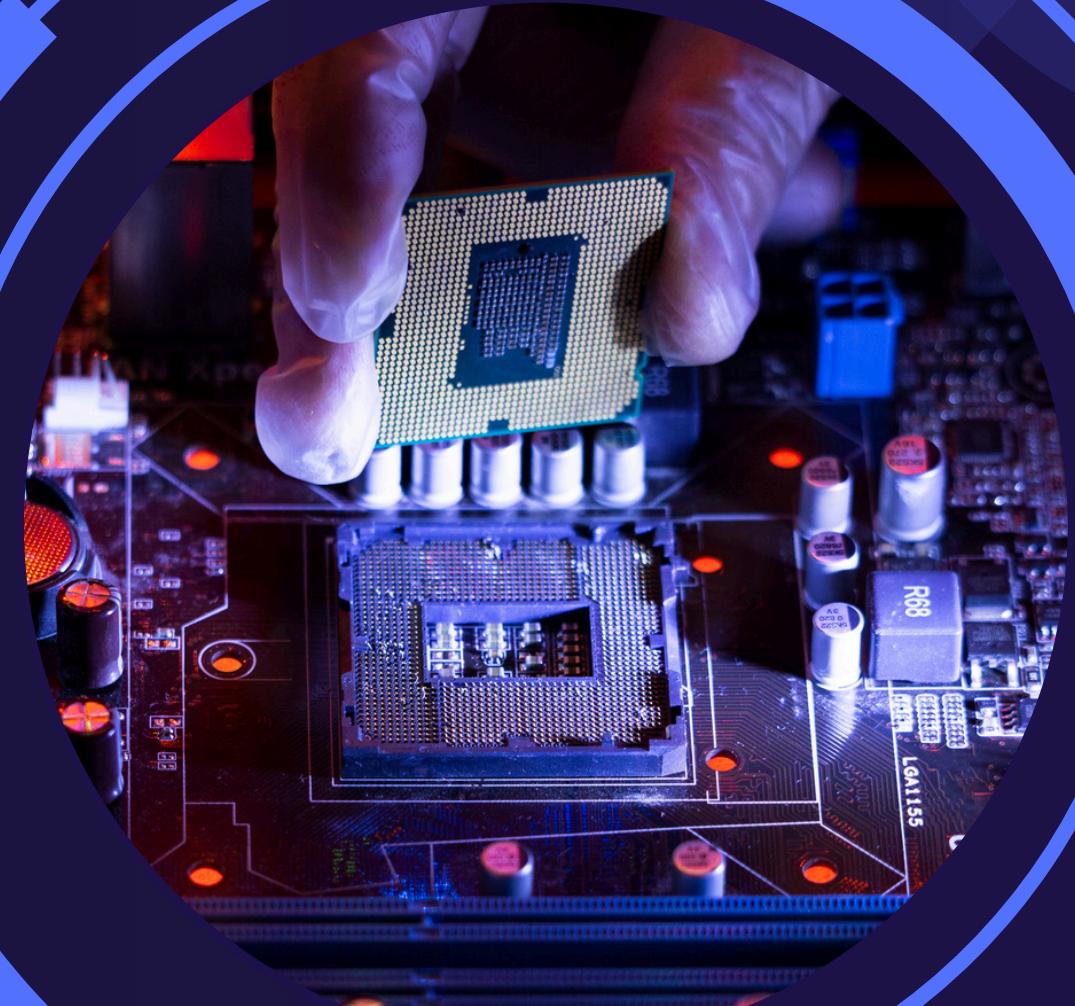


# Machine Learning: Classificação de dígitos manuscritos



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# Objetivo: classificar dígitos manuscritos (0 a 9) a partir de imagens digitalizadas.

**IN**

Vetores de características (features) de imagens de dígitos, onde cada imagem  $8 \times 8$  pixels é transformada em um vetor de 64 dimensões, com cada valor representando a intensidade de um pixel.

**OUT**

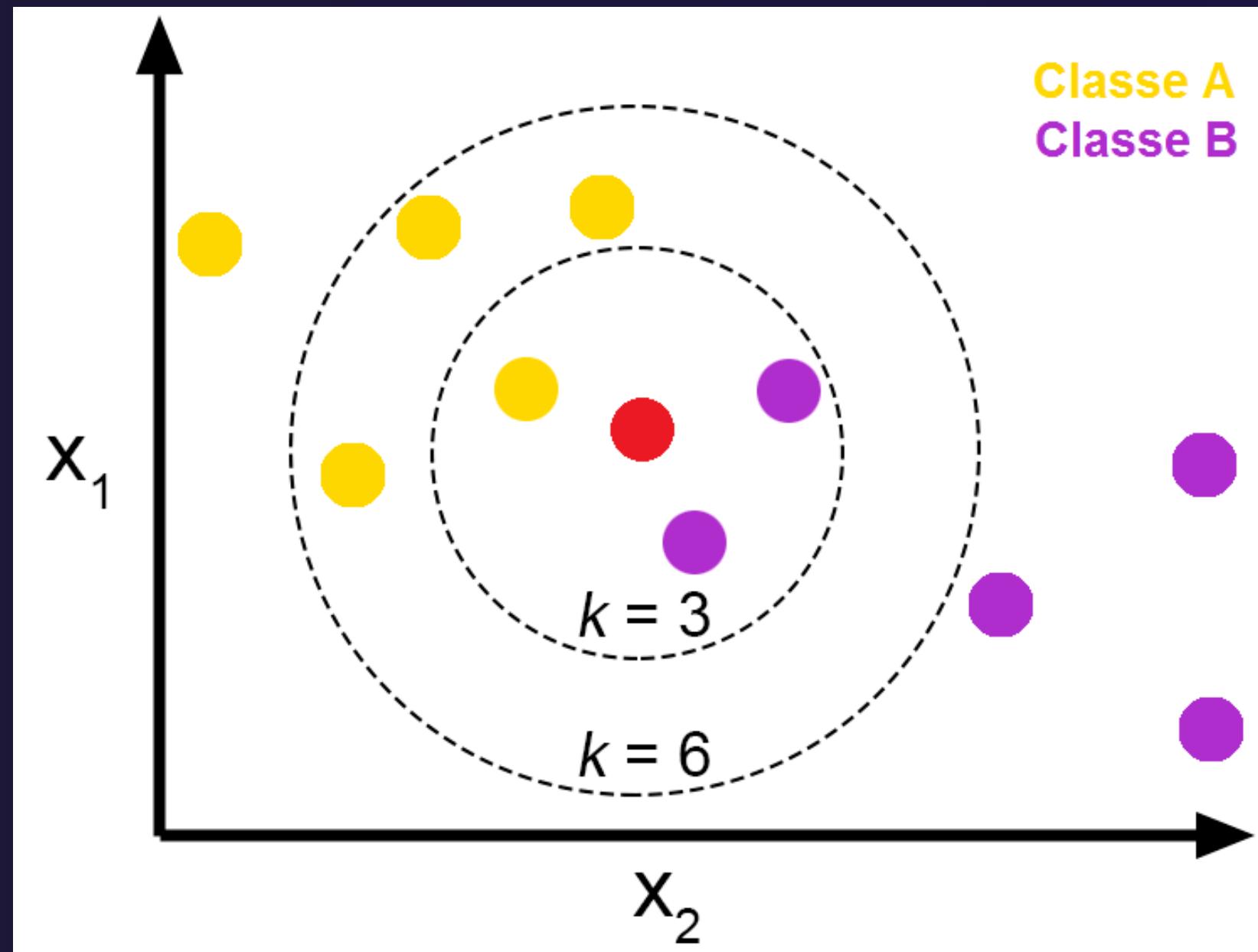
Dígitos corretos (0, 1, 2, ..., 9).

# Ferramentas

- **K-Vizinhos Mais Próximos  
(KNeighborsClassifier)**
- **Floresta Aleatória  
(RandomForestClassifier)**

# K-Vizinhos Mais Próximos

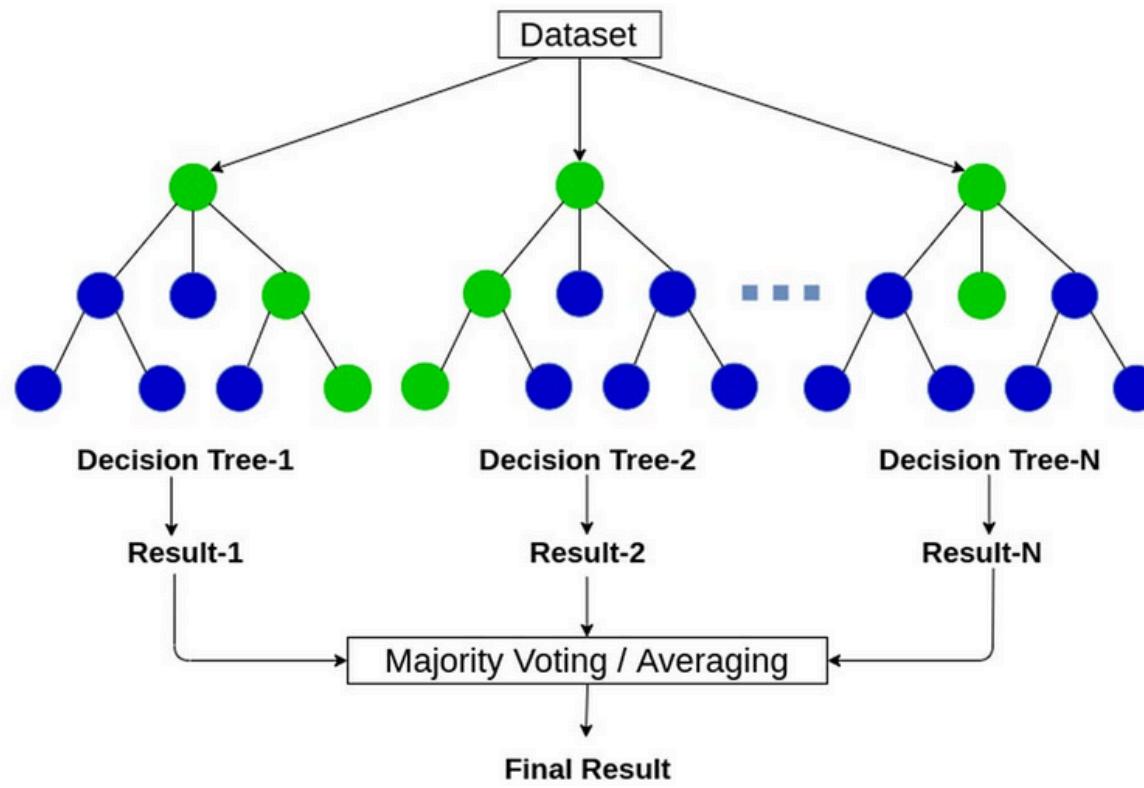
## (KNeighborsClassifier)



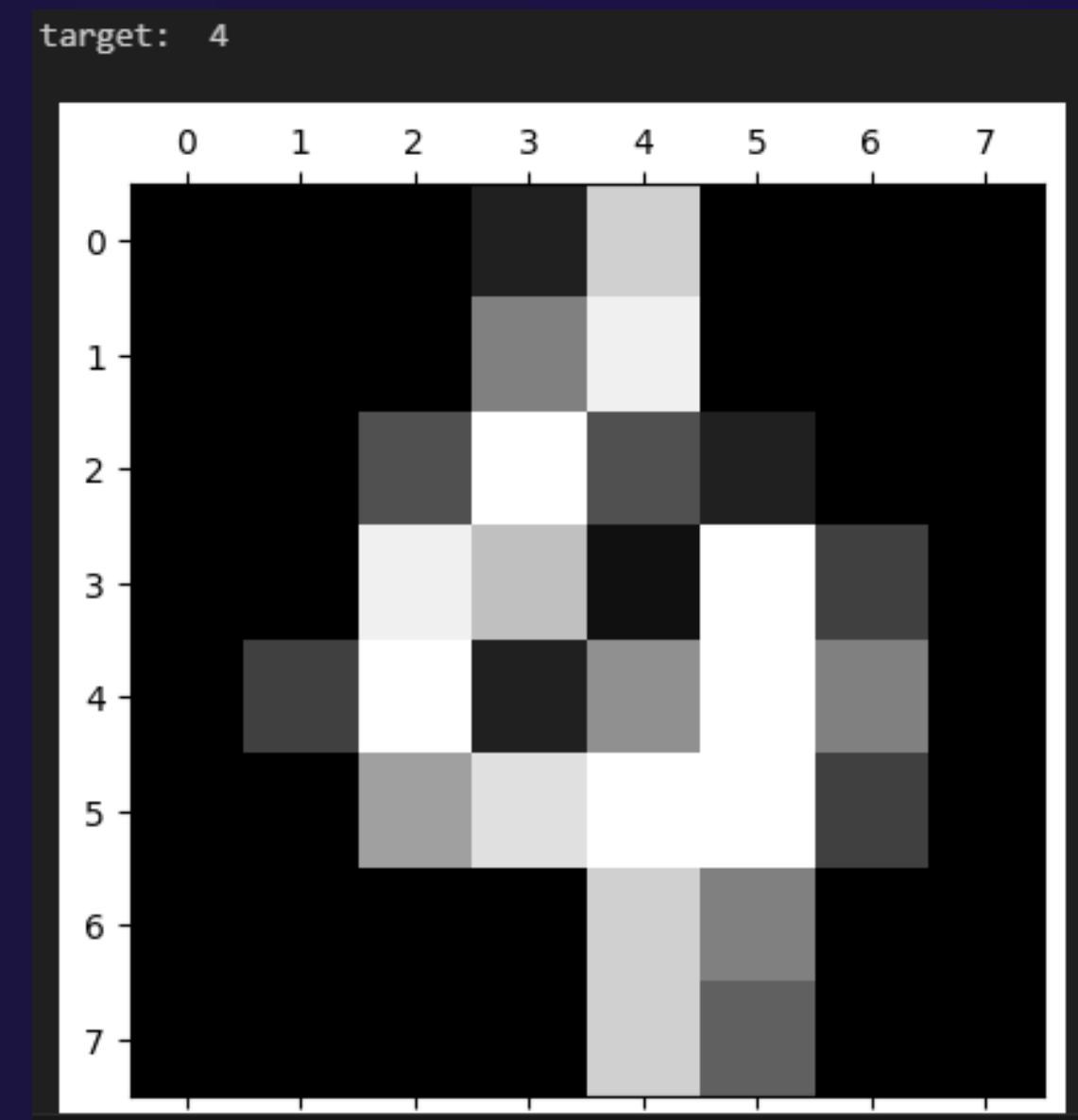
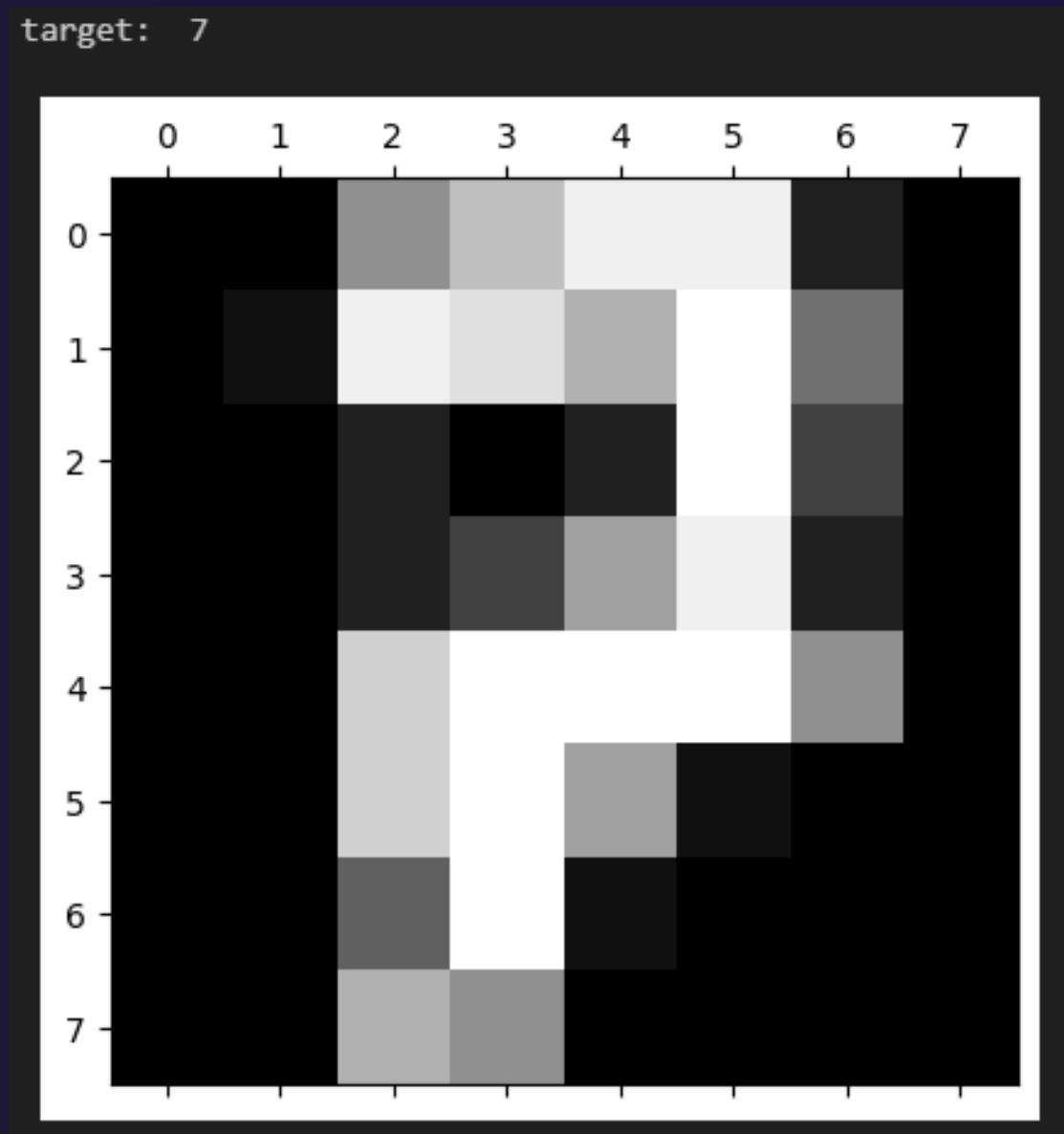
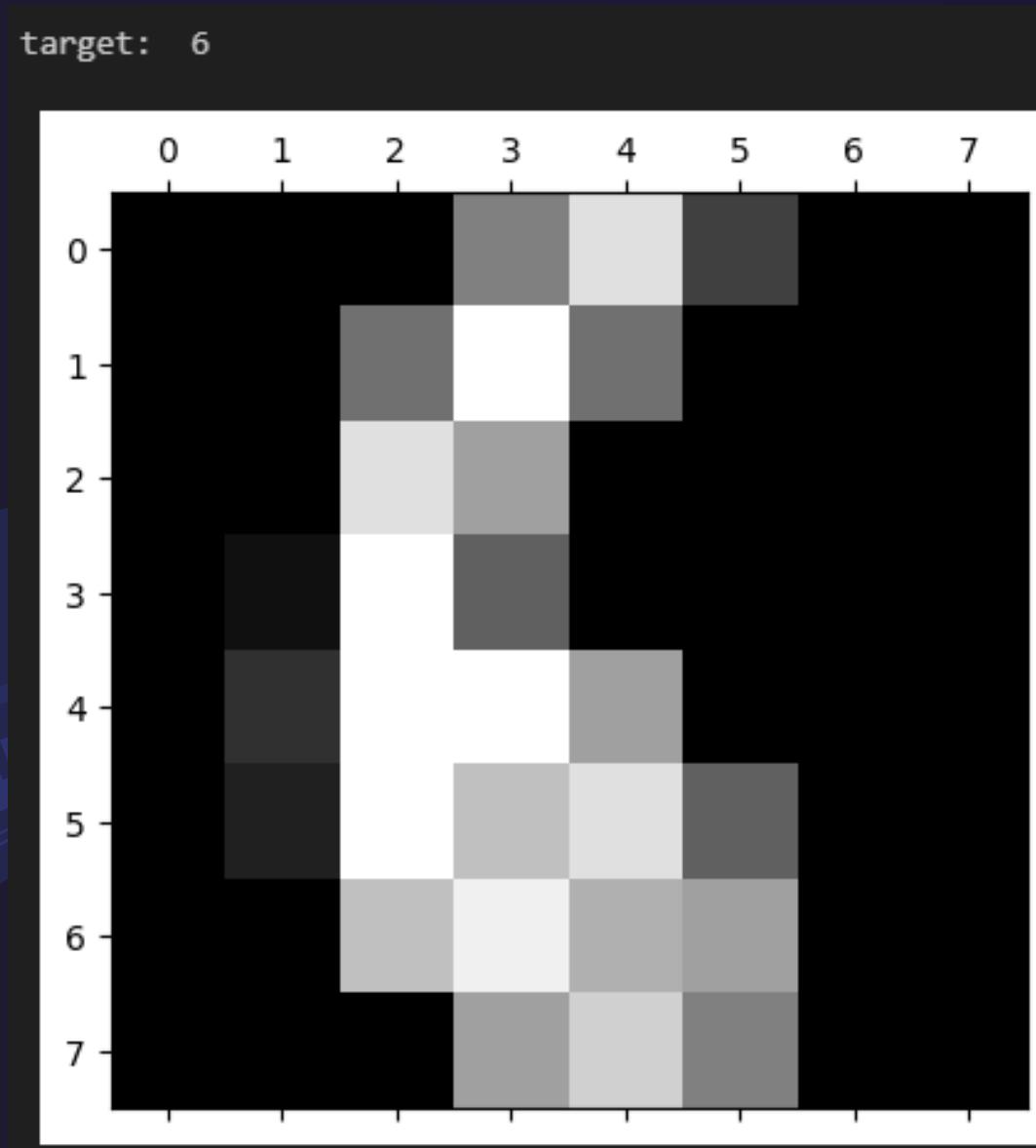
# Floresta Aleatória

## (RandomForestClassifier)

### Random Forest



# Dataset (sk.datasets)



# Dataset (sk.datasets)

```
print(data[0])
print(len(data[0]))
```

✓ 0.0s

```
[ 0.  0.  5. 13.  9.  1.  0.  0.  0. 13. 15. 10. 15.  5.  0.  0.  3.
 15.  2.  0. 11.  8.  0.  0.  4. 12.  0.  0.  8.  8.  0.  0.  5.  8.  0.
 0.  9.  8.  0.  0.  4. 11.  0.  1. 12.  7.  0.  0.  2. 14.  5. 10. 12.
 0.  0.  0.  0.  6. 13. 10.  0.  0.  0.]
```

# Resultados

```
data = digits.data
target = digits.target

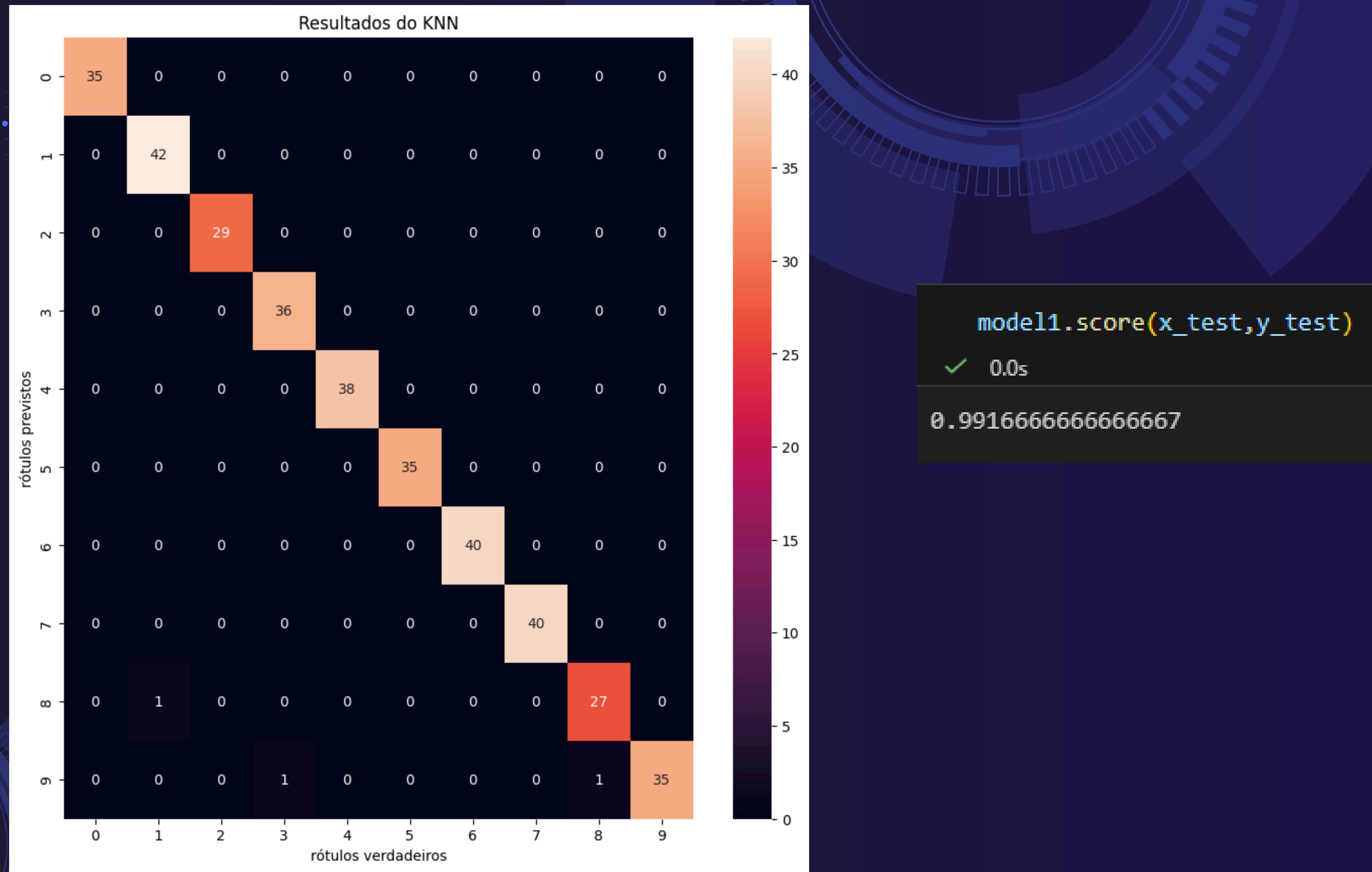
x_train, x_test, y_train, y_test = train_test_split(data,target,test_size=0.2)
✓ 0.0s

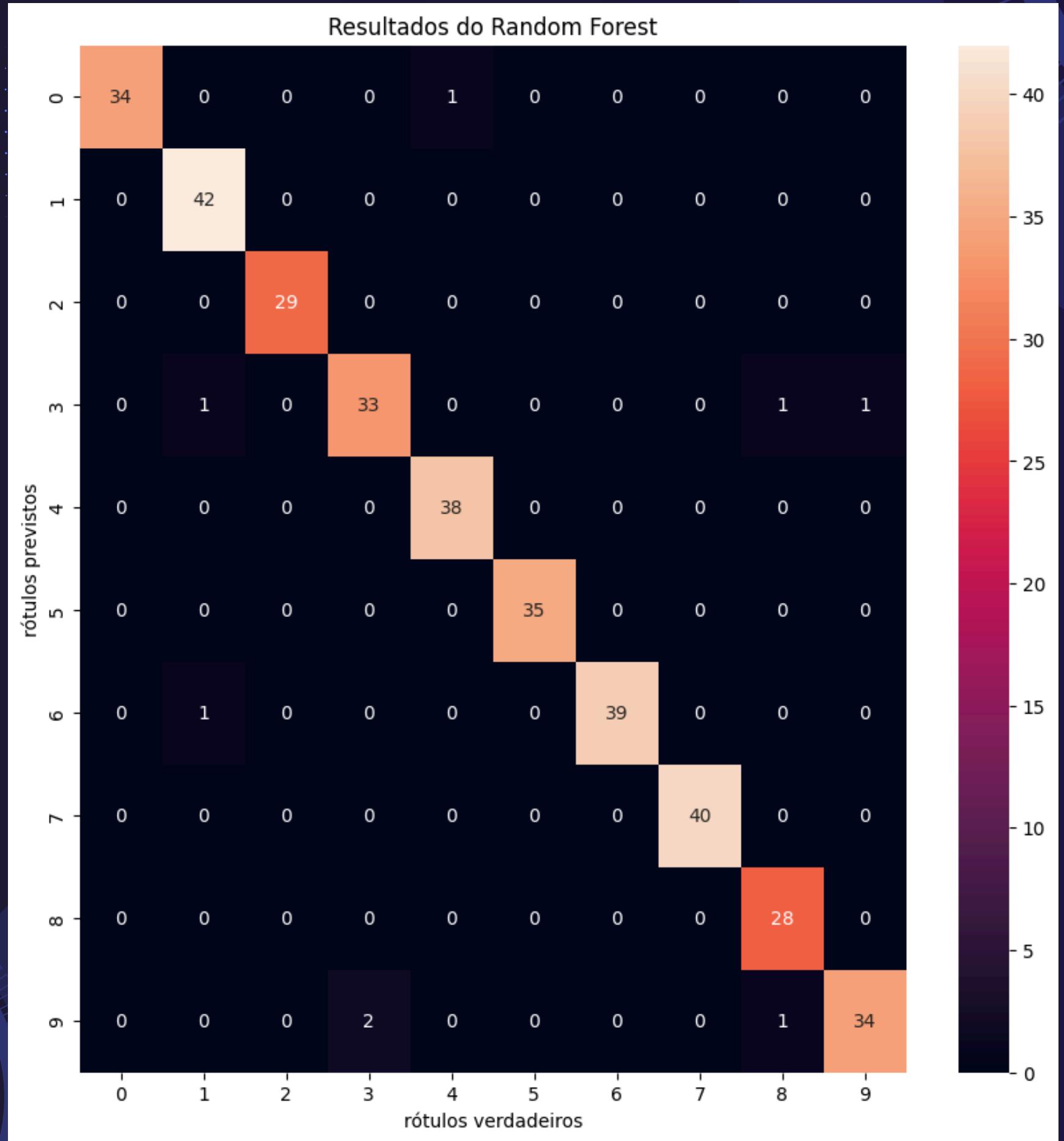
model1 = KNeighborsClassifier()
model1.fit(x_train,y_train)
✓ 0.0s

▼ KNeighborsClassifier ⓘ ⓘ
▶ Parameters

model2 = RandomForestClassifier()
model2.fit(x_train,y_train)
✓ 0.4s

▼ RandomForestClassifier ⓘ ⓘ
▶ Parameters
```





```
model2.score(x_test,y_test)  
✓ 0.0s  
0.9777777777777777
```

# Análise dos resultados

KNN: Acurácia de 99,16%

RF: Acurácia de 97,77%

O KNN, que é um algoritmo baseado em distância, pode ter se beneficiado da proximidade clara entre os vetores de imagens dos mesmos dígitos.

# Obrigado!



Documentação completa:  
<https://scikit-learn.org/stable/index.html>