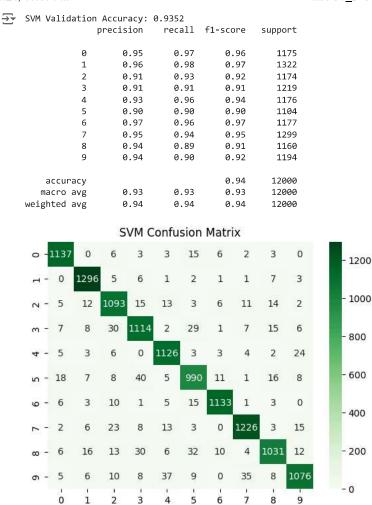
```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from \ sklearn.metrics \ import \ accuracy\_score, \ classification\_report, \ confusion\_matrix
import tensorflow as tf
from tensorflow.keras.datasets import mnist
(X_train, y_train), (X_test, y_test) = mnist.load_data()
print(f'Training data shape: {X train.shape}')
print(f'Test data shape: {X_test.shape}')
 Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz">https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz</a>
     11490434/11490434 -
                                                0s Ous/step
     Training data shape: (60000, 28, 28)
     Test data shape: (10000, 28, 28)
plt.figure(figsize=(10, 5))
for i in range(10):
    plt.subplot(2, 5, i + 1)
    plt.imshow(X_train[i], cmap='gray')
    plt.title(f"Label: {y_train[i]}")
    plt.axis('off')
plt.show()
X_{\text{train\_flat}} = X_{\text{train.reshape}}(X_{\text{train.shape}}[0], -1) / 255.0
X_{\text{test_flat}} = X_{\text{test.reshape}}(X_{\text{test.shape}}[0], -1) / 255.0
print(f'Flattened training data shape: {X_train_flat.shape}')
print(f'Flattened test data shape: {X test flat.shape}')
X_train_split, X_val, y_train_split, y_val = train_test_split(X_train_flat, y_train, test_size=0.2, random_state=42)
print(f'Training set shape: {X_train_split.shape}')
print(f'Validation set shape: {X_val.shape}')
\rightarrow
                                  Label: 0
                                                         Label: 4
                                                                                Label: 1
                                                                                                       Label: 9
            Label: 5
                                                                                Label: 1
            Label: 2
                                  Label: 1
                                                         Label: 3
                                                                                                       Label: 4
     Flattened training data shape: (60000, 784)
     Flattened test data shape: (10000, 784)
     Training set shape: (48000, 784)
```

```
knn_model = KNeighborsClassifier(n_neighbors=3)
knn_model.fit(X_train_split, y_train_split)
y_val_pred_knn = knn_model.predict(X_val)
knn_acc = accuracy_score(y_val, y_val_pred_knn)
print(f'KNN Validation Accuracy: {knn_acc:.4f}')
print(classification_report(y_val, y_val_pred_knn))
sns.heatmap(confusion_matrix(y_val, y_val_pred_knn), annot=True, fmt='d', cmap='Blues')
plt.title("KNN Confusion Matrix")
plt.show()
→ KNN Validation Accuracy: 0.9727
                                 recall f1-score
                   precision
                                                     support
                0
                         0.98
                                   0.99
                                              0.99
                                                        1175
                1
                         0.96
                                   1.00
                                              0.98
                                                        1322
                2
                         0.98
                                   0.97
                                              0.97
                                                        1174
                3
                         0.97
                                   0.97
                                              0.97
                                                        1219
                4
                         0.97
                                   0.97
                                              0.97
                                                        1176
                5
                         0.97
                                   0.97
                                              0.97
                                                        1104
                6
                         0.99
                                   0.99
                                              0.99
                                                        1177
                7
                         0.97
                                   0.98
                                              0.97
                                                        1299
                8
                         9.99
                                   9.94
                                              9.96
                                                        1160
                9
                         0.96
                                   0.96
                                              0.96
                                                        1194
                                              0.97
                                                       12000
         accuracy
                                   0.97
                                                       12000
                         0.97
                                              9.97
        macro avg
     weighted avg
                         0.97
                                   0.97
                                              0.97
                                                       12000
                        KNN Confusion Matrix
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                                            2
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                                                       1
                                                            1
                                                                       1200
            0
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                     1133
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                                                       2
                 11
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      m - 1
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            4
                 4
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                            2
                                 18
                                                 13
                                                       1
            0
                            3
                                            6
                                                       8
                                                            9
svm_model = SVC(kernel='linear')
svm_model.fit(X_train_split, y_train_split)
y_val_pred_svm = svm_model.predict(X_val)
svm_acc = accuracy_score(y_val, y_val_pred_svm)
print(f'SVM Validation Accuracy: {svm_acc:.4f}')
print(classification_report(y_val, y_val_pred_svm))
sns.heatmap(confusion\_matrix(y\_val,\ y\_val\_pred\_svm),\ annot=True,\ fmt='d',\ cmap='Greens')
plt.title("SVM Confusion Matrix")
plt.show()
```



y\_test\_pred\_svm = svm\_model.predict(X\_test\_flat)
test\_acc\_svm = accuracy\_score(y\_test, y\_test\_pred\_svm)
print(f'SVM Test Accuracy: {test\_acc\_svm:.4f}')

→ SVM Test Accuracy: 0.9351