

Project Name: Harvesting Brilliance: A Taxonomic Tale of Pumpkin Seed Varieties

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1. Project Overview

- **Goal:** To classify pumpkin seeds into two distinct varieties ('Çerçevelek' and 'Ürgüp Sivrisi') based on morphological features.
- **Problem Type:** Binary Classification (Supervised Machine Learning). ● **Tech Stack:** Python, Pandas, Scikit-learn, Flask, HTML/CSS.
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2. Data Collection & Preparation

The foundation of the project involved preparing the raw dataset for analysis.

- **Data Source:** Pumpkin_Seeds_Dataset.xlsx containing 2,500 samples with 13 columns.
- **Loading:** The dataset was loaded using pandas.read_excel().
- **Data Cleaning:**
 - **Null Check:** Verified that the dataset had **0 missing values**.
 - **Outlier Removal:** Used the Interquartile Range (IQR) method to remove outliers from the 'Area' column. Rows falling outside the range $[Q1 - 1.5 \times IQR, Q3 + 1.5 \times IQR]$ were dropped.
- **Preprocessing:**
 - **Scaling:** Applied MinMaxScaler to normalize 'Area', 'Perimeter', and 'Major_Axis_Length' between 0 and 1 to improve model performance.
 - **Feature Selection:** Dropped redundant or highly correlated columns: ['Convex_Area', 'Equiv_Diameter', 'Eccentricity', 'Minor_Axis_Length'].
 - **Encoding:** Converted the target variable Class from text to integers (0 and 1) using Label Encoding.

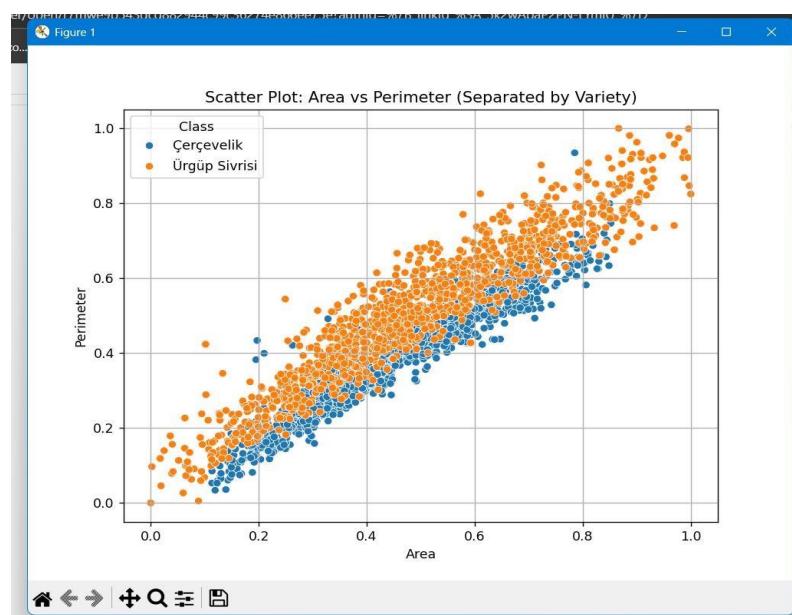
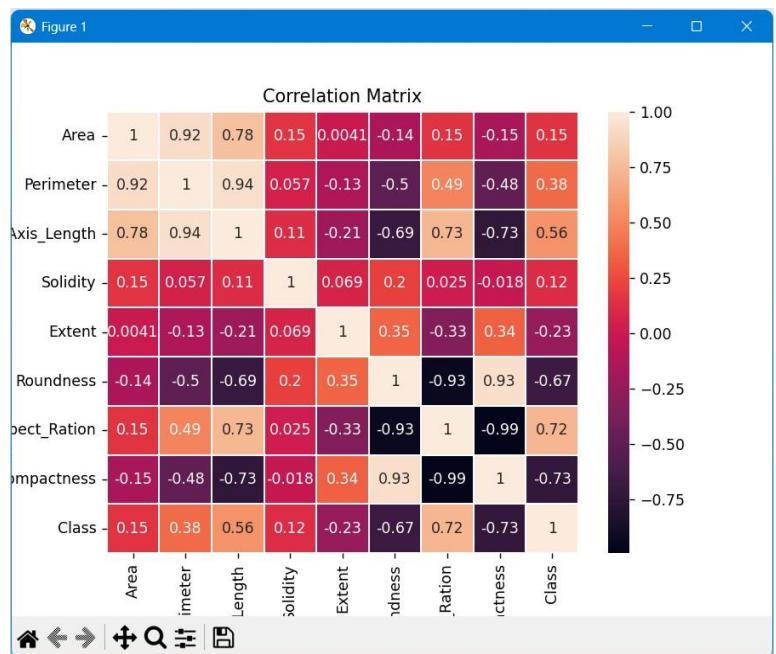
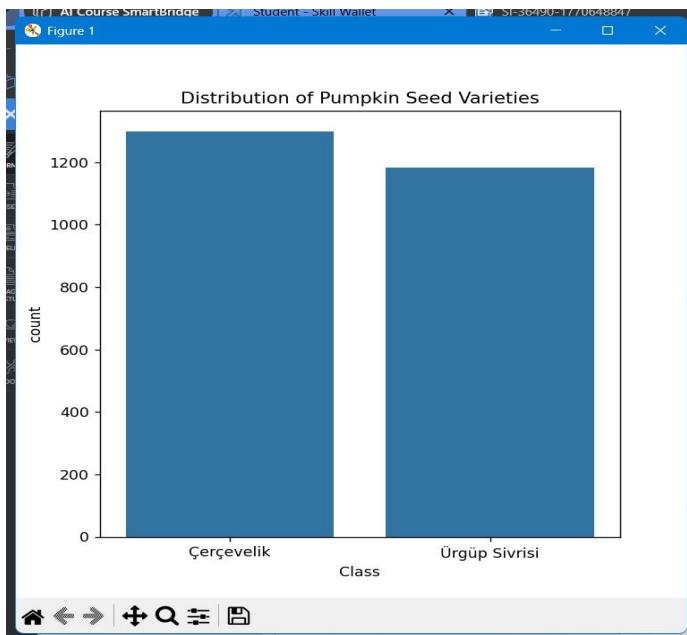
Dropping Columns										
Columns Dropped. New Dataframe Head:										
	Area	Perimeter	Major_Axis_Length	Solidity	Extent	Roundness	Aspect_Ration	Compactness	Class	
0	0.119284	0.033709	0.016192	0.9902	0.7453	0.8963	1.4809	0.8207	Çerçevevik	
1	0.410519	0.340661	0.294143	0.9916	0.7151	0.8440	1.7811	0.7487	Çerçevevik	
2	0.338866	0.365983	0.351048	0.9857	0.7400	0.7674	2.0651	0.6929	Cerçevevik	
3	0.264966	0.210828	0.185370	0.9902	0.7396	0.8486	1.7146	0.7624	Çerçevevik	
4	0.259944	0.221228	0.192467	0.9850	0.6752	0.8338	1.7413	0.7557	Çerçevevik	

3. Exploratory Data Analysis (EDA)

We analyzed the data to understand patterns and separability.

- **Descriptive Statistics:** Generated a statistical summary (Mean, Std, Min, Max) using df.describe() to understand feature distributions.
- **Univariate Analysis:**
 - **Class Balance:** Visualized the target variable using a Countplot, confirming a balanced dataset.
 - **Outliers:** Used Boxplots to visualize the spread of 'Area' before and after cleaning.
- **Bivariate Analysis:** Plotted **Area vs. Perimeter** (Scatter Plot) to observe the positive correlation between seed size and boundary length.
- **Multivariate Analysis:** Created a **Correlation Heatmap** to identify relationships between all features and the target variable





Training Logistic Regression						
Accuracy Score: 0.8611670020120724						
	precision	recall	f1-score	support		
0	0.84	0.91	0.87	257		
1	0.89	0.81	0.85	240		
accuracy			0.86	497		
macro avg	0.86	0.86	0.86	497		
weighted avg	0.86	0.86	0.86	497		
Random Forest						
Accuracy Score: 0.8893360160965795						
	precision	recall	f1-score	support		
0	0.87	0.93	0.90	257		
1	0.92	0.85	0.88	240		
accuracy			0.89	497		
macro avg	0.89	0.89	0.89	497		
weighted avg	0.89	0.89	0.89	497		
Decision Tree						
Accuracy Score: 0.8370221327967807						
	precision	recall	f1-score	support		
0	0.83	0.86	0.84	257		
1	0.84	0.82	0.83	240		
accuracy			0.84	497		
macro avg	0.84	0.84	0.84	497		
weighted avg	0.84	0.84	0.84	497		
Multinomial Naive Bayes						
Accuracy Score: 0.7022132796780685						
	precision	recall	f1-score	support		
0	0.64	0.96	0.77	257		
1	0.92	0.42	0.58	240		
accuracy			0.70	497		
macro avg	0.78	0.69	0.67	497		
weighted avg	0.77	0.70	0.68	497		
Support Vector Machine (SVM)						
Accuracy Score: 0.8651911468812877						
	precision	recall	f1-score	support		
0	0.83	0.92	0.88	257		
1	0.91	0.80	0.85	240		
accuracy			0.87	497		
macro avg	0.87	0.86	0.86	497		
weighted avg	0.87	0.87	0.86	497		

	4	0.239944	0.221228	0.192407	0.9850	0.0732	0.8558	1.7415	0.7357	Çerçevelik
	Area	Perimeter	Major_Axis_Length	Solidity	Extent	Roundness	Aspect_Ration	Compactness		
count	2482.000000	2482.000000	2482.000000	2482.000000	2482.000000	2482.000000	2482.000000	2482.000000	2482.000000	2482.000000
mean	0.463459	0.442677	0.411127	0.989479	0.693502	0.791838	2.039858	0.704435		
std	0.188186	0.180938	0.167586	0.003499	0.060676	0.055916	0.315819	0.053053		
min	0.000000	0.000000	0.000000	0.918600	0.468000	0.554600	1.148700	0.560800		
25%	0.325145	0.307016	0.285968	0.988300	0.659300	0.752325	1.800325	0.663900		
50%	0.443448	0.433898	0.391094	0.990300	0.713250	0.798200	1.982850	0.707900		
75%	0.592092	0.567014	0.522056	0.991500	0.740275	0.834575	2.258775	0.743700		
max	1.000000	1.000000	1.000000	0.994400	0.829600	0.939600	3.144400	0.904900		

We trained multiple supervised learning algorithms to find the best classifier.

- **Data Splitting:** The dataset was split into **Training (80%)** and **Testing (20%)** sets using `train_test_split` with `random_state=30`.
- **Models**
- **Trained:**
 - a. **Logistic Regression:** A baseline linear classifier.
 - b. **Decision Tree:** A non-linear tree-based model.
 - c. **Random Forest:** An ensemble of decision trees (Bagging).
 - d. **Naive Bayes:** Probabilistic classifier.
 - e. **Support Vector Machine (SVM):** A margin-based classifier.
 - f. **Gradient Boosting:** An ensemble boosting technique.

5. Performance Testing & Tuning

We evaluated models to select the most accurate one for deployment.

- **Evaluation Metrics:** Used **Accuracy Score** and **Classification Report** (Precision, Recall, F1Score) for each model.
- **Comparison Results:**
 - Gradient Boosting: **0.885**
 - Random Forest: **0.881**
 - SVM: **0.865**
 - Logistic Regression: **0.861**
 - Naive Bayes: **0.702**
- **Selection:** **Random Forest** and **Gradient Boosting** were the top performers. Random Forest was selected for deployment due to its robustness.
- **Manual Testing:** Performed a manual prediction test with sample data [0.41, 0.34, ...] which correctly predicted "Çerçevelek".

6. Model Deployment

The final phase involved making the model accessible via a web interface.

- **Serialization:** The trained Random Forest model was saved as a serialized file, **model.pkl**, using the pickle library.

- **Web Framework (Flask):** Created `app.py` to handle web requests. It loads `model.pkl` and defines a `/predict` route.
- **Frontend (HTML):**
 - **index.html:** A user-friendly form to accept 8 morphological inputs (Area, Perimeter, Solidity, etc.).
 - **predict.html:** A results page to display the predicted pumpkin seed variety.
- **Integration:** The system successfully takes user input from the browser, processes it through the Python backend, and returns the classification result in real-time.

Project Output :

Harvesting Brilliance

Enter Seed Metrics

Area:	<input type="text" value="e.g. 0.41"/>
Perimeter:	<input type="text" value="e.g. 0.34"/>
Major Axis Length:	<input type="text" value="e.g. 0.29"/>
Solidity:	<input type="text" value="e.g. 0.99"/>
Extent:	<input type="text" value="e.g. 0.71"/>
Roundness:	<input type="text" value="e.g. 0.84"/>
Aspect Ratio:	<input type="text" value="e.g. 1.78"/>
Compactness:	<input type="text" value="e.g. 0.74"/>

Predict Variety

Prediction Result

The Pumpkin Seed Variety is:

**The Pumpkin Seed Variety is
Ürgüp Sivrisi**

Predict Again