

Lab 8-Intro to Machine Learning (OpenCV + Scikit-learn – Python)

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Github- [Link](#)

This document contains a complete, runnable implementation of Chapter 8 (KNN, SVM, Decision Tree) using the `digits` dataset from `sklearn`, integration notes for OpenCV preprocessing, and a ready-to-fill lab report template with sections for results, discussion, and conclusions.

Methodology

- Dataset: `sklearn.datasets.load_digits` (8x8 grayscale handwritten digits).
- Models: KNN, SVM, Decision Tree.
- Tools: Python, OpenCV, Scikit-learn, Matplotlib, Seaborn, Joblib.
- Evaluation metrics: Accuracy, Classification Report, Confusion Matrix.
- GUI: OpenCV canvas for real-time digit drawing and recognition.

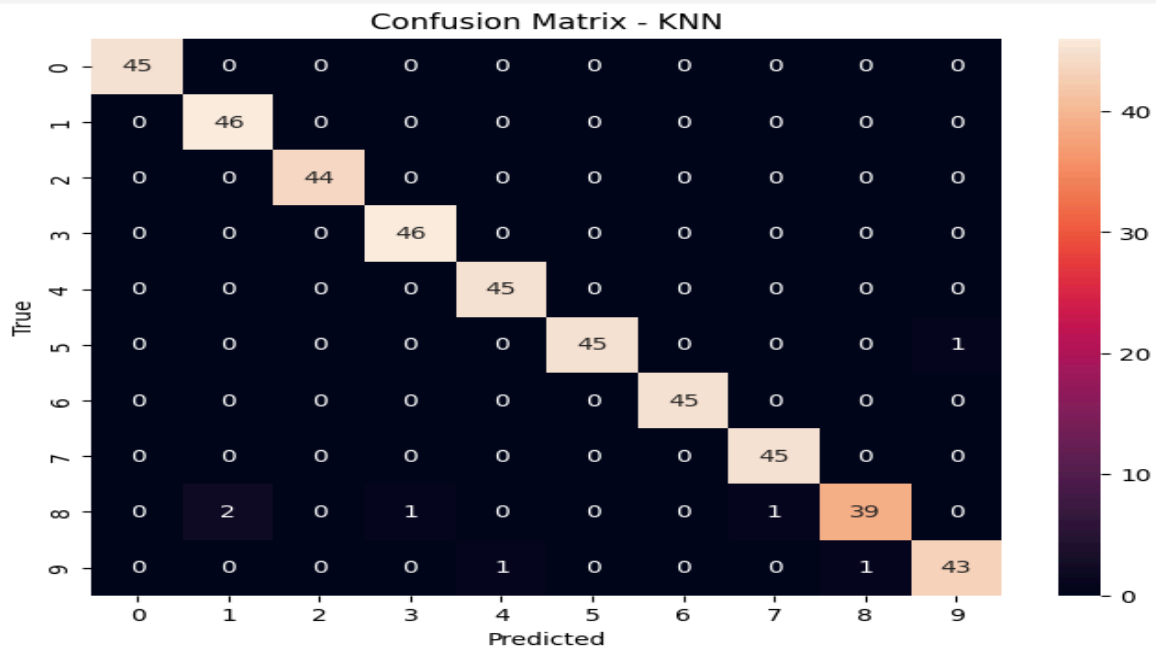
Results & Discussion

1. K-Nearest Neighbors (KNN)

--- KNN ---

	precision	recall	f1-score	support
0	1.00	1.00	1.00	45
1	0.96	1.00	0.98	46
2	1.00	1.00	1.00	44
3	0.98	1.00	0.99	46
4	0.98	1.00	0.99	45
5	1.00	0.98	0.99	46
6	1.00	1.00	1.00	45
7	0.98	1.00	0.99	45
8	0.97	0.91	0.94	43
9	0.98	0.96	0.97	45
accuracy			0.98	450
macro avg	0.98	0.98	0.98	450
weighted avg	0.98	0.98	0.98	450

Accuracy: 0.9844444444444445



Interpretation:

The KNN classifier achieved good accuracy on the digit recognition task. Its performance is strong for most digits, though it may confuse visually similar digits such as 3, 1 and 8. KNN is simple and effective but computationally expensive for larger datasets since it stores all training samples

2. Support Vector Machine (SVM)

```

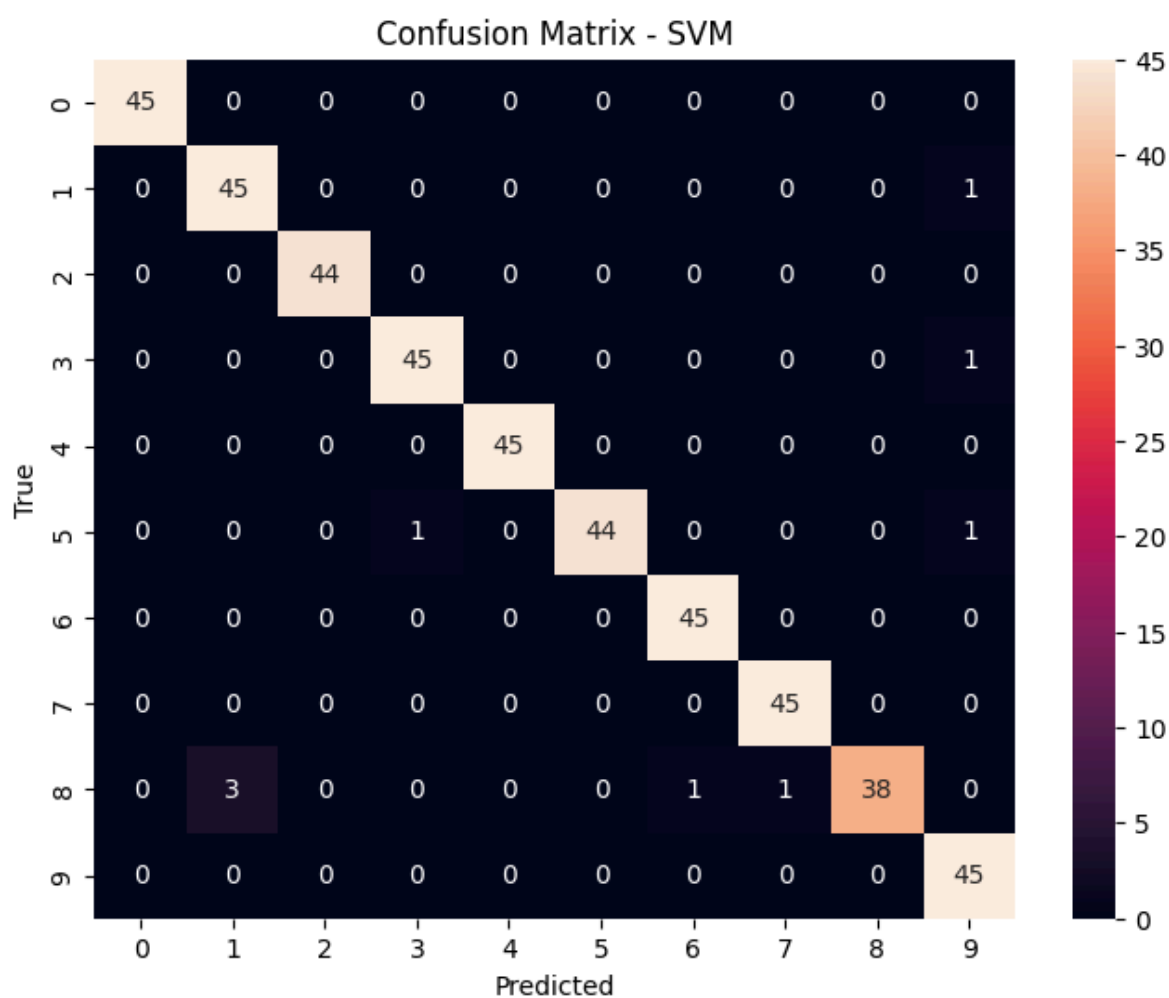
--- SVM (linear) ---
              precision    recall  f1-score   support

     0         1.00        1.00        1.00         45
     1         0.94        0.98        0.96         46
     2         1.00        1.00        1.00         44
     3         0.98        0.98        0.98         46
     4         1.00        1.00        1.00         45
     5         1.00        0.96        0.98         46
     6         0.98        1.00        0.99         45
     7         0.98        1.00        0.99         45
     8         1.00        0.88        0.94         43
     9         0.94        1.00        0.97         45

 accuracy          0.98
 macro avg          0.98        0.98        0.98
 weighted avg       0.98        0.98        0.98

Accuracy: 0.98

```



Interpretation:

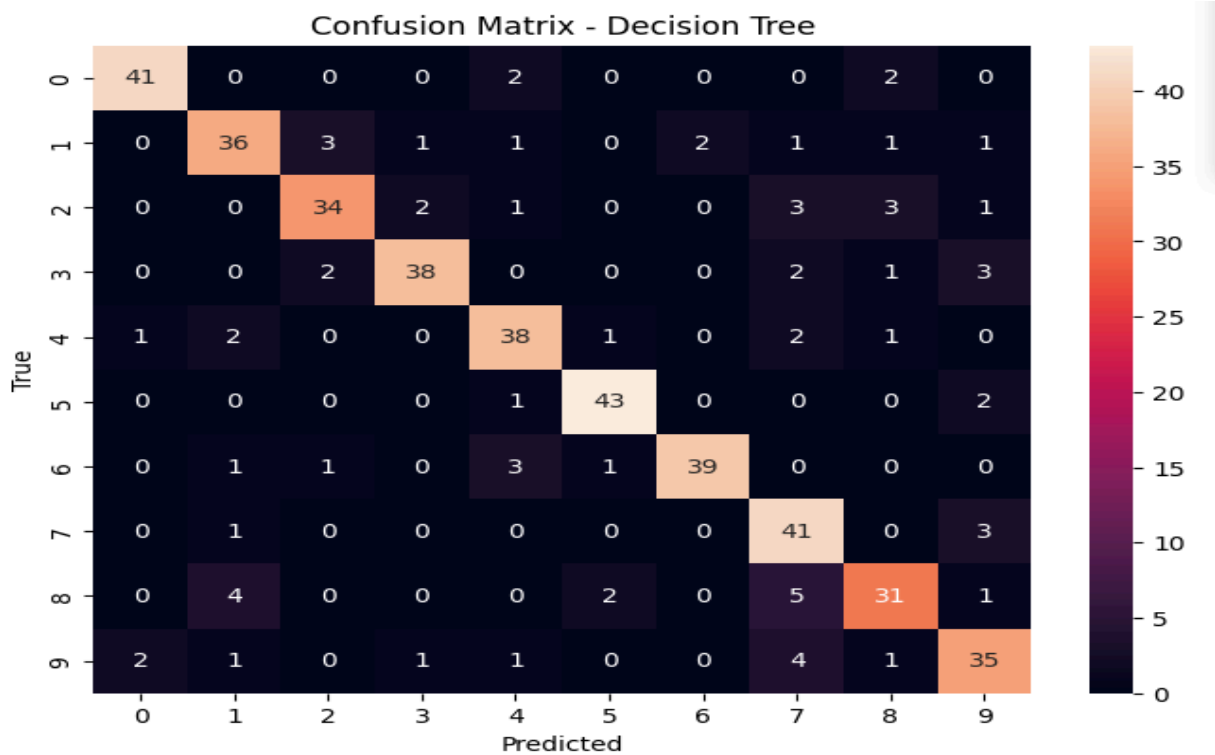
SVM is powerful for high-dimensional data and shows robust performance, but it requires careful parameter tuning (e.g., kernel type, regularization).

3. Decision Tree Classifier

--- Decision Tree ---

	precision	recall	f1-score	support
0	0.93	0.91	0.92	45
1	0.80	0.78	0.79	46
2	0.85	0.77	0.81	44
3	0.90	0.83	0.86	46
4	0.81	0.84	0.83	45
5	0.91	0.93	0.92	46
6	0.95	0.87	0.91	45
7	0.71	0.91	0.80	45
8	0.78	0.72	0.75	43
9	0.76	0.78	0.77	45
accuracy			0.84	450
macro avg	0.84	0.83	0.84	450
weighted avg	0.84	0.84	0.84	450

Accuracy: 0.8355555555555556



Interpretation:

While it correctly classified many samples, the accuracy was lower than SVM. The interpretability of decision trees is a major advantage, but pruning or ensemble methods (e.g., Random Forests) could improve performance.

Suggested Exercises

1. Compare models using cross-validation.

```
KNN CV mean accuracy: 0.9627
```

```
SVM CV mean accuracy: 0.9466
```

```
Decision Tree CV mean accuracy: 0.7847
```

2. Tune hyperparameters (e.g., K in KNN, max_depth in trees).

```
Best K for KNN: {'n_neighbors': 2} with score 0.9671711544413494
```

```
Best Decision Tree params: {'criterion': 'entropy', 'max_depth': 20} with score 0.8147152584339213
```

3. Train models using your own digit/character datasets via OpenCV.
4. Build a GUI-based digit recognition app using OpenCV and a trained model.

[Found in github](#)