

# CODE TO CLASSIFY IMAGES (CIFAR-10) USING CNNs

## STEP 0: PROBLEM STATEMENT

- CIFAR-10 is a dataset that consists of several images divided into the following 10 classes:
  - Airplanes
  - Cars
  - Birds
  - Cats
  - Deer
  - Dogs
  - Frogs
  - Horses
  - Ships
  - Trucks
- The dataset stands for the Canadian Institute For Advanced Research (CIFAR)
- CIFAR-10 is widely used for machine learning and computer vision applications.
- The dataset consists of 60,000 32x32 color images and 6,000 images of each class.
- Images have low resolution (32x32).

## STEP #1: IMPORT LIBRARIES/DATASETS

```
In [14]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn
```

```
In [15]: from keras.datasets import cifar10
(X_train, y_train) , (X_test, y_test) = cifar10.load_data()
```

```
In [16]: X_train.shape
```

```
Out[16]: (50000, 32, 32, 3)
```

```
In [17]: X_test.shape
```

```
Out[17]: (10000, 32, 32, 3)
```

```
In [18]: y_train.shape
```

```
Out[18]: (50000, 1)
```

```
In [19]: y_test.shape
```

```
Out[19]: (10000, 1)
```

## STEP #2: VISUALIZE DATA

```
In [20]: i = 30009
plt.imshow(X_train[i])
print(y_train[i])
```

[1]



```
In [21]: W_grid = 4
L_grid = 4

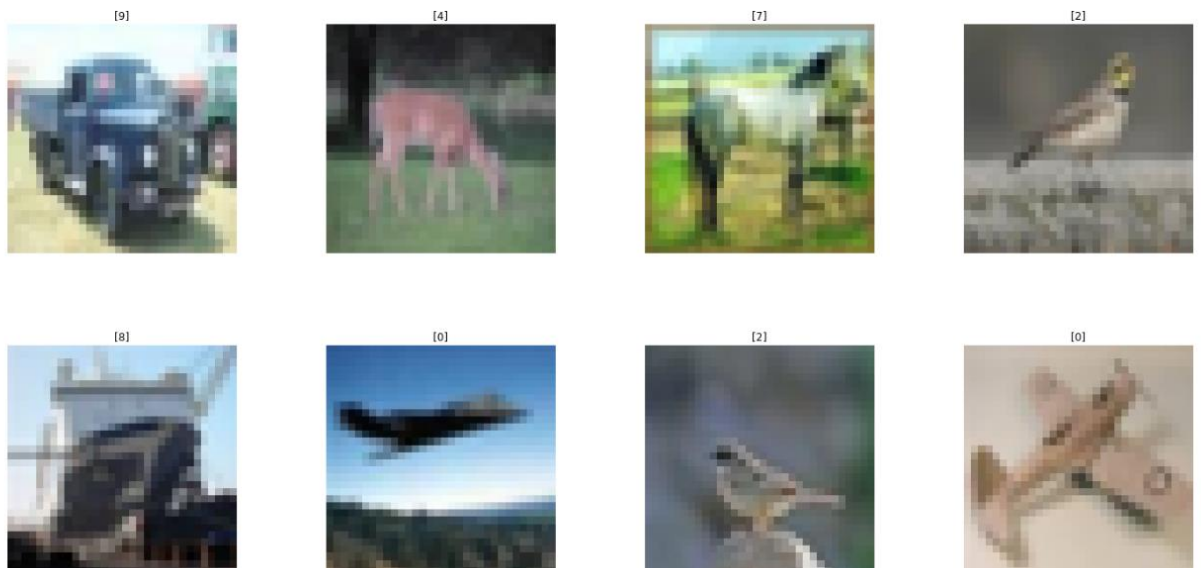
fig, axes = plt.subplots(L_grid, W_grid, figsize = (25, 25))
axes = axes.ravel()

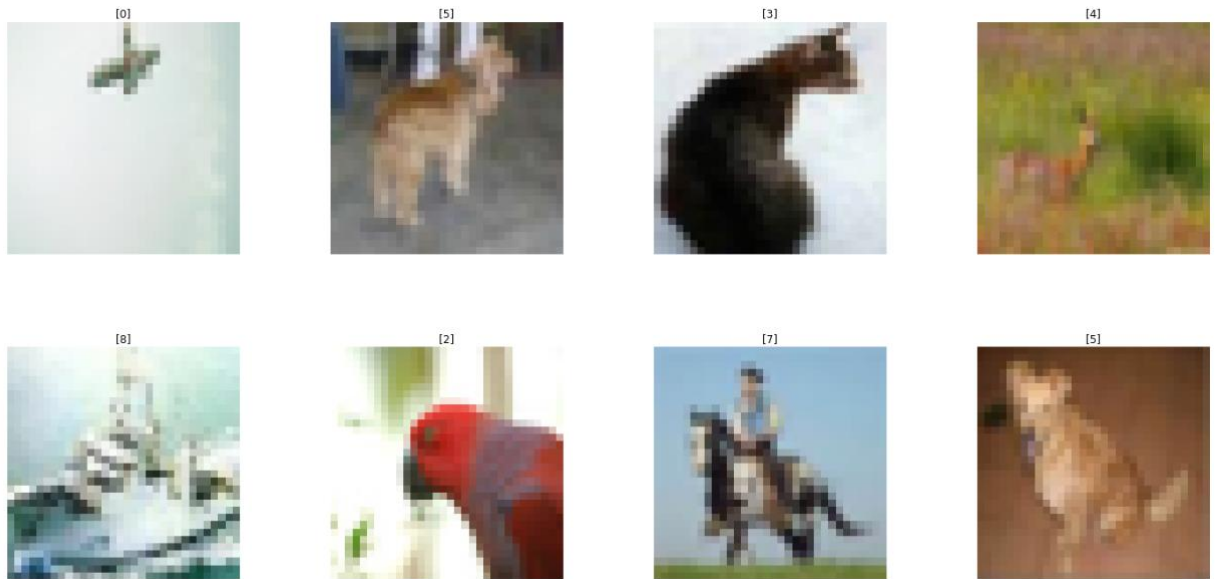
n_training = len(X_train)

for i in np.arange(0, L_grid * W_grid):
    index = np.random.randint(0, n_training) # pick a random number
    axes[i].imshow(X_train[index])
    axes[i].set_title(y_train[index])
    axes[i].axis('off')

plt.subplots_adjust(hspace = 0.4)
```

## Output:





```
In [22]: n_training
```

```
Out[22]: 50000
```

## STEP #3: DATA PREPARATION

```
In [23]: X_train = X_train.astype('float32')
         X_test = X_test.astype('float32')
```

```
In [24]: number_cat = 10
```

```
In [25]: y_train
```

```
Out[25]: array([[6],
               [9],
               [9],
               ...,
               [9],
               [1],
               [1]], dtype=uint8)
```

```
In [26]: import keras
         y_train = keras.utils.to_categorical(y_train, number_cat)
```

```
In [27]: y_train
```

```
Out[27]: array([[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., ..., 0., 0., 0.]], dtype=float32)
```

```
In [28]: y_test = keras.utils.to_categorical(y_test, number_cat)
```

```
In [29]: y_test
```

```
Out[29]: array([[0., 0., 0., ..., 0., 0., 0.],
               [0., 0., 0., ..., 0., 1., 0.],
               [0., 0., 0., ..., 0., 1., 0.],
               ...,
               [0., 0., 0., ..., 0., 0., 0.],
               [0., 1., 0., ..., 0., 0., 0.],
               [0., 0., 0., ..., 1., 0., 0.]], dtype=float32)
```

```
In [30]: X_train = X_train/255
X_test = X_test/255
```

```
In [31]: X_train
```

```
Out[31]: array([[0.23137255, 0.24313726, 0.24705882],
                [0.16862746, 0.18039216, 0.1764706 ],
                [0.19607843, 0.1882353 , 0.16862746],
                ...,
                [0.61960787, 0.5176471 , 0.42352942],
                [0.59607846, 0.49019608, 0.4       ],
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                [[0.0627451 , 0.07843138, 0.07843138],
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                [[0.09803922, 0.09411765, 0.08235294],
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                ...,

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                ...,
                [0.627451 , 0.52156866, 0.27450982],
                [0.21960784, 0.12156863, 0.02745098],
                [0.20784314, 0.13333334, 0.07843138]],

                [[0.7058824 , 0.54509807, 0.3764706 ],
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                [0.3254902 , 0.20784314, 0.13333334]],

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                [0.48235294, 0.36078432, 0.28235295]]],

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                [[0.54901963, 0.627451 , 0.6627451 ],
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                [0.49019608, 0.49019608, 0.4627451 ]],
```

```

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```

```

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```

```

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[0.6431373 , 0.6431373 , 0.63529414],
[0.6392157 , 0.6392157 , 0.6313726 ]]]], dtype=float32)

```

```
In [32]: X_train.shape
```

```
Out[32]: (50000, 32, 32, 3)
```

```
In [33]: Input_shape = X_train.shape[1:]
```

```
In [34]: Input_shape
```

```
Out[34]: (32, 32, 3)
```

## STEP #4: TRAIN THE MODEL

```
In [35]: from keras.models import Sequential
from keras.layers import Conv2D, MaxPooling2D, AveragePooling2D, Dense, Flatten, Dropout
from keras.optimizers import Adam
from keras.callbacks import TensorBoard
```

```
In [36]: cnn_model = Sequential()
cnn_model.add(Conv2D(filters = 64, kernel_size = (3,3), activation = 'relu', input_shape = Input_shape))
cnn_model.add(Conv2D(filters = 64, kernel_size = (3,3), activation = 'relu'))
cnn_model.add(MaxPooling2D(2,2))
cnn_model.add(Dropout(0.4))

cnn_model.add(Conv2D(filters = 128, kernel_size = (3,3), activation = 'relu'))
cnn_model.add(Conv2D(filters = 128, kernel_size = (3,3), activation = 'relu'))
cnn_model.add(MaxPooling2D(2,2))
cnn_model.add(Dropout(0.4))

cnn_model.add(Flatten())

cnn_model.add(Dense(units = 1024, activation = 'relu'))

cnn_model.add(Dense(units = 1024, activation = 'relu'))

cnn_model.add(Dense(units = 10, activation = 'softmax'))
```

```
In [37]: cnn_model.compile(loss = 'categorical_crossentropy', optimizer = keras.optimizers.rmsprop(lr = 0.001), metrics = ['accuracy'])
```

```
In [ ]: history = cnn_model.fit(X_train, y_train, batch_size = 32, epochs = 1, shuffle = True)
```

```
Epoch 1/1
10912/50000 [====>.....] - ETA: 5:32 - loss: 2.0551 - acc: 0.2424
```

## STEP #5: EVALUATE THE MODEL



```

In [ ]: evaluation = cnn_model.evaluate(X_test, y_test)
        print('Test Accuracy: {}'.format(evaluation[1]))

In [ ]: predicted_classes = cnn_model.predict_classes(X_test)
        predicted_classes

In [ ]: y_test

In [ ]: y_test = y_test.argmax(1)

In [ ]: y_test

In [ ]: L = 7
        W = 7
        fig, axes = plt.subplots(L, W, figsize = (12, 12))
        axes = axes.ravel()

        for i in np.arange(0, L*W):
            axes[i].imshow(X_test[i])
            axes[i].set_title('Prediction = {}\n True = {}'.format(predicted_classes[i], y_test[i]))
            axes[i].axis('off')

        plt.subplots_adjust(wspace = 1)

In [ ]: from sklearn.metrics import confusion_matrix
        import seaborn as sns

        cm = confusion_matrix(y_test, predicted_classes)
        cm
        plt.figure(figsize = (10, 10))
        sns.heatmap(cm, annot = True)

```

## STEP #6: SAVING THE MODEL

```

In [ ]: import os
        directory = os.path.join(os.getcwd(), 'saved_models')

        if not os.path.isdir(directory):
            os.makedirs(directory)
        model_path = os.path.join(directory, 'keras_cifar10_trained_model.h5')
        cnn_model.save(model_path)

```

## STEP #7: IMPROVING THE MODEL WITH DATA AUGMENTATION

- Image Augmentation is the process of artificially increasing the variations of the images in the datasets by flipping, enlarging, rotating the original images.
- Augmentations also include shifting and changing the brightness of the images.

### STEP 7.1 DATA AUGMENTATION FOR THE CIFAR-10 DATASET

```

In [ ]: import keras
        from keras.datasets import cifar10
        (X_train, y_train), (X_test, y_test) = cifar10.load_data()

In [ ]: X_train = X_train.astype('float32')
        X_test = X_test.astype('float32')

In [ ]: X_train.shape

In [ ]: n = 8
        X_train_sample = X_train[:n]

In [ ]: X_train_sample.shape

In [ ]: from keras.preprocessing.image import ImageDataGenerator

        # dataget_train = ImageDataGenerator(rotation_range = 90)
        # dataget_train = ImageDataGenerator(vertical_flip=True)
        # dataget_train = ImageDataGenerator(height_shift_range=0.5)
        dataget_train = ImageDataGenerator(brightness_range=(1,3))

        dataget_train.fit(X_train_sample)

In [ ]: from scipy.misc import toimage

        fig = plt.figure(figsize = (20,2))
        for x_batch in dataget_train.flow(X_train_sample, batch_size = n):
            for i in range(0,n):
                ax = fig.add_subplot(1, n, i+1)
                ax.imshow(toimage(x_batch[i]))
            fig.suptitle('Augmented images (rotated 90 degrees)')
            plt.show()
            break;

```

## STEP 7.2 MODEL TRAINING USING AUGEMENTED DATASET

```

In [ ]: from keras.preprocessing.image import ImageDataGenerator

        datagen = ImageDataGenerator(
                    rotation_range = 90,
                    width_shift_range = 0.1,
                    horizontal_flip = True,
                    vertical_flip = True
                )

In [ ]: datagen.fit(X_train)

In [ ]: cnn_model.fit_generator(datagen.flow(X_train, y_train, batch_size = 32), epochs = 2)

In [ ]: score = cnn_model.evaluate(X_test, y_test)
        print('Test accuracy', score[1])

In [ ]: # save the model
        directory = os.path.join(os.getcwd(), 'saved_models')

        if not os.path.isdir(directory):
            os.makedirs(directory)
        model_path = os.path.join(directory, 'keras_cifar10_trained_model_Augmentation.h5')
        cnn_model.save(model_path)

```