

# Group 05-Final project proposal

**Course:** Introduction to Machine Learning

**Topics:** Titanic: Machine Learning from Disaster (binary classification)

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## 1. Dataset and Task Description

**Source and overview :**

The dataset used in this project is “**Titanic: Machine Learning from Disaster**”, available on the Kaggle platform. It provides demographic and travel information for a subset of passengers aboard the RMS Titanic.

**Learning task :**

The goal is to solve a **binary classification** problem. We want to predict whether a passenger **survived (Survived = 1)** or **did not survive (Survived = 0)** based on their personal and travel characteristics such as age, gender, and passenger class.

**Data characteristics :**

- **Type :** Tabular data.
- **Size :** The training set contains **891 samples**, and the test set contains **418 samples**.
- **Features:**  
The dataset includes variables : **Pclass** (Passenger class), **Sex** (Gender), **Age** (Age in years), **SibSp** (Number of siblings/spouses aboard), **Parch** (Number of parents/children aboard), **Ticket** (Ticket number), **Fare** (Ticket fare), **Cabin** (Cabin number) and **Embarked** (Port of embarkation)

## 2. Preprocessing & Feature Engineering

**1. Handle missing values :**

Several features contain missing values, including ‘**Age**’, ‘**Cabin**’, and ‘**Embarked**’. For ‘**Age**’ and ‘**Embarked**’, we will evaluate filling missing values using measures such as the **median**, **mean**, or **mode**.

Since ‘**Cabin**’ has a high proportion of missing data, we may consider **dropping** this feature or extracting partial information (e.g., cabin prefix).

**2. Feature transformation :**

- Convert categorical features such as ‘**Sex**’ and ‘**Embarked**’ into numerical form using one-hot encoding.

- Extract titles (e.g., Mr., Mrs., Miss) from the '**Name**' feature to create a new feature '**Title**', which may reflect social status and survival likelihood.
- Combine '**SibSp**' and '**Parch**' into a new feature '**FamilySize**' to represent family presence aboard.

### 3. Normalize the data:

Apply **standardization** or **normalization** to numerical variables such as '**Age**' and '**Fare**' to prevent models from being biased by differing feature scales.

## 3. Model Implementation & Comparison

To meet the project requirements, we plan to implement and compare at least **three machine learning algorithms**:

1. **Logistic Regression** – a baseline linear classification model.
2. **Decision Tree / Random Forest** – to evaluate the performance of tree-based models.
3. **Support Vector Machine (SVM)** or **K-Nearest Neighbors (KNN)** – to explore alternative classification approaches.

## Evaluation Metrics

Since this is a classification task, we will use the following metrics:

- **Accuracy** – as the primary performance indicator.
- **F1-score** – to balance precision and recall for imbalanced outcomes.
- **Confusion Matrix** – to visualize model predictions and analyze classification errors in more detail.