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# Lab: 9: Image Classification using CNN for CIFAR-10 **Data**

#### **Baseline Model**

```
In [2]:
import pandas
In [3]:
import tensorflow as tf
In [4]:
import keras
from keras.layers import Dense, Dropout, Activation, Flatten
from keras.layers import Conv2D, MaxPooling2D, Dropout, Flatten
In [5]:
                                                                                        H
from __future__ import print_function
In [6]:
from keras.preprocessing.image import ImageDataGenerator
from keras.models import Sequential
from keras.backend import categorical_crossentropy
Dataset
In [8]:
                                                                                        H
from keras.datasets import cifar10
In [9]:
                                                                                        H
(X_train, y_train), (X_test, y_test) = cifar10.load_data()
```

M

```
In [10]:

print(X_train.shape[0], 'train samples')
print(X_test.shape[0], 'test samples')
```

50000 train samples 10000 test samples

#### **Print the shape and Display**

```
In [11]:

X_train[444].shape

Out[11]:
(32, 32, 3)

In [12]:

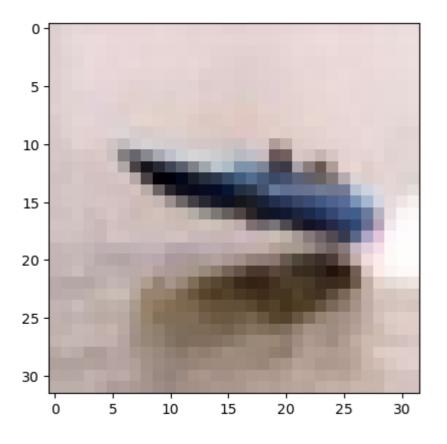
import matplotlib.pyplot as plt
%matplotlib inline

In [13]:

plt.imshow(X_train[441])
```

## Out[13]:

<matplotlib.image.AxesImage at 0x1868d5871d0>



#### Convert y\_train and y\_test

```
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In [14]:
num classes = 10
y_train = tf.keras.utils.to_categorical(y_train, num_classes)
y_test = tf.keras.utils.to_categorical(y_test, num_classes)
In [15]:
                                                                                       M
y_train[444]
Out[15]:
array([0., 0., 0., 0., 0., 0., 0., 0., 1.], dtype=float32)
Convert train data
In [16]:
                                                                                       M
X_train = X_train.astype('float32')
X_test = X_test.astype('float32')
X_train /= 255
X_test /= 255
Build CNN Model
                                                                                       H
In [17]:
from tensorflow.keras.optimizers import RMSprop
from tensorflow.keras.optimizers import Adam
                                                                                       H
In [18]:
model = Sequential()
model.add(Conv2D(32, (5, 5), strides=(2, 2), activation='relu', padding='same', input_st
model.add(MaxPooling2D((2, 2)))
model.add(Conv2D(64, (3, 3), strides=(2, 2), activation='relu', padding='same'))
model.add(MaxPooling2D((2, 2)))
model.add(Flatten())
model.add(Dropout(0.25))
model.add(Dense(128, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(512, activation='relu'))
model.add(Dense(num_classes, activation='softmax'))
In [19]:
                                                                                       H
from tensorflow.keras.optimizers import legacy as legacy_optimizers
```

In [20]: ▶

optimizer =tf.keras.optimizers.legacy.RMSprop(learning\_rate=0.0005, decay=1e-6)
model.compile(optimizer=optimizer, loss='categorical\_crossentropy', metrics=['accuracy']

In [21]:

model.compile(optimizer=optimizer, loss='categorical\_crossentropy', metrics=['accuracy']

### **Summary and Verify Czonfiguration**

In [22]: ▶

model.summary()

### Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 16, 16, 32)	2432
<pre>max_pooling2d (MaxPooling2 D)</pre>	(None, 8, 8, 32)	0
conv2d_1 (Conv2D)	(None, 4, 4, 64)	18496
<pre>max_pooling2d_1 (MaxPoolin g2D)</pre>	(None, 2, 2, 64)	0
flatten (Flatten)	(None, 256)	0
dropout (Dropout)	(None, 256)	0
dense (Dense)	(None, 128)	32896
dropout_1 (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 512)	66048
dense_2 (Dense)	(None, 10)	5130

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Total params: 125002 (488.29 KB)
Trainable params: 125002 (488.29 KB)
Non-trainable params: 0 (0.00 Byte)

Epoch 1/15

In [23]: ▶

```
batch_size = 32
epochs = 15
model.fit(X_train, y_train, batch_size=batch_size, epochs=epochs, validation_split=0.1,
```

```
6 - accuracy: 0.3064 - val_loss: 1.7087 - val_accuracy: 0.3842
Epoch 2/15
1407/1407 [============= - - 88s 63ms/step - loss: 1.5969
- accuracy: 0.4125 - val loss: 1.4127 - val accuracy: 0.4812
Epoch 3/15
1407/1407 [============== ] - 72s 51ms/step - loss: 1.4905
- accuracy: 0.4505 - val_loss: 1.3783 - val_accuracy: 0.4920
Epoch 4/15
1407/1407 [=============== ] - 80s 57ms/step - loss: 1.4273
- accuracy: 0.4799 - val_loss: 1.3714 - val_accuracy: 0.4908
1407/1407 [============= ] - 88s 63ms/step - loss: 1.3721
- accuracy: 0.5048 - val_loss: 1.2587 - val_accuracy: 0.5400
Epoch 6/15
- accuracy: 0.5155 - val_loss: 1.3183 - val_accuracy: 0.5160
Epoch 7/15
1407/1407 [============== ] - 80s 57ms/step - loss: 1.3065
- accuracy: 0.5290 - val_loss: 1.4107 - val_accuracy: 0.5164
- accuracy: 0.5397 - val_loss: 1.1555 - val_accuracy: 0.5796
Epoch 9/15
- accuracy: 0.5500 - val_loss: 1.1758 - val_accuracy: 0.5686
Epoch 10/15
1407/1407 [============= ] - 79s 56ms/step - loss: 1.2333
- accuracy: 0.5619 - val_loss: 1.2297 - val_accuracy: 0.5560
Epoch 11/15
- accuracy: 0.5725 - val_loss: 1.1479 - val_accuracy: 0.5856
Epoch 12/15
- accuracy: 0.5742 - val loss: 1.0688 - val accuracy: 0.6108
Epoch 13/15
- accuracy: 0.5799 - val_loss: 1.0881 - val_accuracy: 0.6150
Epoch 14/15
1407/1407 [============== ] - 86s 61ms/step - loss: 1.1820
- accuracy: 0.5868 - val loss: 1.1481 - val accuracy: 0.5932
Epoch 15/15
- accuracy: 0.5874 - val_loss: 1.0898 - val_accuracy: 0.6078
```

#### Out[23]:

<keras.src.callbacks.History at 0x1868d50b1d0>

```
In [24]:

test_loss, test_accuracy = model.evaluate(X_test, y_test)
print(f"Test accuracy: {test_accuracy}")
```

accuracy: 0.6111

Test accuracy: 0.6111000180244446

### **Model Improvements**

```
In [25]:
model1 = Sequential()
```

```
model1 = Sequential()
model1.add(Conv2D(filters=32, kernel_size=(5,5), strides=1, padding='same', activation='model1.add(Conv2D(filters=32, kernel_size=(5,5), strides=1, padding='same', activation='model1.add(MaxPooling2D(pool_size=(2,2)))
model1.add(Conv2D(filters=64, kernel_size=(5,5), strides=1, padding='same', activation='model1.add(Conv2D(filters=64, kernel_size=(5,5), strides=1, padding='same', activation='model1.add(MaxPooling2D(pool_size=(2,2)))
model1.add(Dropout(0.25))
model1.add(Dropout(0.25))
model1.add(Dense(512, activation='relu'))
model1.add(Dense(10, activation='softmax'))
model1.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
```

```
In [31]:
```

history01 = model1.fit(X\_train, y\_train, shuffle=True, epochs=5, batch\_size=32, validati
score01 = model1.evaluate(X\_test, y\_test, verbose=0)

In [34]:

model1.summary()

# Model: "sequential\_4"

Layer (type)	Output Shape	Param #
conv2d_10 (Conv2D)	(None, 32, 32, 32)	2432
conv2d_11 (Conv2D)	(None, 32, 32, 32)	25632
<pre>max_pooling2d_6 (MaxPoolin g2D)</pre>	(None, 16, 16, 32)	0
conv2d_12 (Conv2D)	(None, 16, 16, 64)	51264
conv2d_13 (Conv2D)	(None, 16, 16, 64)	102464
<pre>max_pooling2d_7 (MaxPoolin g2D)</pre>	(None, 8, 8, 64)	0
dropout_6 (Dropout)	(None, 8, 8, 64)	0
flatten_3 (Flatten)	(None, 4096)	0
dense_7 (Dense)	(None, 512)	2097664
dropout_7 (Dropout)	(None, 512)	0
dense_8 (Dense)	(None, 10)	5130

Trainable params: 2284586 (8.72 MB) Non-trainable params: 0 (0.00 Byte)

In [35]: H

```
print('Test loss:', score01[0])
print('Test accuracy:', score01[1])
```

Test loss: 0.8771228790283203 Test accuracy: 0.7027999758720398

In [ ]: M