lab5.2

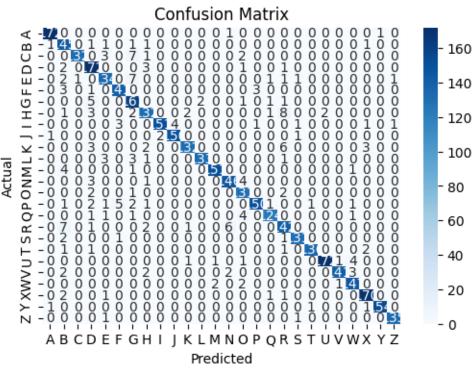
January 20, 2025

```
[1]: import numpy as np
       import pandas as pd
       from sklearn.model_selection import train_test_split
       from sklearn.svm import SVC
       from sklearn.metrics import accuracy_score, confusion_matrix,_
        aclassification_report,roc_auc_score,precision_score,recall_score,f1_score
       import matplotlib.pyplot as plt
       import seaborn as sns
       %matplotlib inline
  []: df = pd.read_csv("./letter-recognition.csv")
  [3]: from sklearn.preprocessing import StandardScaler
       from sklearn.model_selection import train_test_split
       scaler = StandardScaler()
  [4]: encoding = {chr(i): i - 64 for i in range(65, 91)}
       df['encoded_alphabet'] = df['letter'].map(encoding)
  []: df.columns = df.columns.str.strip()
[96]: X = L
        →df[['xbox','ybox','width','height','onpix','xbar','ybar','x2bar','y2bar','xybar','x2ybar',
        'xy2bar', 'xedge', 'xedgey', 'yedge', 'yedgex']]
       y= df[['encoded_alphabet']].to_numpy()
[97]: X_train, X_test, y_train, y_test = train_test_split(X, y, __

¬random_state=22053747, test_size=0.2)
[98]: y_test = y_test.ravel()
       y_train = y_train.ravel()
[100]: scaler X = StandardScaler()
       X_train_scaled = scaler.fit_transform(X_train)
[103]: | svm_clf = SVC(kernel='rbf', C=1.0, gamma='scale', random_state=42)
```

```
[104]: svm_clf.fit(X_train_scaled, y_train)
[104]: SVC(random state=42)
[106]: X test scaled = scaler.transform(X test)
[107]: y_pred = svm_clf.predict(X_test_scaled)
[108]: accuracy_score(y_pred=y_pred, y_true=y_test)
[108]: 0.94625
[112]: class_names = list(encoding.keys())
       cm = confusion_matrix(y_test, y_pred)
       plt.figure(figsize=(6, 4))
       sns.heatmap(cm, annot=True, fmt="d", cmap="Blues",
                   xticklabels=class_names, yticklabels=class_names)
       plt.xlabel("Predicted")
       plt.ylabel("Actual")
       plt.title("Confusion Matrix")
       plt.show()
```

Confusion Matrix



```
[114]: print("Classification Report:\n", classification_report(y_test, y_pred, __ otarget_names=class_names))
```

Classification Report:

| Report: | | | | |
|-----------|--|--|--|--|
| precision | recall | f1-score | support | |
| 0.98 | 0.99 | 0.99 | 174 | |
| 0.84 | 0.96 | 0.90 | 151 | |
| 0.99 | 0.91 | 0.95 | 145 | |
| 0.88 | 0.96 | 0.92 | 179 | |
| 0.92 | 0.91 | 0.91 | 148 | |
| 0.94 | 0.95 | 0.95 | 148 | |
| 0.87 | 0.94 | 0.90 | 171 | |
| 0.90 | 0.87 | 0.89 | 151 | |
| 0.99 | 0.93 | 0.96 | 162 | |
| 0.97 | 0.97 | 0.97 | 154 | |
| 0.97 | 0.90 | 0.94 | 146 | |
| 0.99 | 0.94 | 0.96 | 141 | |
| 0.98 | 0.96 | 0.97 | 157 | |
| 0.94 | 0.95 | 0.94 | 154 | |
| 0.90 | 0.96 | 0.93 | 136 | |
| 0.97 | 0.91 | 0.94 | 165 | |
| 0.96 | 0.94 | 0.95 | 132 | |
| 0.86 | 0.89 | 0.88 | 159 | |
| 0.97 | 0.97 | 0.97 | 135 | |
| 0.99 | 0.96 | 0.97 | 141 | |
| 0.99 | 0.96 | 0.97 | 180 | |
| 0.99 | 0.94 | 0.96 | 150 | |
| 0.94 | 0.97 | 0.95 | 152 | |
| 0.96 | 0.97 | 0.96 | 175 | |
| 0.99 | 0.98 | 0.99 | 157 | |
| 0.99 | 0.99 | 0.99 | 137 | |
| | | 0.95 | 4000 | |
| 0.95 | 0.95 | 0.95 | 4000 | |
| 0.95 | 0.95 | 0.95 | 4000 | |
| | 0.98 0.84 0.99 0.88 0.92 0.94 0.87 0.90 0.97 0.99 0.98 0.94 0.90 0.97 0.96 0.86 0.97 0.99 0.99 0.99 0.99 0.99 | precision recall 0.98 0.99 0.84 0.96 0.99 0.91 0.88 0.96 0.92 0.91 0.94 0.95 0.87 0.94 0.90 0.87 0.99 0.93 0.97 0.97 0.99 0.94 0.99 0.94 0.98 0.96 0.97 0.91 0.96 0.94 0.86 0.89 0.97 0.97 0.99 0.96 0.99 0.96 0.99 0.94 0.99 0.94 0.99 0.94 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 0.99 | precision recall f1-score 0.98 0.99 0.99 0.84 0.96 0.90 0.99 0.91 0.95 0.88 0.96 0.92 0.92 0.91 0.91 0.92 0.91 0.91 0.92 0.91 0.91 0.92 0.91 0.91 0.92 0.91 0.91 0.94 0.95 0.95 0.87 0.99 0.93 0.99 0.93 0.96 0.97 0.97 0.97 0.99 0.94 0.96 0.94 0.95 0.94 0.90 0.96 0.93 0.97 0.91 0.94 0.96 0.93 0.94 0.96 0.97 0.97 0.99 0.96 0.97 0.99 0.96 0.97 0.99 0.96 0.97 0.99 0.99 0.99 | |

```
[115]: from sklearn.model_selection import GridSearchCV

param_grid = {
    'C': [0.1, 1, 10],
    'gamma': ['scale', 1e-2, 1e-3],
    'kernel': ['rbf', 'poly']
}

grid_search = GridSearchCV(
```

```
estimator=SVC(random_state=42),
           param_grid=param_grid,
           scoring='accuracy',
           cv=5,
           n_{jobs=-1}
       grid_search.fit(X_train_scaled, y_train)
       print("Best params:", grid_search.best_params_)
       print("Best CV score:", grid_search.best_score_)
      Best params: {'C': 10, 'gamma': 'scale', 'kernel': 'rbf'}
      Best CV score: 0.9669375
[140]: print(precision)
       print(f1)
      print(recall)
      0.9481826224448482
      0.9466014788213704
      0.94625
[135]: print(classification_report(y_test, y_pred))
       classification_report
```

| | precision | recall | f1-score | support |
|----|-----------|--------|----------|---------|
| 1 | 0.98 | 0.99 | 0.99 | 174 |
| 2 | 0.84 | 0.96 | 0.90 | 151 |
| 3 | 0.99 | 0.91 | 0.95 | 145 |
| 4 | 0.88 | 0.96 | 0.92 | 179 |
| 5 | 0.92 | 0.91 | 0.91 | 148 |
| 6 | 0.94 | 0.95 | 0.95 | 148 |
| 7 | 0.87 | 0.94 | 0.90 | 171 |
| 8 | 0.90 | 0.87 | 0.89 | 151 |
| 9 | 0.99 | 0.93 | 0.96 | 162 |
| 10 | 0.97 | 0.97 | 0.97 | 154 |
| 11 | 0.97 | 0.90 | 0.94 | 146 |
| 12 | 0.99 | 0.94 | 0.96 | 141 |
| 13 | 0.98 | 0.96 | 0.97 | 157 |
| 14 | 0.94 | 0.95 | 0.94 | 154 |
| 15 | 0.90 | 0.96 | 0.93 | 136 |
| 16 | 0.97 | 0.91 | 0.94 | 165 |
| 17 | 0.96 | 0.94 | 0.95 | 132 |
| 18 | 0.86 | 0.89 | 0.88 | 159 |
| 19 | 0.97 | 0.97 | 0.97 | 135 |
| 20 | 0.99 | 0.96 | 0.97 | 141 |
| 21 | 0.99 | 0.96 | 0.97 | 180 |

| | 22 | 0.99 | 0.94 | 0.96 | 150 |
|------------|----|------|------|------|------|
| : | 23 | 0.94 | 0.97 | 0.95 | 152 |
| : | 24 | 0.96 | 0.97 | 0.96 | 175 |
| : | 25 | 0.99 | 0.98 | 0.99 | 157 |
| : | 26 | 0.99 | 0.99 | 0.99 | 137 |
| | | | | | |
| accura | су | | | 0.95 | 4000 |
| macro a | vg | 0.95 | 0.95 | 0.95 | 4000 |
| weighted a | vg | 0.95 | 0.95 | 0.95 | 4000 |

[135]: <function sklearn.metrics._classification.classification_report(y_true, y_pred,
 *, labels=None, target_names=None, sample_weight=None, digits=2,
 output_dict=False, zero_division='warn')>