

```
pip install ucimlrepo
```

```
Requirement already satisfied: ucimlrepo in /usr/local/lib/python3.12/dist-packages (0.0.7)
Requirement already satisfied: pandas>=1.0.0 in /usr/local/lib/python3.12/dist-packages (from ucimlrepo) (2.2.2)
Requirement already satisfied: certifi>=2020.12.5 in /usr/local/lib/python3.12/dist-packages (from ucimlrepo) (2025.11.0)
Requirement already satisfied: numpy>=1.26.0 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.0.2)
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.9.0)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.2)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.12/dist-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
```

```
from ucimlrepo import fetch_ucirepo
```

```
# fetch dataset
abalone = fetch_ucirepo(id=1)
```

```
# data (as pandas dataframes)
X = abalone.data.features
y = abalone.data.targets
```

```
# metadata
print(abalone.metadata)
```

```
# variable information
print(abalone.variables)
```

```
{'uci_id': 1, 'name': 'Abalone', 'repository_url': 'https://archive.ics.uci.edu/dataset/1/abalone', 'data_url': 'http://archive.ics.uci.edu/ml/machine-learning-databases/abalone/abalone.data', 'description': 'The Abalone dataset is a collection of measurements taken on the shells of abalone. The measurements are: Sex, Length, Diameter, Height, Whole_weight, Shucked_weight, Viscera_weight, Shell_weight, and Rings. The target variable is Rings, which represents the age of the abalone in years. The dataset is divided into three classes based on the number of rings: 0 (infant), 1 (juvenile), and 2 (adult). The dataset is available in both CSV and ARFF formats.'}

name      role      type      demographic \
0      Sex      Feature      Categorical      None
1      Length      Feature      Continuous      None
2      Diameter      Feature      Continuous      None
3      Height      Feature      Continuous      None
4      Whole_weight      Feature      Continuous      None
5      Shucked_weight      Feature      Continuous      None
6      Viscera_weight      Feature      Continuous      None
7      Shell_weight      Feature      Continuous      None
8      Rings      Target      Integer      None

description      units      missing_values
0      M, F, and I (infant)      None      no
1      Longest shell measurement      mm      no
2      perpendicular to length      mm      no
3      with meat in shell      mm      no
4      whole abalone      grams      no
5      weight of meat      grams      no
6      gut weight (after bleeding)      grams      no
7      after being dried      grams      no
8      +1.5 gives the age in years      None      no
```

```
# Lab 1: Pandas Library
# Heading: Data Frame Operations with Pandas
```

```
import pandas as pd
from ucimlrepo import fetch_ucirepo
```

```
# Fetch Abalone dataset
abalone = fetch_ucirepo(id=1)
```

```
# Convert to DataFrame
X = abalone.data.features
y = abalone.data.targets
```

```
# Display first few rows
print(X.head())
print(y.head())
```

```
Sex      Length      Diameter      Height      Whole_weight      Shucked_weight      Viscera_weight \
0      M      0.455      0.365      0.095      0.5140      0.2245      0.1010
1      M      0.350      0.265      0.090      0.2255      0.0995      0.0485
2      F      0.530      0.420      0.135      0.6770      0.2565      0.1415
3      M      0.440      0.365      0.125      0.5160      0.2155      0.1140
4      I      0.330      0.255      0.080      0.2050      0.0895      0.0395

Shell_weight
0      0.150
1      0.070
2      0.210
3      0.155
4      0.055

Rings
```

```
0    15
1     7
2     9
3    10
4     7
```

```
# Lab 1: Numpy Library
# Heading: Numpy Array Operations

import numpy as np

# Convert Pandas DataFrame to Numpy Array
X_array = X.values

# Display array shape and first row
print(f"Shape of X: {X_array.shape}")
print("First row of X:", X_array[0])
```

```
Shape of X: (4177, 8)
First row of X: ['M' 0.455 0.365 0.095 0.514 0.2245 0.101 0.15]
```

```
# Lab 1: Scikit Learn Library
# Heading: Data Preparation with Scikit Learn

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler, OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline

# Split into train and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Identify categorical and numerical columns
categorical_features = ['Sex']
numerical_features = X.select_dtypes(include=np.number).columns.tolist()

# Create a column transformer for one-hot encoding and scaling
preprocessor = ColumnTransformer(
    transformers=[
        ('num', StandardScaler(), numerical_features),
        ('cat', OneHotEncoder(handle_unknown='ignore'), categorical_features)
    ],
    remainder='passthrough' # Keep other columns (if any)
)

# Apply the preprocessing to the training and testing data
X_train_processed = preprocessor.fit_transform(X_train)
X_test_processed = preprocessor.transform(X_test)

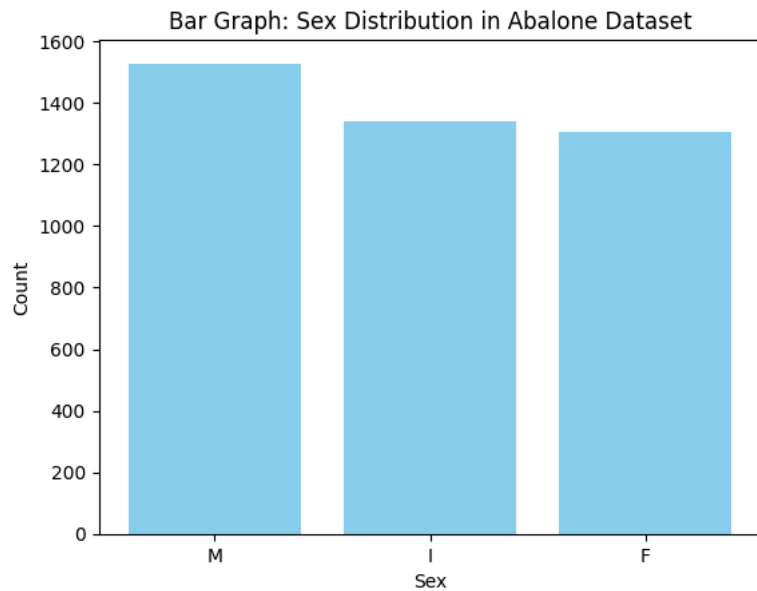
print("Processed X_train shape:", X_train_processed.shape)
```

```
Processed X_train shape: (3341, 10)
```

```
# Lab 1: Matplotlib Library
# Bar Graph: Sex Distribution in Abalone Dataset

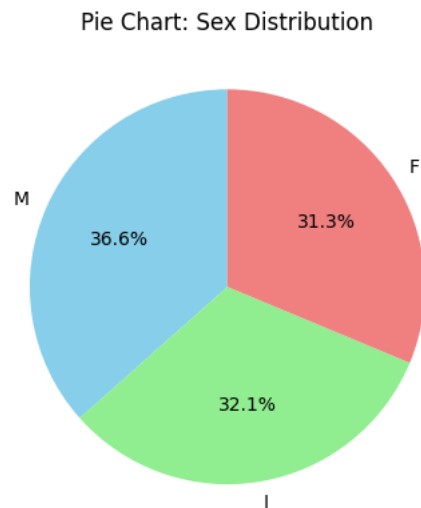
import matplotlib.pyplot as plt

plt.bar(X['Sex'].value_counts().index, X['Sex'].value_counts().values, color='skyblue')
plt.title('Bar Graph: Sex Distribution in Abalone Dataset')
plt.xlabel('Sex')
plt.ylabel('Count')
plt.show()
```



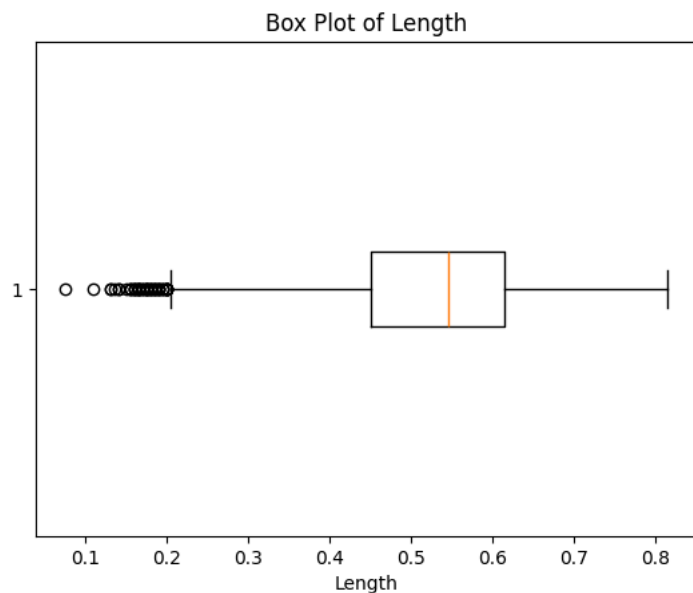
```
# Lab 1: Matplotlib Library
# Pie Chart: Sex Distribution

plt.pie(X['Sex'].value_counts().values, labels=X['Sex'].value_counts().index, autopct='%1.1f%%',
        startangle=90, colors=['skyblue', 'lightgreen', 'lightcoral'])
plt.title('Pie Chart: Sex Distribution')
plt.show()
```



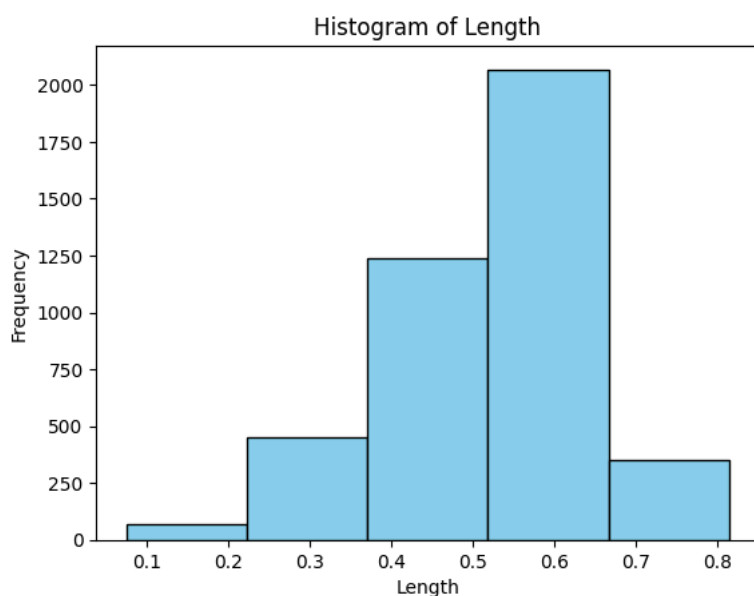
```
# Lab 1: Matplotlib Library
# Box Plot: Length

plt.boxplot(X['Length'], vert=False)
plt.title('Box Plot of Length')
plt.xlabel('Length')
plt.show()
```



```
# Lab 1: Matplotlib Library
# Histogram: Length

plt.hist(X['Length'], bins=5, color='skyblue', edgecolor='black')
plt.title('Histogram of Length')
plt.xlabel('Length')
plt.ylabel('Frequency')
plt.show()
```



```
# Lab 1: Matplotlib Library
# Line Chart and Subplots: Length & Diameter

fig, axes = plt.subplots(1, 2, figsize=(10, 5))
axes[0].plot(X['Length'], color='blue')
axes[0].set_title('Line Chart: Length')
axes[0].set_xlabel('Index')
axes[0].set_ylabel('Length')

axes[1].plot(X['Diameter'], color='green')
axes[1].set_title('Line Chart: Diameter')
axes[1].set_xlabel('Index')
axes[1].set_ylabel('Diameter')

plt.tight_layout()
plt.show()
```

