pr-9

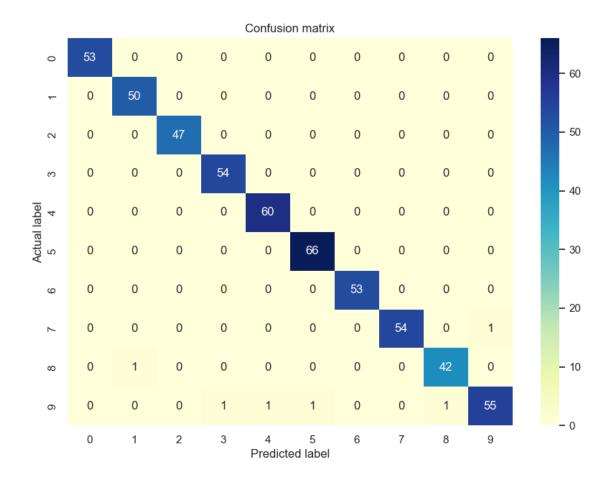
October 15, 2023

1 Classification algorithms: KNN, Logistic Regression, Decision Tree, Random Forest, SVM, Naive Bayes, Neural Network.

```
[17]: # Import necessary libraries
      from sklearn import datasets
      from sklearn.model_selection import train_test_split
      from sklearn.neighbors import KNeighborsClassifier
      from sklearn.metrics import confusion matrix, accuracy score
      import matplotlib.pyplot as plt
      import seaborn as sns; sns.set() # for plot styling
      import pandas as pd
      # Load iris dataset
      iris = datasets.load_digits()
      X = iris.data
      y = iris.target
      # Split dataset into training set and test set
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,__
       →random_state=42) # 70% training and 30% testing
      # Create KNN Classifier
      knn = KNeighborsClassifier(n_neighbors=3) # 3 is an example, you might want to⊔
       \hookrightarrowuse cross-validation to find the best parameter
      # Train the model using the training sets
      knn.fit(X_train, y_train)
      # Predict the response for test dataset
      y_pred = knn.predict(X_test)
      # Model evaluation
      conf_matrix = confusion_matrix(y_test, y_pred)
      accuracy = accuracy_score(y_test, y_pred)
      print(f"Confusion Matrix:\n{conf_matrix}")
      print(f"Accuracy: {accuracy}")
```

```
[[53 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 50 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 47 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 54 0 0 0 0 0 0 0 0]
[ 0 0 0 0 66 0 0 0 0 0 0]
[ 0 0 0 0 0 66 0 0 0 0 0 0]
[ 0 0 0 0 0 0 53 0 0 0]
[ 0 0 0 0 0 0 0 54 0 1]
[ 0 1 0 0 0 0 0 0 0 42 0]
[ 0 0 0 1 1 1 1 0 0 1 55]
```

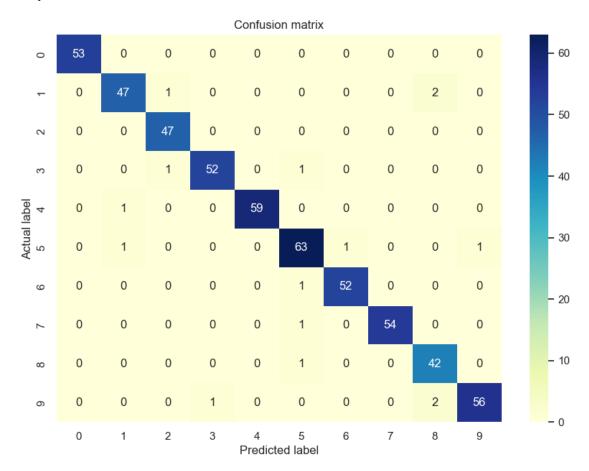
Accuracy: 0.988888888888889



```
[18]: # Import necessary libraries
      from sklearn import datasets
      from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LogisticRegression
      from sklearn.metrics import confusion_matrix, accuracy_score
      import matplotlib.pyplot as plt
      import seaborn as sns; sns.set() # for plot styling
      import pandas as pd
      # Load iris dataset
      iris = datasets.load_digits()
      X = iris.data
      y = iris.target
      # Split dataset into training set and test set
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,__
       →random_state=42) # 70% training and 30% testing
      # Create Logistic Regression classifier
```

```
logistic_regression = LogisticRegression(max_iter=1000) # Increased max_iter_
 may be necessary as logistic regression may need more iterations to converge
# Train the model using the training sets
logistic_regression.fit(X_train, y_train)
# Predict the response for test dataset
y_pred = logistic_regression.predict(X_test)
# Model Evaluation
conf_matrix = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
print(f"Confusion Matrix:\n{conf_matrix}")
print(f"Accuracy: {accuracy}")
# Plotting confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(pd.DataFrame(conf_matrix, index=iris.target_names, columns=iris.
 →target_names), annot=True, cmap="YlGnBu", fmt='g')
plt.title('Confusion matrix')
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
plt.show()
/opt/homebrew/lib/python3.11/site-
packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: lbfgs failed
to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as shown in:
   https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
   https://scikit-learn.org/stable/modules/linear_model.html#logistic-
regression
 n_iter_i = _check_optimize_result(
Confusion Matrix:
[[53 0 0 0 0 0 0 0 0]
 [047 1 0 0 0 0 0 2
                            0]
 [0 0 47 0 0 0 0 0 0
                            01
 [0 0 1 52 0 1 0 0 0 0]
 [0 1 0 0 59 0 0 0 0
                            07
 [0 1 0 0 0 63 1 0 0
                            17
 [0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 52 \ 0 \ 0 \ 0]
 [0 0 0 0 0 1 0 54 0 0]
 [0000010420]
 [0 0 0 1 0 0 0 0 2 56]]
```

Accuracy: 0.97222222222222



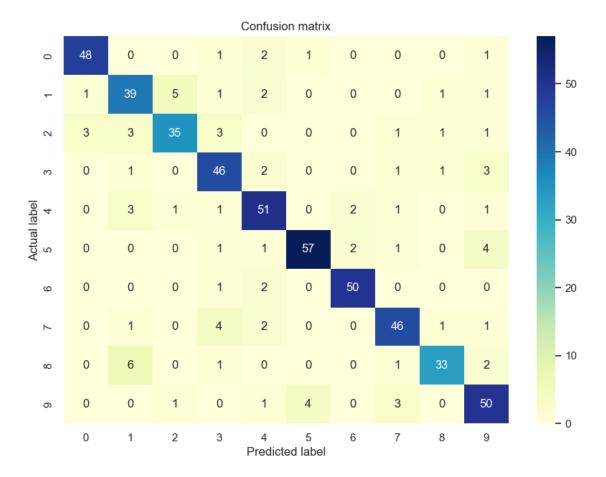
```
[20]: # Import necessary libraries
from sklearn.datasets import load_digits
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import confusion_matrix, accuracy_score
import matplotlib.pyplot as plt
from sklearn import tree
import seaborn as sns
import pandas as pd

# Load iris dataset
iris = load_digits()
X = iris.data
y = iris.target

# Split dataset into training set and test set
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,_
⊖random_state=42) # 70% training and 30% testing
# Create Decision Tree Classifier
clf = DecisionTreeClassifier(random_state=42)
# Train the model using the training sets
clf.fit(X_train, y_train)
# Predict the response for test dataset
y_pred = clf.predict(X_test)
# Model Evaluation
conf_matrix = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
print(f"Confusion Matrix:\n{conf_matrix}")
print(f"Accuracy: {accuracy}")
# Plotting confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(pd.DataFrame(conf_matrix, index=iris.target_names, columns=iris.
 →target_names), annot=True, cmap="YlGnBu", fmt='g')
plt.title('Confusion matrix')
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
plt.show()
```

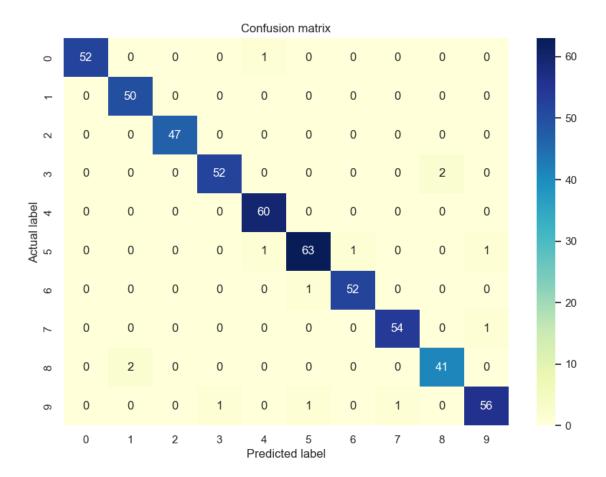
```
[[48  0  0  1  2  1  0  0  0  1]
[ 1  39  5  1  2  0  0  0  1  1]
[ 3  3  35  3  0  0  0  1  1  1]
[ 0  1  0  46  2  0  0  1  1  1]
[ 0  3  1  1  51  0  2  1  0  1]
[ 0  0  0  1  1  57  2  1  0  4]
[ 0  0  0  1  2  0  50  0  0  0]
[ 0  1  0  4  2  0  0  46  1  1]
[ 0  6  0  1  0  0  0  1  33  2]
[ 0  0  1  0  1  4  0  3  0  50]]
Accuracy: 0.8425925925925926
```



```
[21]: # Import necessary libraries
      from sklearn.datasets import load_digits
      from sklearn.model_selection import train_test_split
      from sklearn.ensemble import RandomForestClassifier
      from sklearn.metrics import confusion_matrix, accuracy_score
      import matplotlib.pyplot as plt
      import seaborn as sns
      import pandas as pd
      # Load iris dataset
      iris = load_digits()
      X = iris.data
      y = iris.target
      # Split dataset into training set and test set
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,__
       →random_state=42) # 70% training and 30% testing
      # Create a Random Forest Classifier
```

```
rf = RandomForestClassifier(n_estimators=100, random_state=42) # n_estimators_{\square}
→= number of trees in the forest
# Train the model using the training sets
rf.fit(X_train, y_train)
# Predict the response for test dataset
y_pred = rf.predict(X_test)
# Model Evaluation
conf_matrix = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
print(f"Confusion Matrix:\n{conf_matrix}")
print(f"Accuracy: {accuracy}")
# Plotting confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(pd.DataFrame(conf_matrix, index=iris.target_names, columns=iris.
⇔target_names), annot=True, cmap="YlGnBu", fmt='g')
plt.title('Confusion matrix')
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
plt.show()
```

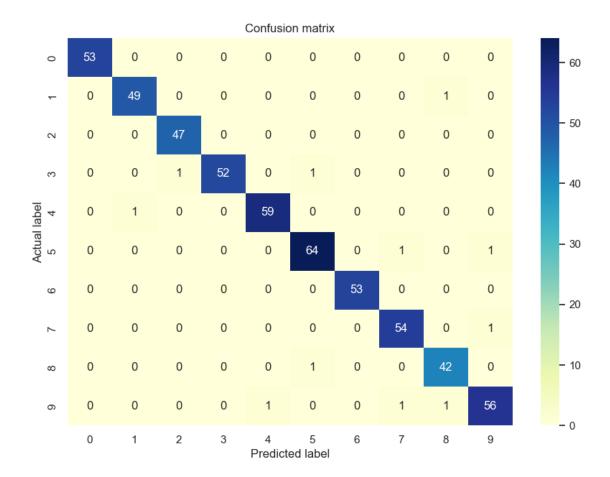
[[52 0 0 0 1 0 0 0 0 0]
[0 50 0 0 0 0 0 0 0 0 0]
[0 0 47 0 0 0 0 0 0 0 0]
[0 0 0 52 0 0 0 0 0 2 0]
[0 0 0 0 60 0 0 0 0 0 0]
[0 0 0 0 1 63 1 0 0 1]
[0 0 0 0 0 0 1 52 0 0 0]
[0 0 0 0 0 0 1 52 0 0 0]
[0 0 0 0 0 0 0 0 54 0 1]
[0 0 0 0 1 0 1 0 1 0 56]]
Accuracy: 0.975925925925926



```
[22]: # Import necessary libraries
      from sklearn.datasets import load_digits
      from sklearn.model_selection import train_test_split
      from sklearn.svm import SVC
      from sklearn.metrics import confusion_matrix, accuracy_score
      import matplotlib.pyplot as plt
      import seaborn as sns
      import pandas as pd
      # Load iris dataset
      iris = load_digits()
      X = iris.data
      y = iris.target
      # Split dataset into training set and test set
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,__
       →random_state=42) # 70% training and 30% testing
      # Create a Support Vector Machine Classifier
```

```
svc = SVC(kernel='linear') # You can try other kernels like 'rbf', 'poly', etc.
# Train the model using the training sets
svc.fit(X_train, y_train)
# Predict the response for test dataset
y_pred = svc.predict(X_test)
# Model Evaluation
conf_matrix = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
print(f"Confusion Matrix:\n{conf_matrix}")
print(f"Accuracy: {accuracy}")
# Plotting confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(pd.DataFrame(conf_matrix, index=iris.target_names, columns=iris.
 →target_names), annot=True, cmap="YlGnBu", fmt='g')
plt.title('Confusion matrix')
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
plt.show()
```

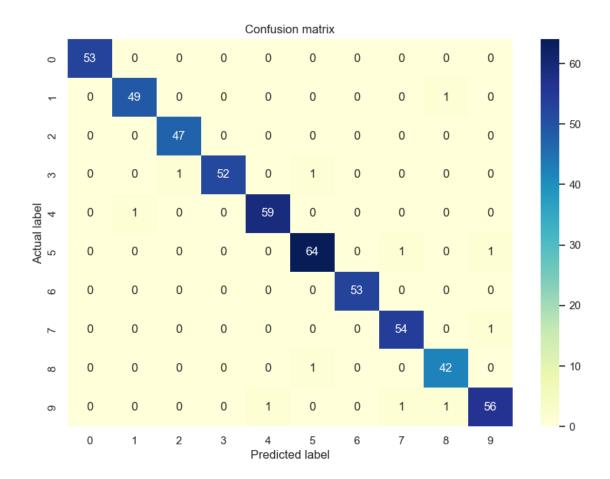
[[53 0 0 0 0 0 0 0 0 0 0 0]
[0 49 0 0 0 0 0 0 0 1 0]
[0 0 47 0 0 0 0 0 0 0 0]
[0 1 52 0 1 0 0 0 0 0]
[0 1 0 0 59 0 0 0 0 0]
[0 0 0 0 0 64 0 1 0 1]
[0 0 0 0 0 0 53 0 0 0]
[0 0 0 0 0 54 0 1]
[0 0 0 0 0 1 0 0 42 0]
[0 0 0 0 0 1 0 0 1 1 56]]
Accuracy: 0.9796296296296296



```
[23]: # Import necessary libraries
      from sklearn.datasets import load_iris
      from sklearn.model_selection import train_test_split
      from sklearn.svm import SVC
      from sklearn.metrics import confusion_matrix, accuracy_score
      import matplotlib.pyplot as plt
      import seaborn as sns
      import pandas as pd
      # Load iris dataset
      iris = load_digits()
      X = iris.data
      y = iris.target
      # Split dataset into training set and test set
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,__
       →random_state=42) # 70% training and 30% testing
      # Create a Support Vector Machine Classifier
```

```
svc = SVC(kernel='linear') # You can try other kernels like 'rbf', 'poly', etc.
# Train the model using the training sets
svc.fit(X_train, y_train)
# Predict the response for test dataset
y_pred = svc.predict(X_test)
# Model Evaluation
conf_matrix = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
print(f"Confusion Matrix:\n{conf_matrix}")
print(f"Accuracy: {accuracy}")
# Plotting confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(pd.DataFrame(conf_matrix, index=iris.target_names, columns=iris.
 →target_names), annot=True, cmap="YlGnBu", fmt='g')
plt.title('Confusion matrix')
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
plt.show()
```

[[53 0 0 0 0 0 0 0 0 0 0 0]
[0 49 0 0 0 0 0 0 0 1 0]
[0 0 47 0 0 0 0 0 0 0 0]
[0 1 52 0 1 0 0 0 0 0]
[0 1 0 0 59 0 0 0 0 0]
[0 0 0 0 0 64 0 1 0 1]
[0 0 0 0 0 0 53 0 0 0]
[0 0 0 0 0 54 0 1]
[0 0 0 0 0 1 0 0 42 0]
[0 0 0 0 0 1 0 0 1 1 56]]
Accuracy: 0.9796296296296296



```
[24]: # Import necessary libraries
      from sklearn.datasets import load_iris
      from sklearn.model_selection import train_test_split
      from sklearn.naive_bayes import GaussianNB # for Gaussian Naive Bayes
      from sklearn.metrics import confusion_matrix, accuracy_score
      import matplotlib.pyplot as plt
      import seaborn as sns
      import pandas as pd
      # Load iris dataset
      iris = load_digits()
      X = iris.data
      y = iris.target
      # Split dataset into training set and test set
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,__
       →random_state=42) # 70% training and 30% testing
      # Create a Gaussian Naive Bayes Classifier
```

```
gnb = GaussianNB()
# Train the model using the training sets
gnb.fit(X_train, y_train)
# Predict the response for test dataset
y_pred = gnb.predict(X_test)
# Model Evaluation
conf_matrix = confusion_matrix(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
print(f"Confusion Matrix:\n{conf_matrix}")
print(f"Accuracy: {accuracy}")
# Plotting confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(pd.DataFrame(conf_matrix, index=iris.target_names, columns=iris.
⇔target_names), annot=True, cmap="YlGnBu", fmt='g')
plt.title('Confusion matrix')
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
plt.show()
```

[[52 0 0 0 0 0 0 1 0 0]
[0 37 2 0 0 0 0 1 0 12 0]
[0 3 31 0 0 0 1 0 12 0]
[0 0 2 41 0 0 1 0 8 2]
[0 0 0 0 51 0 2 7 0 0]
[0 0 0 1 0 62 1 2 0 0]
[0 0 0 0 1 1 51 0 0 0]
[0 0 0 0 0 1 1 51 0 0 0]
[0 0 0 0 0 1 0 54 0 0]
[0 1 1 1 0 2 1 7 4 42]]
Accuracy: 0.8518518518519

