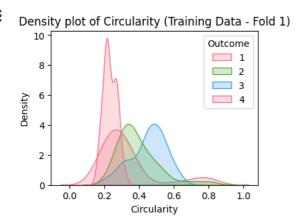
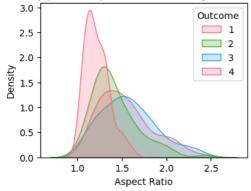
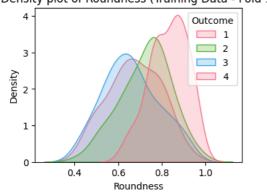
```
#Entire dataset
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import KFold
import numpy as np
# Step 1: Load the dataset
data = pd.read_excel("newww_weed1 (3).xlsx")
# Step 2: Set up k-fold cross-validation
kf = KFold(n_splits=5, shuffle=True, random_state=42)
# Step 3: Perform k-fold cross-validation
accuracies = []
predi=[]
for fold, (train_index, test_index) in enumerate(kf.split(data), 1):
    train_data, test_data = data.iloc[train_index].copy(), data.iloc[test_index].copy()
    # Generate density plots for training data
    features = train_data.drop(columns=['Outcome'])
    for feature in features.columns:
      plt.figure(figsize=(4, 3))
      sns.kdeplot(data=train_data, x=feature, hue='Outcome', fill=True, common_norm=False, palette="husl")
      plt.title(f'Density plot of {feature} (Training Data - Fold {fold})')
      plt.show()
      print()
    # Calculate spread factor for training data
    spread_factors = train_data.groupby('Outcome').apply(lambda x: x.mean().tolist())
    print("Spread factor for fold",fold)
   print(list(spread_factors))
    # Find correct threshold for training data
    threshold multiplier = 1.5
    thresholds = {cls: [spread_factors.loc[cls][i] * threshold_multiplier for i in range(len(spread_factors.iloc[0]))] for cls in spread_
    print("Thresholds for fold",fold)
    print(thresholds)
    def find_class(row, thresholds):
        max_distance = float('-inf')
        predicted_class = 'Unknown'
        for cls, class_thresholds in thresholds.items():
            distance = sum(1 for i, value in enumerate(row) if value < class_thresholds[i])</pre>
            if distance > max_distance:
                max_distance = distance
               predicted_class = cls
        return predicted_class
    predictions = test_data.drop(columns='Outcome').apply(lambda row: find_class(row, thresholds), axis=1)
   print("Predictions")
    print(predictions)
    # Evaluate the predictions
    test_labels = test_data['Outcome']
    accuracy = sum(1 for pred, label in zip(predictions, test_labels) if pred == label) / len(test_labels)
    accuracies.append(accuracy)
    predi.append(predictions)
    acc=max(accuracies)
    print()
```



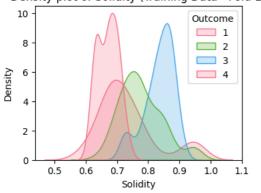




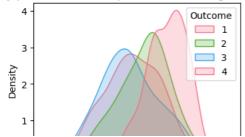
# Density plot of Roundness (Training Data - Fold 1)

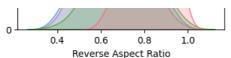


## Density plot of Solidity (Training Data - Fold 1)

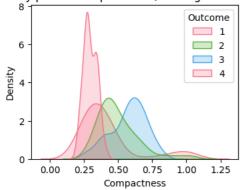


### Density plot of Reverse Aspect Ratio (Training Data - Fold 1)

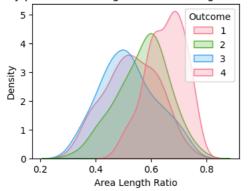




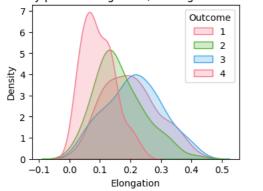




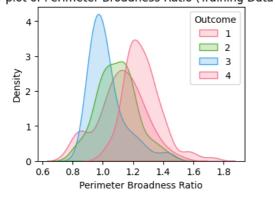
Density plot of Area Length Ratio (Training Data - Fold 1)



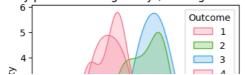
Density plot of Elongation (Training Data - Fold 1)

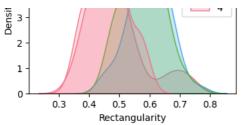


Density plot of Perimeter Broadness Ratio (Training Data - Fold 1)

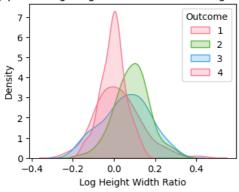


Density plot of Rectangularity (Training Data - Fold 1)

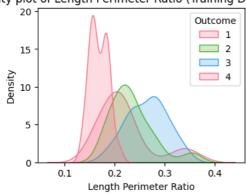




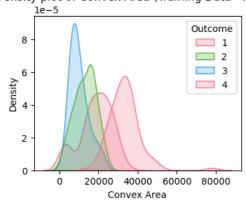
Density plot of Log Height Width Ratio (Training Data - Fold 1)



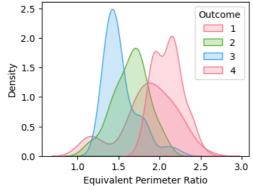
Density plot of Length Perimeter Ratio (Training Data - Fold 1)



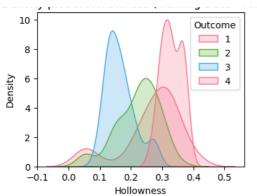
Density plot of Convex Area (Training Data - Fold 1)



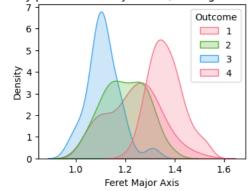
Density plot of Equivalent Perimeter Ratio (Training Data - Fold 1)



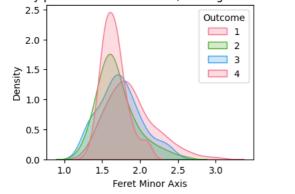
Density plot of Hollowness (Training Data - Fold 1)



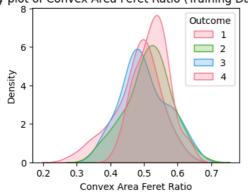
Density plot of Feret Major Axis (Training Data - Fold 1)



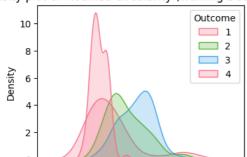
Density plot of Feret Minor Axis (Training Data - Fold 1)

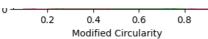


Density plot of Convex Area Feret Ratio (Training Data - Fold 1)

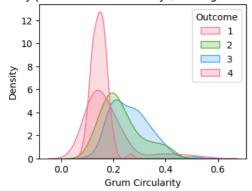


Density plot of Modified Circularity (Training Data - Fold 1)

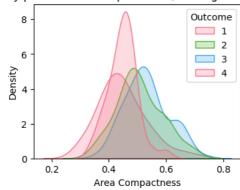








### Density plot of Area Compactness (Training Data - Fold 1)



Spread factor for fold 1

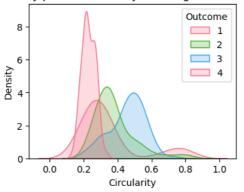
[[0.23187654320987655, 1.2181975308641977, 0.8302716049382716, 0.6704074074074075, 0.8302839506172838, 0.2952, 0.6521024691358025, 0.70 Thresholds for fold 1

 $\{1: [0.3478148148148148, 1.8272962962962964, 1.2454074074074073, 1.0056111111111112, 1.2454259259259257, 0.4428, 0.9781537037037038, Predictions \}$ 

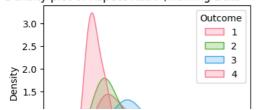
420

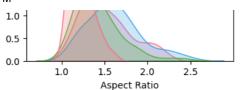
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### Density plot of Circularity (Training Data - Fold 2)

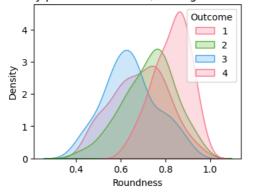


### Density plot of Aspect Ratio (Training Data - Fold 2)

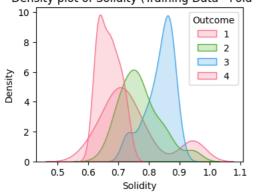




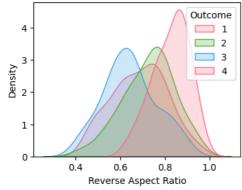
Density plot of Roundness (Training Data - Fold 2)



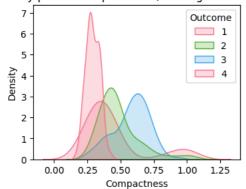
Density plot of Solidity (Training Data - Fold 2)



Density plot of Reverse Aspect Ratio (Training Data - Fold 2)

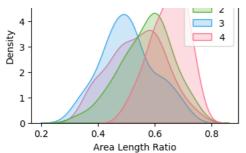


Density plot of Compactness (Training Data - Fold 2)

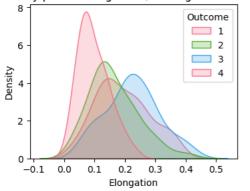


Density plot of Area Length Ratio (Training Data - Fold 2)

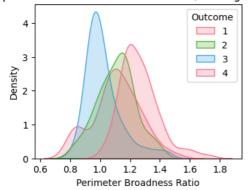




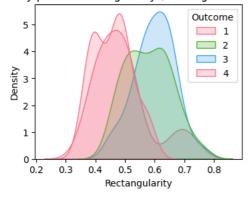
Density plot of Elongation (Training Data - Fold 2)



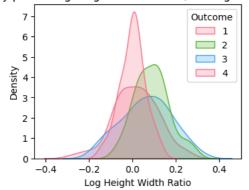
Density plot of Perimeter Broadness Ratio (Training Data - Fold 2)



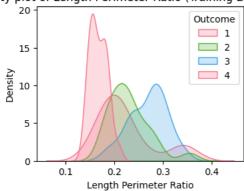
Density plot of Rectangularity (Training Data - Fold 2)



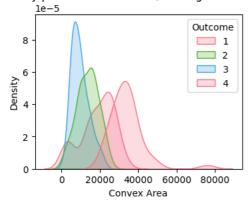
Density plot of Log Height Width Ratio (Training Data - Fold 2)



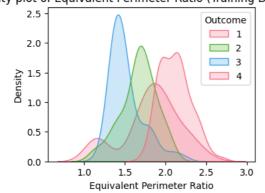
Density plot of Length Perimeter Ratio (Training Data - Fold 2)



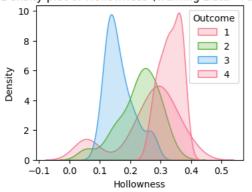
Density plot of Convex Area (Training Data - Fold 2)



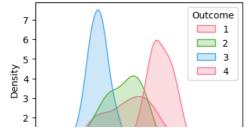
Density plot of Equivalent Perimeter Ratio (Training Data - Fold 2)

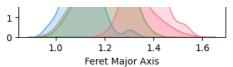


Density plot of Hollowness (Training Data - Fold 2)

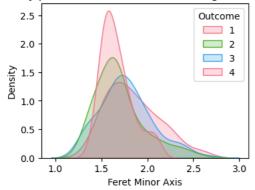


Density plot of Feret Major Axis (Training Data - Fold 2)

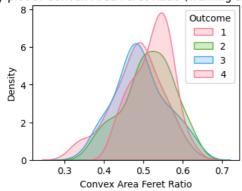




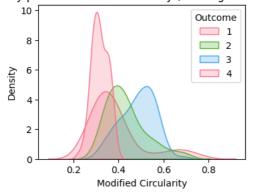
Density plot of Feret Minor Axis (Training Data - Fold 2)



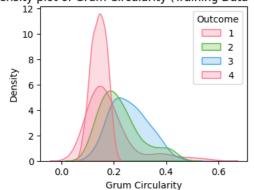
Density plot of Convex Area Feret Ratio (Training Data - Fold 2)



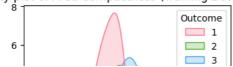
Density plot of Modified Circularity (Training Data - Fold 2)

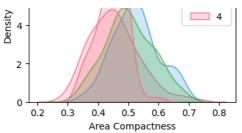


Density plot of Grum Circularity (Training Data - Fold 2)



Density plot of Area Compactness (Training Data - Fold 2)



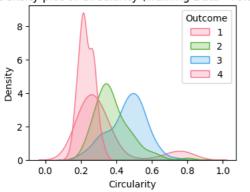


Spread factor for fold 2 [[0.22846341463414632, 1.2224146341463415, 0.8266341463414633, 0.6679999999999, 0.8266560975609757, 0.29083292682926826, 0.649252 Thresholds for fold 2

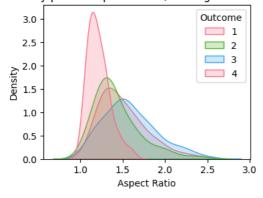
{1: [0.3426951219512195, 1.8336219512195124, 1.239951219512195, 1.0019999999999, 1.2399841463414636, 0.4362493902439024, 0.973878 Predictions

Length: 85, dtype: int64

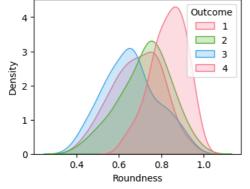
### Density plot of Circularity (Training Data - Fold 3)



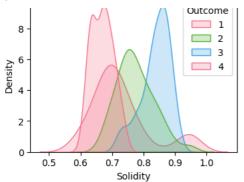
### Density plot of Aspect Ratio (Training Data - Fold 3)



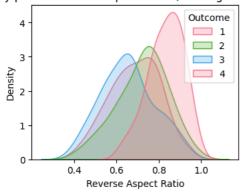
# Density plot of Roundness (Training Data - Fold 3)



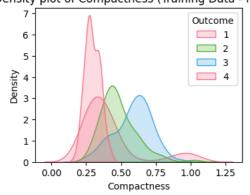
# Density plot of Solidity (Training Data - Fold 3)



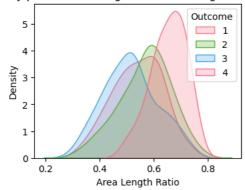
Density plot of Reverse Aspect Ratio (Training Data - Fold 3)



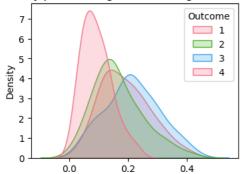
Density plot of Compactness (Training Data - Fold 3)



Density plot of Area Length Ratio (Training Data - Fold 3)

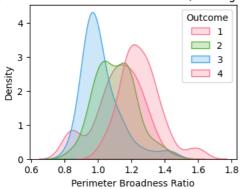


Density plot of Elongation (Training Data - Fold 3)

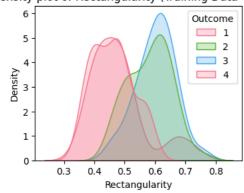


Elongation

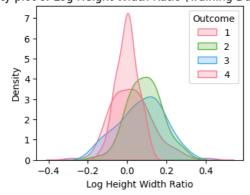
Density plot of Perimeter Broadness Ratio (Training Data - Fold 3)



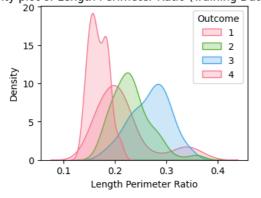
Density plot of Rectangularity (Training Data - Fold 3)



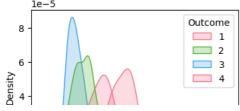
Density plot of Log Height Width Ratio (Training Data - Fold 3)

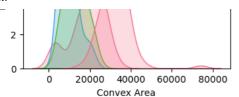


Density plot of Length Perimeter Ratio (Training Data - Fold 3)

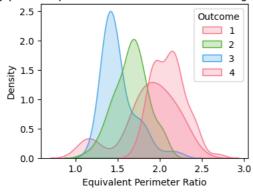


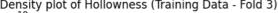
Density plot of Convex Area (Training Data - Fold 3)

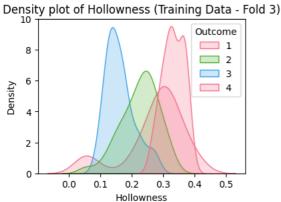




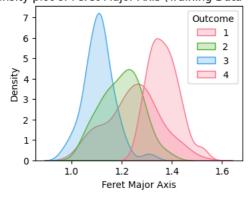
Density plot of Equivalent Perimeter Ratio (Training Data - Fold 3)



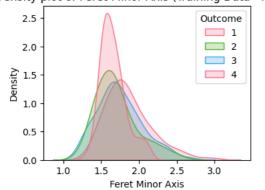




Density plot of Feret Major Axis (Training Data - Fold 3)

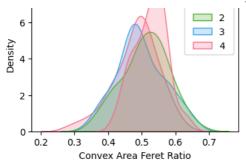


Density plot of Feret Minor Axis (Training Data - Fold 3)

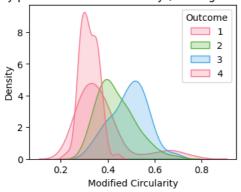


Density plot of Convex Area Feret Ratio (Training Data - Fold 3)

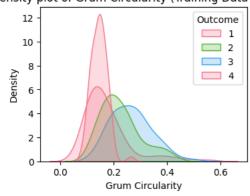




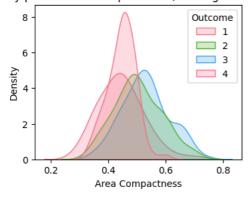
### Density plot of Modified Circularity (Training Data - Fold 3)



### Density plot of Grum Circularity (Training Data - Fold 3)



### Density plot of Area Compactness (Training Data - Fold 3)

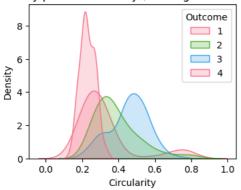


Spread factor for fold 3 [[0.23190789473684215, 1.2138026315789474, 0.8326184210526316, 0.6686578947368421, 0.8326092105263158, 0.29524736842105265, 0.653928 Thresholds for fold 3

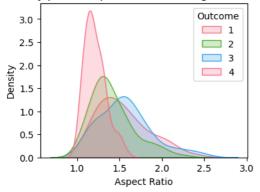
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```
2
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11
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401
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404
405
408
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418
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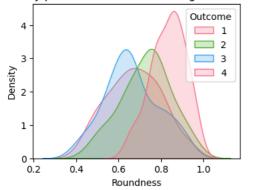
Density plot of Circularity (Training Data - Fold 4)



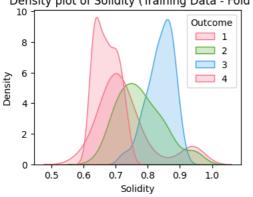
Density plot of Aspect Ratio (Training Data - Fold 4)



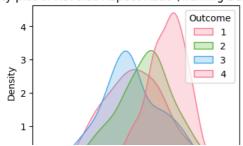
Density plot of Roundness (Training Data - Fold 4)

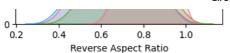


Density plot of Solidity (Training Data - Fold 4)

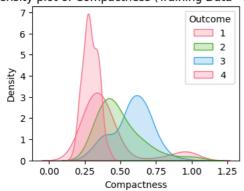


Density plot of Reverse Aspect Ratio (Training Data - Fold 4)

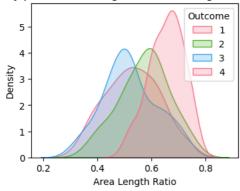




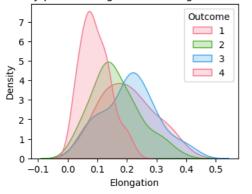




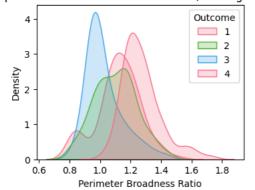
Density plot of Area Length Ratio (Training Data - Fold 4)



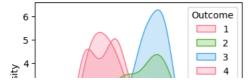
Density plot of Elongation (Training Data - Fold 4)

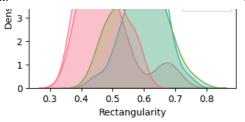


Density plot of Perimeter Broadness Ratio (Training Data - Fold 4)

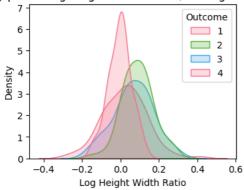


Density plot of Rectangularity (Training Data - Fold 4)

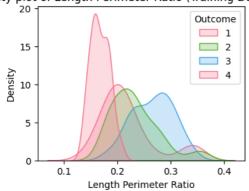




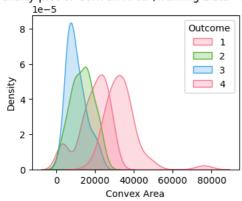
Density plot of Log Height Width Ratio (Training Data - Fold 4)



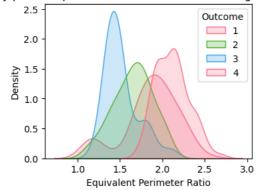
Density plot of Length Perimeter Ratio (Training Data - Fold 4)



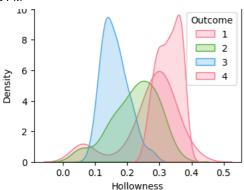
Density plot of Convex Area (Training Data - Fold 4)



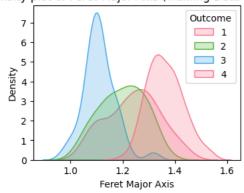
Density plot of Equivalent Perimeter Ratio (Training Data - Fold 4)



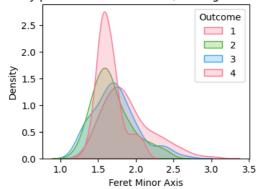
Density plot of Hollowness (Training Data - Fold 4)



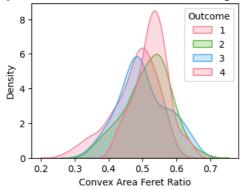
Density plot of Feret Major Axis (Training Data - Fold 4)



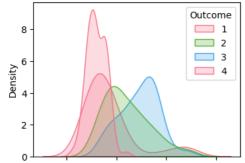
Density plot of Feret Minor Axis (Training Data - Fold 4)



Density plot of Convex Area Feret Ratio (Training Data - Fold 4)

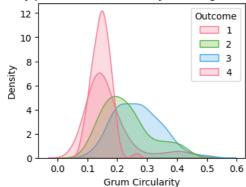


Density plot of Modified Circularity (Training Data - Fold 4)

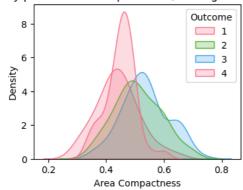


0.2 0.4 0.6 0.8 Modified Circularity





### Density plot of Area Compactness (Training Data - Fold 4)



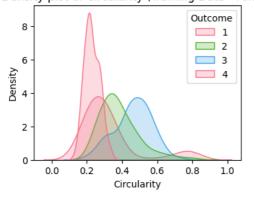
Spread factor for fold 4

[[0.2295421686746988, 1.2127108433734939, 0.8328674698795182, 0.6715180722891567, 0.8328506024096385, 0.2921915662650602, 0.65411927 Thresholds for fold 4

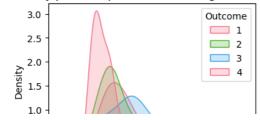
411

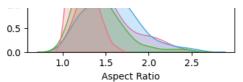
413 2 Length: 85, dtype: int64

### Density plot of Circularity (Training Data - Fold 5)

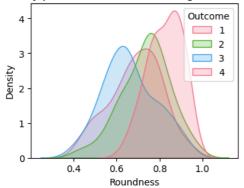


### Density plot of Aspect Ratio (Training Data - Fold 5)

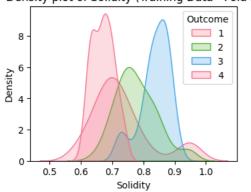




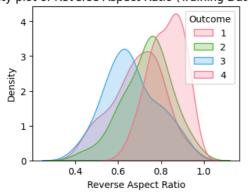




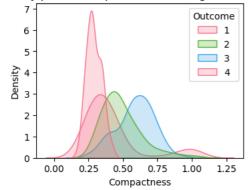
Density plot of Solidity (Training Data - Fold 5)



Density plot of Reverse Aspect Ratio (Training Data - Fold 5)

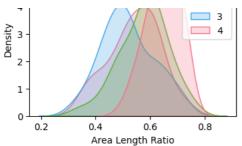


Density plot of Compactness (Training Data - Fold 5)

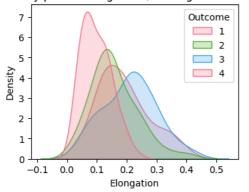


Density plot of Area Length Ratio (Training Data - Fold 5)

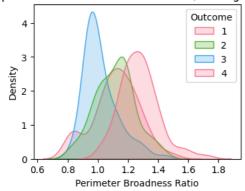




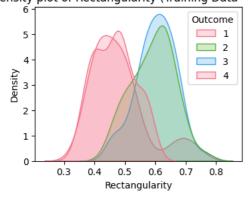
Density plot of Elongation (Training Data - Fold 5)



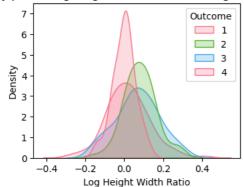
Density plot of Perimeter Broadness Ratio (Training Data - Fold 5)



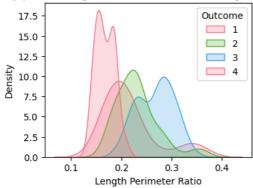
Density plot of Rectangularity (Training Data - Fold 5)



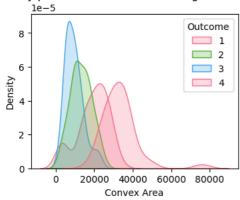
Density plot of Log Height Width Ratio (Training Data - Fold 5)



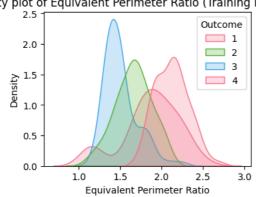
Density plot of Length Perimeter Ratio (Training Data - Fold 5)



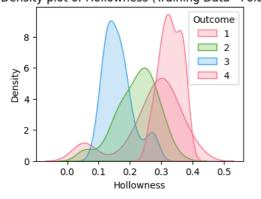
Density plot of Convex Area (Training Data - Fold 5)



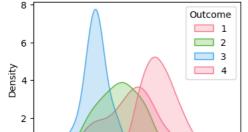
Density plot of Equivalent Perimeter Ratio (Training Data - Fold 5)

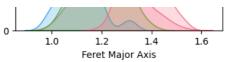


Density plot of Hollowness (Training Data - Fold 5)

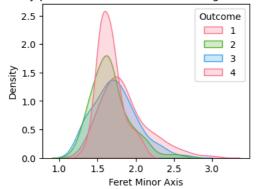


Density plot of Feret Major Axis (Training Data - Fold 5)

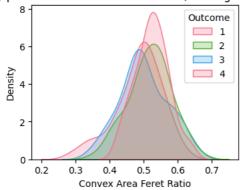




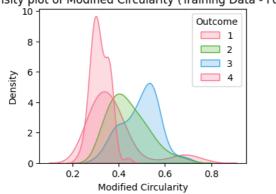
Density plot of Feret Minor Axis (Training Data - Fold 5)



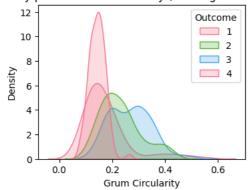
Density plot of Convex Area Feret Ratio (Training Data - Fold 5)



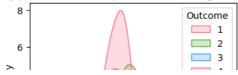
Density plot of Modified Circularity (Training Data - Fold 5)



Density plot of Grum Circularity (Training Data - Fold 5)



Density plot of Area Compactness (Training Data - Fold 5)



```
2 - 0.4 0.6 0.8 Area Compactness
```

```
Spread factor for fold 5
           [[0.2281923076923077, 1.2242564102564104, 0.8250384615384615, 0.6705000000000002, 0.8250346153846152, 0.29052051282051283, 0.6479807 Thresholds for fold 5
            \{1: [0.34228846153846154, \ 1.8363846153846155, \ 1.2375576923076923, \ 1.0057500000000004, \ 1.2375519230769227, \ 0.43578076923076925, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.9711323076923, \ 0.97113230076923, \ 0.97113230076923, \ 0.97113230076923, \ 0.97113230076923, \ 0
           Predictions
            1
                           1
            13
                           2
            20
                           2
            21
                           3
            34
                           1
           409
                           2
           414
                           2
           417
                           1
            421
                           1
            423
           Length: 84, dtype: int64
# Print the average accuracy across all folds
print(f"Overall Accuracy: {acc*100:.3f} %")
            Overall Accuracy: 60.714 %
pred_ind=accuracies.index(acc)
final_prediction=predi[pred_ind]
# Initialize dictionaries to store class-wise counts
correct_counts = {cls: 0 for cls in set(test_labels)}
total_counts = {cls: 0 for cls in set(test_labels)}
# Calculate overall accuracy and class-wise counts
for pred, label in zip(final_prediction, test_labels):
         if pred == label:
                 correct_counts[label] += 1
         total_counts[label] += 1
# Calculate class-wise accuracies
class_accuracies = {cls: correct_counts[cls] / total_counts[cls] if total_counts[cls] != 0 else 0 for cls in correct_counts}
print("Class-wise Accuracies:")
for cls, accuracy in class_accuracies.items():
         print(f"{cls}: {accuracy*100:.3f} %")
print("Accuracy for 1 and 2: ",(correct\_counts[1]+correct\_counts[2])/(total\_counts[1]+total\_counts[2])*100," \ \%")
print("Accuracy for 3 and 4: ",(correct_counts[3]+correct_counts[4])/(total_counts[3]+total_counts[4])*100," %")
           Class-wise Accuracies:
           1: 90.909 %
            2: 72.222 %
           3: 42.308 %
           4: 38.889 %
           Accuracy for 1 and 2: 82.5 %
           Accuracy for 3 and 4: 40.909090909090914 %
```

```
#Seperate dataset for common lambsquarters and common purslane
# Step 1: Load the dataset
data = pd.read_excel("direct1and2 (2).xlsx")
\# Step 2: Set up k-fold cross-validation
kf = KFold(n_splits=5, shuffle=True, random_state=42)
\# Step 3: Perform k-fold cross-validation
accuracies = []
for fold, (train_index, test_index) in enumerate(kf.split(data), 1):
   train_data, test_data = data.iloc[train_index].copy(), data.iloc[test_index].copy()
   # Generate density plots for training data
   features = train_data.drop(columns=['Outcome'])
   # Calculate spread factor for training data
   spread_factors = train_data.groupby('Outcome').apply(lambda x: x.mean().tolist())
   # Find correct threshold for training data
    threshold_multiplier = 1.5
    thresholds = {cls: [spread_factors.loc[cls][i] * threshold_multiplier for i in range(len(spread_factors.iloc[0]))] for cls in spread_factors.
    def find_class(row, thresholds):
       max distance = float('-inf')
       predicted_class = 'Unknown'
       for cls, class_thresholds in thresholds.items():
            distance = sum(1 for i, value in enumerate(row) if value < class_thresholds[i])</pre>
            if distance > max_distance:
                max distance = distance
                predicted_class = cls
        return predicted_class
   predictions = test_data.drop(columns='Outcome').apply(lambda row: find_class(row, thresholds), axis=1)
   # Evaluate the predictions
   test labels = test data['Outcome']
   accuracy = sum(1 for pred, label in zip(predictions, test_labels) if pred == label) / len(test_labels)
    accuracies.append(accuracy)
   acc=max(accuracies)
# Print the average accuracy across all folds
print(f"Accuracy: {acc*100:.3f} %")
     Accuracy: 97.561 %
```

```
#Seperate dataset for horseweed and redroot pigweed
# Step 1: Load the dataset
data = pd.read_excel("newww_weed1 (4).xlsx")
\# Step 2: Set up k-fold cross-validation
kf = KFold(n_splits=5, shuffle=True, random_state=42)
\# Step 3: Perform k-fold cross-validation
accuracies = []
for fold, (train_index, test_index) in enumerate(kf.split(data), 1):
   train_data, test_data = data.iloc[train_index].copy(), data.iloc[test_index].copy()
   # Generate density plots for training data
   features = train_data.drop(columns=['Outcome'])
   # Calculate spread factor for training data
   spread_factors = train_data.groupby('Outcome').apply(lambda x: x.mean().tolist())
   # Find correct threshold for training data
    threshold_multiplier = 1.5
    thresholds = {cls: [spread_factors.loc[cls][i] * threshold_multiplier for i in range(len(spread_factors.iloc[0]))] for cls in spread_factors.
    def find_class(row, thresholds):
       max distance = float('-inf')
       predicted_class = 'Unknown'
       for cls, class_thresholds in thresholds.items():
            distance = sum(1 for i, value in enumerate(row) if value < class_thresholds[i])</pre>
            if distance > max_distance:
                max distance = distance
                predicted_class = cls
        return predicted_class
   predictions = test_data.drop(columns='Outcome').apply(lambda row: find_class(row, thresholds), axis=1)
   # Evaluate the predictions
   test labels = test data['Outcome']
   accuracy = sum(1 for pred, label in zip(predictions, test_labels) if pred == label) / len(test_labels)
    accuracies.append(accuracy)
   acc=max(accuracies)
# Print the average accuracy across all folds
print(f"Accuracy: {acc*100:.3f} %")
     Accuracy: 88.636 %
```