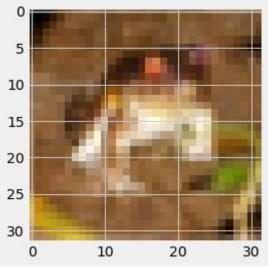
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**Project by:L.CHRISTINA SHERIN
                                           Project by:L.CHRISTINA SHERIN Date:29/12/21
Date:29/12/21**
import tensorflow as tf
from tensorflow import keras
from keras.models import Sequential
from keras.layers import Dense, Flatten, Conv2D, MaxPooling2D, Dropout
from tensorflow.keras import layers
from tensorflow.keras.utils import to_categorical
import numpy as np
import matplotlib.pyplot as plt
plt.style.use('fivethirtyeight')
#Load the data
from keras.datasets import cifar10
(x_train,y_train),(x_test,y_test) = cifar10.load_data()
    Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
    #Look at the data types of variables
print(type(x train))
print(type(y_train))
print(type(x test))
print(type(y_test))
    <class 'numpy.ndarray'>
    <class 'numpy.ndarray'>
    <class 'numpy.ndarray'>
    <class 'numpy.ndarray'>
#get the shape of the arrays
print('x_train shape:',x_train.shape)
print('y_train shape:',y_train.shape)
print('x_test shape:',x_test.shape)
print('y_test shape:',y_test.shape)
    x train shape: (50000, 32, 32, 3)
    y_train shape: (50000, 1)
    x_test shape: (10000, 32, 32, 3)
    y_test shape: (10000, 1)
```

```
index=0
x_train[index]
     array([[[ 59, 62, 63],
             [ 43, 46, 45],
             [50, 48, 43],
             . . . ,
             [158, 132, 108],
             [152, 125, 102],
             [148, 124, 103]],
                   20, 20],
            [[ 16,
             [ 0,
                     0,
                          0],
             [ 18,
                     8,
                          0],
             . . . ,
                    88, 55],
             [123]
             [119,
                    83, 50],
                    87,
                         57]],
             [122,
            [[ 25,
                    24, 21],
                    7,
             [ 16,
                          0],
             [ 49,
                    27,
                          8],
             . . . ,
                    84, 50],
             [118,
                    84, 50],
             [120,
             [109,
                    73, 42]],
            . . . ,
            [[208, 170, 96],
             [201, 153, 34],
             [198, 161, 26],
             ...,
             [160, 133,
                        70],
                          7],
             [ 56, 31,
                    34, 20]],
             [ 53,
            [[180, 139, 96],
             [173, 123,
                         42],
             [186, 144,
                        30],
             . . . ,
             [184, 148, 94],
             [ 97, 62,
                         34],
             [ 83,
                    53,
                         34]],
            [[177, 144, 116],
             [168, 129, 94],
             [179, 142, 87],
             . . . ,
             [216, 184, 140],
             [151, 118, 84],
             [123, 92, 72]]], dtype=uint8)
```

#take a look at the first image as an array

img = plt.imshow(x_train[index])



```
#get the image label
print('The image labelis:',y_train[index])
     The image labelis: [6]
#get the image classification
classification = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'tr
#print the image class
print('The image class is:',classification[y_train[index][0]])
     The image class is: frog
#convert the labels into a set of 10 numbers to input into the neural network
y train one hot = to categorical(y train)
y_test_one_hot = to_categorical(y_test)
#print the new labels
print(y train one hot)
     [[0. 0. 0. ... 0. 0. 0.]
      [0. 0. 0. ... 0. 0. 1.]
      [0. 0. 0. ... 0. 0. 1.]
      [0. 0. 0. ... 0. 0. 1.]
      [0. 1. 0. ... 0. 0. 0.]
      [0. 1. 0. \dots 0. 0. 0.]
#print the new label of the current image/picture
print('The one hot label is:',y_train_one_hot[index])
```

https://colab.research.google.com/drive/1Wpr6eGo hf7ld5CUaUNL4VBFl9n8sZW7#scrollTo=Hqz-TusbghT &printMode=true

The one hot label is: [0. 0. 0. 0. 0. 0. 1. 0. 0. 0.]

```
#normalize the pixels to be values between 0 and 1 x_{train} = x_{train} / 255 x_{test} = x_{test} / 255
```

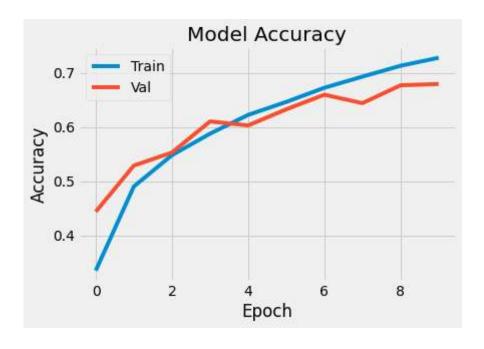
x_train[index]

```
array([[[0.23137255, 0.24313725, 0.24705882],
        [0.16862745, 0.18039216, 0.17647059],
        [0.19607843, 0.18823529, 0.16862745],
        [0.61960784, 0.51764706, 0.42352941],
        [0.59607843, 0.49019608, 0.4
        [0.58039216, 0.48627451, 0.40392157]],
       [[0.0627451 , 0.07843137, 0.07843137],
                         , 0.
                  , 0.
                                           ],
        [0.07058824, 0.03137255, 0.
        [0.48235294, 0.34509804, 0.21568627],
        [0.46666667, 0.3254902, 0.19607843],
        [0.47843137, 0.34117647, 0.22352941]],
       [[0.09803922, 0.09411765, 0.08235294],
        [0.0627451 , 0.02745098 , 0.
        [0.19215686, 0.10588235, 0.03137255],
        [0.4627451, 0.32941176, 0.19607843],
        [0.47058824, 0.32941176, 0.19607843],
        [0.42745098, 0.28627451, 0.16470588]],
       . . . ,
       [[0.81568627, 0.66666667, 0.37647059],
        [0.78823529, 0.6
                          , 0.13333333],
        [0.77647059, 0.63137255, 0.10196078],
        [0.62745098, 0.52156863, 0.2745098],
        [0.21960784, 0.12156863, 0.02745098],
        [0.20784314, 0.13333333, 0.07843137]],
       [[0.70588235, 0.54509804, 0.37647059],
        [0.67843137, 0.48235294, 0.16470588],
        [0.72941176, 0.56470588, 0.11764706],
        [0.72156863, 0.58039216, 0.36862745],
        [0.38039216, 0.24313725, 0.13333333],
        [0.3254902, 0.20784314, 0.13333333]],
       [[0.69411765, 0.56470588, 0.45490196],
        [0.65882353, 0.50588235, 0.36862745],
        [0.70196078, 0.55686275, 0.34117647],
        [0.84705882, 0.72156863, 0.54901961],
```

```
[0.59215686, 0.4627451, 0.32941176],
          [0.48235294, 0.36078431, 0.28235294]]])
#create the model architecture
model = Sequential()
#add first layer
model.add(Conv2D(32,(5,5),activation='relu',input_shape=(32,32,3)))
#add a pooling layer
model.add(MaxPooling2D(pool size = (2,2)))
#add another convolution layer
model.add(Conv2D(32,(5,5),activation='relu',input_shape=(32,32,3)))
#add another pooling layer
model.add(MaxPooling2D(pool_size = (2,2)))
#add a flattening layer
model.add(Flatten())
#add a layer with 1000 layers
model.add(Dense(1000,activation='relu'))
#add a drop out layer
model.add(Dropout(0.5))
#add a layer with 500 layers
model.add(Dense(1000,activation='relu'))
#add a drop out layer
model.add(Dropout(0.5))
#add a layer with 250 layers
model.add(Dense(250,activation='relu'))
#add a layer with 10 layers
model.add(Dense(10,activation='softmax'))
#compile the model
model.compile(loss = 'categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
#train the model
hist = model.fit(x_train,y_train_one_hot,batch_size=256,epochs=10,validation_split=0.2)
   Epoch 1/10
   Epoch 2/10
   Epoch 3/10
   Epoch 4/10
```

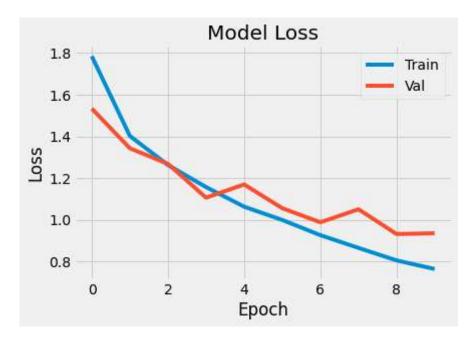
```
model.evaluate(x_test,y_test_one_hot)[1]
```

```
plt.plot(hist.history['accuracy'])
plt.plot(hist.history['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['Train','Val'],loc='upper left')
plt.show()
```



#visalise the models loss

```
plt.plot(hist.history['loss'])
plt.plot(hist.history['val_loss'])
plt.title('Model Loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train','Val'],loc='upper right')
plt.show()
```



#test the model with an exmple
from google.colab import files
uploaded = files.upload()

Choose Files No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

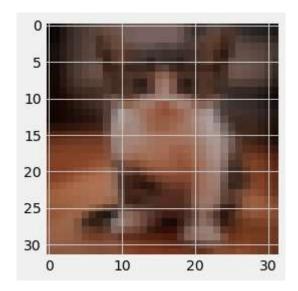
Saving king to king

```
#show the image
new_image = plt.imread('k.jpg')
img = plt.imshow(new image)
```

#resize the image

predictions

```
from skimage.transform import resize
resized_image = resize(new_image,(32,32,3))
img = plt.imshow(resized_image)
```



```
#get the models prediction
predictions = model.predict(np.array([resized_image]))
#show the predictions
```

```
array([[0.02222436, 0.00546992, 0.05909045, 0.34055468, 0.0881229, 0.30987534, 0.0596545, 0.10063497, 0.00726401, 0.00710883]], dtype=float32)
```

```
#sort predictions from least to greatest
list_index = [0,1,2,3,4,5,6,7,8,9]
x = predictions
for i in range(10):
    for j in range(10):
        if x[0][list_index[i]] > x[0][list_index[j]]:
            temp = list_index[i]
            list_index[i] = list_index[j]
            list_index[j] = temp
```

#show the sorted labels in order
print(list_index)

```
[3, 5, 7, 4, 6, 2, 0, 8, 9, 1]
```

```
#print the first 5 predictions
for i in range(5):
    print(classification[list_index[i]],':',round(predictions[0][list_index[i]]*100,2),'%')
        cat : 34.06 %
        dog : 30.99 %
        horse : 10.06 %
        deer : 8.81 %
        frog : 5.97 %
```

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