# Dog, Cat Image Classification

## Step 1: Download dataset

1 - For this first you need to downland the Kaggel API (kaggle.json file) from your kaggle account. Then put that Kaggle.jason file in .kaggle folder in you PC. After this step run the following code to download dataset directly from kaggle.

#### 2 - Make sure to install kaggle libray [pip install kaggle]

```
In [ ]: #download dataset from kaggle, using Kaggle API
   !kaggle datasets download -d salader/dogs-vs-cats
```

## Step 2: Unzipe the dataset

```
In [ ]: #Unzip the downLaoded dataset
import zipfile
zip_ref = zipfile.ZipFile('dogs-vs-cats.zip', 'r')
zip_ref.extractall()
zip_ref.close()
```

# **Step 3: Import libraries**

```
In [ ]: #Import tensorflow libraries for creating model and classification (CNN Model)
import tensorflow as tf
from tensorflow import keras
from keras import Sequential
from keras.layers import Dense,Conv2D,MaxPooling2D,Flatten,BatchNormalization,Dr
```

# Step 4: Data Preprocessing

#### 1 - Assign lables, sample from data, set image size

```
In []: # generators
    train_ds = keras.utils.image_dataset_from_directory(
        directory = 'train',
        labels='inferred',
        label_mode = 'int',
        batch_size=32,
        image_size=(256,256)
)

validation_ds = keras.utils.image_dataset_from_directory(
        directory = 'test',
        labels='inferred',
        label_mode = 'int',
        batch_size=32,
        image_size=(256,256)
)
```

Found 20000 files belonging to 2 classes. Found 5000 files belonging to 2 classes.

#### 2 - Normalize the data values in betweeen 0 & 1 [0 - 1]

```
In [ ]: # Normalize
    def process(image,label):
        image = tf.cast(image/255. ,tf.float32)
        return image,label

    train_ds = train_ds.map(process)
    validation_ds = validation_ds.map(process)
```

#### Step 5: Model training

```
In [ ]: # create CNN model
        model = Sequential()
        model.add(Conv2D(32,kernel_size=(3,3),padding='valid',activation='relu',input_sh
        model.add(BatchNormalization())
        model.add(MaxPooling2D(pool_size=(2,2),strides=2,padding='valid'))
        model.add(Conv2D(64,kernel_size=(3,3),padding='valid',activation='relu'))
        model.add(BatchNormalization())
        model.add(MaxPooling2D(pool_size=(2,2),strides=2,padding='valid'))
        model.add(Conv2D(128,kernel_size=(3,3),padding='valid',activation='relu'))
        model.add(BatchNormalization())
        model.add(MaxPooling2D(pool_size=(2,2),strides=2,padding='valid'))
        model.add(Flatten())
        model.add(Dense(128,activation='relu'))
        model.add(Dropout(0.1))
        model.add(Dense(64,activation='relu'))
        model.add(Dropout(0.1))
        model.add(Dense(1,activation='sigmoid'))
In [ ]: model.compile(optimizer='adam',loss='binary crossentropy',metrics=['accuracy'])
In [ ]: history = model.fit(train_ds,epochs=10,validation_data=validation_ds)
In [ ]: import matplotlib.pyplot as plt
        plt.plot(history.history['accuracy'],color='red',label='train')
        plt.plot(history.history['val accuracy'],color='blue',label='validation')
        plt.legend()
        plt.show()
        plt.plot(history.history['accuracy'],color='red',label='train')
        plt.plot(history.history['val_accuracy'],color='blue',label='validation')
        plt.legend()
        plt.show()
        plt.plot(history.history['loss'],color='red',label='train')
        plt.plot(history.history['val_loss'],color='blue',label='validation')
```

```
plt.legend()
plt.show()

In []: plt.plot(history.history['loss'],color='red',label='train')
plt.plot(history.history['val_loss'],color='blue',label='validation')
plt.legend()
plt.show()
```