

## Assignment 1 (Individual Assignment: Supervised Learning – Coding) - Due

Nov 2, 2025 11:59 PM

Fall 2025 Machine Learning (DAMO-640-10)

# Assignment 1 (Individual Assignment: Supervised Learning – Coding)

Implement a full ML pipeline in Python using Modules 1–4 on the Haberman's Survival dataset (binary classification).

### Dataset URL:

<https://archive.ics.uci.edu/dataset/43/haberman+s+survival>

### Column Names:

1. age (in years)
2. operation\_year (year of operation, 1958–1969)
3. axillary\_nodes (# of positive lymph nodes detected)
4. survival\_status (1 = survived  $\geq$  5 years, 2 = died within 5 years)

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Your **Jupyter notebook** must perform each of the following steps:

## 1. Data Loading & EDA (Module 1)

- Load the CSV (no header) and assign the four column names.
- Display first five rows, summary statistics, and the class distribution (survival\_status).

- Check for missing or invalid values.

## **2. Preprocessing (Modules 1 & 3)**

- Convert survival\_status to 0/1 (0 = no, 1 = yes).
- Split into 75 % train / 25 % test (random\_state=42).
- Standardize the three feature columns (age, operation\_year, axillary\_nodes) using StandardScaler.

## **3. Dimensionality Reduction (Module 4)**

- Apply PCA on the training features.
- Plot cumulative explained variance vs. number of components.
- Choose the smallest number of components that retain  $\geq 90$  % of variance and transform both train and test sets.

## **4. Supervised Learning (Module 2)**

- Train Logistic Regression and Decision Tree classifiers on the PCA-transformed training data.
- For each model, compare two hyperparameter settings:
  - Logistic Regression:  $C = [0.1, 1.0]$
  - Decision Tree:  $\text{max\_depth} = [3, \text{None}]$
- Use 5-fold cross-validation on the training set to compute mean accuracy for each setting.

## **5. Model Evaluation & Optimization (Module 3)**

- On the test set (PCA-transformed), for each model's best hyperparameter setting, compute:
  - Accuracy, precision, recall,  $F_1$ -score
  - ROC curve and AUC
- Plot both models' ROC curves on the same axes and discuss which is preferable and why.

## 6. Brief Report

In a Markdown cell or separate PDF ( $\leq 2$  pages), summarize:

- Key EDA findings (class balance, feature distributions)
  - PCA results (chosen components and variance retained)
  - Cross-validation accuracies for each hyperparameter setting
  - Test-set metrics and ROC comparison
  - Final recommendation
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## Submission

The following needs to be submitted:

- Notebook (.ipynb), reproducible top to bottom.
- Report (embedded Markdown or separate PDF,  $\leq 2$  pages).

Submit both (or a single .zip) via the LMS by the deadline.



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Due November 2 at 11:59 PM



Starts Oct 19, 2025 12:01 AM Ends Nov 2, 2025 11:59 PM