

Machine Learning Project

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Project Overview

Instructor: Dr. Salwa Emam.

Project description: The project focuses on implementing and evaluating machine learning models for classification and regression tasks on two datasets:

- Oxford 102 Flowers Dataset (Image Classification)
- Credit Risk Dataset (Numerical Regression)

KNN & Logistic and Linear Regression

KNN:

- **Nature:** Non-parametric model (no assumption on data distribution).
- **Training:** No explicit training phase; stores the dataset and computes distances during prediction.
- **Interpretability:** Less interpretable due to reliance on nearest neighbors without a clear decision boundary.
- **Performance:** Can capture more complex decision boundaries and is more flexible for high-dimensional image data.
- **Scalability:** Computationally expensive during inference, especially with large datasets, since it requires comparing every test point to all training points.

Logistic Regression:

- **Nature:** Parametric model (requires assumption about the underlying distribution of data).
- **Training:** Faster since it involves finding optimal weights for features.
- **Interpretability:** Good, as the model provides the influence of each feature on the prediction.
- **Performance:** Works well for simpler patterns, but struggles with highly complex or non-linear image data.

- **Scalability:** More scalable to large datasets due to lower computational cost during inference.

Linear Regression:

- **Nature:** Parametric model (assumes a linear relationship between features and target).
- **Training:** Faster to train, as it directly calculates the optimal weights using closed-form solutions (or gradient descent).
- **Interpretability:** High, as it shows the linear relationship between the input features and target variable.
- **Performance:** Works well when the relationship between features and target is truly linear. Struggles with non-linear relationships.
- **Scalability:** Scalable to large datasets, as it has low computational cost after training.

Numerical Dataset Overview

1. **Dataset name:** Credit risk dataset.

2. **Total Rows:** 32,581.

3. **Total Columns:** 12.

4. **Columns:**

- `person_age`: Age of the person.
- `person_income`: Annual income of the person.
- `person_home_ownership`: Type of home ownership (e.g., RENT, OWN, MORTGAGE).
- `person_emp_length`: Length of employment (in years), with some missing values (895).
- `loan_intent`: Purpose of the loan (e.g., PERSONAL, EDUCATION, MEDICAL).
- `loan_grade`: Loan grade (categorical, e.g., A, B, C, D).
- `loan_amnt`: Loan amount.
- `loan_int_rate`: Interest rate of the loan, with missing values (3,116).
- `loan_status`: Loan status (0 or 1).
- `loan_percent_income`: Loan amount as a percentage of income.
- `cb_person_default_on_file`: Whether the person has a history of default (Y/N).
- `cb_person_cred_hist_length`: Credit history length.

5. **Missing Values:**

- `person_emp_length`: 895 missing values.

- loan_int_rate: 3,116 missing values.

6. Sample Of Data (First five rows):

Person_age	Person_income	Person_home_o	Person_emp_ler	Loan_intent	Loan_grade	Loan_amnt	Loan_int_rate	Loan_status	Loan_percent_i	Cb_person_defa	Cb_person_cred
22	59000	RENT	123	PERSONAL	D	35000	16.02	1	0.59	Y	3
21	9600	OWN	5	EDUCATION	B	1000	11.14	0	0.1	N	2
25	9600	MORTGAGE	1	MEDICAL	C	5500	12.87	1	0.57	N	3
23	65500	RENT	4	MEDICAL	C	35000	15.23	1	0.53	N	2
24	54400	RENT	8	MEDICAL	C	35000	14.27	1	0.55	Y	4

KNN & Linear Regression On The Numerical Dataset

Metric	Linear Regression	KNN
Mean Squared Error (MSE)	1.8691	1.9021
Mean Absolute Error (MAE)	1.0395	1.0399
R ² Score	0.8051	0.8017

1. Mean Squared Error (MSE):

- Linear Regression (1.8691) has a slightly lower MSE than KNN (1.9021), meaning it makes smaller squared prediction errors on average.
- Lower MSE is better, so Linear Regression performs better here.

2. Mean Absolute Error (MAE):

- Both algorithms have almost identical MAE: Linear Regression (1.0395) vs. KNN (1.0399).
- MAE measures average absolute errors, and in this case, both algorithms are performing nearly equally well.

3. R² Score (Coefficient of Determination):

- Linear Regression: 0.8051
- KNN: 0.8017
- R² indicates how well the model explains the variability in the target variable. Higher is better.
- Linear Regression has a slightly better R² score, meaning it explains a marginally higher proportion of the variance in the data compared to KNN.

• Conclusion:

Linear Regression performs slightly better overall in terms of MSE and R² score, while MAE is nearly identical. Linear Regression is preferred for simplicity and interpretability.

Image Dataset Overview

1. **Dataset name:** Oxford 102 Flowers Dataset.
2. **Total Rows:** 8,189 images.
3. **Total Classes:** 102 flower categories.
4. **Image Sizes:** Images are of variable sizes, but all are in JPEG format.
5. **Class Distribution:** Each class (flower type) has between 40 and 258 images.
6. The dataset comes with class labels (1 to 102), corresponding to the flower types.
7. **Sample Of Data (First two images):**

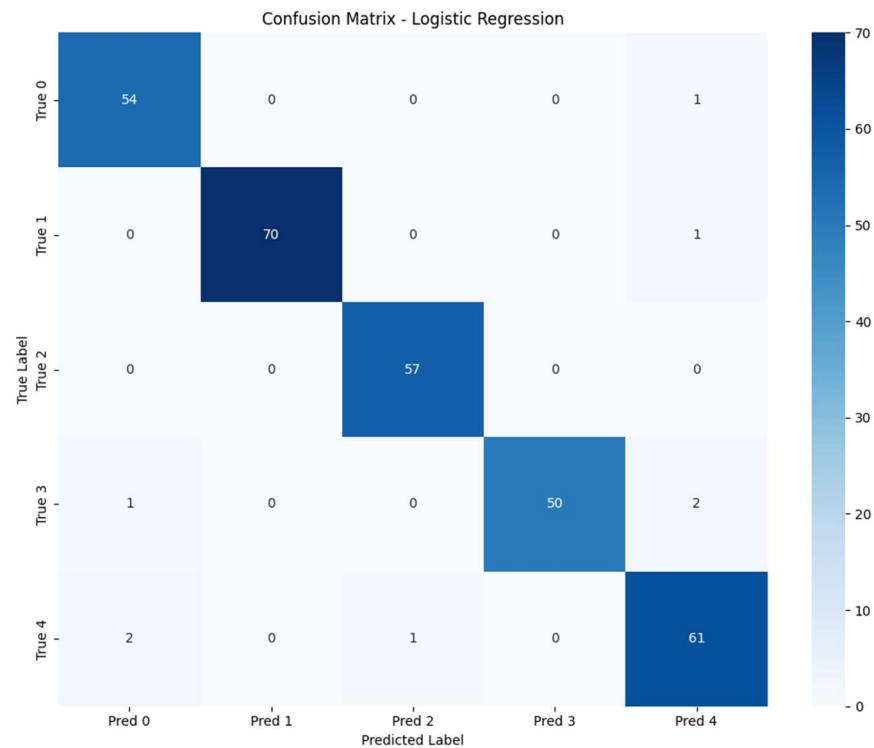


KNN & Logistic on The Image Dataset

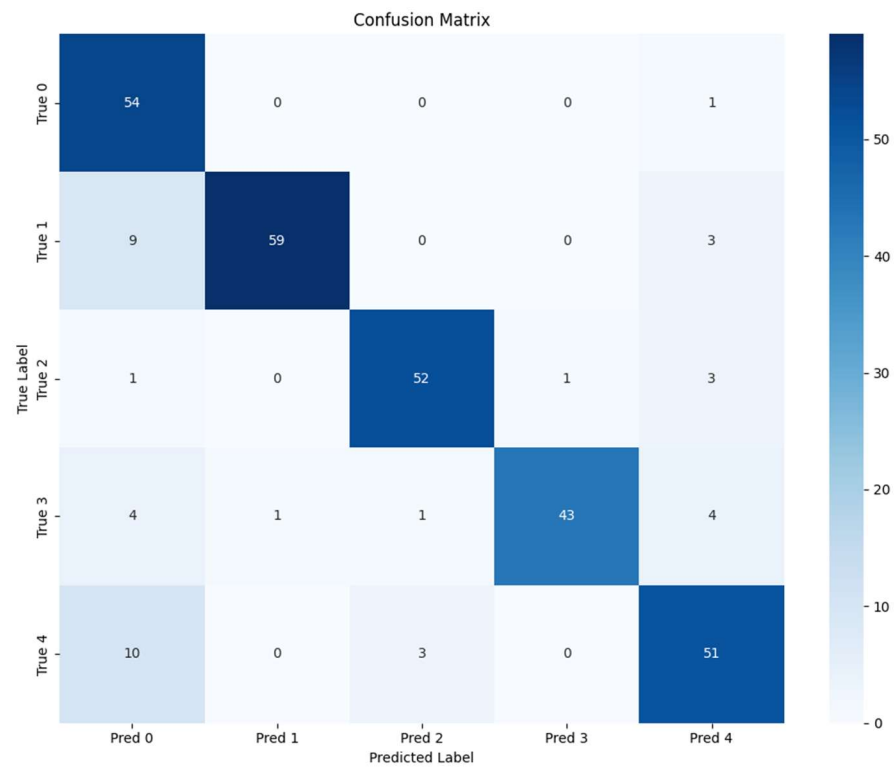
Metric	Logistic	KNN (K = 4)
Accuracy Score	0.9733333333333334	0.8633333333333333
Loss Value	0.1570	1.0681
F1-Score	0.97	0.88
Recall	0.97	0.87
Precision	0.97	0.88

1. Confusion Matrix:

- Logistic:

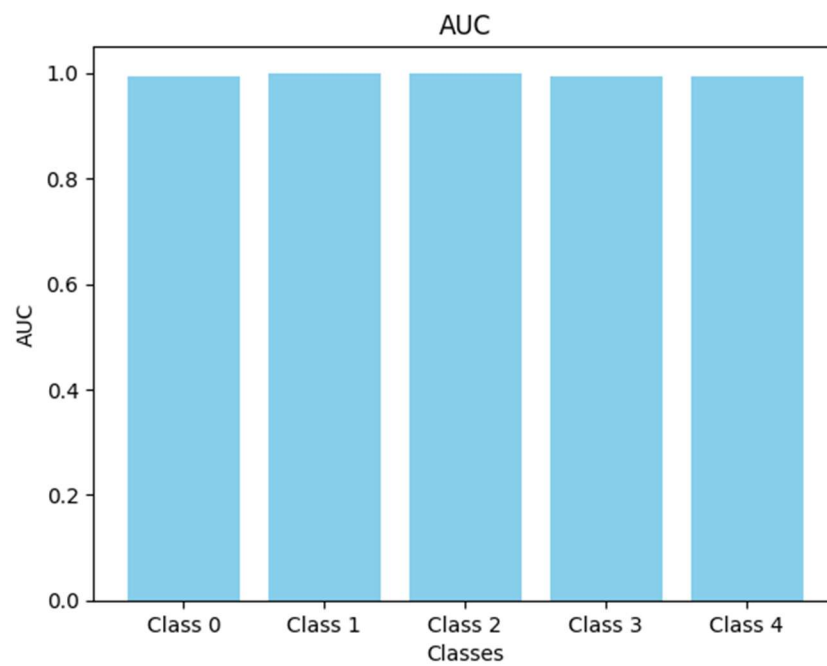


- KNN:

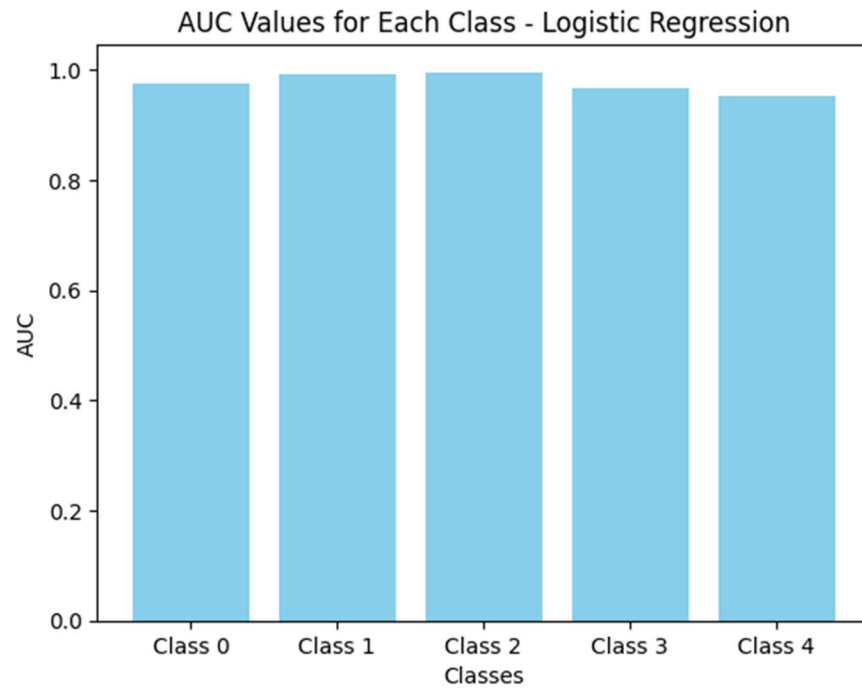


2. AUC:

- Logistic

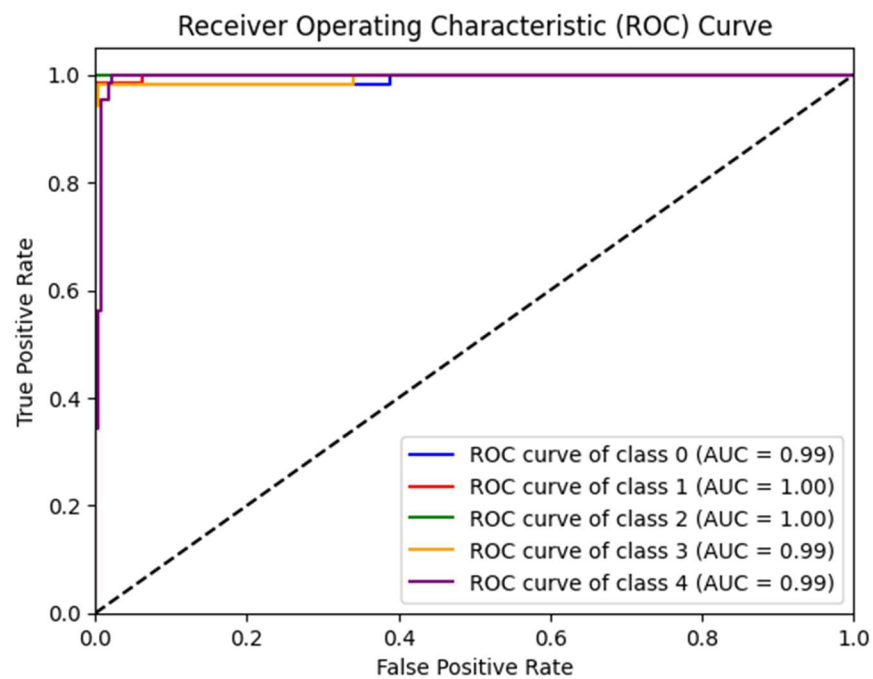


- KNN

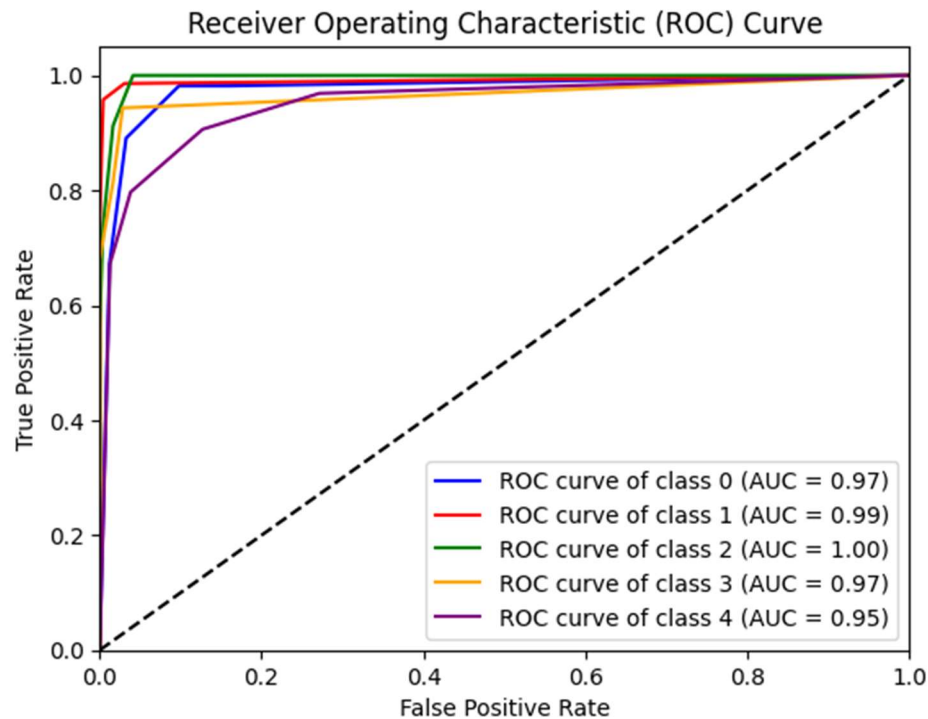


3. ROC:

- Logistic



- KNN



- **Accuracy:** Logistic Regression achieves 97.3% accuracy, significantly outperforming KNN, which achieves 86.3% accuracy.
- **Precision, Recall, F1-Score:**
 - Logistic Regression shows higher scores across all metrics for each class, indicating better overall performance.
 - KNN struggles slightly in Class 3 and Class 4 compared to Logistic Regression.
- **Precision, Recall, F1-Score:**
 - Logistic Regression shows higher scores across all metrics for each class, indicating better overall performance.
 - KNN struggles slightly in Class 3 and Class 4 compared to Logistic Regression.

- **Conclusion:**

Logistic Regression clearly outperforms KNN in terms of accuracy, precision, recall, and F1-score. It generalizes better on the Oxford 102 Flowers dataset, making it the preferred model for this classification task.

Tools & Techniques Used

- `os`: Interact with the operating system (file paths, directories).
- `numpy`: Numerical operations on arrays and matrices.
- `matplotlib.pyplot`: Data visualization library for plots and graphs.
- `seaborn`: Simplifies statistical data visualization.
- `classification_report`: Generates precision, recall, and F1-score for classification.
- `confusion_matrix`: Evaluates model performance with a confusion matrix.
- `roc_curve` / `auc`: Plots ROC curve and calculates the area under it.
- `precision_recall_curve` / `average_precision_score`: Precision-recall tradeoff analysis.
- `to_categorical`: Converts labels to one-hot encoded format.
- `KNeighborsClassifier`: KNN algorithm for classification.
- `train_test_split`: Splits data into training and testing sets.
- `learning_curve`: Analyzes model learning performance.
- `PIL.Image`: Loads and processes images.
- `loadmat`: Loads MATLAB .mat files.
- `resample`: Balances datasets through sampling.
- `StandardScaler`: Standardizes features for consistent scaling.
- `label_binarize`: Converts labels into binary format.

Thank you