



SAKARYA
ÜNİVERSİTESİ

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**HAŞİM GÜRDAMAR BİLGİSAYAR
BİLİŞİM BİLİMLERİ FAKÜLTESİ
BİLGİSAYAR MÜHENDİSLİĞİ**

DERİN ÖĞRENME VE ERİŞİMLİ SİNİR AĞLARI

Ödev 2

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1.Eğitimde Kullanılan Sınıflar :

beetle:



can:



house:



mouse:



pickup_truck:



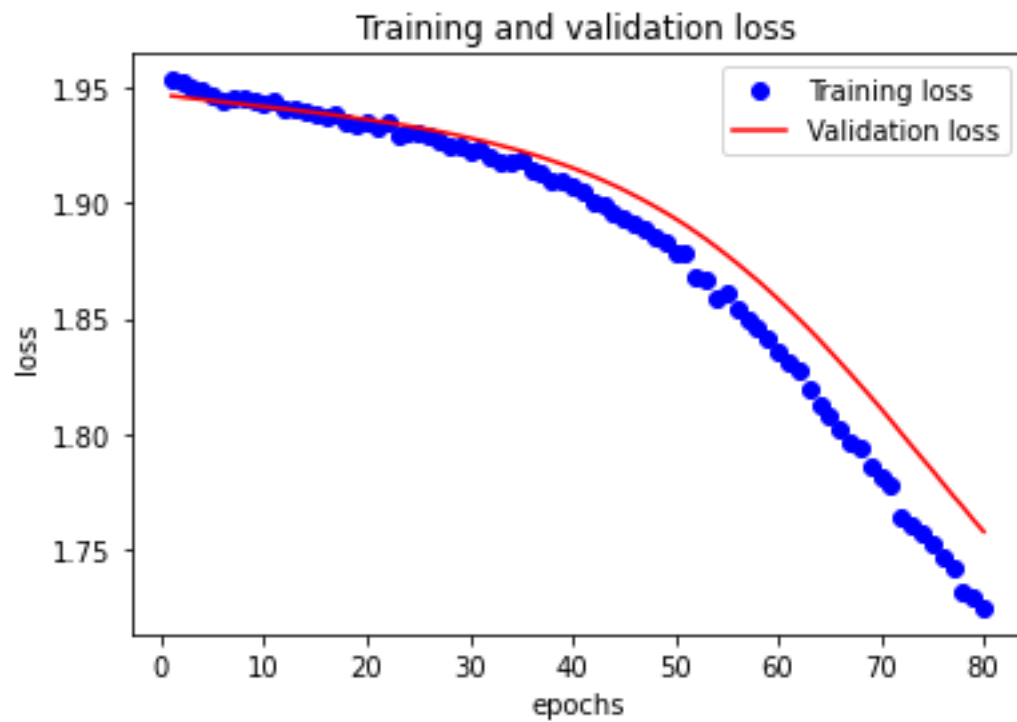
shark:



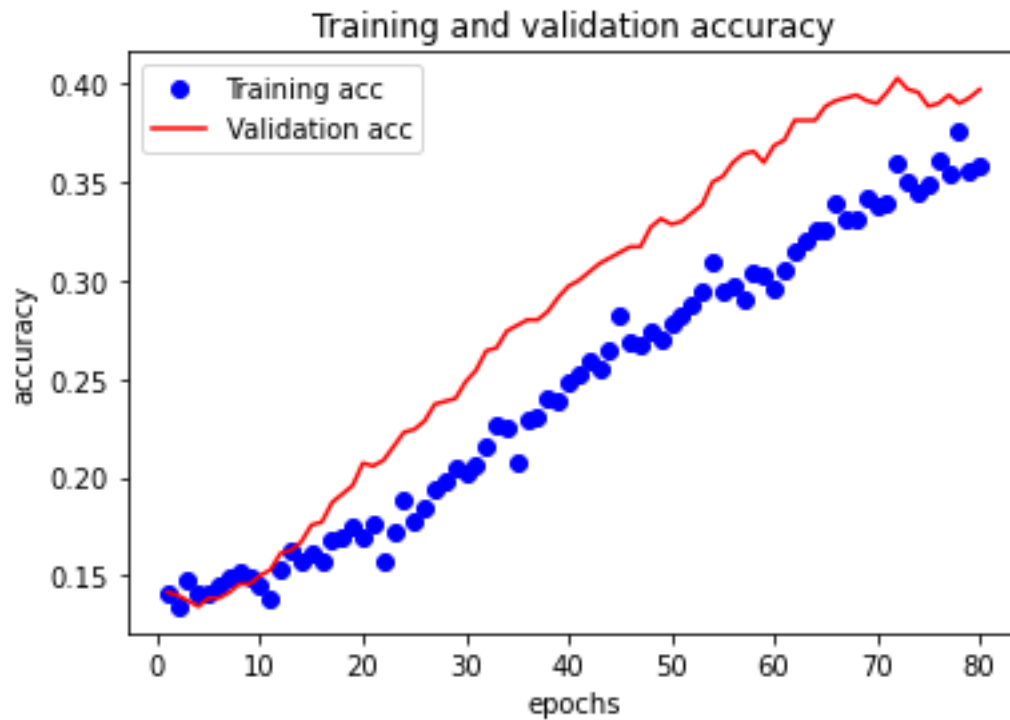
sweet_pepper:



Kayıp Grafiği:



Accuracy Grafiği:



Epoch, Loss ve Accuracy:

```
110/110 [=====] - 6s 53ms/step - loss: 1.8022 - acc: 0.3391 - val_loss: 1.8311 - val_acc: 0.3914
Epoch 67/80
110/110 [=====] - 6s 54ms/step - loss: 1.7969 - acc: 0.3309 - val_loss: 1.8262 - val_acc: 0.3929
Epoch 68/80
110/110 [=====] - 6s 53ms/step - loss: 1.7936 - acc: 0.3309 - val_loss: 1.8212 - val_acc: 0.3943
Epoch 69/80
110/110 [=====] - 6s 55ms/step - loss: 1.7858 - acc: 0.3420 - val_loss: 1.8162 - val_acc: 0.3914
Epoch 70/80
110/110 [=====] - 6s 54ms/step - loss: 1.7808 - acc: 0.3386 - val_loss: 1.8110 - val_acc: 0.3900
Epoch 71/80
110/110 [=====] - 6s 53ms/step - loss: 1.7774 - acc: 0.3397 - val_loss: 1.8057 - val_acc: 0.3957
Epoch 72/80
110/110 [=====] - 6s 53ms/step - loss: 1.7640 - acc: 0.3600 - val_loss: 1.8002 - val_acc: 0.4029
Epoch 73/80
110/110 [=====] - 6s 52ms/step - loss: 1.7601 - acc: 0.3506 - val_loss: 1.7949 - val_acc: 0.3971
Epoch 74/80
110/110 [=====] - 6s 52ms/step - loss: 1.7575 - acc: 0.3446 - val_loss: 1.7895 - val_acc: 0.3957
Epoch 75/80
110/110 [=====] - 6s 53ms/step - loss: 1.7521 - acc: 0.3491 - val_loss: 1.7843 - val_acc: 0.3886
Epoch 76/80
110/110 [=====] - 6s 52ms/step - loss: 1.7467 - acc: 0.3611 - val_loss: 1.7788 - val_acc: 0.3900
Epoch 77/80
110/110 [=====] - 6s 52ms/step - loss: 1.7418 - acc: 0.3540 - val_loss: 1.7734 - val_acc: 0.3943
Epoch 78/80
110/110 [=====] - 6s 53ms/step - loss: 1.7313 - acc: 0.3763 - val_loss: 1.7683 - val_acc: 0.3900
Epoch 79/80
110/110 [=====] - 6s 54ms/step - loss: 1.7288 - acc: 0.3557 - val_loss: 1.7628 - val_acc: 0.3929
Epoch 80/80
110/110 [=====] - 6s 52ms/step - loss: 1.7248 - acc: 0.3589 - val_loss: 1.7577 - val_acc: 0.3971
22/22 [=====] - 0s 13ms/step - loss: 1.7577 - acc: 0.3971
Loss = % 175.77122449874878
Accuracy = % 39.71428573131561
```

Confusion Matrix :

```
Console 1/A x
Confusion Matrix
[[58  5  5  1  0  2 29]
 [53  5 10  5  0 10 17]
 [26  7 59  1  0  6  1]
 [48  6  4 12  2  6 22]
 [23 16 27  3  4  8 19]
 [ 8  4  7  3  0 76  2]
 [27  3  5  0  0  1 64]]

Classification Report
precision    recall  f1-score   support

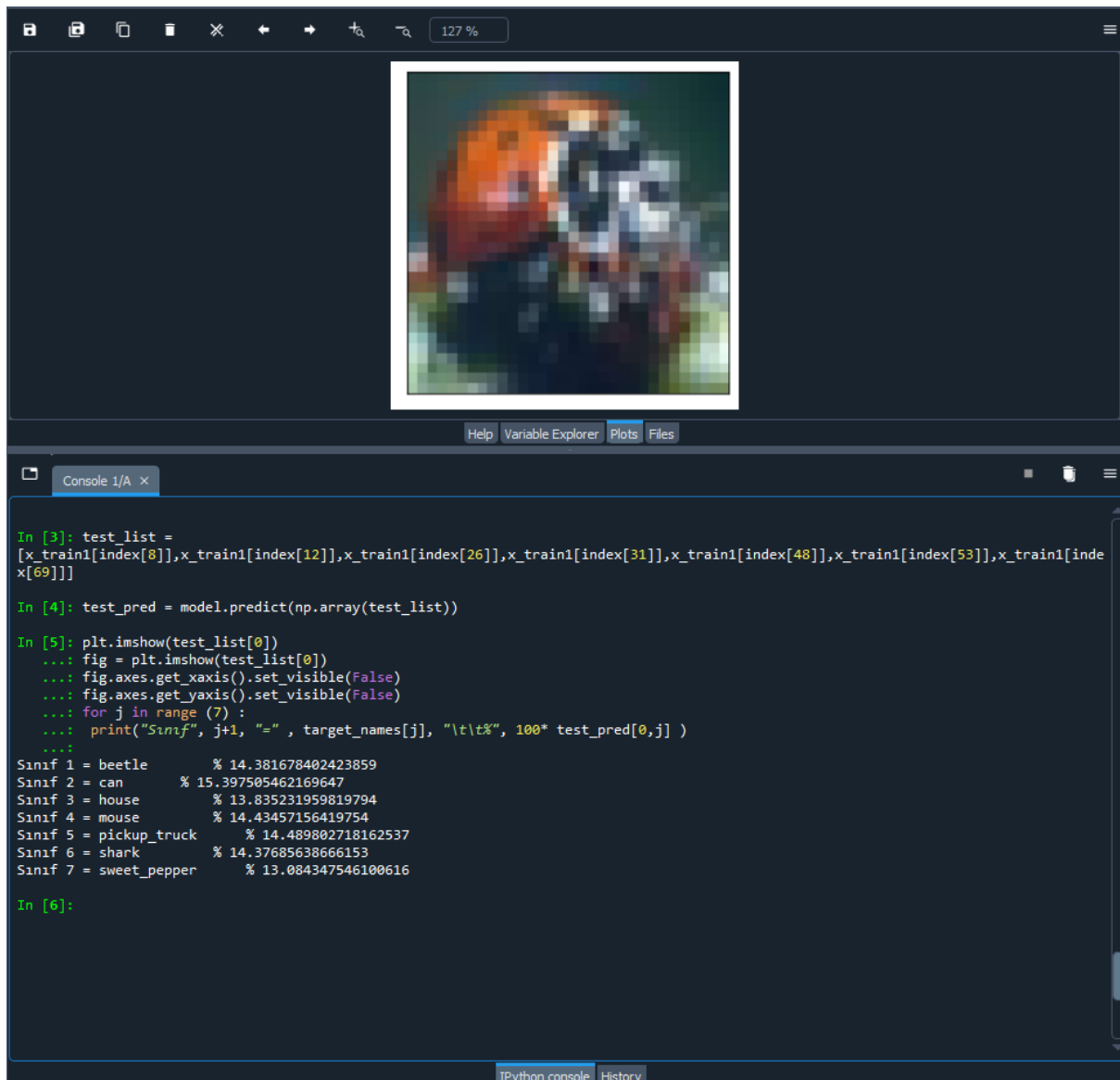
 beetle      0.24      0.58      0.34      100
  can        0.11      0.05      0.07      100
  house      0.50      0.59      0.54      100
  mouse      0.48      0.12      0.19      100
pickup_truck 0.67      0.04      0.08      100
  shark      0.70      0.76      0.73      100
sweet_pepper 0.42      0.64      0.50      100

 micro avg   0.40      0.40      0.40      700
 macro avg   0.44      0.40      0.35      700
weighted avg   0.44      0.40      0.35      700
 samples avg   0.40      0.40      0.40      700

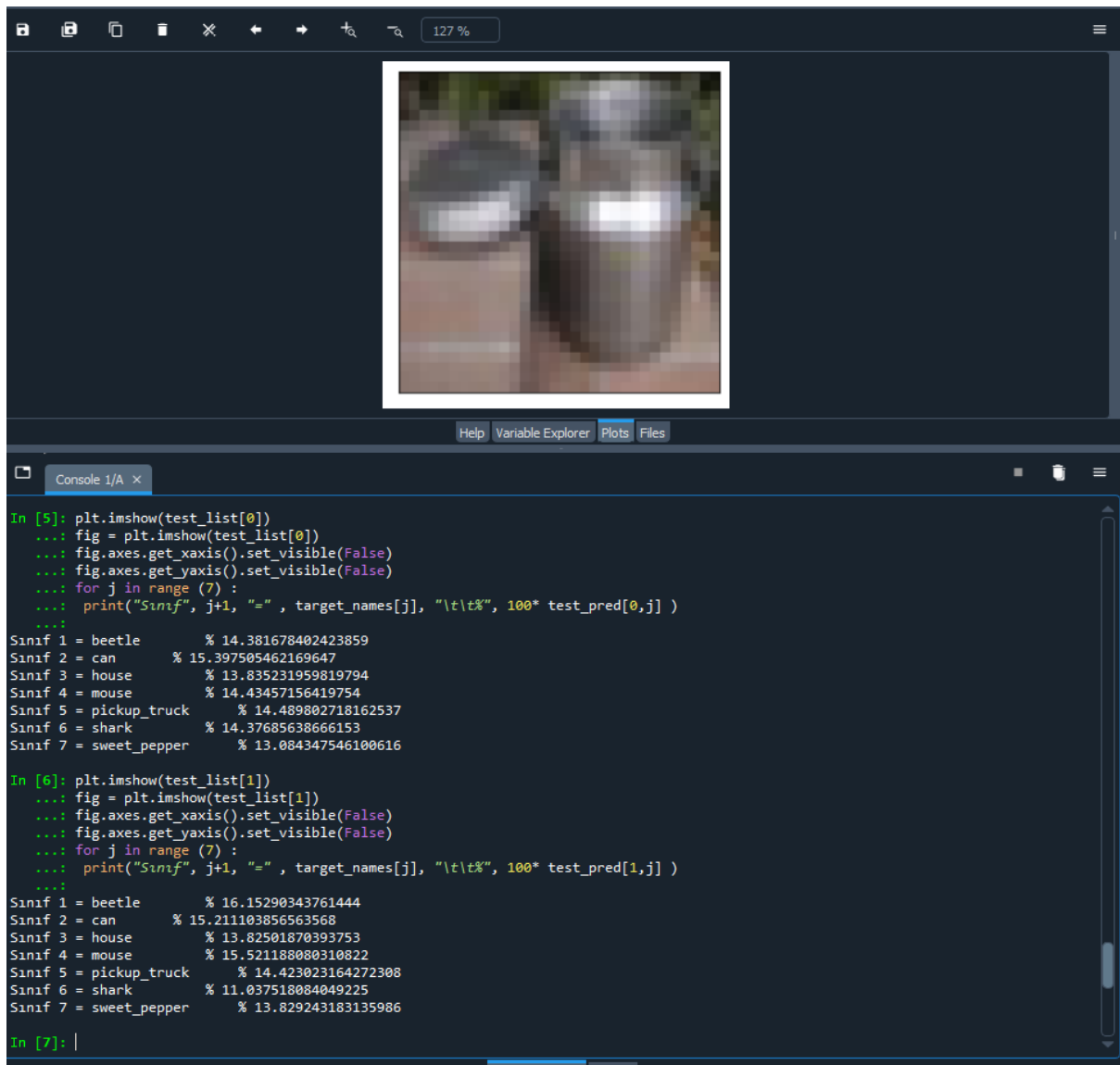
In [3]:
```

Sınıf örnekleri:

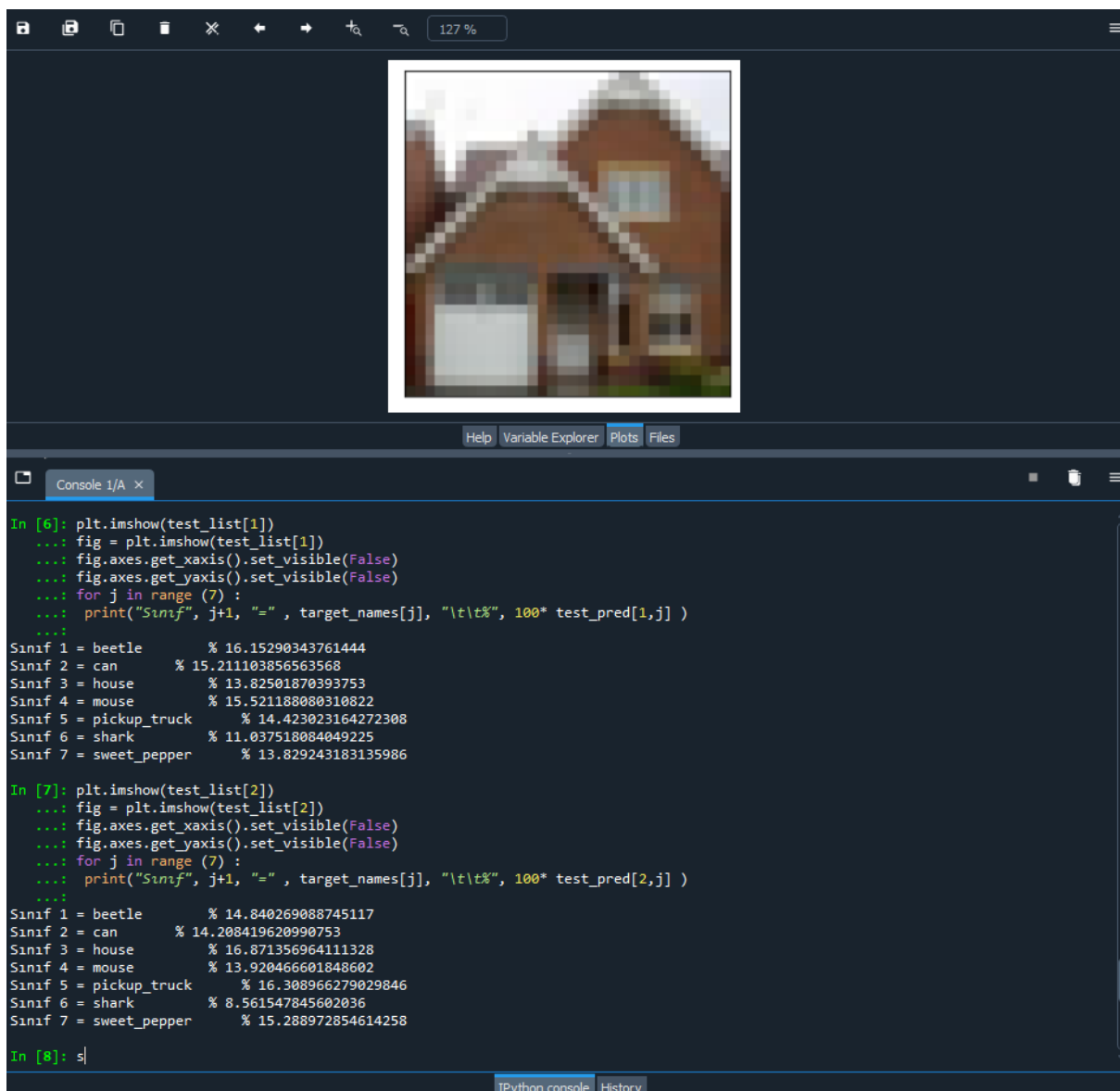
Beetle :



Can :



House :



127 %

Help Variable Explorer Plots Files

Console 1/A X

```
In [6]: plt.imshow(test_list[1])
...: fig = plt.imshow(test_list[1])
...: fig.axes.get_xaxis().set_visible(False)
...: fig.axes.get_yaxis().set_visible(False)
...: for j in range(7):
...:     print("Sinif", j+1, "=", target_names[j], "\t\t\t", 100* test_pred[1,j] )
...:
```

Sinif 1 = beetle	% 16.15290343761444
Sinif 2 = can	% 15.211103856563568
Sinif 3 = house	% 13.82501870393753
Sinif 4 = mouse	% 15.521188080310822
Sinif 5 = pickup_truck	% 14.423023164272308
Sinif 6 = shark	% 11.037518084049225
Sinif 7 = sweet_pepper	% 13.829243183135986

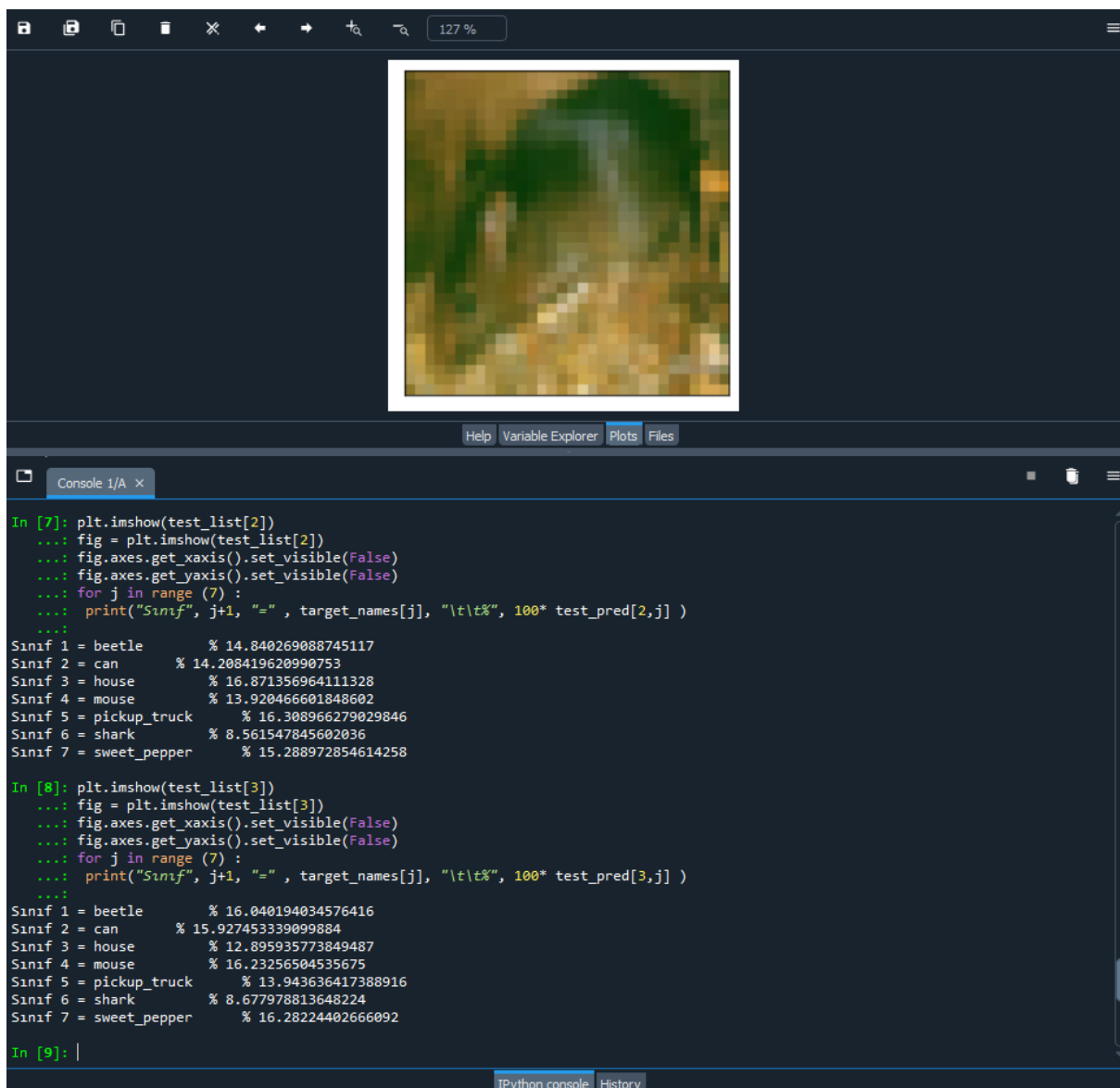
```
In [7]: plt.imshow(test_list[2])
...: fig = plt.imshow(test_list[2])
...: fig.axes.get_xaxis().set_visible(False)
...: fig.axes.get_yaxis().set_visible(False)
...: for j in range(7):
...:     print("Sinif", j+1, "=", target_names[j], "\t\t\t", 100* test_pred[2,j] )
...:
```

Sinif 1 = beetle	% 14.840269088745117
Sinif 2 = can	% 14.208419620990753
Sinif 3 = house	% 16.871356964111328
Sinif 4 = mouse	% 13.920466601848602
Sinif 5 = pickup_truck	% 16.308966279029846
Sinif 6 = shark	% 8.561547845602036
Sinif 7 = sweet_pepper	% 15.288972854614258

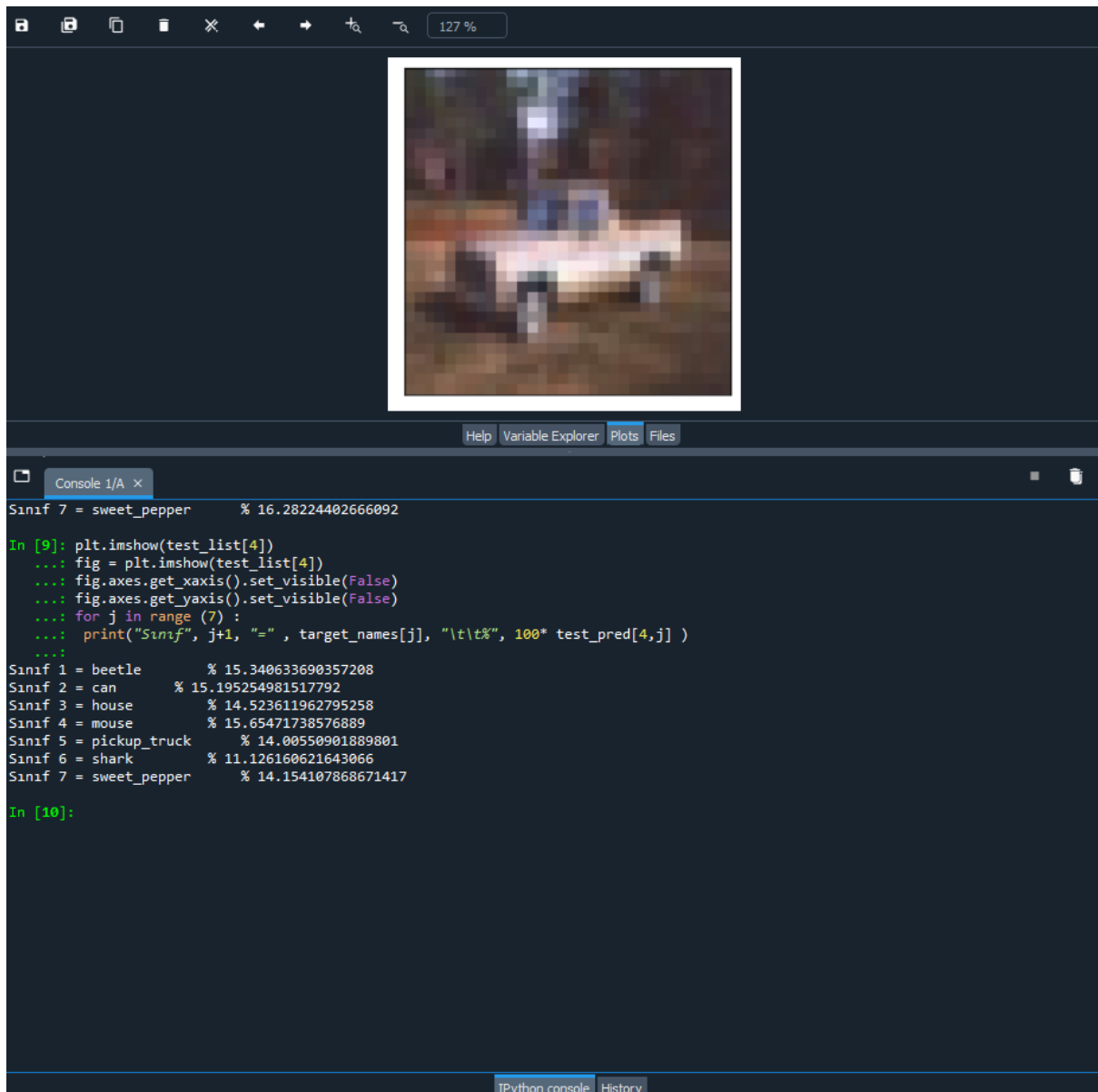
```
In [8]: s|
```

IPython console History

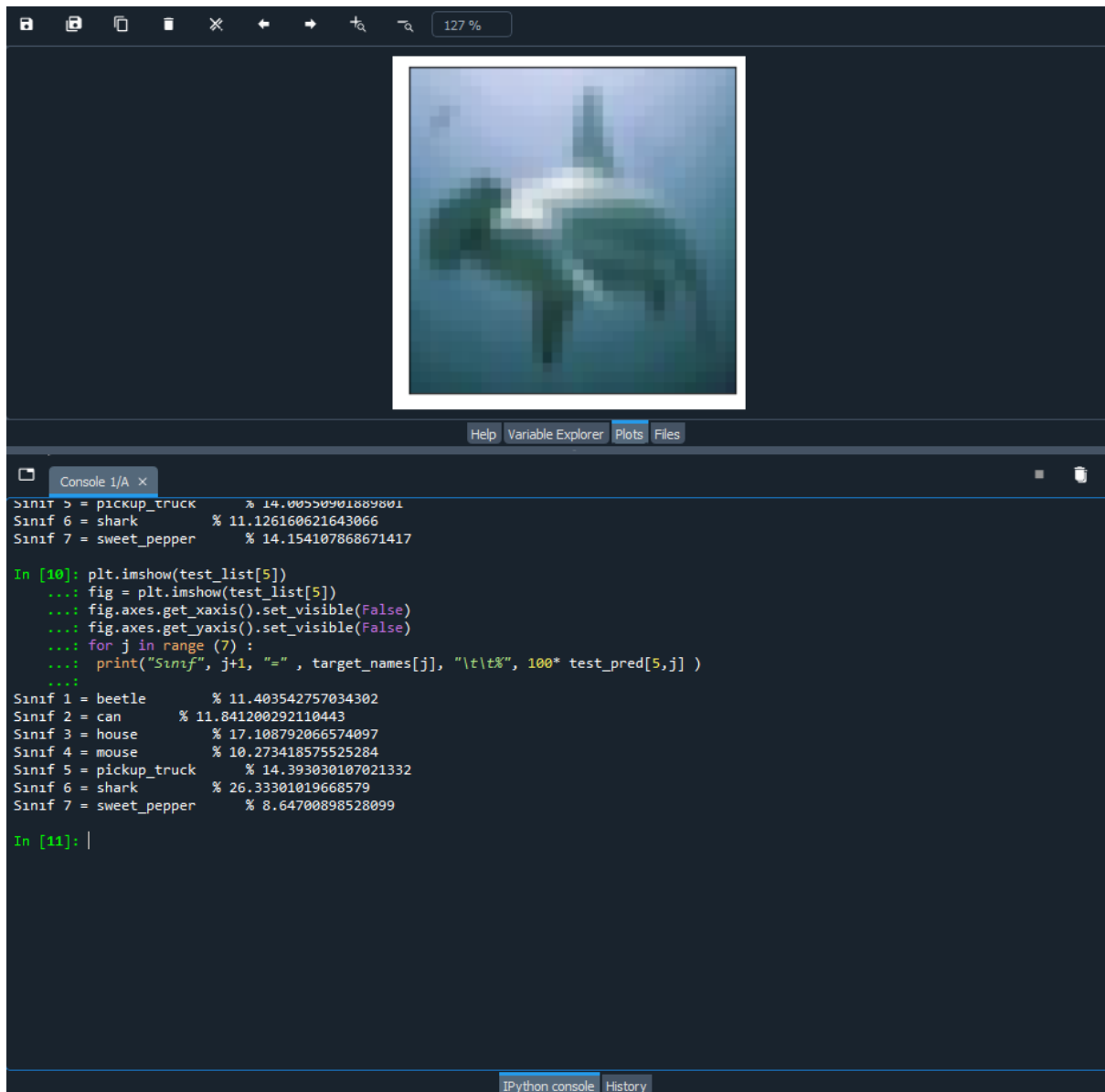
Mouse :



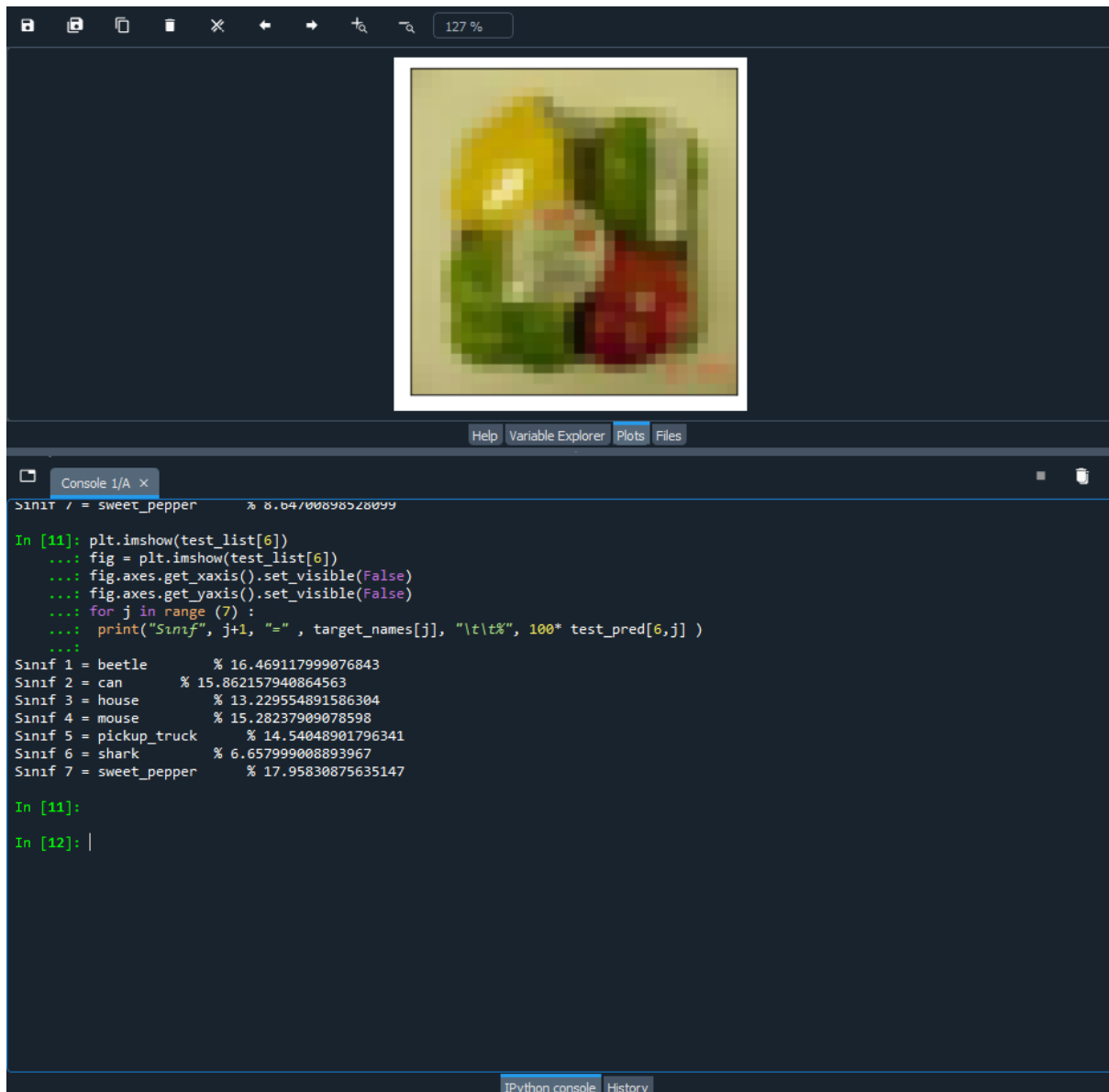
pickup_truck :



Shark :



sweet_pepper :



Kynak Kodu(.py dosyası) :

```
# -*- coding: utf-8 -*-
```

```
from keras.datasets import cifar100
```

```
from keras.models import Sequential
```

```
from keras import layers
```

```
import matplotlib.pyplot as plt
```

```
from tensorflow.keras.utils import to_categorical
```

```
import numpy as np
```

```
(x_train, y_train), (x_test, y_test) = cifar100.load_data()
```

```
y_train1 = []
```

```
x_train1 = []
```

```
y_test1 = []
```

```
x_test1 = []
```

```
for i in range (50000):
```

```
    if y_train[i] == 7 :
```

```
        y_train1.append(0)
```

```
        x_train1.append(x_train[i])
```

```
elif y_train[i] == 16 :  
    y_train1.append(1)  
    x_train1.append(x_train[i])
```

```
elif y_train[i] == 37 :  
    y_train1.append(2)  
    x_train1.append(x_train[i])
```

```
elif y_train[i] == 50 :  
    y_train1.append(3)  
    x_train1.append(x_train[i])
```

```
elif y_train[i] == 58 :  
    y_train1.append(4)  
    x_train1.append(x_train[i])
```

```
elif y_train[i] == 73 :  
    y_train1.append(5)  
    x_train1.append(x_train[i])
```

```
elif y_train[i] == 83 :  
    y_train1.append(6)  
    x_train1.append(x_train[i])
```

```
y_train1= np.array(y_train1)
x_train1= np.array(x_train1)
```

```
for i in range (10000):
    if y_test[i] == 7 :
        y_test1.append(0)
        x_test1.append(x_test[i])

    elif y_test[i] == 16 :
        y_test1.append(1)
        x_test1.append(x_test[i])

    elif y_test[i] == 37 :
        y_test1.append(2)
        x_test1.append(x_test[i])

    elif y_test[i] == 50 :
        y_test1.append(3)
        x_test1.append(x_test[i])

    elif y_test[i] == 58 :
        y_test1.append(4)
        x_test1.append(x_test[i])
```

```
elif y_test[i] == 73 :  
    y_test1.append(5)  
    x_test1.append(x_test[i])
```

```
elif y_test[i] == 83 :  
    y_test1.append(6)  
    x_test1.append(x_test[i])
```

```
y_test1= np.array(y_test1)  
x_test1= np.array(x_test1)
```

```
index=[]
```

```
enum=0
```

```
for i in range (2500):
```

```
    if y_train1[i] == 0 and enum < 10:
```

```
        index.append(i)
```

```
        enum=enum+1
```

```
        if enum == 10 : i = 0
```

```
    elif y_train1[i] == 1 and enum >= 10 and enum < 20:
```

```
        index.append(i)
```

```
        enum=enum+1
```

```
        if enum == 20 : i = 0
```

```
    elif y_train1[i] == 2 and enum >= 20 and enum < 30:
```

```
        index.append(i)
```

```
enum=enum+1
if enum == 30 : i = 0
elif y_train1[i] == 3 and enum >= 30 and enum < 40:
    index.append(i)
    enum=enum+1
if enum == 40 : i = 0
elif y_train1[i] == 4 and enum >= 40 and enum < 50:
    index.append(i)
    enum=enum+1

if enum == 50 : i = 0
elif y_train1[i] == 5 and enum >= 50 and enum < 60:
    index.append(i)
    enum=enum+1

if enum == 60 : i = 0
elif y_train1[i] == 6 and enum >= 60 and enum < 70:
    index.append(i)
    enum=enum+1

for i in range(0,70):
    plt.subplot(7,10,i+1)
    plt.imshow(x_train1[index[i]])
    fig = plt.imshow(x_train1[index[i]])
```



```
fig.axes.get_xaxis().set_visible(False)
```

```
fig.axes.get_yaxis().set_visible(False)
```

```
x_train1=x_train1.astype('float32')/255.0
```

```
x_test1=x_test1.astype('float32')/255
```

```
y_train1=to_categorical(y_train1,7)
```

```
y_test1=to_categorical(y_test1,7)
```

```
model=Sequential()
```

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

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

```
model.add(layers.MaxPool2D())
```

```
model.add(layers.Dropout(0.25))
```

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

```
model.add(layers.MaxPool2D())
```

```
model.add(layers.Dropout(0.25))
```

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

```
model.add(layers.MaxPool2D())
```

```
model.add(layers.Dropout(0.4))
```

```
model.add(layers.Flatten())
```

```
model.add(layers.Dense(512,activation='relu'))
```

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

```
model.summary()
```

```
from tensorflow.keras import optimizers
```

```
from keras import losses
```

```
model.compile(loss=losses.CategoricalCrossentropy(),
```

```
               optimizer=optimizers.Adam(lr=1e-6),
```

```
               metrics=['acc'])
```

```
history=model.fit(x_train1,
```

```
                  y_train1,
```

```
                  epochs=80,
```

```
                  validation_data=(x_test1,y_test1))
```

```
test_loss, test_acc = model.evaluate(x_test1,y_test1)
```

```
print(" Loss = %",100*test_loss,"\n","Accuracy = %",100*test_acc)
```

```
acc = history.history['acc']
```

```
val_acc = history.history['val_acc']
```

```
loss = history.history['loss']
```

```
val_loss = history.history['val_loss']
```

```
epochs = range(1,len(acc)+1)
```

```
import matplotlib.pyplot as plt
```

```
plt.figure()
```

```
plt.plot(epochs, acc, 'bo', label='Training acc')  
plt.plot(epochs, val_acc, 'r', label='Validation acc')  
plt.title('Training and validation accuracy')  
plt.xlabel('epochs')  
plt.ylabel('accuracy')  
plt.legend()
```

```
plt.show()
```

```
plt.figure()
```

```
plt.plot(epochs, loss, 'bo', label='Training loss')  
plt.plot(epochs, val_loss, 'r', label='Validation loss')  
plt.title('Training and validation loss')  
plt.xlabel('epochs')  
plt.ylabel('loss')  
plt.legend()
```

```
plt.show()
```

```
pred=model.predict(x_test1)
```

```

for i in range(0,700):
    maxindex= np.argmax(pred[i], axis=None, out=None)
    for y in range(0,7):
        if y == maxindex:
            pred[i,y] = 1
        else :
            pred[i,y] = 0

from sklearn.metrics import confusion_matrix, classification_report

print('Confusion Matrix')
print(confusion_matrix(y_test1.argmax(axis=1),
pred.argmax(axis=1)))
print("\n",'Classification Report')
target_names = ["beetle", "can", "house", "mouse", "pickup_truck",
"shark", "sweet_pepper"]
print(classification_report(y_test1, pred,
target_names=target_names))

# test kodu :

# test_list =
[x_train1[index[8]],x_train1[index[12]],x_train1[index[26]],x_train1[i

```

```
index[31]],x_train1[index[48]],x_train1[index[53]],x_train1[index[69]]
]
```

```
# test_pred = model.predict(np.array(test_list))
```

```
# plt.imshow(test_list[0])
```

```
# fig = plt.imshow(test_list[0])
```

```
# fig.axes.get_xaxis().set_visible(False)
```

```
# fig.axes.get_yaxis().set_visible(False)
```

```
# for j in range (7) :
```

```
# print("Sınıf", j+1, "=", target_names[j], "\t\t%", 100*  
test_pred[0,j] )
```