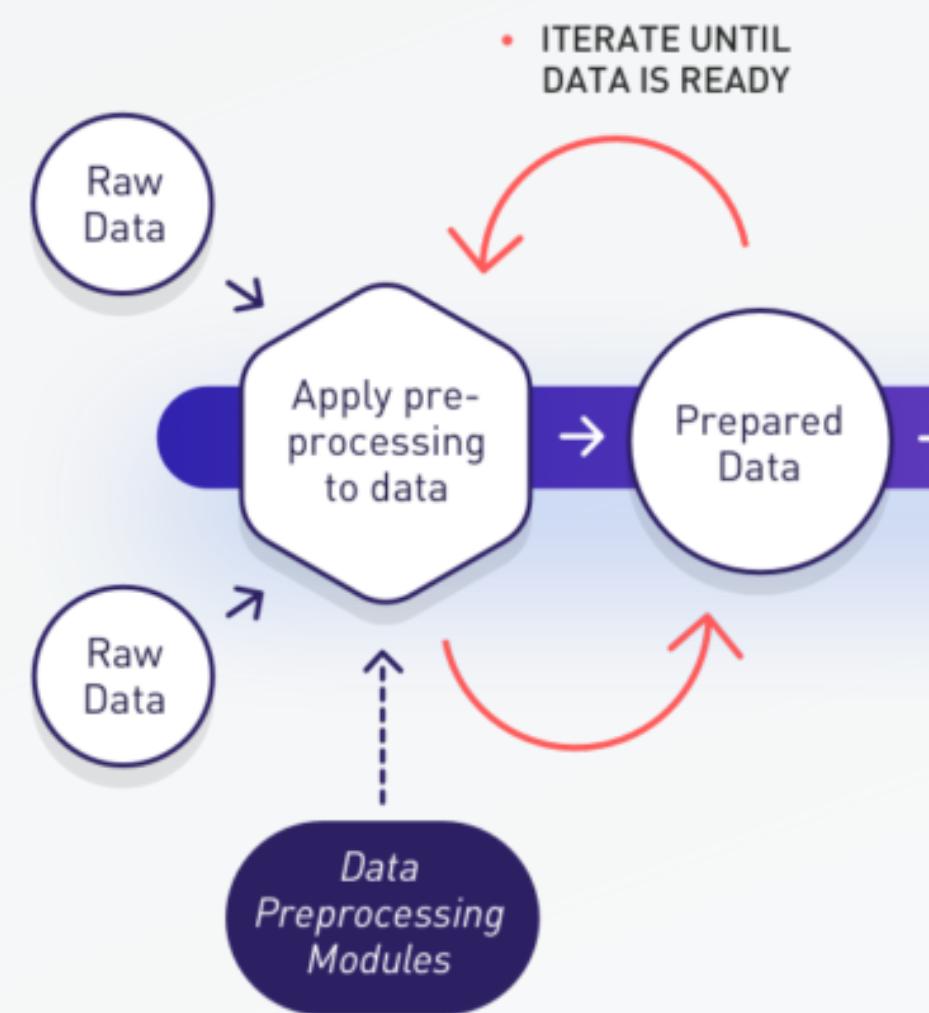
It features various classification, regression, and clustering algorithms including svm, random forests, gradient boosting.

Mainly used for working with human language data widely used in the field of NLP

# WEEK-2

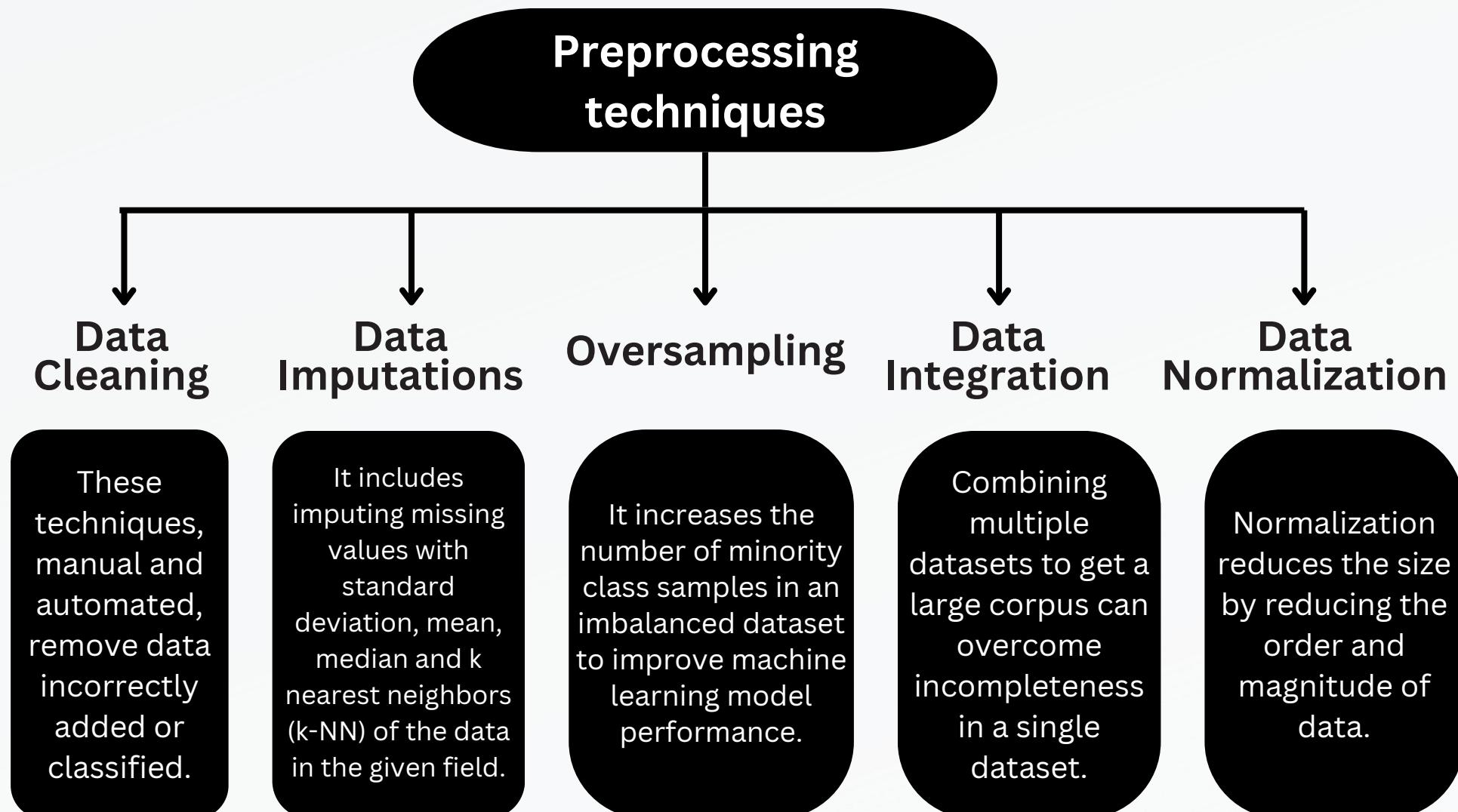
## Data collection

Collecting data for training the ML model is the basic step in the machine learning pipeline. The predictions made by ML systems can only be as good as the data on which they have been trained.



## Data Preprocessing

Real-world raw data and images are often incomplete, inconsistent and lacking in certain behaviors or trends. They are also likely to contain many errors. So, once collected, they are pre-processed into a format the machine learning algorithm can use for the model.





# WEEK - 3

## One Hot Encoding

Converts categorical data (words) into a binary format that machine learning algorithms can process.

## Word Embedding

Represents words in continuous vector space where semantically similar words are closer together.

## WordtoVec

It is a popular technique in NLP that converts words into numerical vectors, capturing semantic meanings

## Feature Extraction

**process of transforming raw data into numerical features that can be used by machine learning algorithms.**

## Bag of words

Represents text data by the frequency (or presence) of words, ignoring grammar and word order.

## Regex

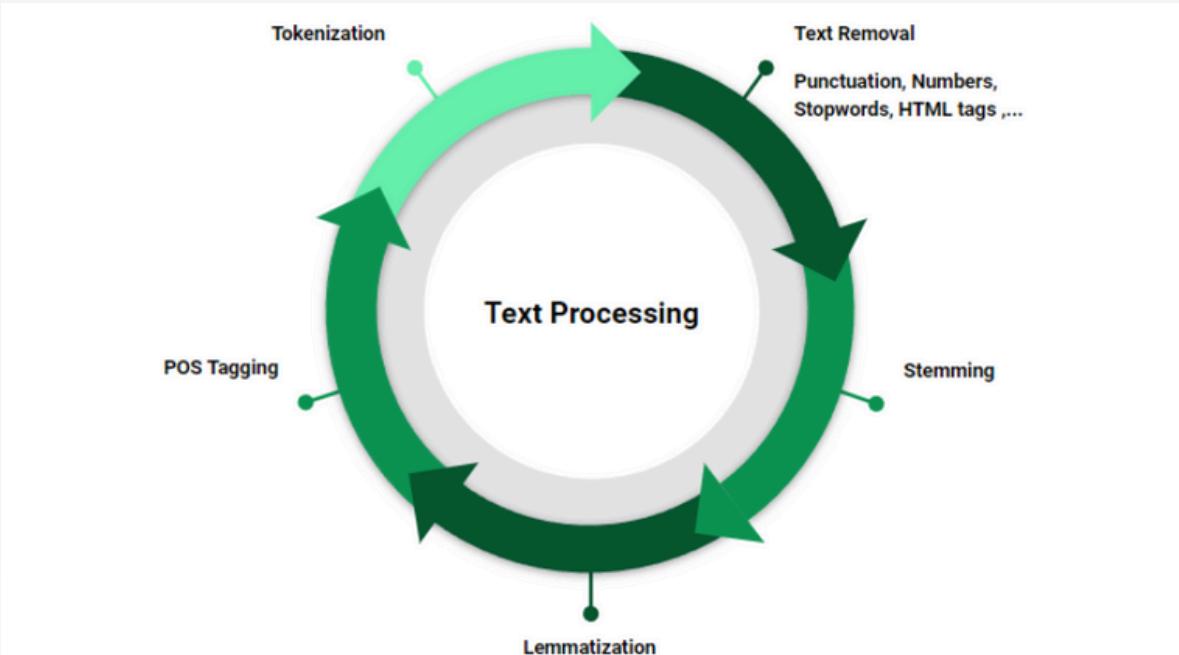
- Extracts relevant text patterns, like dates, emails, and keywords, from raw data.
- Removes unwanted text, like special characters or noise, to prepare data for analysis.
- Converts unstructured text into structured data by capturing specific information, enhancing data processing and analysis.

## TF-IDF

Enhances the Bag of Words model by reducing the weight of commonly occurring words and emphasizing rarer but potentially more informative words.

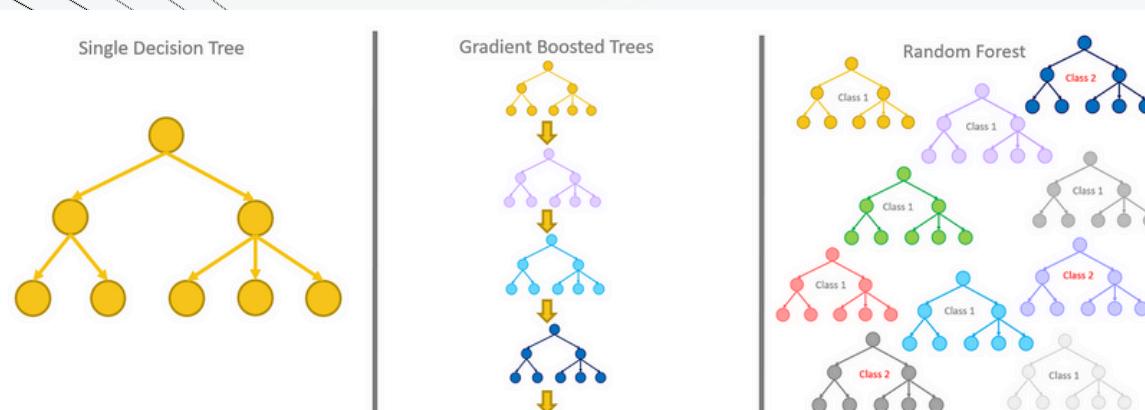
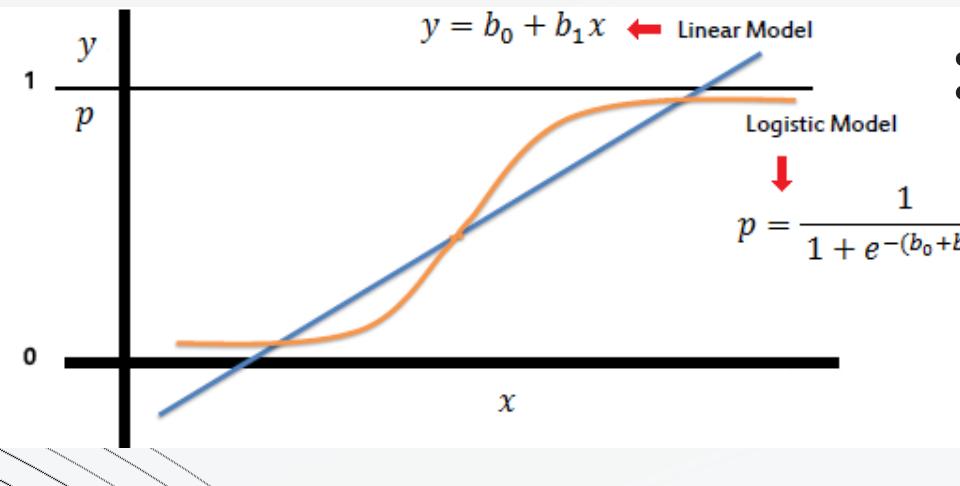


# WEEK - 4



## Text cleaning Techniques

Suitable models that can be used for our project



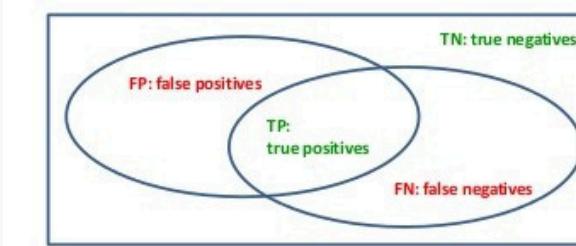
## Logistic Regression

Decision Tree  
Gradient Boosting  
Random Forest Classifier  
and their main similarities  
differences

## Evaluation metrics

| True Class      |          |    |
|-----------------|----------|----|
| Predicted Class | Positive |    |
|                 | Negative |    |
| Positive        | TP       | FP |
| Negative        | FN       | TN |

Accuracy, Precision, Recall,  
and F-measure



$$\text{Precision: } p = \frac{TP}{TP + FP}$$

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

$$\text{F-measure: } \text{Harmonic mean of precision and recall}$$
$$F = \frac{1}{\frac{1}{2}(\frac{1}{p} + \frac{1}{r})} = \frac{2pr}{p+r}$$

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

## Confusion Matrix

## Evaluation metric formulae



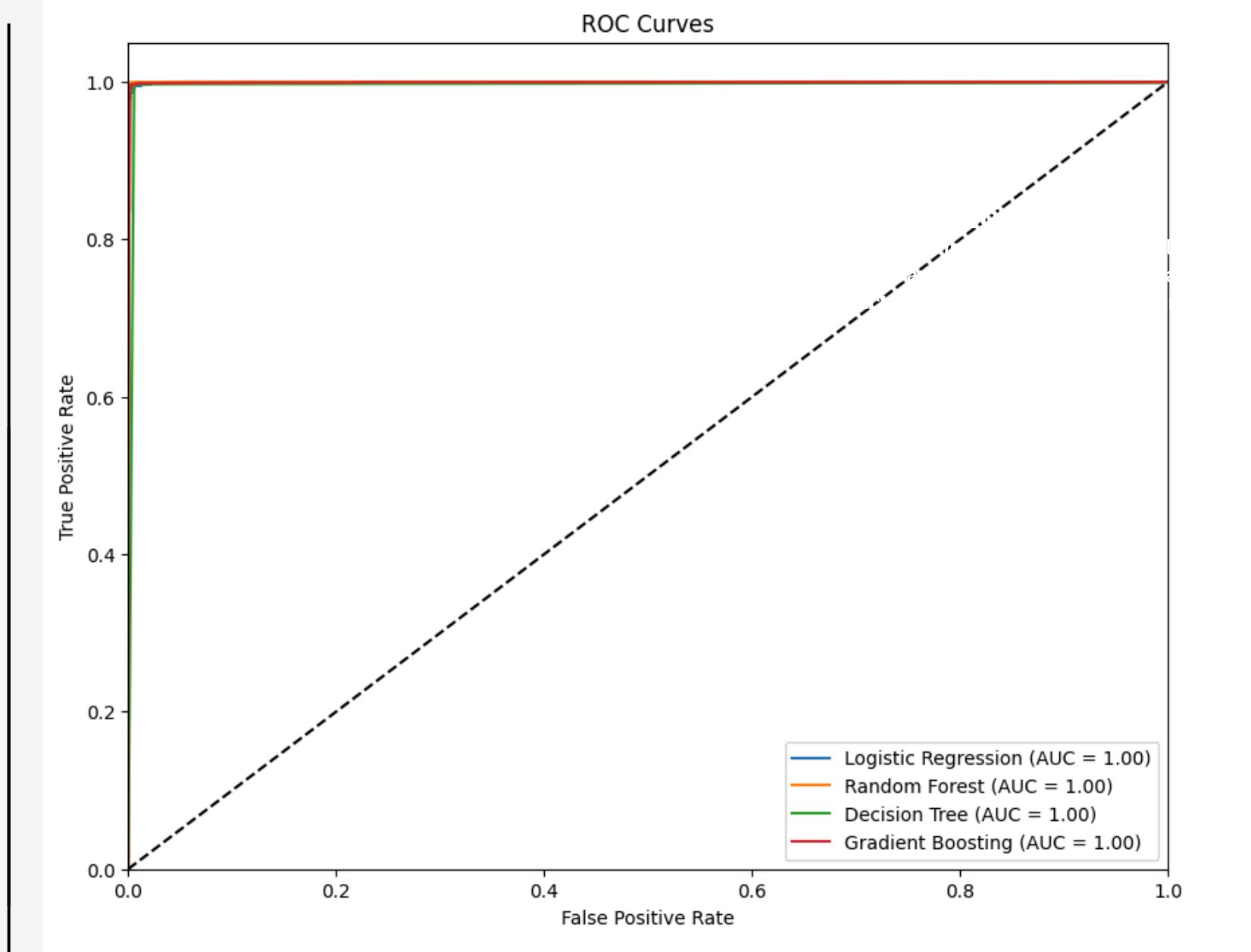
# WEEK - 5

## Model Evaluation and Comparison:

**Model evaluation assesses a model's performance using metrics like accuracy, precision, recall, F1-score, ROC-AUC, and confusion matrix to ensure it generalizes well to new data.**

| Model               | Accuracy | Precision | Recall | F1-score | ROC-AUC |
|---------------------|----------|-----------|--------|----------|---------|
| Logistic Regression | 0.9938   | 0.9925    | 0.9944 | 0.9935   | 0.9998  |
| Random Forest       | 0.9984   | 0.9984    | 0.9984 | 0.9984   | 0.9999  |
| Decision Tree       | 0.9957   | 0.9937    | 0.9972 | 0.9955   | 0.9952  |
| Gradient Boosting   | 0.996    | 0.9949    | 0.9967 | 0.9958   | 1.0     |

**Model comparison involves evaluating multiple models on the same dataset to determine the best-performing one, using key metrics to guide the selection process.**



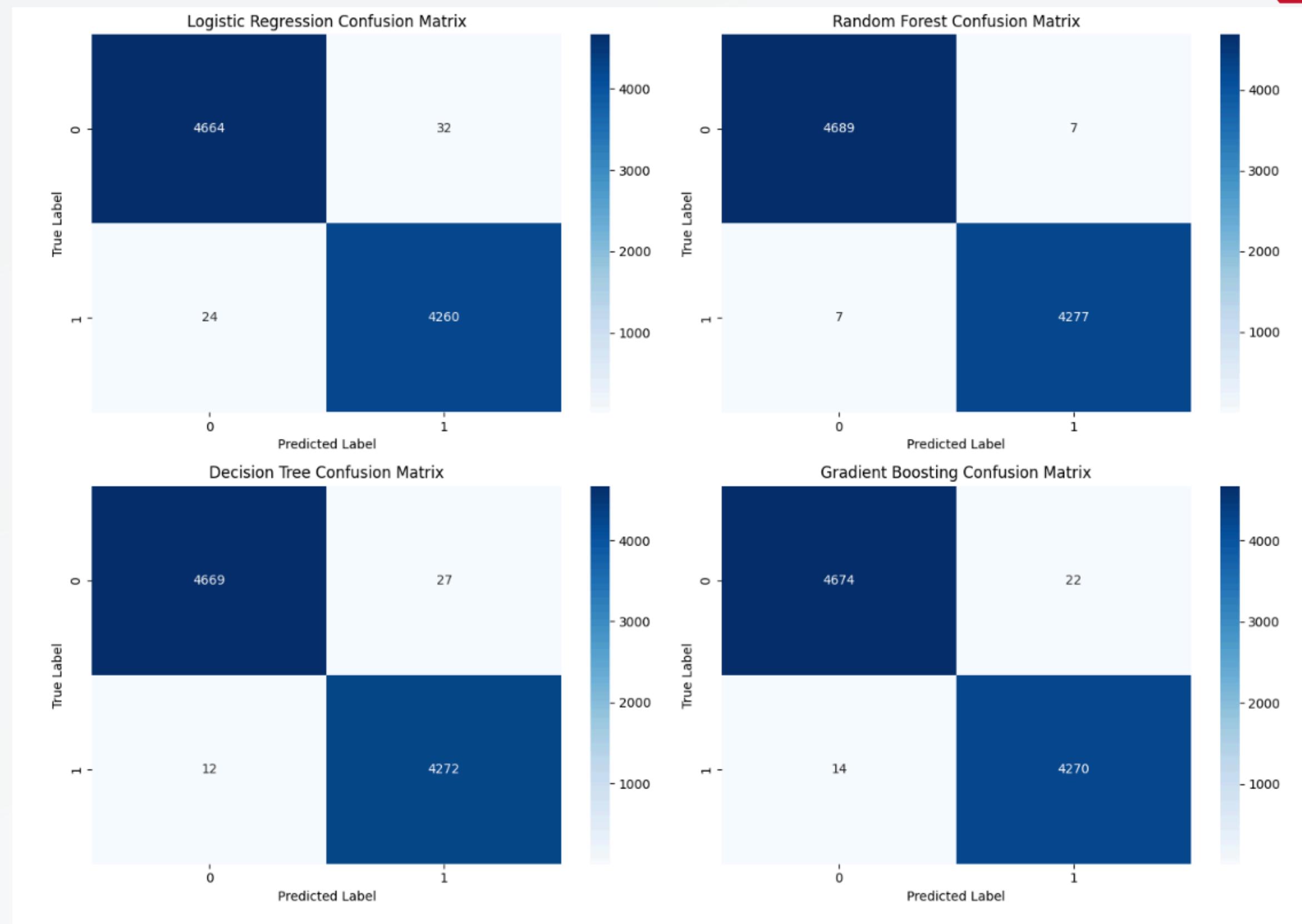
This ROC curve demonstrates that all four models (Logistic Regression, Random Forest, Decision Tree, and Gradient Boosting) are performing perfectly on the given dataset, with an AUC of 1.00, indicating that they all have perfect classification abilities on this specific dataset. This could be a sign of overfitting, where the models perform exceptionally well on the training/testing data but may not generalize to new, unseen data.



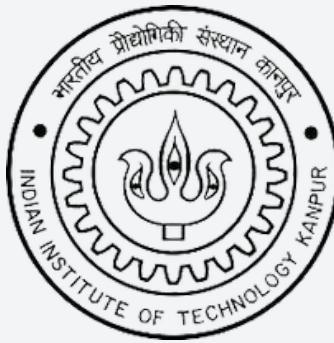
# WEEK - 5

Public Policy  
& Opinion Cell

Indian Institute of Technology, Kanpur



Each models Confusion Matrix



# THANK YOU!

