

Derin Öğrenme Ve Evrişimli Sinir Ağları

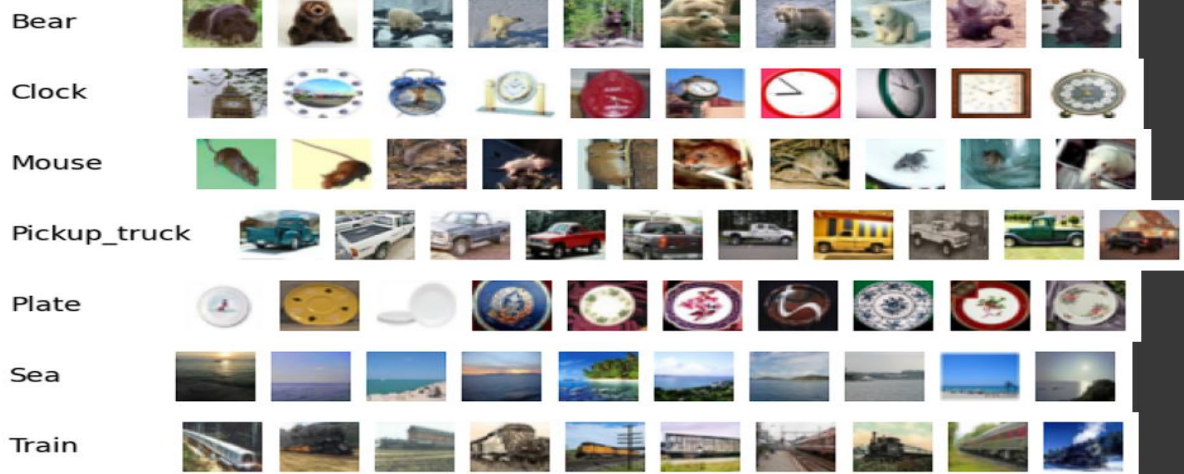
Proje Ödevi

Öğrenci Adı: Enes Buğra Turğut

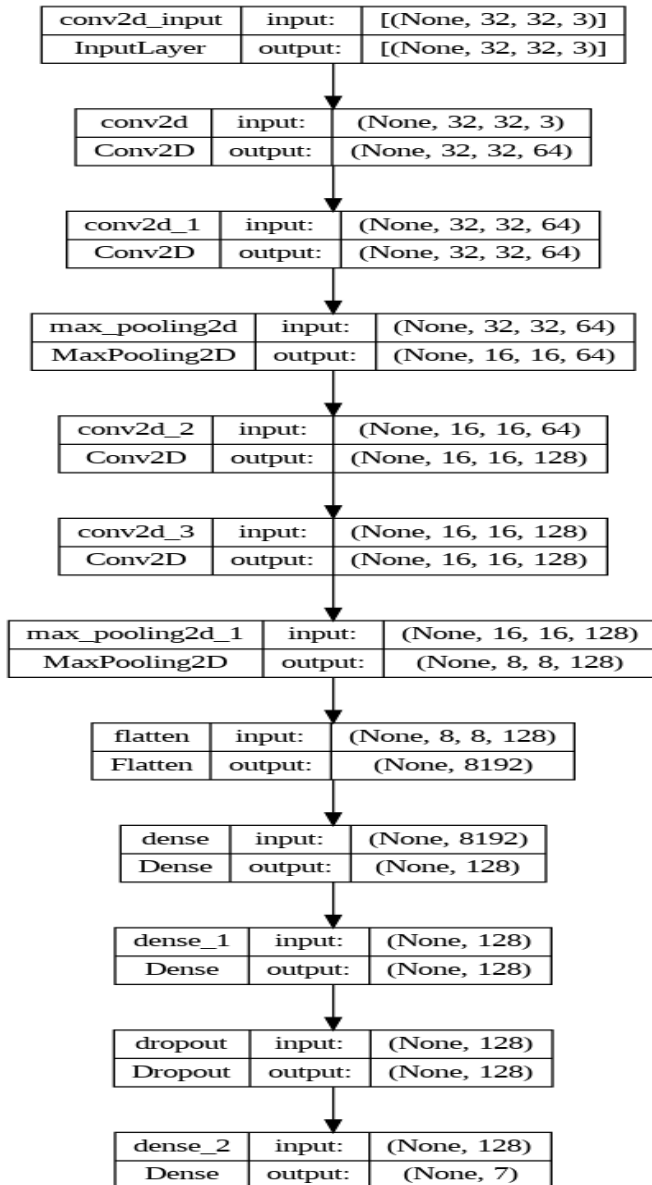
Öğrenci Numarası: G211210372

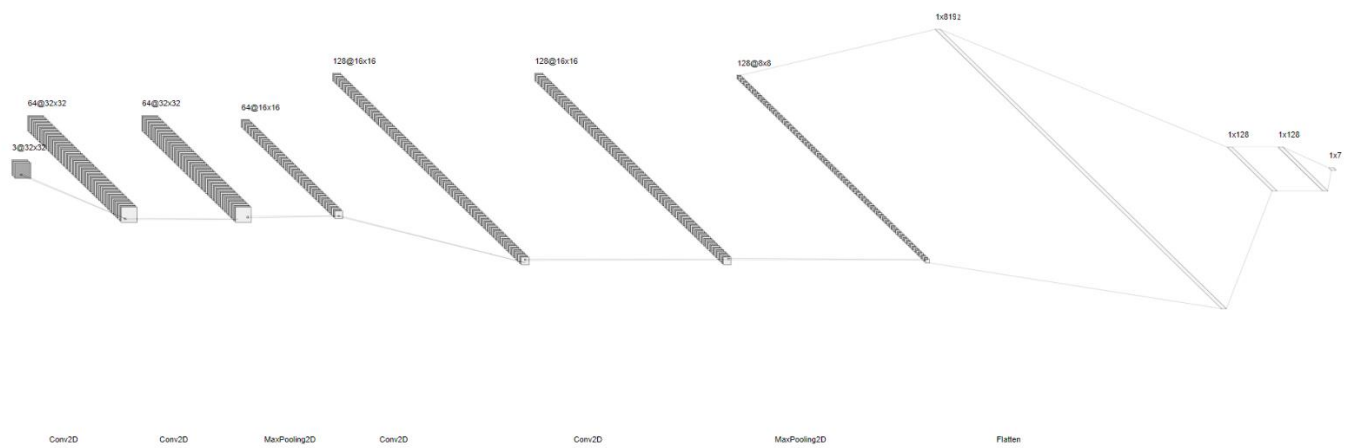
Öğrenim Türü: 2.Öğretim A grubu

1.Kullanılan sınıflar için örnek görüntüler

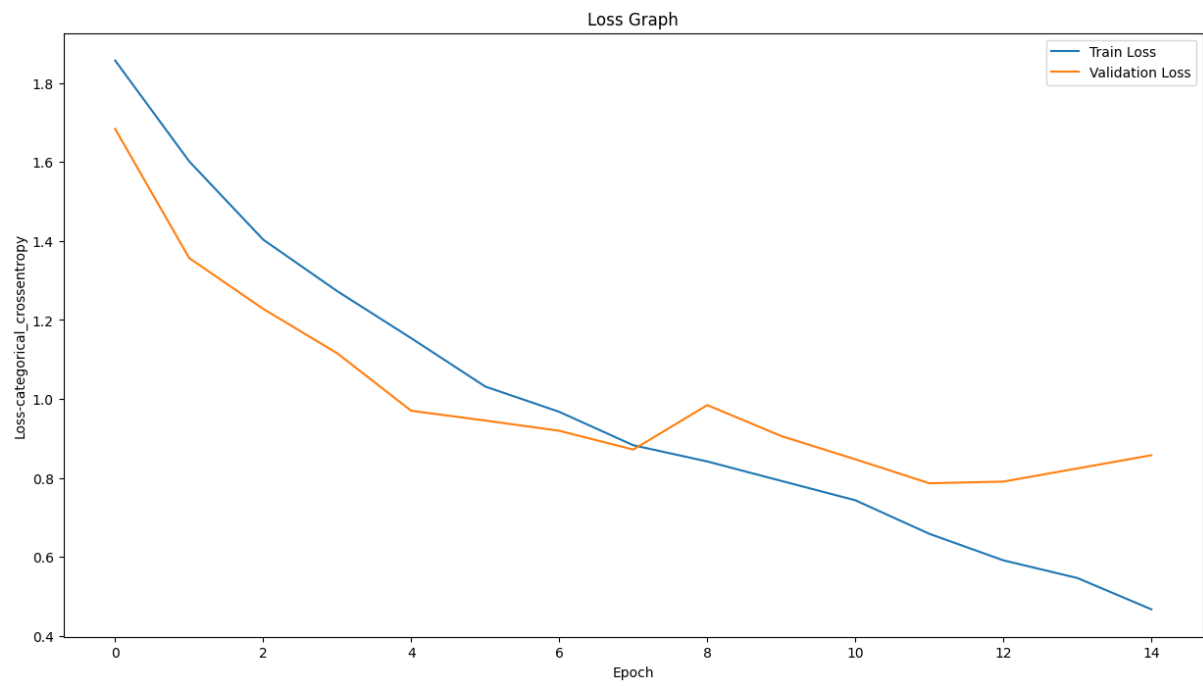


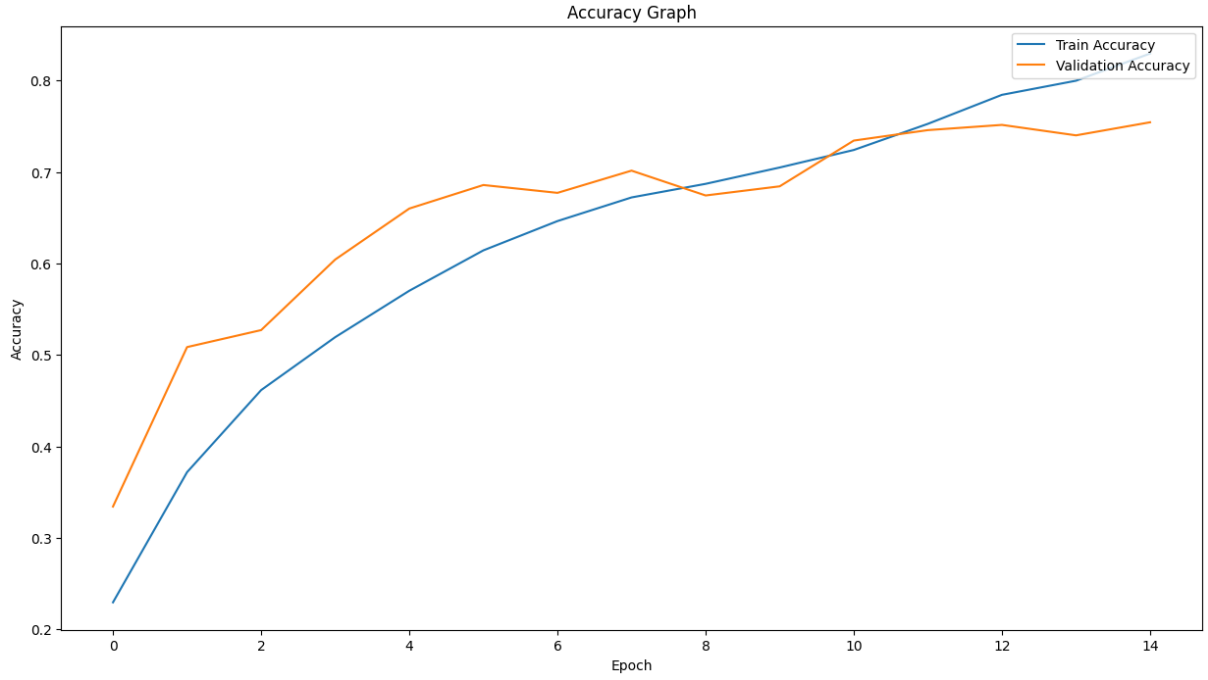
2.Blok Şeması





3.Grafikler





Modelin kayıp oranı her epoch ta düşerek öğrenmenin git gide iyiye ilerlediğini gösterir. Bununla birlikte validation verisinin de kayıp oranı düşmektedir fakat 10.epochtan sonra val_loss değeri iyileşmemiş ve model overfit olmaya başlamıştır.

Accuracy grafiğinde ise model öğrenmesine devam ederken yine 10.epochtan sonra model overfit olmaya başlamıştır. Yani eğitim verilerini ezberliyor görmediği verilere karşı kötü sonuçlar çıkarıyor.

Modeli eğitirken erken durdurma fonksiyonu kullandım. Burada val_loss değerini takip ederek bu değer 3 epoch iyileşmemesi durumunda eğitimi durdurdum ve en iyi ağırlıklarla modeli kaydettim. Ve model 12. Epochta durdurulup buradaki ağırlıklarla kaydedilmiştir.

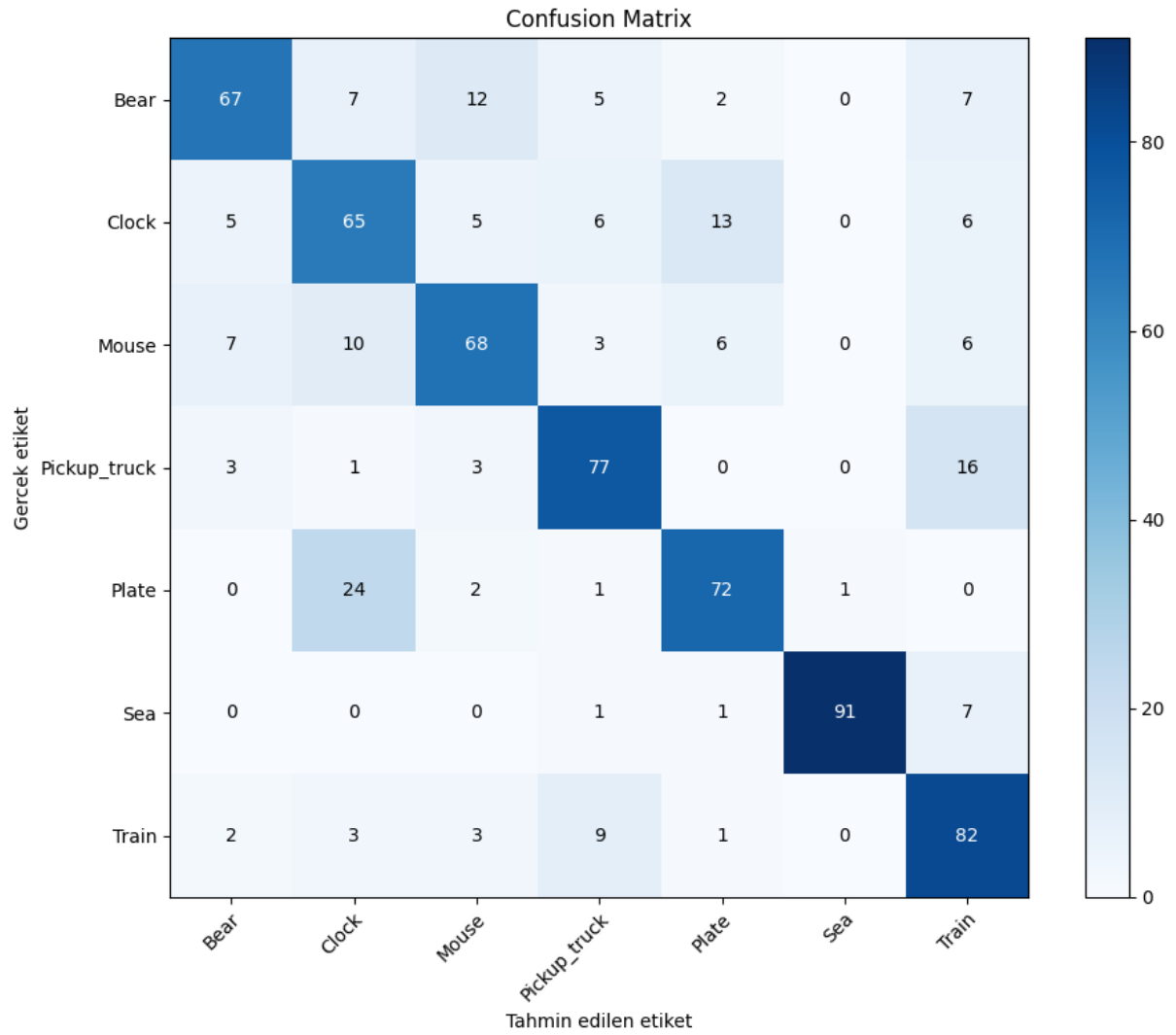
Modelin,

`train_accuracy = 0.7516`

`test_accuracy = 0.7457`

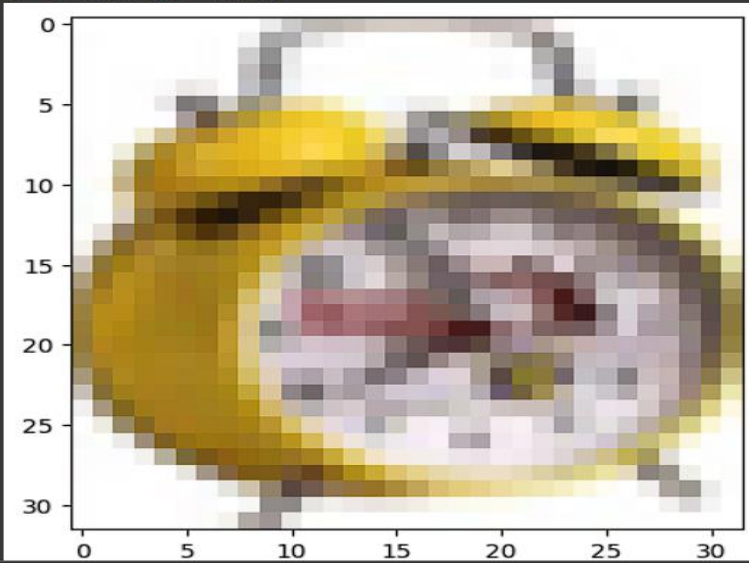
değeri elde edilmiştir. Model hiç görmediği verilere karşı eğitim doğruluğuna çok yakın bir sonuç vermiştir. Modelin hiç görmediği verilere karşı doğruluk performansı %74.57 diyebiliriz.

4.Confusion Matrix

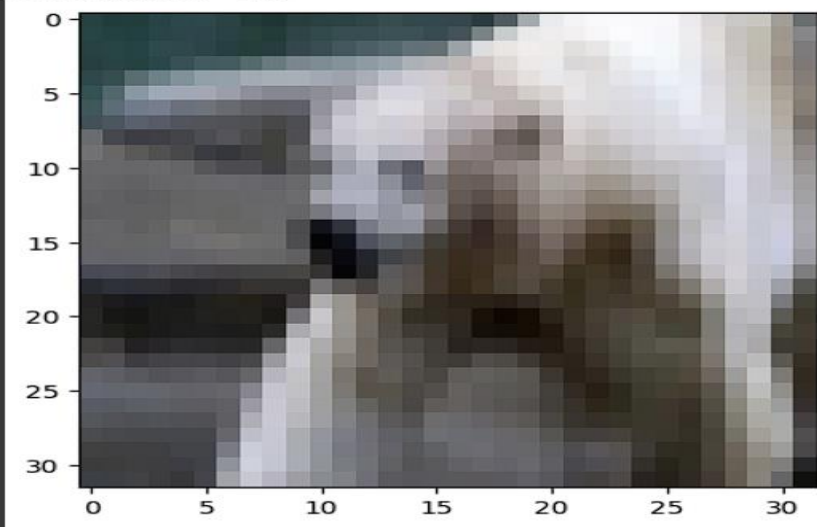


5.Predict

```
Vektor: [[1.0261596e-05 8.5142934e-01 7.0804270e-04 3.6879668e-05 1.4777270e-01  
2.9934572e-06 3.9830098e-05]]  
Tahmin sonucu: Clock  
Gercek sonuc: Clock
```



```
Vektor: [[0.91693324 0.01057601 0.03122309 0.00498996 0.001155 0.00096455  
0.03415816]]  
Tahmin sonucu: Bear  
Gercek sonuc: Bear
```

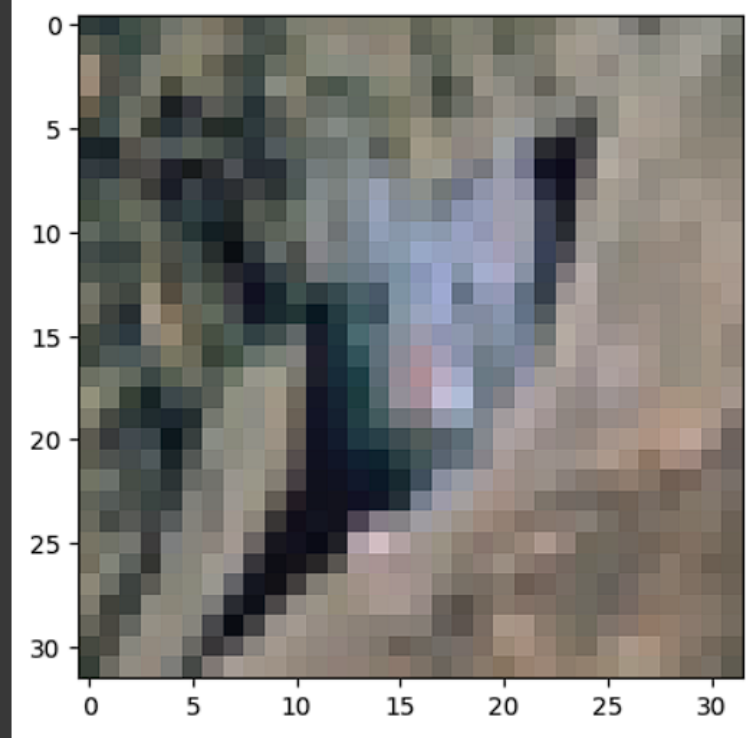


1/1 [=====] - 0s 19ms/step

Vektor: $\begin{bmatrix} 0.14043815 & 0.08516384 & 0.05403446 & 0.5049885 & 0.0278754 & 0.01110991 \\ 0.17638974 \end{bmatrix}$

Tahmin sonucu: Pickup Truck

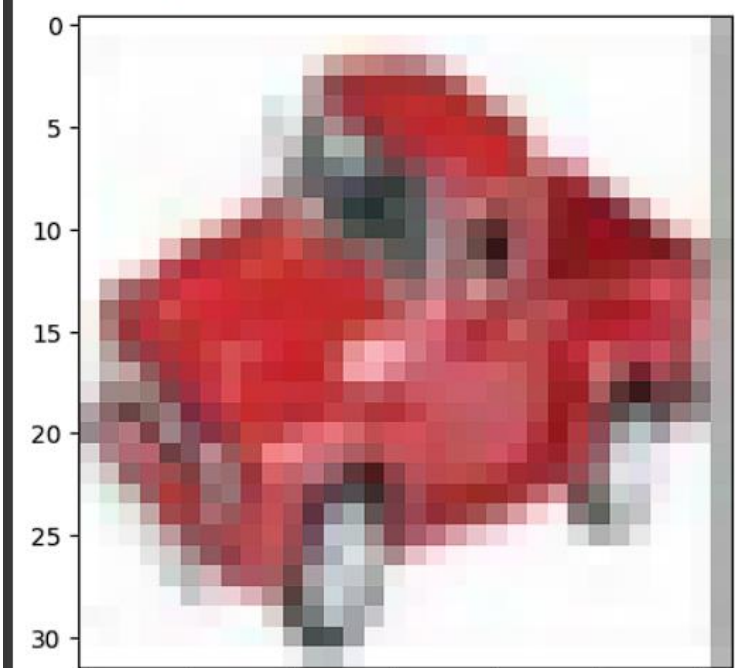
Gerçek sonuç: Mouse



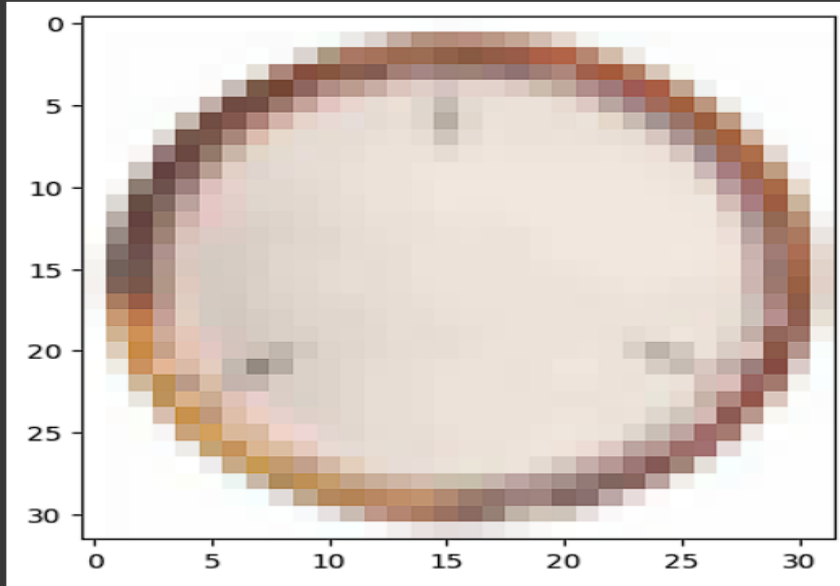
Vektor: $\begin{bmatrix} 1.9234882e-05 & 8.3243139e-03 & 6.7555825e-06 & 9.8529053e-01 & 2.4228402e-04 \\ 1.2558074e-06 & 6.1154654e-03 \end{bmatrix}$

Tahmin sonucu: Pickup Truck

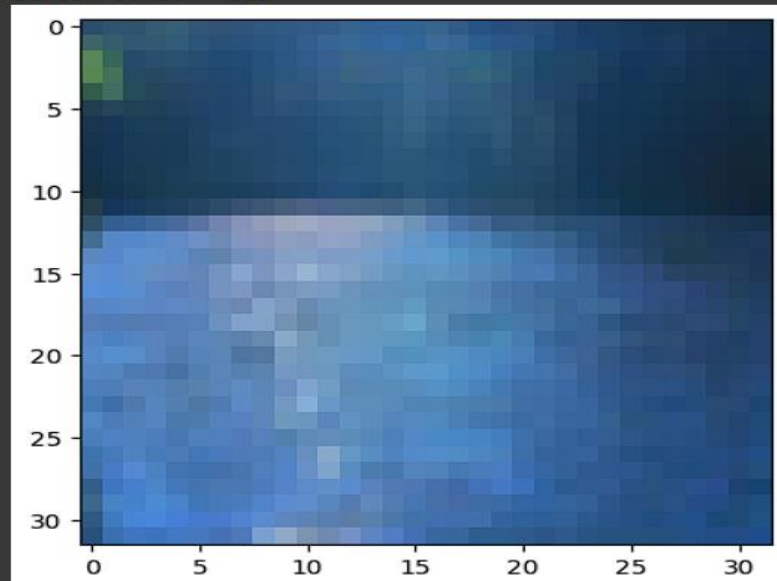
Gerçek sonuç: Pickup Truck



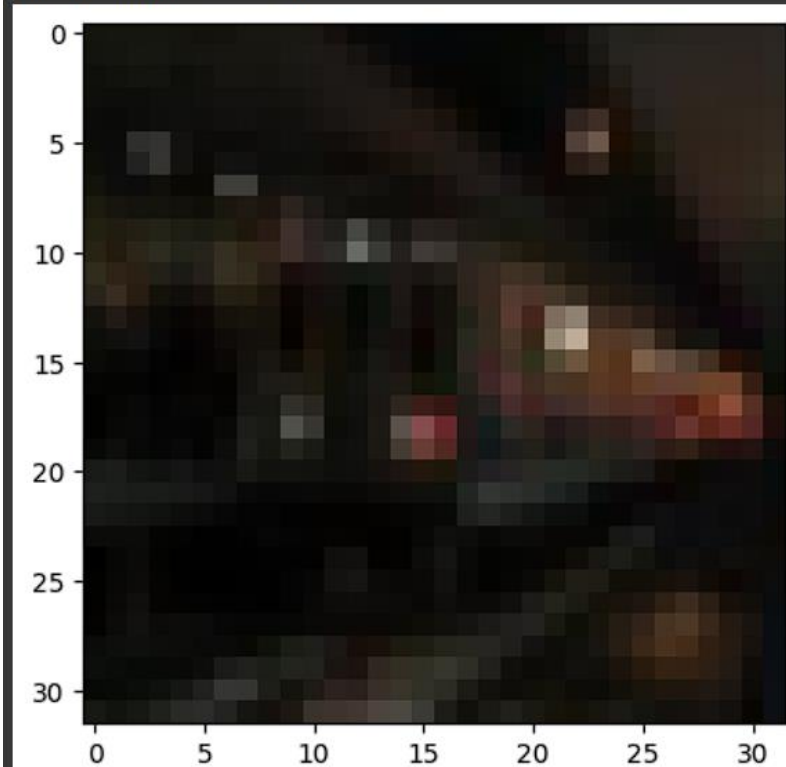
Vektor: $\begin{bmatrix} 3.04047482e-11 & 1.51648875e-02 & 4.37750280e-09 & 6.84158930e-15 \\ 9.84835148e-01 & 1.09809839e-13 & 1.49307053e-13 \end{bmatrix}$
Tahmin sonucu: Plate
Gerçek sonuc: Plate



Vektor: $\begin{bmatrix} 1.7570563e-03 & 7.7384501e-04 & 4.5205551e-04 & 6.0251619e-05 & 6.8244158e-04 \\ 9.9593687e-01 & 3.3758619e-04 \end{bmatrix}$
Tahmin sonucu: Sea
Gerçek sonuc: Sea




```
Vektor: [[0.17398056 0.02756971 0.20546351 0.03810682 0.00407665 0.00464054
 0.5461622 ]]
Tahmin sonucu: Train
Gerçek sonuc: Train
```



6.Kodlar

```
# Library
import numpy as np
from keras import layers
from keras import models
from keras.utils import to_categorical
from keras.datasets import cifar100
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix
import tensorflow as tf

# Load Data
(train_images,train_labels),(test_images,test_labels) = cifar100.load_data()

# Categories
### 3=bear 22=clock 50=mouse 58=pickup_truck 61=plate 71=sea 90=train

selected_classes = [3, 22, 50, 58, 61, 71, 90]

train_images_filtered, train_labels_filtered = [], []
test_images_filtered, test_labels_filtered = [], []
```

```

for i in range(len(train_labels)):
    if train_labels[i][0] in selected_classes:
        train_images_filtered.append(train_images[i])
        train_labels_filtered.append(train_labels[i][0])

for i in range(len(test_labels)):
    if test_labels[i][0] in selected_classes:
        test_images_filtered.append(test_images[i])
        test_labels_filtered.append(test_labels[i][0])

train_images_filtered = np.array(train_images_filtered)
train_labels_filtered = np.array(train_labels_filtered)
test_images_filtered = np.array(test_images_filtered)
test_labels_filtered = np.array(test_labels_filtered)

print("Filtrelenmis egitim veri seti boyutu:", train_images_filtered.shape)
print("Filtrelenmis test veri seti boyutu:", test_images_filtered.shape)

# Visualization
categories = ["Bear", "Clock", "Mouse", "Pickup_truck", "Plate", "Sea", "Train"]
selected_classes = [3, 22, 50, 58, 61, 71, 90]
def show(classes, images, labels, categories, num_images=10):
    for i in range(len(classes)):
        category_images = [images[j] for j in range(len(labels)) if labels[j]
== classes[i]]

        fig, axes = plt.subplots(1, num_images, figsize=(7, 2))
        fig.text(0.01, 0.5, categories[i], ha='center', va='center',
fontsize=12)

        for k in range(min(num_images, len(category_images))): #secilen 10dan
az olabilir bu yuzden min alinir.
            axes[k].imshow(category_images[k])
            axes[k].axis('off')
        plt.show()

show(selected_classes, train_images_filtered, train_labels_filtered,
categories)

# Data Augmentation
datagen =tf.keras.preprocessing.image.ImageDataGenerator(
    rotation_range=90,
    brightness_range=[0.4,1.5],
    width_shift_range=0.10,
    height_shift_range=0.10,
    shear_range=0.1,
    zoom_range=0.1,

```

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        horizontal_flip=True,
        vertical_flip=True,
        fill_mode='nearest'
    )

augmented_data =
datagen.flow(train_images_filtered,train_labels_filtered,batch_size=350)
train_images_augmented = train_images_filtered.copy()
train_labels_augmented = train_labels_filtered.copy()

for i in range(10):
    images, labels = next(augmented_data)
    train_images_augmented = np.concatenate([train_images_augmented, images])
    train_labels_augmented = np.concatenate([train_labels_augmented, labels])
train_images_augmented.shape

# Data Preprocessing

label_mapping = {3: 0, 22: 1, 50: 2, 58: 3, 61: 4, 71: 5, 90: 6}

train_labels_augmented = np.array([label_mapping[label] for label in
train_labels_augmented])
test_labels_filtered = np.array([label_mapping[label] for label in
test_labels_filtered])

train_images_augmented = train_images_augmented.reshape((7000,32,32,3))
train_images_augmented = train_images_augmented.astype('float32') /255

test_images_filtered = test_images_filtered.reshape((700,32,32,3))
test_images_filtered = test_images_filtered.astype('float32') /255

train_labels_augmented = to_categorical(train_labels_augmented)
test_labels_filtered = to_categorical(test_labels_filtered)

# Model
model = models.Sequential()

model.add(layers.Conv2D(
    64,
    (3,3),
    padding='same',
    input_shape=(32,32,3),
    activation='relu'
))
model.add(layers.Conv2D(
    64,

```

```

        (3,3),
        padding='same',
        activation='relu'
    ))

model.add(layers.MaxPool2D((2,2)))

model.add(layers.Conv2D(
    128,
    (3,3),
    padding='same',
    activation='relu'
))
model.add(layers.Conv2D(
    128,
    (3,3),
    padding='same',
    activation='relu'
))

model.add(layers.MaxPool2D((2,2)))

model.add(layers.Flatten())

model.add(layers.Dense(128,activation='relu'))
model.add(layers.Dense(128,activation='relu'))
model.add(layers.Dropout(0.5))

model.add(layers.Dense(7,activation='softmax'))

model.summary()

# Block Schema
import keras
keras.utils.plot_model(model,"model.png",show_shapes=True)

# Model Compile
early_stop = tf.keras.callbacks.EarlyStopping(monitor='val_loss',
                                                mode='min',
                                                patience = 3,
                                                verbose=1,
                                                restore_best_weights = True)

model.compile(
    optimizer='adam',
    loss='categorical_crossentropy',
    metrics=['accuracy', 'AUC', 'Recall']

```

```

)

# Model Fit

history = model.fit(train_images_augmented
                    ,train_labels_augmented
                    ,batch_size=256,
                    epochs=200,
                    validation_data=(test_images_filtered,test_labels_filtered
),
                    callbacks=[early_stop]
                    )

model.save("model.h5")

# Loss Val Loss Graph

plt.figure(figsize=(15,8))
plt.plot(history.history['loss'],label = 'Train Loss')
plt.plot(history.history['val_loss'],label = 'Validation Loss')
plt.ylabel('Loss-categorical_crossentropy')
plt.xlabel('Epoch')
plt.legend(loc='upper right')
plt.title('Loss Graph')
plt.show()

# Accuracy Val Accuracy Graph
plt.figure(figsize=(15,8))
plt.plot(history.history['accuracy'],label='Train Accuracy')
plt.plot(history.history['val_accuracy'],label='Validation Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(loc='upper right')
plt.title('Accuracy Graph')
plt.show()

# Model Evaluate
test_loss,test_acc,test_auc,test_recall =
model.evaluate(test_images_filtered,test_labels_filtered)
print("Test loss: ",test_loss)
print("Test acc: ",test_acc)
print("Test auc: ",test_auc)
print("Test recall: ",test_recall)

# Confusion Matrix
predictions = model.predict(test_images_filtered)
predicted_classes = np.argmax(predictions, axis=1)

true_classes = np.argmax(test_labels_filtered, axis=1)

```

```

cm = confusion_matrix(true_classes, predicted_classes)

class_labels = ["Bear", "Clock", "Mouse", "Pickup_truck", "Plate", "Sea", "Train"]

fig, ax = plt.subplots(figsize=(10, 8))
im = ax.imshow(cm, interpolation='nearest', cmap='Blues')
ax.figure.colorbar(im, ax=ax)
ax.set(xticks=np.arange(cm.shape[1]),
       yticks=np.arange(cm.shape[0]),
       xticklabels=class_labels, yticklabels=class_labels,
       title='Confusion Matrix',
       ylabel='Gerçek etiket',
       xlabel='Tahmin edilen etiket')

plt.setp(ax.get_xticklabels(), rotation=45, ha="right",
rotation_mode="anchor")
ax.set_xlim(-0.5, len(np.unique(class_labels))-0.5)
ax.set_ylim(len(np.unique(class_labels))-0.5, -0.5)

for i in range(cm.shape[0]):
    for j in range(cm.shape[1]):
        ax.text(j, i, format(cm[i, j], 'd'),
                ha="center", va="center",
                color="white" if cm[i, j] > cm.max() / 2. else "black")
plt.tight_layout()
plt.show()

# Model Test
map_list = {0: 'Bear', 1: 'Clock', 2: 'Mouse', 3: 'Pickup
Truck', 4: 'Plate', 5: 'Sea', 6: 'Train'}
def testModel(test_images_filtered, test_labels_filtered):
    rand = np.random.randint(0, len(test_images_filtered) - 1)
    test = test_images_filtered[rand]
    plt.imshow(test)
    test = test.reshape(1, 32, 32, 3)
    tahmin = model.predict(test)
    sonuc = np.argmax(tahmin)
    print("Vektor: ", tahmin)
    print("Tahmin sonucu: ", map_list[sonuc])

    gercek = test_labels_filtered[rand]
    gercek = np.argmax(gercek)
    print("Gerçek sonucu: ", map_list[gercek])

testModel(test_images_filtered, test_labels_filtered)

```