

Untitled_changed

November 17, 2024

[1]: #AŞAMA 1: GEREKLİ KÜTÜPHANELERİN YÜKLENMESİ

[2]: !pip install keras

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: keras in
c:\users\sulta\appdata\roaming\python\python312\site-packages (3.6.0)
Requirement already satisfied: absl-py in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras)
(2.1.0)
Requirement already satisfied: numpy in c:\programdata\anaconda3\lib\site-
packages (from keras) (1.26.4)
Requirement already satisfied: rich in c:\programdata\anaconda3\lib\site-
packages (from keras) (13.7.1)
Requirement already satisfied: namex in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras)
(0.0.8)
Requirement already satisfied: h5py in c:\programdata\anaconda3\lib\site-
packages (from keras) (3.11.0)
Requirement already satisfied: optree in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras)
(0.13.1)
Requirement already satisfied: ml-dtypes in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras)
(0.4.1)
Requirement already satisfied: packaging in c:\programdata\anaconda3\lib\site-
packages (from keras) (24.1)
Requirement already satisfied: typing-extensions>=4.5.0 in
c:\programdata\anaconda3\lib\site-packages (from optree->keras) (4.11.0)
Requirement already satisfied: markdown-it-py>=2.2.0 in
c:\programdata\anaconda3\lib\site-packages (from rich->keras) (2.2.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in
c:\programdata\anaconda3\lib\site-packages (from rich->keras) (2.15.1)
Requirement already satisfied: mdurl~=0.1 in c:\programdata\anaconda3\lib\site-
packages (from markdown-it-py>=2.2.0->rich->keras) (0.1.0)
```

[3]: !pip install tensorflow

```
Defaulting to user installation because normal site-packages is not writeable
```

Requirement already satisfied: tensorflow in
c:\users\sulta\appdata\roaming\python\python312\site-packages (2.18.0)

Requirement already satisfied: tensorflow-intel==2.18.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow)
(2.18.0)

Requirement already satisfied: absl-py>=1.0.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (2.1.0)

Requirement already satisfied: astunparse>=1.6.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (1.6.3)

Requirement already satisfied: flatbuffers>=24.3.25 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (24.3.25)

Requirement already satisfied: gast!=0.5.0,!0.5.1,!0.5.2,>=0.2.1 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (0.6.0)

Requirement already satisfied: google-pasta>=0.1.1 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (0.2.0)

Requirement already satisfied: libclang>=13.0.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (18.1.1)

Requirement already satisfied: opt-einsum>=2.3.2 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (3.4.0)

Requirement already satisfied: packaging in c:\programdata\anaconda3\lib\site-
packages (from tensorflow-intel==2.18.0->tensorflow) (24.1)

Requirement already satisfied:
protobuf!=4.21.0,!4.21.1,!4.21.2,!4.21.3,!4.21.4,!4.21.5,<6.0.0dev,>=3.20.3
in c:\programdata\anaconda3\lib\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (4.25.3)

Requirement already satisfied: requests<3,>=2.21.0 in
c:\programdata\anaconda3\lib\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (2.32.3)

Requirement already satisfied: setuptools in c:\programdata\anaconda3\lib\site-
packages (from tensorflow-intel==2.18.0->tensorflow) (75.1.0)

Requirement already satisfied: six>=1.12.0 in c:\programdata\anaconda3\lib\site-
packages (from tensorflow-intel==2.18.0->tensorflow) (1.16.0)

Requirement already satisfied: termcolor>=1.1.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (2.5.0)

Requirement already satisfied: typing-extensions>=3.6.6 in
c:\programdata\anaconda3\lib\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (4.11.0)

Requirement already satisfied: wrapt>=1.11.0 in
c:\programdata\anaconda3\lib\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (1.14.1)

Requirement already satisfied: grpcio<2.0,>=1.24.3 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (1.67.1)

Requirement already satisfied: tensorboard<2.19,>=2.18 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (2.18.0)

Requirement already satisfied: keras>=3.5.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (3.6.0)

Requirement already satisfied: numpy<2.1.0,>=1.26.0 in
c:\programdata\anaconda3\lib\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (1.26.4)

Requirement already satisfied: h5py>=3.11.0 in
c:\programdata\anaconda3\lib\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (3.11.0)

Requirement already satisfied: ml-dtypes<0.5.0,>=0.4.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from tensorflow-
intel==2.18.0->tensorflow) (0.4.1)

Requirement already satisfied: wheel<1.0,>=0.23.0 in
c:\programdata\anaconda3\lib\site-packages (from astunparse>=1.6.0->tensorflow-
intel==2.18.0->tensorflow) (0.44.0)

Requirement already satisfied: rich in c:\programdata\anaconda3\lib\site-
packages (from keras>=3.5.0->tensorflow-intel==2.18.0->tensorflow) (13.7.1)

Requirement already satisfied: namex in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from
keras>=3.5.0->tensorflow-intel==2.18.0->tensorflow) (0.0.8)

Requirement already satisfied: optree in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from
keras>=3.5.0->tensorflow-intel==2.18.0->tensorflow) (0.13.1)

Requirement already satisfied: charset-normalizer<4,>=2 in
c:\programdata\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorflow-intel==2.18.0->tensorflow) (3.3.2)

Requirement already satisfied: idna<4,>=2.5 in
c:\programdata\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorflow-intel==2.18.0->tensorflow) (3.7)

Requirement already satisfied: urllib3<3,>=1.21.1 in
c:\programdata\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorflow-intel==2.18.0->tensorflow) (2.2.3)

Requirement already satisfied: certifi>=2017.4.17 in
c:\programdata\anaconda3\lib\site-packages (from
requests<3,>=2.21.0->tensorflow-intel==2.18.0->tensorflow) (2024.8.30)

Requirement already satisfied: markdown>=2.6.8 in
c:\programdata\anaconda3\lib\site-packages (from
tensorboard<2.19,>=2.18->tensorflow-intel==2.18.0->tensorflow) (3.4.1)

Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0 in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from
tensorboard<2.19,>=2.18->tensorflow-intel==2.18.0->tensorflow) (0.7.2)

Requirement already satisfied: werkzeug>=1.0.1 in

```
c:\programdata\anaconda3\lib\site-packages (from
tensorboard<2.19,>=2.18->tensorflow-intel==2.18.0->tensorflow) (3.0.3)
Requirement already satisfied: MarkupSafe>=2.1.1 in
c:\programdata\anaconda3\lib\site-packages (from
werkzeug>=1.0.1->tensorboard<2.19,>=2.18->tensorflow-intel==2.18.0->tensorflow)
(2.1.3)
Requirement already satisfied: markdown-it-py>=2.2.0 in
c:\programdata\anaconda3\lib\site-packages (from rich->keras>=3.5.0->tensorflow-
intel==2.18.0->tensorflow) (2.2.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in
c:\programdata\anaconda3\lib\site-packages (from rich->keras>=3.5.0->tensorflow-
intel==2.18.0->tensorflow) (2.15.1)
Requirement already satisfied: mdurl~=0.1 in c:\programdata\anaconda3\lib\site-
packages (from markdown-it-py>=2.2.0->rich->keras>=3.5.0->tensorflow-
intel==2.18.0->tensorflow) (0.1.0)
```

```
[4]: # MLP ve RBF modelleri için
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Input
```

```
[5]: !pip install pandas
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: pandas in c:\programdata\anaconda3\lib\site-
packages (2.2.2)
Requirement already satisfied: numpy>=1.26.0 in
c:\programdata\anaconda3\lib\site-packages (from pandas) (1.26.4)
Requirement already satisfied: python-dateutil>=2.8.2 in
c:\programdata\anaconda3\lib\site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in
c:\programdata\anaconda3\lib\site-packages (from pandas) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in
c:\programdata\anaconda3\lib\site-packages (from pandas) (2023.3)
Requirement already satisfied: six>=1.5 in c:\programdata\anaconda3\lib\site-
packages (from python-dateutil>=2.8.2->pandas) (1.16.0)
```

```
[6]: !pip install numpy scikit-learn matplotlib seaborn keras
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: numpy in c:\programdata\anaconda3\lib\site-
packages (1.26.4)
Requirement already satisfied: scikit-learn in
c:\programdata\anaconda3\lib\site-packages (1.5.1)
Requirement already satisfied: matplotlib in c:\programdata\anaconda3\lib\site-
packages (3.9.2)
Requirement already satisfied: seaborn in c:\programdata\anaconda3\lib\site-
packages (0.13.2)
Requirement already satisfied: keras in
c:\users\sulta\appdata\roaming\python\python312\site-packages (3.6.0)
```

Requirement already satisfied: scipy>=1.6.0 in
c:\programdata\anaconda3\lib\site-packages (from scikit-learn) (1.13.1)

Requirement already satisfied: joblib>=1.2.0 in
c:\programdata\anaconda3\lib\site-packages (from scikit-learn) (1.4.2)

Requirement already satisfied: threadpoolctl>=3.1.0 in
c:\programdata\anaconda3\lib\site-packages (from scikit-learn) (3.5.0)

Requirement already satisfied: contourpy>=1.0.1 in
c:\programdata\anaconda3\lib\site-packages (from matplotlib) (1.2.0)

Requirement already satisfied: cyclor>=0.10 in
c:\programdata\anaconda3\lib\site-packages (from matplotlib) (0.11.0)

Requirement already satisfied: fonttools>=4.22.0 in
c:\programdata\anaconda3\lib\site-packages (from matplotlib) (4.51.0)

Requirement already satisfied: kiwisolver>=1.3.1 in
c:\programdata\anaconda3\lib\site-packages (from matplotlib) (1.4.4)

Requirement already satisfied: packaging>=20.0 in
c:\programdata\anaconda3\lib\site-packages (from matplotlib) (24.1)

Requirement already satisfied: pillow>=8 in c:\programdata\anaconda3\lib\site-packages (from matplotlib) (10.4.0)

Requirement already satisfied: pyparsing>=2.3.1 in
c:\programdata\anaconda3\lib\site-packages (from matplotlib) (3.1.2)

Requirement already satisfied: python-dateutil>=2.7 in
c:\programdata\anaconda3\lib\site-packages (from matplotlib) (2.9.0.post0)

Requirement already satisfied: pandas>=1.2 in c:\programdata\anaconda3\lib\site-packages (from seaborn) (2.2.2)

Requirement already satisfied: absl-py in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras) (2.1.0)

Requirement already satisfied: rich in c:\programdata\anaconda3\lib\site-packages (from keras) (13.7.1)

Requirement already satisfied: namex in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras) (0.0.8)

Requirement already satisfied: h5py in c:\programdata\anaconda3\lib\site-packages (from keras) (3.11.0)

Requirement already satisfied: optree in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras) (0.13.1)

Requirement already satisfied: ml-dtypes in
c:\users\sulta\appdata\roaming\python\python312\site-packages (from keras) (0.4.1)

Requirement already satisfied: pytz>=2020.1 in
c:\programdata\anaconda3\lib\site-packages (from pandas>=1.2->seaborn) (2024.1)

Requirement already satisfied: tzdata>=2022.7 in
c:\programdata\anaconda3\lib\site-packages (from pandas>=1.2->seaborn) (2023.3)

Requirement already satisfied: six>=1.5 in c:\programdata\anaconda3\lib\site-packages (from python-dateutil>=2.7->matplotlib) (1.16.0)

Requirement already satisfied: typing-extensions>=4.5.0 in
c:\programdata\anaconda3\lib\site-packages (from optree->keras) (4.11.0)

Requirement already satisfied: markdown-it-py>=2.2.0 in
c:\programdata\anaconda3\lib\site-packages (from rich->keras) (2.2.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in
c:\programdata\anaconda3\lib\site-packages (from rich->keras) (2.15.1)
Requirement already satisfied: mdurl~=0.1 in c:\programdata\anaconda3\lib\site-
packages (from markdown-it-py>=2.2.0->rich->keras) (0.1.0)

```
[7]: # Veri işleme için
import pandas as pd
import numpy as np

# Model oluşturma ve veri ayırma için
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler

# Görselleştirme için
import matplotlib.pyplot as plt
import seaborn as sns

# MLP ve RBF modelleri için
from sklearn.neural_network import MLPClassifier
from sklearn.metrics import confusion_matrix, classification_report
from keras.models import Sequential
from keras.layers import Dense, Input
```

```
[8]: #AŞAMA 2: VERİ SETİ YÜKLEME VE İŞLEME
```

```
[9]: import pandas as pd

# CSV dosyalarını yükleme
train_labels = pd.read_csv('data/Train.csv')
test_labels = pd.read_csv('data/Test.csv')

# İlk birkaç satırı görüntüleme
print("Train Labels:")
print(train_labels.head())
print("\nTest Labels:")
print(test_labels.head())
```

Train Labels:

	Width	Height	Roi.X1	Roi.Y1	Roi.X2	Roi.Y2	ClassId	\
0	27	26	5	5	22	20	20	
1	28	27	5	6	23	22	20	
2	29	26	6	5	24	21	20	
3	28	27	5	6	23	22	20	
4	28	26	5	5	23	21	20	

Path

```

0 Train/20/00020_00000_00000.png
1 Train/20/00020_00000_00001.png
2 Train/20/00020_00000_00002.png
3 Train/20/00020_00000_00003.png
4 Train/20/00020_00000_00004.png

```

Test Labels:

	Width	Height	Roi.X1	Roi.Y1	Roi.X2	Roi.Y2	ClassId	Path
0	53	54	6	5	48	49	16	Test/00000.png
1	42	45	5	5	36	40	1	Test/00001.png
2	48	52	6	6	43	47	38	Test/00002.png
3	27	29	5	5	22	24	33	Test/00003.png
4	60	57	5	5	55	52	11	Test/00004.png

[10]: #AŞAMA 3: GÖRÜNTÜ YÜKLEME VE İŞLEME

```

[11]: from PIL import Image
import numpy as np

# Eğitim görüntülerini ve etiketlerini yükleme
train_images = []
train_labels_list = []

for index, row in train_labels.iterrows():
    image_path = 'data/' + row['Path'] # Görüntü dosyasının yolu
    image = Image.open(image_path).convert('RGB') # Görüntüyü RGB formatında
    ↪yükleyin
    image = image.resize((32, 32)) # Boyutları model için ayarlayın (örneğin,
    ↪32x32)
    train_images.append(np.array(image)) # Görüntüyü listeye ekleyin
    train_labels_list.append(row['ClassId']) # Sınıf etiketini ekleyin

# Görüntü ve etiketleri numpy dizilerine çevirme
train_images = np.array(train_images)
train_labels_list = np.array(train_labels_list)

print("Eğitim görüntüleri boyutu:", train_images.shape)
print("Eğitim etiketleri boyutu:", train_labels_list.shape)

```

Eğitim görüntüleri boyutu: (39209, 32, 32, 3)

Eğitim etiketleri boyutu: (39209,)

[12]: #AŞAMA 4: MODEL OLUŞTURMA VE EĞİTME

```

[13]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten, Input
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.layers import Dropout

```

```

# Veriyi Hazırlama
train_labels_categorical = to_categorical(train_labels_list)

# MLP Modelini Tanımlama
model = Sequential()
model.add(Input(shape=(32, 32, 3))) # Giriş katmanı
model.add(Flatten()) # Düzleştirme katmanı
model.add(Dense(64, activation='relu')) # Gizli katman
model.add(Dropout(0.4)) # Dropout eklenmesi (aşırı öğrenmeyi engellemeye,
↳ yardımcı olur)
model.add(Dense(43, activation='softmax')) # Çıkış katmanı

# Modeli derleme (örneğin, optimizer ve loss belirleyerek)
model.compile(optimizer='adam', loss='categorical_crossentropy',
↳ metrics=['accuracy'])

# Modelin özetini görmek için
model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
flatten (Flatten)	(None, 3072)	0
dense (Dense)	(None, 64)	196,672
dropout (Dropout)	(None, 64)	0
dense_1 (Dense)	(None, 43)	2,795

Total params: 199,467 (779.17 KB)

Trainable params: 199,467 (779.17 KB)

Non-trainable params: 0 (0.00 B)

[14]: #VERİYİ STANDARTLAŞTIRMA VEYA NORMALLEŞTİRME

[15]: from sklearn.preprocessing import StandardScaler


```
scaler = StandardScaler()
train_images_scaled = scaler.fit_transform(train_images.reshape(-1, 32 * 32 * 3))
train_images_scaled = train_images_scaled.reshape(-1, 32, 32, 3)
```

```
[16]: #Veriyi Eğitim ve Doğrulama için Bölme:
from sklearn.model_selection import train_test_split

X_train, X_val, y_train, y_val = train_test_split(train_images_scaled,
train_labels_categorical, test_size=0.2, random_state=42)
```

```
[17]: #Modeli Eğitme
```

```
[18]: history = model.fit(X_train, y_train,
                        validation_data=(X_val, y_val),
                        epochs=10,
                        batch_size=32)
```

Epoch 1/10

981/981 2s 2ms/step -
accuracy: 0.3558 - loss: 2.7025 - val_accuracy: 0.7805 - val_loss: 0.8491

Epoch 2/10

981/981 2s 2ms/step -
accuracy: 0.6568 - loss: 1.1772 - val_accuracy: 0.8370 - val_loss: 0.5730

Epoch 3/10

981/981 2s 2ms/step -
accuracy: 0.7318 - loss: 0.9191 - val_accuracy: 0.8558 - val_loss: 0.5266

Epoch 4/10

981/981 1s 1ms/step -
accuracy: 0.7685 - loss: 0.7904 - val_accuracy: 0.8796 - val_loss: 0.4681

Epoch 5/10

981/981 2s 2ms/step -
accuracy: 0.7724 - loss: 0.7741 - val_accuracy: 0.8908 - val_loss: 0.4185

Epoch 6/10

981/981 2s 2ms/step -
accuracy: 0.7948 - loss: 0.6908 - val_accuracy: 0.8944 - val_loss: 0.3804

Epoch 7/10

981/981 2s 2ms/step -
accuracy: 0.8102 - loss: 0.6314 - val_accuracy: 0.8907 - val_loss: 0.3903

Epoch 8/10

981/981 2s 2ms/step -
accuracy: 0.8140 - loss: 0.6392 - val_accuracy: 0.9116 - val_loss: 0.3708

Epoch 9/10

981/981 2s 2ms/step -
accuracy: 0.8261 - loss: 0.6058 - val_accuracy: 0.9151 - val_loss: 0.3437

Epoch 10/10

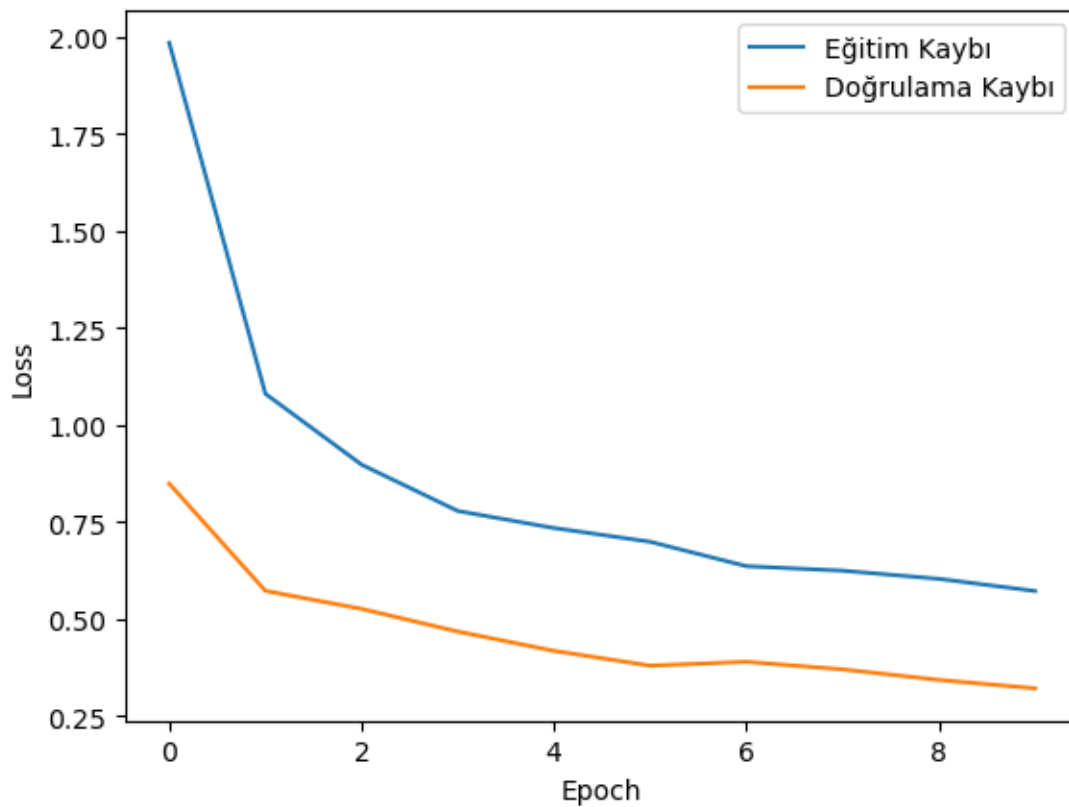
981/981 2s 2ms/step -
accuracy: 0.8250 - loss: 0.6069 - val_accuracy: 0.9111 - val_loss: 0.3219

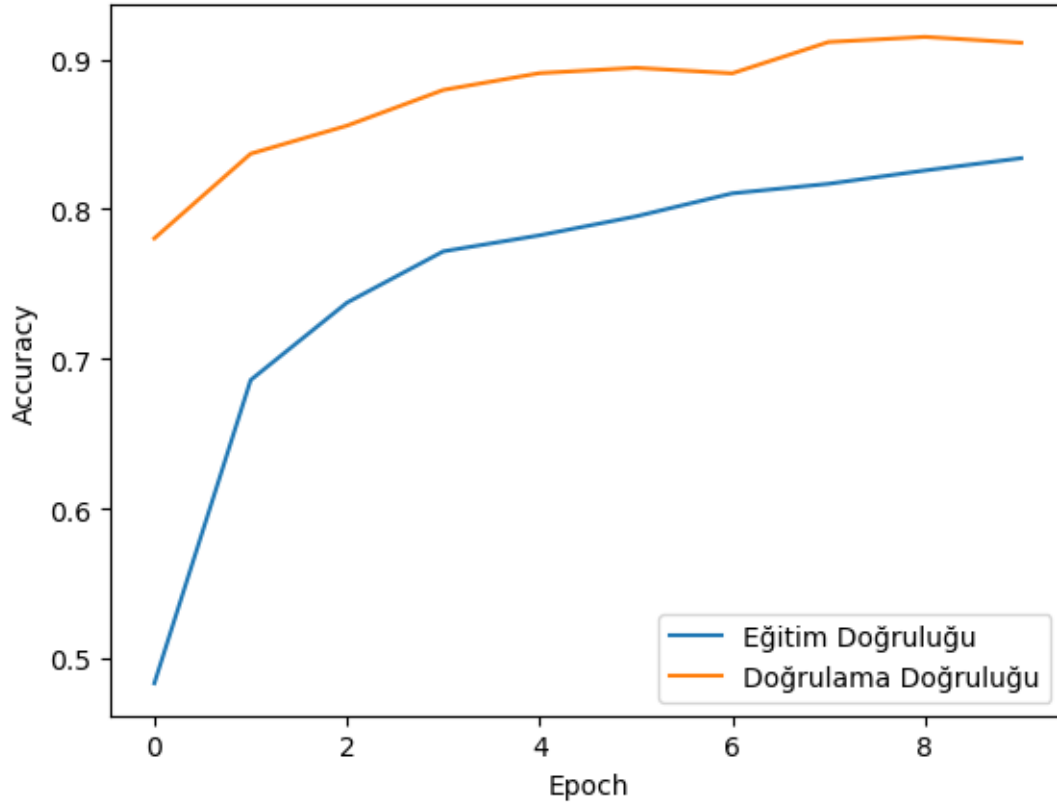
```
[19]: #EĞİTİM SONUÇLARINI GÖRÜNTÜLEME
```

```
[20]: import matplotlib.pyplot as plt
```

```
# Kayıp grafiği
plt.plot(history.history['loss'], label='Eğitim Kaybı')
plt.plot(history.history['val_loss'], label='Doğrulama Kaybı')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()

# Doğruluk grafiği
plt.plot(history.history['accuracy'], label='Eğitim Doğruluğu')
plt.plot(history.history['val_accuracy'], label='Doğrulama Doğruluğu')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```





```
[21]: import pandas as pd
from PIL import Image
import numpy as np

# Test etiket dosyasını yeniden yükleyin (CSV formatında olduğunu varsayıyoruz)
test_labels = pd.read_csv('data/Test.csv')

# Test görüntülerini ve etiketlerini yüklemek için boş listeler oluşturun
test_images = []
test_labels_list = []

# Test etiket dosyanızı yükleyin
for index, row in test_labels.iterrows():
    image_path = 'data/' + row['Path'] # Görüntü dosyasının yolu
    image = Image.open(image_path).convert('RGB') # Görüntüyü RGB formatında
    ↪yükleyin
    image = image.resize((32, 32)) # Boyutları model için ayarlayın (32x32)
    test_images.append(np.array(image)) # Görüntüyü listeye ekleyin
    test_labels_list.append(row['ClassId']) # Sınıf etiketini ekleyin

# Görüntüleri ve etiketleri numpy dizilerine çevirin
```

```
test_images = np.array(test_images)
test_labels_list = np.array(test_labels_list)
```

```
[22]: from tensorflow.keras.utils import to_categorical

# Test etiketlerini one-hot encoding yaparak dönüştürme
test_labels_list = to_categorical(test_labels_list)
```

```
[23]: # Modeli test verisi üzerinde değerlendirme
test_loss, test_accuracy = model.evaluate(test_images, test_labels_list)

print("Test Kayıp (Loss):", test_loss)
print("Test Doğruluğu (Accuracy):", test_accuracy)
```

```
395/395          0s 958us/step -
accuracy: 0.8038 - loss: 68.9993
Test Kayıp (Loss): 69.03533935546875
Test Doğruluğu (Accuracy): 0.8061757683753967
```

```
[24]: # Modeli Değerlendirme ve Analiz Etme
```

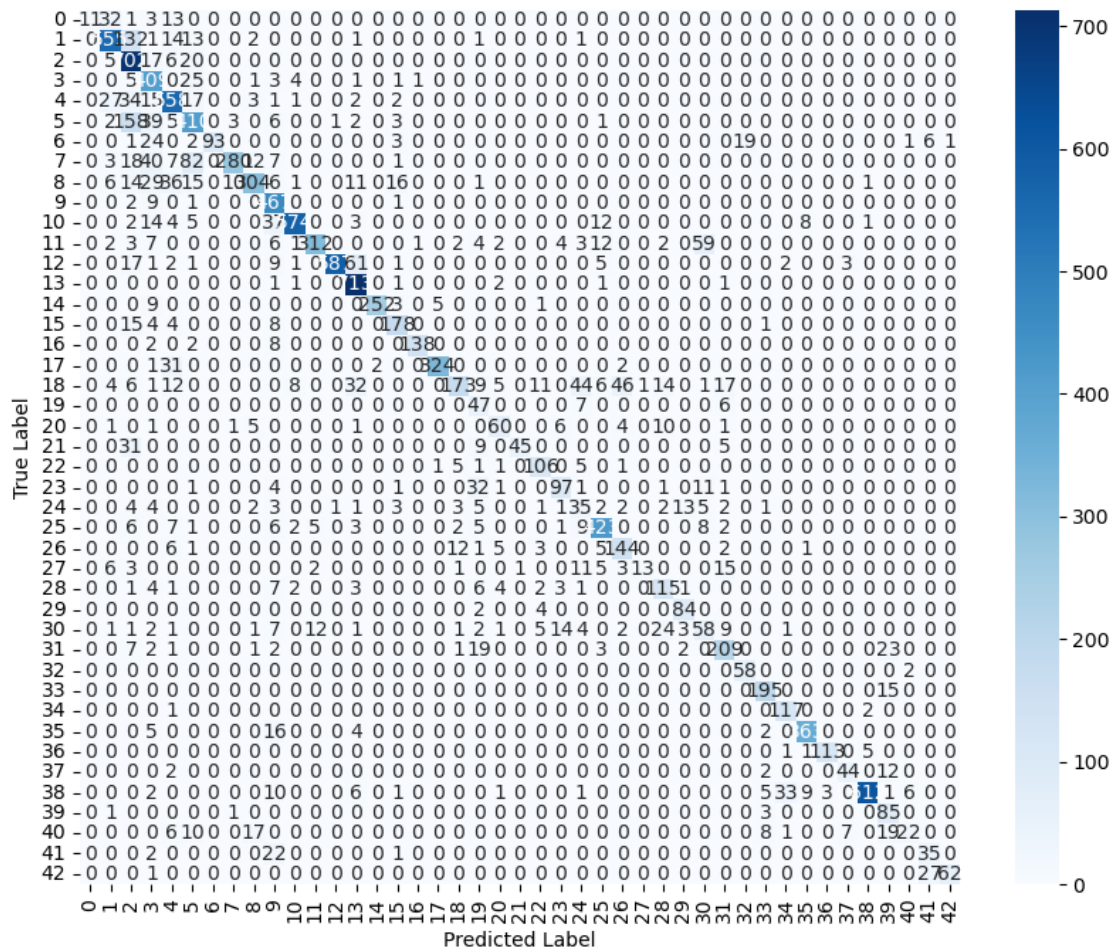
```
[25]: from sklearn.metrics import confusion_matrix
import seaborn as sns
import matplotlib.pyplot as plt

# Test verisi üzerinde tahmin yap
y_pred = model.predict(test_images)
y_pred_classes = np.argmax(y_pred, axis=1) # Her tahmin için en yüksek
↳ olasılık verilen sınıfı seç

# Confusion matrix oluştur
cm = confusion_matrix(np.argmax(test_labels_list, axis=1), y_pred_classes)

# Görselleştir
plt.figure(figsize=(10, 8))
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.show()
```

```
395/395          0s 905us/step
```



```
[26]: from sklearn.metrics import classification_report
print(classification_report(np.argmax(test_labels_list, axis=1),
↪ y_pred_classes))
```

	precision	recall	f1-score	support
0	1.00	0.18	0.31	60
1	0.86	0.77	0.81	720
2	0.60	0.94	0.73	750
3	0.63	0.91	0.74	450
4	0.78	0.85	0.81	660
5	0.68	0.65	0.66	630
6	1.00	0.62	0.77	150
7	0.95	0.62	0.75	450
8	0.87	0.68	0.76	450
9	0.73	0.97	0.84	480
10	0.96	0.87	0.91	660
11	0.94	0.74	0.83	420

12	1.00	0.85	0.92	690
13	0.84	0.99	0.91	720
14	0.99	0.93	0.96	270
15	0.82	0.85	0.84	210
16	0.99	0.92	0.95	150
17	0.98	0.90	0.94	360
18	0.86	0.44	0.59	390
19	0.33	0.78	0.46	60
20	0.73	0.67	0.70	90
21	0.98	0.50	0.66	90
22	0.80	0.88	0.84	120
23	0.77	0.65	0.70	150
24	0.29	0.39	0.33	90
25	0.89	0.88	0.89	480
26	0.71	0.80	0.75	180
27	0.93	0.22	0.35	60
28	0.68	0.77	0.72	150
29	0.82	0.93	0.87	90
30	0.41	0.39	0.40	150
31	0.77	0.77	0.77	270
32	0.75	0.97	0.85	60
33	0.90	0.93	0.91	210
34	0.75	0.97	0.85	120
35	0.95	0.93	0.94	390
36	0.97	0.94	0.96	120
37	0.81	0.73	0.77	60
38	0.99	0.89	0.93	690
39	0.55	0.94	0.69	90
40	0.71	0.24	0.36	90
41	0.51	0.58	0.55	60
42	0.98	0.69	0.81	90
accuracy				0.81 12630
macro avg				0.80 0.75 0.75 12630
weighted avg				0.83 0.81 0.81 12630

```
[27]: import os
      print(os.listdir('data'))
```

```
['archive.zip', 'Meta', 'Meta.csv', 'Test', 'Test.csv', 'Train', 'Train.csv']
```

```
[28]: train_labels_list = pd.read_csv('data/Train.csv')['ClassId'].values
      test_labels_list = pd.read_csv('data/Test.csv')['ClassId'].values
```

```
[29]: #Model Eğitimi
```

```
[30]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten
from tensorflow.keras.utils import to_categorical

# Eğitim etiketlerini kategorik hale getirin
train_labels = to_categorical(train_labels_list)
test_labels = to_categorical(test_labels_list)

# Modeli tanımlayın
model = Sequential([
    Flatten(input_shape=(32, 32, 3)), # Giriş katmanı
    Dense(128, activation='relu'),    # Gizli katman
    Dense(len(train_labels[0]), activation='softmax') # Çıkış katmanı
])

# Modeli derleyin
model.compile(optimizer='adam', loss='categorical_crossentropy',
    metrics=['accuracy'])

# Modeli eğitin
history = model.fit(train_images, train_labels, epochs=10,
    validation_data=(test_images, test_labels))
```

Epoch 1/10

C:\Users\sulta\AppData\Roaming\Python\Python312\site-packages\keras\src\layers\reshaping\flatten.py:37: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.
 super().__init__(**kwargs)

1226/1226 3s 2ms/step -
 accuracy: 0.0517 - loss: 17.9656 - val_accuracy: 0.0570 - val_loss: 3.5262

Epoch 2/10

1226/1226 3s 2ms/step -
 accuracy: 0.0569 - loss: 3.5205 - val_accuracy: 0.0594 - val_loss: 3.4763

Epoch 3/10

1226/1226 3s 2ms/step -
 accuracy: 0.0593 - loss: 3.4910 - val_accuracy: 0.0570 - val_loss: 3.4666

Epoch 4/10

1226/1226 3s 2ms/step -
 accuracy: 0.0580 - loss: 3.4874 - val_accuracy: 0.0570 - val_loss: 3.4641

Epoch 5/10

1226/1226 3s 2ms/step -
 accuracy: 0.0565 - loss: 3.4842 - val_accuracy: 0.0594 - val_loss: 3.4633

Epoch 6/10

1226/1226 3s 2ms/step -
 accuracy: 0.0542 - loss: 3.4823 - val_accuracy: 0.0594 - val_loss: 3.4630

Epoch 7/10

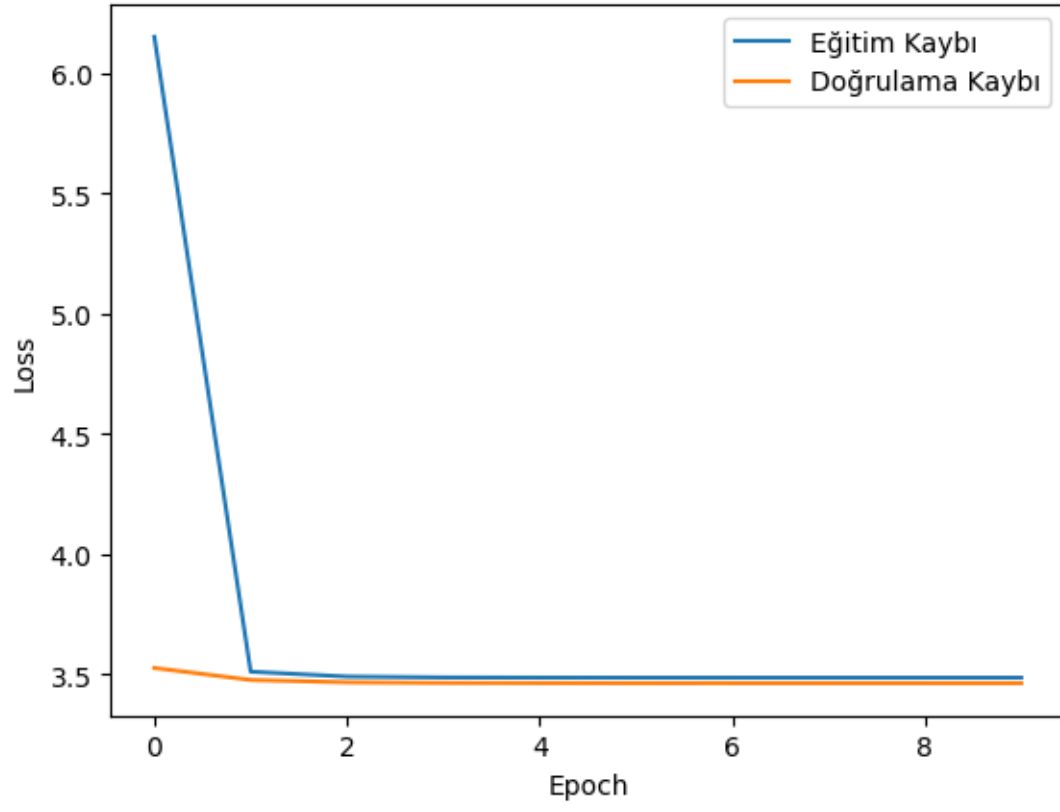
```
1226/1226          3s 2ms/step -
accuracy: 0.0577 - loss: 3.4851 - val_accuracy: 0.0594 - val_loss: 3.4629
Epoch 8/10
1226/1226          3s 2ms/step -
accuracy: 0.0555 - loss: 3.4811 - val_accuracy: 0.0594 - val_loss: 3.4629
Epoch 9/10
1226/1226          3s 2ms/step -
accuracy: 0.0581 - loss: 3.4808 - val_accuracy: 0.0594 - val_loss: 3.4629
Epoch 10/10
1226/1226          3s 2ms/step -
accuracy: 0.0584 - loss: 3.4834 - val_accuracy: 0.0570 - val_loss: 3.4628
```

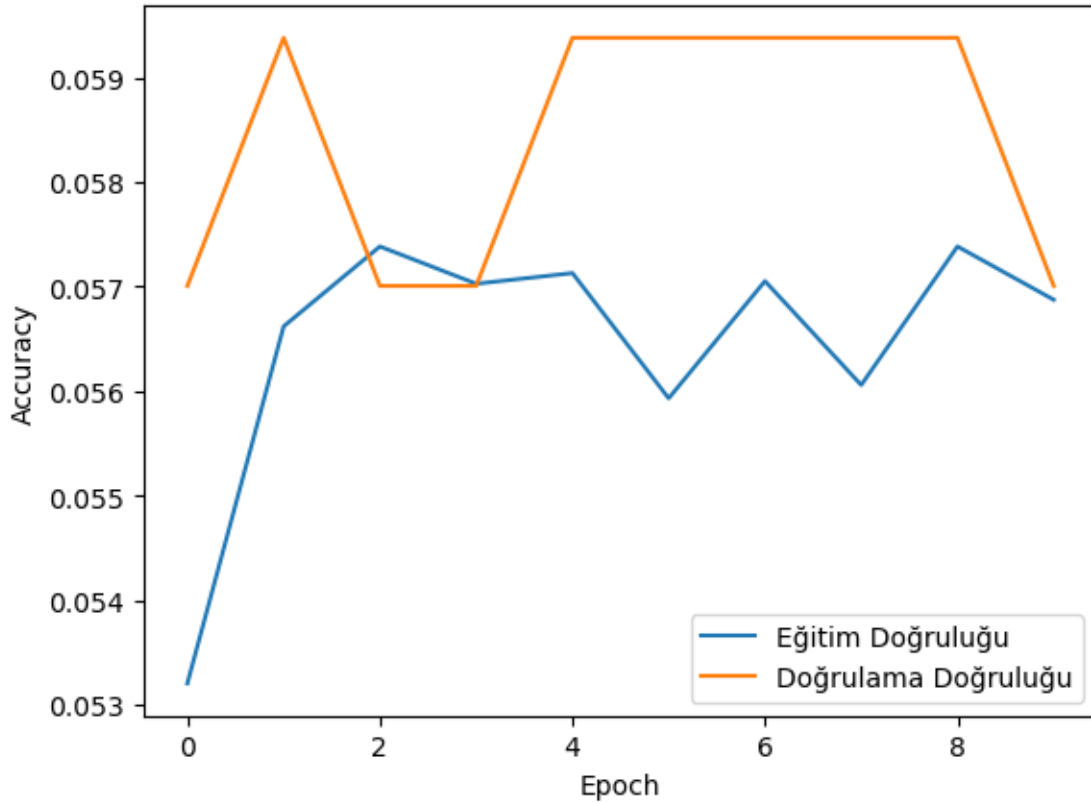
[31]: *#EĞİTİM VE DOĞRULAMA GRAFİKLERİ ÇİZME*

```
[32]: import matplotlib.pyplot as plt

# Kayıp grafiği
plt.plot(history.history['loss'], label='Eğitim Kaybı')
plt.plot(history.history['val_loss'], label='Doğrulama Kaybı')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()

# Doğruluk grafiği
plt.plot(history.history['accuracy'], label='Eğitim Doğruluğu')
plt.plot(history.history['val_accuracy'], label='Doğrulama Doğruluğu')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```



```
[33]: test_loss, test_accuracy = model.evaluate(test_images, test_labels)
print("Test Kayıp (Loss):", test_loss)
print("Test Doğruluğu (Accuracy):", test_accuracy)
```

```
395/395          0s 1ms/step -
accuracy: 0.0582 - loss: 3.4651
Test Kayıp (Loss): 3.4628419876098633
Test Doğruluğu (Accuracy): 0.05700712651014328
```

```
[34]: from sklearn.metrics import confusion_matrix
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np

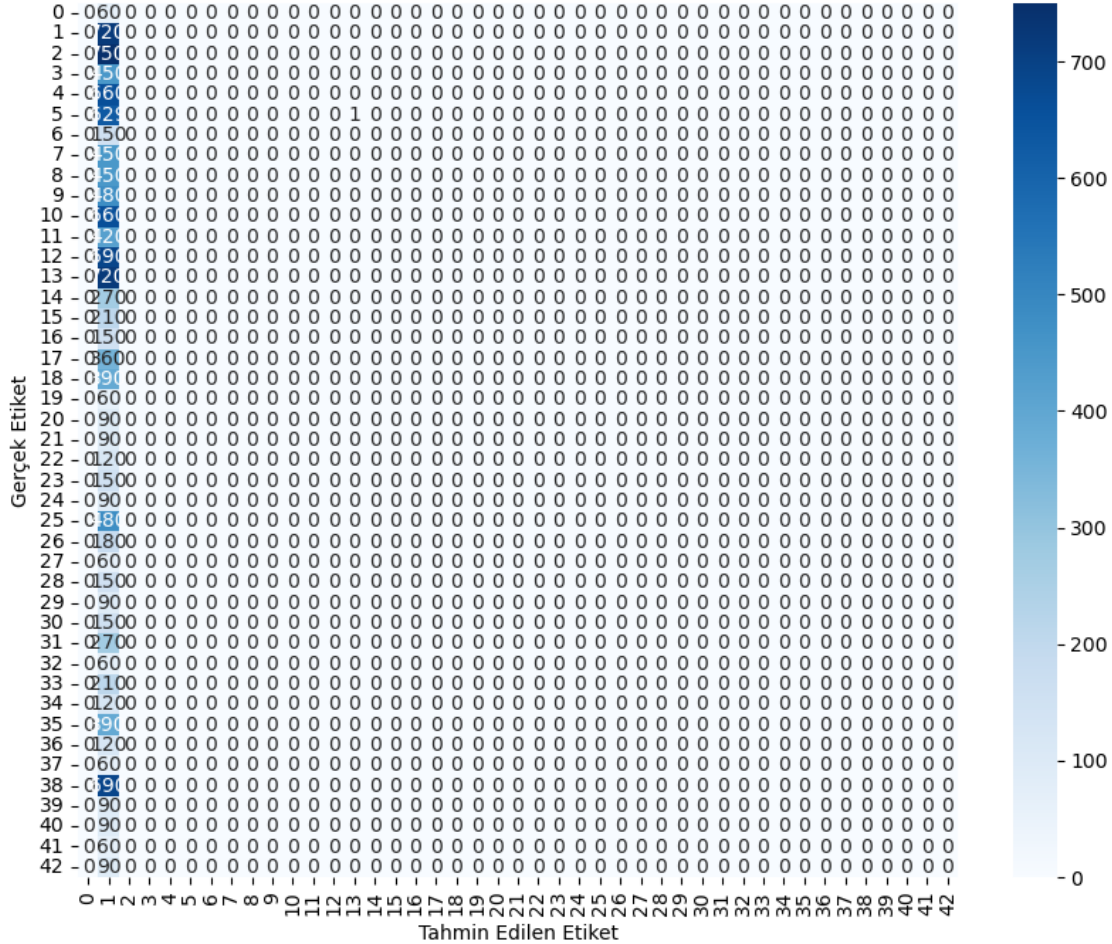
# Tahminleri al
y_pred = model.predict(test_images)
y_pred_classes = np.argmax(y_pred, axis=1) # Tahmin edilen sınıf etiketleri
y_true = np.argmax(test_labels, axis=1)    # Gerçek sınıf etiketleri

# Karışıklık matrisi oluştur
cm = confusion_matrix(y_true, y_pred_classes)
```

```
# Görselleştir
plt.figure(figsize=(10, 8))
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.xlabel("Tahmin Edilen Etiket")
plt.ylabel("Gerçek Etiket")
plt.show()
```

395/395

1s 2ms/step



```
[35]: import matplotlib.pyplot as plt
from PIL import Image

# Örnek bir görsel yükle (Path'i veri setindeki bir görüntüyle değiştir)
image_path = 'C:/Users/sulta/Desktop/GC/02-uodev/data/Train/1/00001_00000_00000.
            ↪png' # Örnek görüntü yolu
original_image = Image.open(image_path).convert('RGB')
```

```

# Görüntüyü yeniden boyutlandır
resized_image = original_image.resize((32, 32))

# Görselleri yan yana göster
plt.figure(figsize=(10, 5))

# Orijinal görüntü
plt.subplot(1, 2, 1)
plt.imshow(original_image)
plt.title("Orijinal Görüntü")
plt.axis('off')

# Yeniden boyutlandırılmış görüntü
plt.subplot(1, 2, 2)
plt.imshow(resized_image)
plt.title("32x32 Yeniden Boyutlandırılmış Görüntü")
plt.axis('off')

plt.show()

```

Orijinal Görüntü



32x32 Yeniden Boyutlandırılmış Görüntü



```

[36]: import matplotlib.pyplot as plt
import numpy as np
from PIL import Image

# Görüntü yolunu belirleyin
image_path = 'C:/Users/sulta/Desktop/GC/02-uodev/data/Train/1/00001_00000_00000.
↳png' # Örnek görüntü

```

```

# Görüntüyü yükle
original_image = Image.open(image_path)

# Görüntüyü numpy dizisine çevir
original_array = np.array(original_image)

# Görüntüyü normalleştir (0-1 aralığına dönüştür)
normalized_array = original_array / 255.0

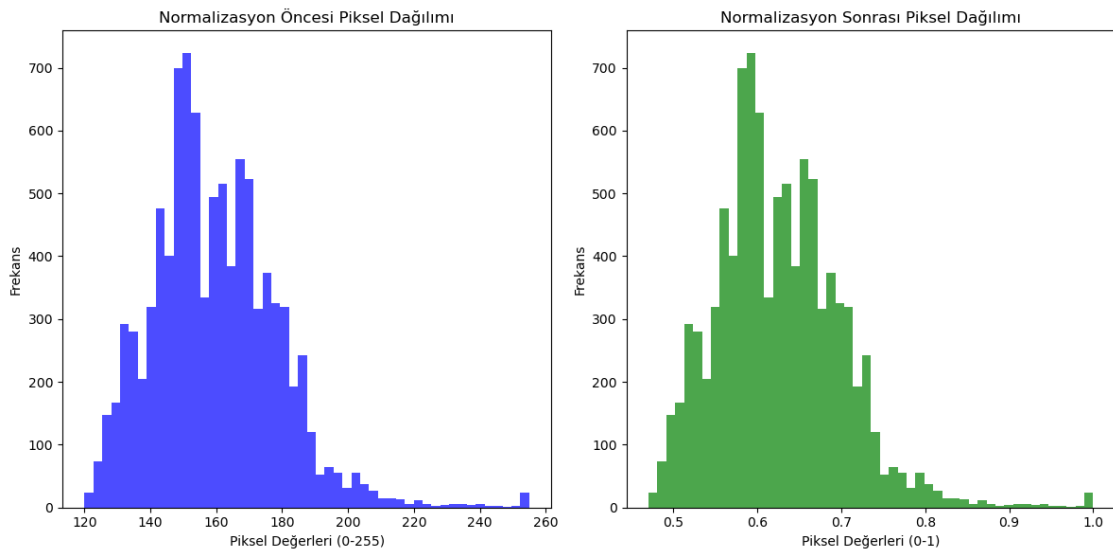
# Histogramları çiz
plt.figure(figsize=(12, 6))

# Normalizasyon öncesi
plt.subplot(1, 2, 1)
plt.hist(original_array.flatten(), bins=50, color='blue', alpha=0.7)
plt.title('Normalizasyon Öncesi Piksel Dağılımı')
plt.xlabel('Piksel Değerleri (0-255)')
plt.ylabel('Frekans')

# Normalizasyon sonrası
plt.subplot(1, 2, 2)
plt.hist(normalized_array.flatten(), bins=50, color='green', alpha=0.7)
plt.title('Normalizasyon Sonrası Piksel Dağılımı')
plt.xlabel('Piksel Değerleri (0-1)')
plt.ylabel('Frekans')

plt.tight_layout()
plt.show()

```



```
[37]: import numpy as np
import matplotlib.pyplot as plt
from PIL import Image

# Örnek bir görsel yolu (kendi veri yolunuza göre güncelleyin)
image_path = 'C:
↪\\Users\\sulta\\Desktop\\GC\\02-uodev\\data\\Train\\1\\00001_00000_00000.png'

# Görseli yükle ve piksel değerlerini al
original_image = Image.open(image_path).convert('RGB')
original_pixels = np.array(original_image).flatten() # Görüntüyü düzleştir

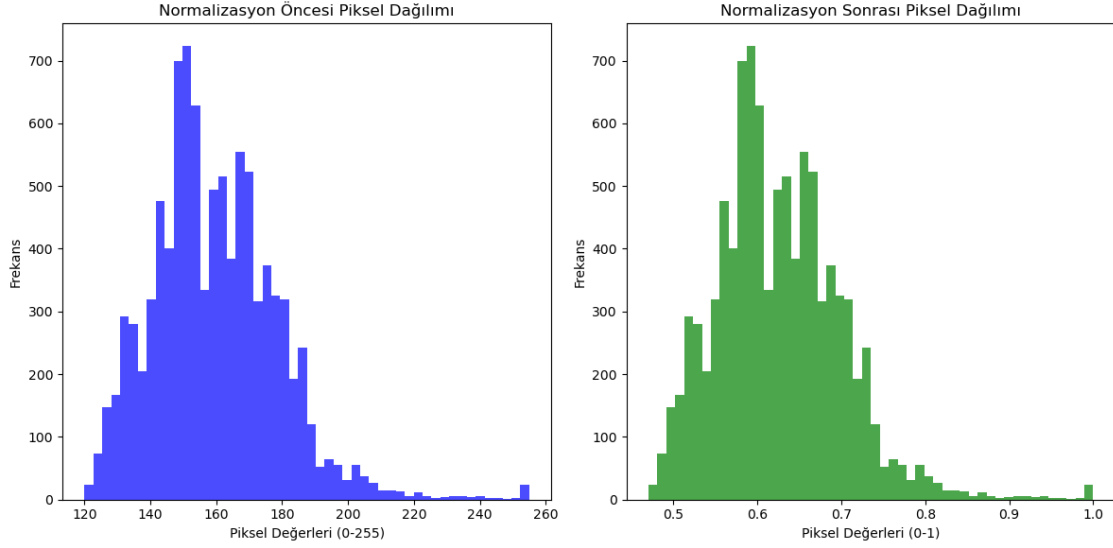
# Görseli normalize et (0-1 aralığına getir)
normalized_pixels = original_pixels / 255.0

# Grafik oluştur
plt.figure(figsize=(12, 6))

# Normalizasyon öncesi histogram
plt.subplot(1, 2, 1)
plt.hist(original_pixels, bins=50, color='blue', alpha=0.7)
plt.title("Normalizasyon Öncesi Piksel Dağılımı")
plt.xlabel("Piksel Değerleri (0-255)")
plt.ylabel("Frekans")

# Normalizasyon sonrası histogram
plt.subplot(1, 2, 2)
plt.hist(normalized_pixels, bins=50, color='green', alpha=0.7)
plt.title("Normalizasyon Sonrası Piksel Dağılımı")
plt.xlabel("Piksel Değerleri (0-1)")
plt.ylabel("Frekans")

plt.tight_layout()
plt.show()
```



```
[38]: model.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
flatten_1 (Flatten)	(None, 3072)	0
dense_2 (Dense)	(None, 128)	393,344
dense_3 (Dense)	(None, 43)	5,547

Total params: 1,196,675 (4.56 MB)

Trainable params: 398,891 (1.52 MB)

Non-trainable params: 0 (0.00 B)

Optimizer params: 797,784 (3.04 MB)

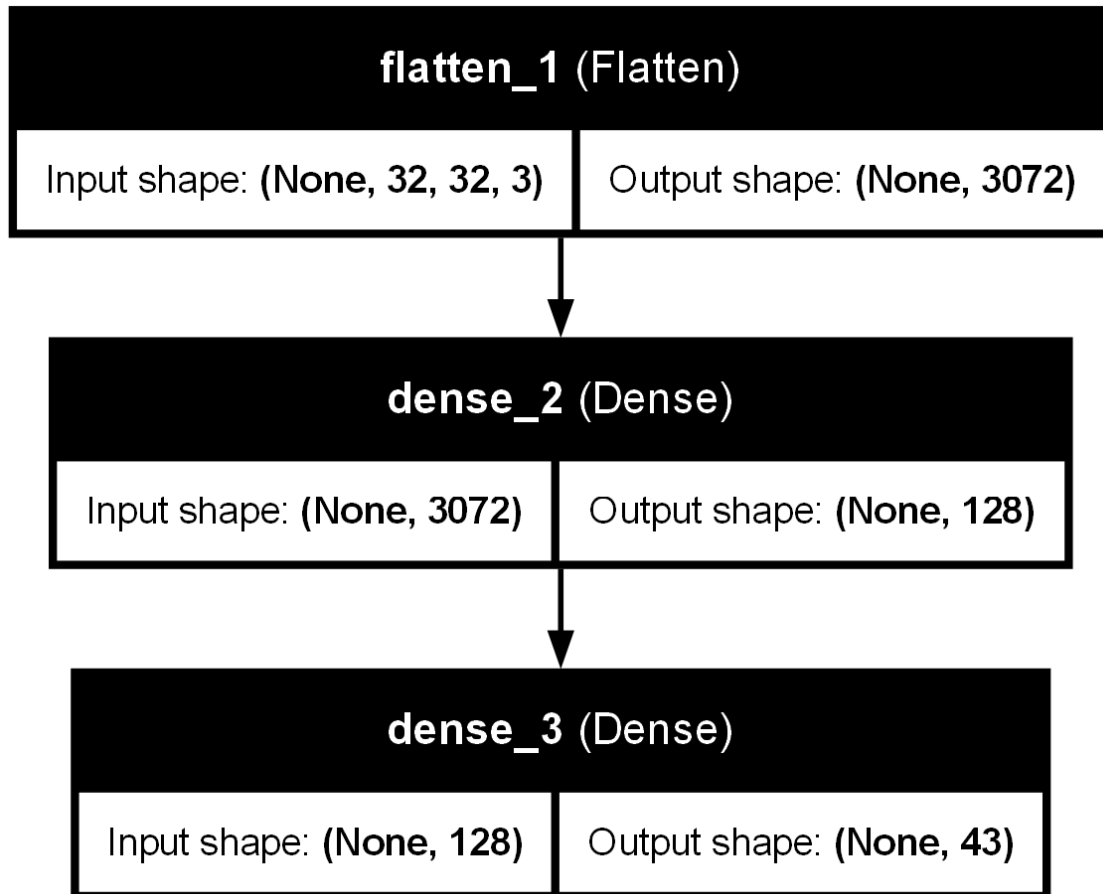
```
[39]: pip install pydot
```

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: pydot in

```
c:\users\sulta\appdata\roaming\python\python312\site-packages (3.0.2)
Requirement already satisfied: pyparsing>=3.0.9 in
c:\programdata\anaconda3\lib\site-packages (from pydot) (3.1.2)
Note: you may need to restart the kernel to use updated packages.
```

```
[40]: from tensorflow.keras.utils import plot_model
      plot_model(model, to_file='model.png', show_shapes=True, show_layer_names=True)
```

[40]:

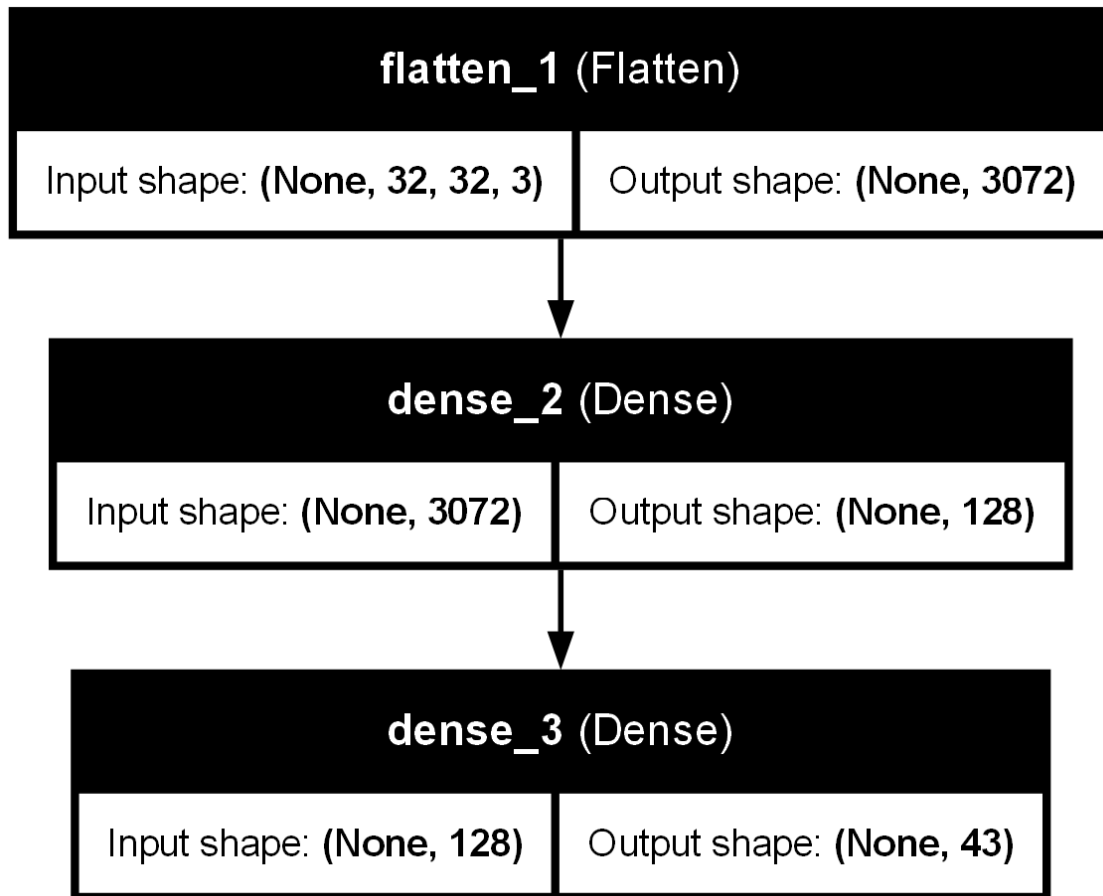


```
[41]: pip install graphviz
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: graphviz in
c:\users\sulta\appdata\roaming\python\python312\site-packages (0.20.3)
Note: you may need to restart the kernel to use updated packages.
```

```
[42]: from tensorflow.keras.utils import plot_model
      plot_model(model, to_file='model.png', show_shapes=True, show_layer_names=True)
```

[42]:



```
[43]: from keras.models import Sequential
      from keras.layers import Dense, Flatten
      from keras.optimizers import Adam
      import pandas as pd

      # Hiperparametre denemeleri için değerler
      epochs_list = [10, 20, 30]
      learning_rates = [0.001, 0.01, 0.1]
      batch_sizes = [32, 64, 128]

      # Sonuçları saklamak için bir liste
      results = []

      # Hiperparametre denemeleri
      for epochs in epochs_list:
          for lr in learning_rates:
              for batch_size in batch_sizes:
```

```

print(f"Epochs: {epochs}, Learning Rate: {lr}, Batch Size: {batch_size}")

# Modeli oluřtur
model = Sequential([
    Flatten(input_shape=(32, 32, 3)),
    Dense(128, activation='relu'),
    Dense(len(train_labels[0]), activation='softmax')
])

# Modeli derle
optimizer = Adam(learning_rate=lr)
model.compile(optimizer=optimizer, loss='categorical_crossentropy',
metrics=['accuracy'])

# Modeli eęit
history = model.fit(
    train_images, train_labels,
    epochs=epochs,
    batch_size=batch_size,
    validation_data=(test_images, test_labels),
    verbose=0
)

# Sonuęları kaydet
final_train_acc = history.history['accuracy'][-1]
final_val_acc = history.history['val_accuracy'][-1]
final_train_loss = history.history['loss'][-1]
final_val_loss = history.history['val_loss'][-1]

results.append({
    'Epochs': epochs,
    'Learning Rate': lr,
    'Batch Size': batch_size,
    'Train Accuracy': final_train_acc,
    'Validation Accuracy': final_val_acc,
    'Train Loss': final_train_loss,
    'Validation Loss': final_val_loss
})

# Sonuęları tabloya ęevir
results_df = pd.DataFrame(results)

```

Epochs: 10, Learning Rate: 0.001, Batch Size: 32

C:\Users\sulta\AppData\Roaming\Python\Python312\site-packages\keras\src\layers\reshaping\flatten.py:37: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models,

```
prefer using an `Input(shape)` object as the first layer in the model instead.
super().__init__(**kwargs)
```

```
Epochs: 10, Learning Rate: 0.001, Batch Size: 64
Epochs: 10, Learning Rate: 0.001, Batch Size: 128
Epochs: 10, Learning Rate: 0.01, Batch Size: 32
Epochs: 10, Learning Rate: 0.01, Batch Size: 64
Epochs: 10, Learning Rate: 0.01, Batch Size: 128
Epochs: 10, Learning Rate: 0.1, Batch Size: 32
Epochs: 10, Learning Rate: 0.1, Batch Size: 64
Epochs: 10, Learning Rate: 0.1, Batch Size: 128
Epochs: 20, Learning Rate: 0.001, Batch Size: 32
Epochs: 20, Learning Rate: 0.001, Batch Size: 64
Epochs: 20, Learning Rate: 0.001, Batch Size: 128
Epochs: 20, Learning Rate: 0.01, Batch Size: 32
Epochs: 20, Learning Rate: 0.01, Batch Size: 64
Epochs: 20, Learning Rate: 0.01, Batch Size: 128
Epochs: 20, Learning Rate: 0.1, Batch Size: 32
Epochs: 20, Learning Rate: 0.1, Batch Size: 64
Epochs: 20, Learning Rate: 0.1, Batch Size: 128
Epochs: 30, Learning Rate: 0.001, Batch Size: 32
Epochs: 30, Learning Rate: 0.001, Batch Size: 64
Epochs: 30, Learning Rate: 0.001, Batch Size: 128
Epochs: 30, Learning Rate: 0.01, Batch Size: 32
Epochs: 30, Learning Rate: 0.01, Batch Size: 64
Epochs: 30, Learning Rate: 0.01, Batch Size: 128
Epochs: 30, Learning Rate: 0.1, Batch Size: 32
Epochs: 30, Learning Rate: 0.1, Batch Size: 64
Epochs: 30, Learning Rate: 0.1, Batch Size: 128
```

```
[44]: # İlk 5 sonucu görüntüleyin
print(results_df.head())

# Tabloyu kaydet
results_df.to_csv("hyperparameter_results.csv", index=False)
```

	Epochs	Learning Rate	Batch Size	Train Accuracy	Validation Accuracy	\
0	10	0.001	32	0.057028	0.059382	
1	10	0.001	64	0.057461	0.059382	
2	10	0.001	128	0.101405	0.105938	
3	10	0.010	32	0.055880	0.057007	
4	10	0.010	64	0.055395	0.059382	

	Train Loss	Validation Loss
0	3.485910	3.463017
1	3.485574	3.464263
2	3.254200	3.217836
3	3.488465	3.464034
4	3.487822	3.464427

```
[45]: # En yüksek Validation Accuracy'e göre sıralama
best_result = results_df.loc[results_df['Validation Accuracy'].idxmax()]
print("En iyi hiperparametre kombinasyonu:")
print(best_result)
```

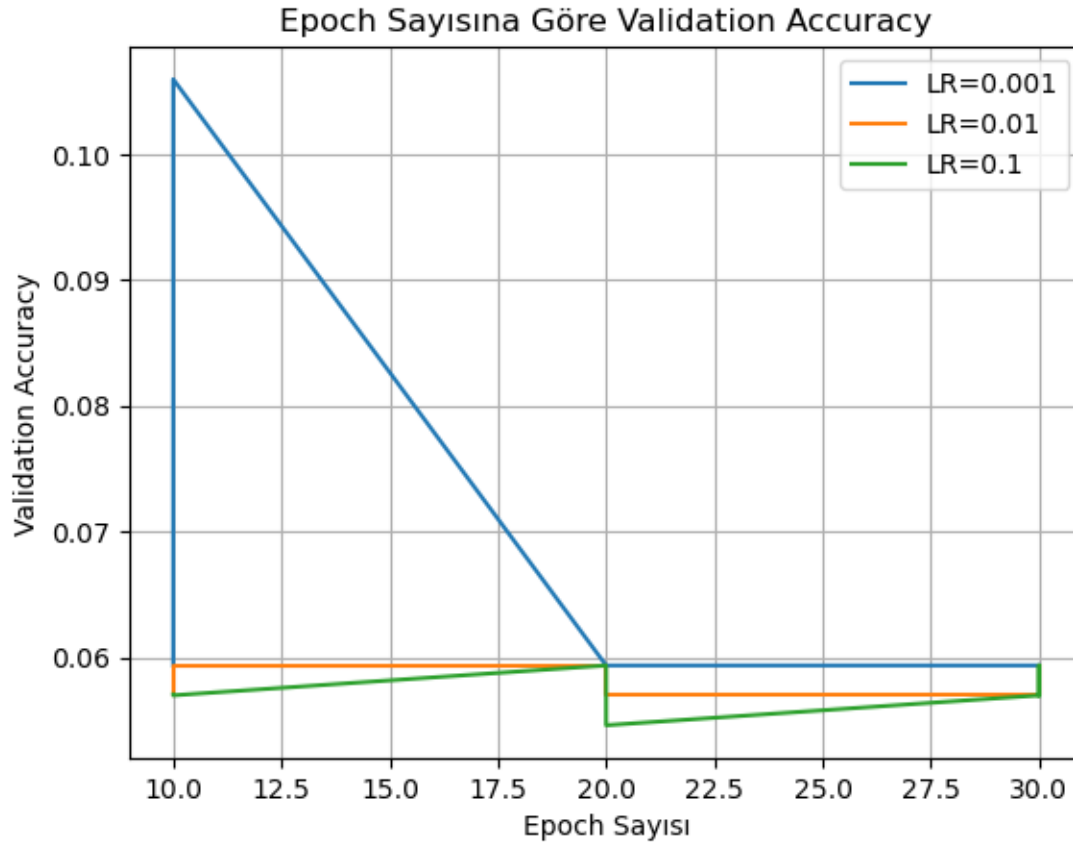
En iyi hiperparametre kombinasyonu:

Epochs	10.000000
Learning Rate	0.001000
Batch Size	128.000000
Train Accuracy	0.101405
Validation Accuracy	0.105938
Train Loss	3.254200
Validation Loss	3.217836
Name: 2, dtype: float64	

```
[46]: import matplotlib.pyplot as plt

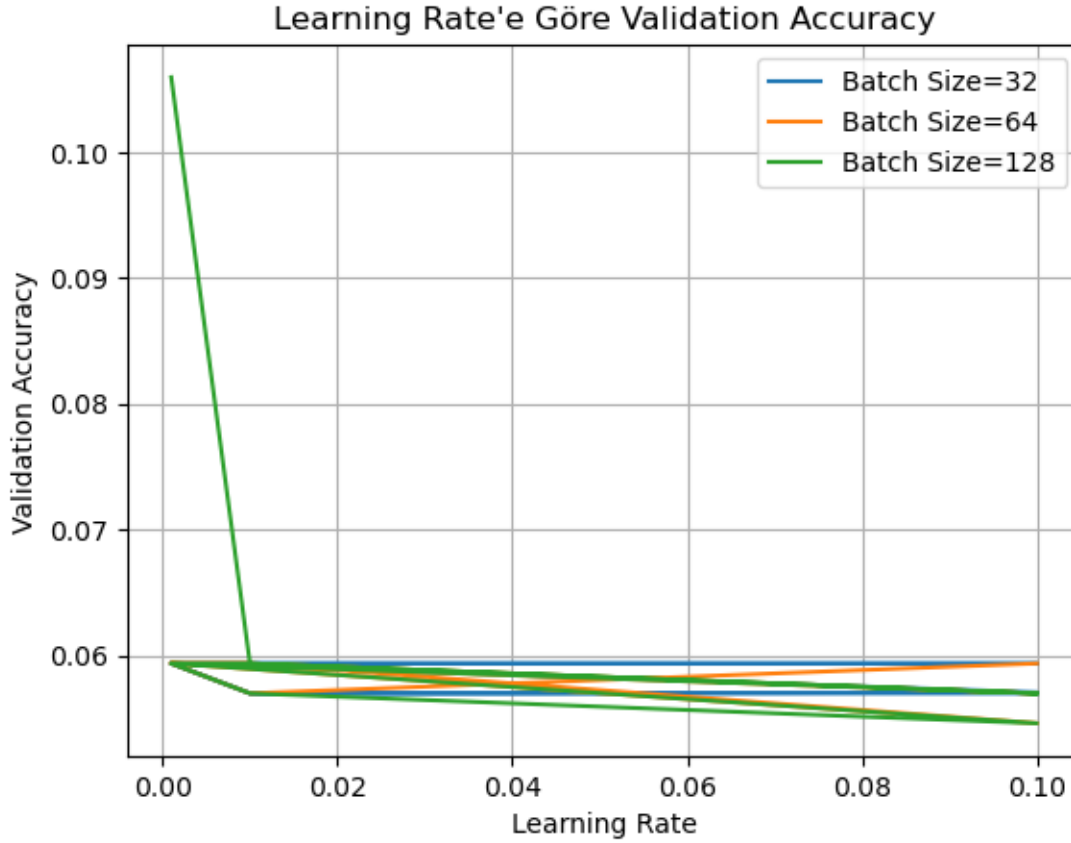
# Epoch - Validation Accuracy grafiği
for lr in learning_rates:
    subset = results_df[results_df['Learning Rate'] == lr]
    plt.plot(subset['Epochs'], subset['Validation Accuracy'], label=f"LR={lr}")

plt.title("Epoch Sayısına Göre Validation Accuracy")
plt.xlabel("Epoch Sayısı")
plt.ylabel("Validation Accuracy")
plt.legend()
plt.grid(True)
plt.show()
```



```
[47]: # Learning Rate - Validation Accuracy grafiği
for batch_size in batch_sizes:
    subset = results_df[results_df['Batch Size'] == batch_size]
    plt.plot(subset['Learning Rate'], subset['Validation Accuracy'],
             label=f"Batch Size={batch_size}")

plt.title("Learning Rate'e Göre Validation Accuracy")
plt.xlabel("Learning Rate")
plt.ylabel("Validation Accuracy")
plt.legend()
plt.grid(True)
plt.show()
```



```
[48]: import matplotlib.pyplot as plt

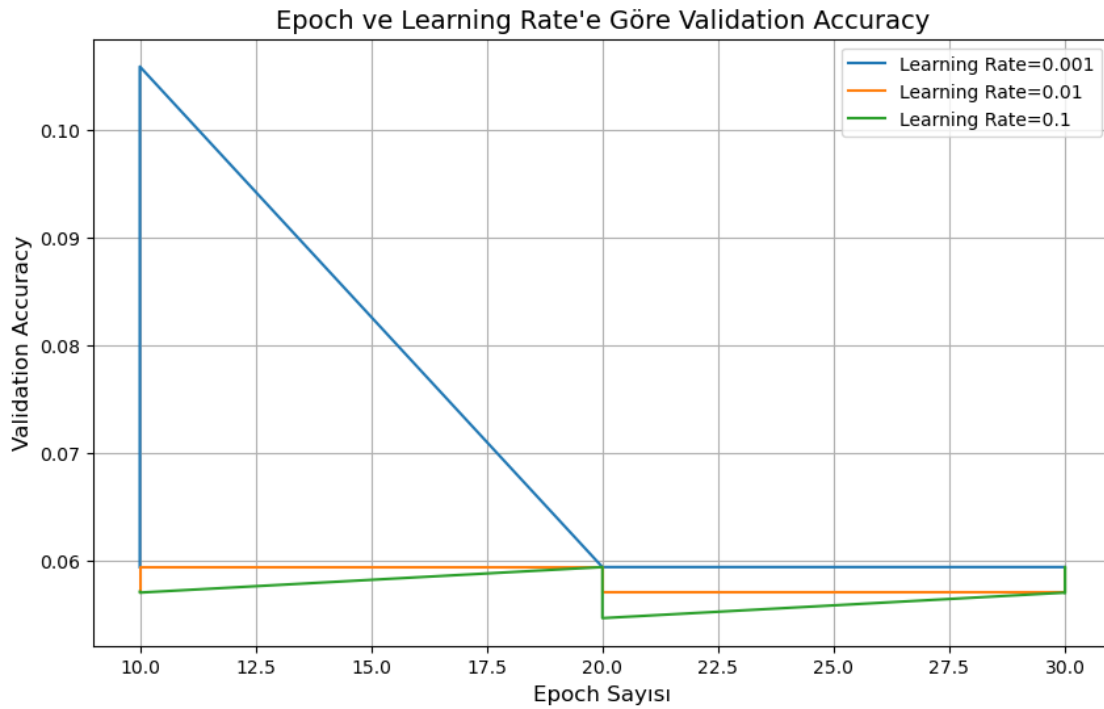
# Hiperparametre sonuçlarının dataframe'inden farklı learning rate'leri filtrele
learning_rates = results_df['Learning Rate'].unique()

# Grafik oluştur
plt.figure(figsize=(10, 6))

# Her bir learning rate için epoch doğruluğunu çizdir
for lr in learning_rates:
    subset = results_df[results_df['Learning Rate'] == lr]
    plt.plot(subset['Epochs'], subset['Validation Accuracy'], label=f'Learning_Rate={lr}')

# Grafik ayarları
plt.title('Epoch ve Learning Rate\'e Göre Validation Accuracy', fontsize=14)
plt.xlabel('Epoch Sayısı', fontsize=12)
plt.ylabel('Validation Accuracy', fontsize=12)
plt.legend()
plt.grid(True)
```

```
plt.show()
```



```
[49]: from sklearn.svm import SVC
from sklearn.preprocessing import StandardScaler
from sklearn.pipeline import Pipeline
from sklearn.metrics import classification_report, confusion_matrix
import numpy as np

# Veri standardizasyonu için pipeline oluştur
pipeline = Pipeline([
    ('scaler', StandardScaler()),
    ('svm', SVC(kernel='rbf', gamma='scale'))
])

# Modeli eđit
pipeline.fit(train_images.reshape(len(train_images), -1), np.
    argmax(train_labels, axis=1))

# Tahmin yap
predictions = pipeline.predict(test_images.reshape(len(test_images), -1))

# Performans deęerlendirme
print("Classification Report:")
print(classification_report(np.argmax(test_labels, axis=1), predictions))
```

```
print("Confusion Matrix:")
print(confusion_matrix(np.argmax(test_labels, axis=1), predictions))
```

Classification Report:

	precision	recall	f1-score	support
0	0.92	0.20	0.33	60
1	0.63	0.84	0.72	720
2	0.63	0.69	0.66	750
3	0.45	0.47	0.46	450
4	0.59	0.67	0.63	660
5	0.46	0.63	0.53	630
6	0.53	0.41	0.47	150
7	0.81	0.58	0.68	450
8	0.58	0.77	0.66	450
9	0.92	0.71	0.80	480
10	0.75	0.90	0.82	660
11	0.67	0.83	0.74	420
12	0.90	0.91	0.91	690
13	0.97	0.98	0.98	720
14	0.96	0.79	0.87	270
15	0.91	0.68	0.78	210
16	0.97	0.67	0.79	150
17	0.97	0.79	0.87	360
18	0.61	0.49	0.55	390
19	0.73	0.50	0.59	60
20	0.78	0.59	0.67	90
21	0.94	0.34	0.50	90
22	0.89	0.71	0.79	120
23	0.37	0.26	0.31	150
24	0.76	0.14	0.24	90
25	0.72	0.88	0.79	480
26	0.74	0.73	0.73	180
27	0.86	0.10	0.18	60
28	0.49	0.45	0.47	150
29	0.77	0.44	0.56	90
30	0.16	0.05	0.07	150
31	0.65	0.78	0.71	270
32	0.43	0.43	0.43	60
33	0.94	0.87	0.90	210
34	0.99	0.82	0.90	120
35	0.88	0.64	0.74	390
36	0.96	0.68	0.79	120
37	0.95	0.67	0.78	60
38	0.77	0.92	0.84	690
39	1.00	0.66	0.79	90
40	0.98	0.63	0.77	90

41	0.58	0.58	0.58	60
42	0.68	0.31	0.43	90
accuracy			0.72	12630
macro avg	0.75	0.61	0.65	12630
weighted avg	0.73	0.72	0.71	12630

Confusion Matrix:

```
[[ 12  43   2 ...   0   0   0]
 [   0 608  67 ...   0   0   0]
 [   0   7 519 ...   0   0   0]
 ...
 [   0   0   2 ...  57   0   0]
 [   0   0   0 ...   0  35   1]
 [   0   5   0 ...   0  19  28]]
```

```
[50]: import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix
import numpy as np

# Confusion Matrix oluřtur
cm = confusion_matrix(np.argmax(test_labels, axis=1), predictions)

# Grselleřtir
plt.figure(figsize=(10, 8))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=range(len(np.
    unique(np.argmax(test_labels, axis=1)))),
            yticklabels=range(len(np.unique(np.argmax(test_labels, axis=1)))))
plt.xlabel('Tahmin Edilen Etiket')
plt.ylabel('Gerek Etiket')
plt.title('Confusion Matrix')
plt.show()
```

