```
In [ ]: import tensorflow as tf
             from tensorflow import keras
            from keras.models import Sequential
            from keras.layers import Dense, Flatten, Conv2D, MaxPooling2D, Dropout, Input, MaxPool2D, GlobalAveragePooling2D
             from tensorflow.keras import layers
             #from keras.utils import to_categorical
             import numpy as np
            import pandas as pd
            import matplotlib.pyplot as plt
             plt.style.use('fivethirtyeight')
             from tensorflow.keras import datasets, layers, models
             from tensorflow.keras.metrics import Precision, Recall, CategoricalAccuracy
            from tensorflow.keras.models import load_model, Model
            from tensorflow.keras.callbacks import EarlyStopping, Callback
             import cv2
            from os import listdir
            from os.path import isfile, join
            from google.colab import files
            from tensorflow.keras.preprocessing.image import ImageDataGenerator
            from keras.callbacks import ReduceLROnPlateau, ModelCheckpoint, EarlyStopping
            from keras.regularizers import 12
             from sklearn.metrics import accuracy_score
            from sklearn.metrics import confusion_matrix
            from tensorflow.keras.applications.mobilenet import MobileNet
            from tensorflow.keras.layers import GlobalAveragePooling2D
In [ ]: from google.colab import drive
             drive.mount('/content/drive')
             Mounted at /content/drive
In [ ]: %cd "/content/drive/MyDrive/proj_DS"
            /content/drive/MyDrive/proj_DS
In [ ]: DIR = '/content/drive/MyDrive/proj_DS/train'
             test_DIR = '/content/drive/MyDrive/proj_DS/test'
In [ ]: !unzip foods_final.zip
In [ ]: train_datagen = ImageDataGenerator(validation_split=0.3, rescale = 1. / 255)
             test_datagen = ImageDataGenerator(rescale = 1. / 255)
In [ ]: train_generator = train_datagen.flow_from_directory(
              target_size = (160, 160),
              color_mode = 'rgb',
               batch\_size = 32,
               class_mode = 'categorical',
               shuffle = True,
               seed = 42,
               subset = "training"
             validation_generator = train_datagen.flow_from_directory(
              target_size = (160, 160),
              color_mode = 'rgb',
              batch\_size = 32,
               class_mode = 'categorical',
               shuffle = True,
               seed = 42,
               subset = "validation"
             test_generator = test_datagen.flow_from_directory(
              '/content/drive/MyDrive/proj_DS/test',
              target_size = (160, 160),
              color_mode = 'rgb',
               batch\_size = 10,
              class_mode = 'categorical',
              shuffle = False,
               seed =42
             Found 22228 images belonging to 6 classes.
            Found 9523 images belonging to 6 classes.
            Found 3525 images belonging to 1 classes.
In [ ]: # lables the model will give to each class
             label_map = (train_generator.class_indices)
            label_map
Out[ ]: {'cakes_cupcakes_snack_cakes': 0,
               'candy': 1,
               'chips_pretzels_snacks': 2,
               'chocolate': 3,
               'cookies_biscuits': 4,
               'popcorn_peanuts_seeds_related_snacks': 5}
             Train model
In [ ]: base_model = MobileNet(weights='imagenet', include_top=False,input_shape = (160, 160, 3))
            1 = base_model.output
            1 = GlobalAveragePooling2D()(1)
            out = Dense(6, activation="softmax")(1)
             model = Model(inputs = base_model.input, outputs = out)
            #callbacks
            checkpointer = ModelCheckpoint(filepath='model.hdf5', verbose=1, save_best_only=True, save_weights_only=True)
            earlystopping = EarlyStopping(monitor='val_loss', min_delta=0.01, patience=20, mode='auto')
            reduceLR = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=10, mode='auto')
             model.compile(loss="categorical_crossentropy",optimizer='adam', metrics=['accuracy'])
            Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/mobilenet/mobilenet_1_0_160_tf_no_
            top.h5
            In [ ]: history = model.fit(train_generator, steps_per_epoch=2250/64,
                                                           validation_data=validation_generator,validation_steps=750/64,
                                                           epochs=100, callbacks=[checkpointer, reduceLR, earlystopping])
            Evaluate model
In [ ]: def plot_hist(history):
                   f,ax = plt.subplots(2,1,figsize=(15,10))
                   ax[0].plot(history.history['accuracy'],c='C2')
                   ax[0].plot(history.history['val_accuracy'], c='C3')
                   ax[0].set_title('Model accuracy')
                   ax[0].set_ylabel('Accuracy')
                   ax[0].set_xlabel('Epoch')
                   ax[0].legend(['Train', 'Validation'], loc='upper left')
                   # summarize history for loss
                   ax[1].plot(history.history['loss'],c='C0')
                   ax[1].plot(history.history['val_loss'],c='C1')
                   ax[1].set_title('Model loss')
                   ax[1].set_ylabel('Loss')
                   ax[1].set_xlabel('Epoch')
                   ax[1].legend(['Train', 'Validation'], loc='upper left')
             plot_hist(history)
                                                                                                 Model accuracy
                       - Train
               궁 0.6
                  0.5
                  0.3
                                                                                                    Model loss
                           Validation
                                                                                                        Epoch
In [ ]: x_test, y_test = validation_generator.next()
            y_pred_conf = model.predict(x_test) #return probabilities of each class
            y_pred = np.argmax(y_pred_conf,axis=1)
            y_label = np.argmax(y_test,axis=1)
             def plot_confusion_matrix(cm):
                   plt.imshow(cm, interpolation='nearest', cmap=plt.cm.Blues)
                   plt.title('Confusion matrix', fontsize=15)
                   plt.colorbar()
                   classes= ["cakes_cupcakes_snack_cakes", "candy", "chips_pretzels_snacks", "chocolate" , "cookies_biscuits", "pop
             corn_peanuts_seeds_related_snacks"]
                   plt.xticks([0,1,2,3,4,5], classes, fontsize=10, rotation='vertical')
                   plt.yticks([0,1,2,3,4,5], classes, fontsize=10, verticalalignment="center", rotation='horizontal')
                   for i in range(cm.shape[0]):
                         for j in range(cm.shape[1]):
                                plt.text(j, i, format(cm[i, j], 'd'), horizontalalignment="center", color="white" if cm[i, j] > np.max(color="white" if
             m)/2. else "black")
                   plt.xlabel('Predicted label', fontsize=15)
                   plt.ylabel('True label', fontsize=15)
             plot_confusion_matrix(confusion_matrix(y_label,y_pred))
                                                               Confusion matrix
                           cakes_cupcakes_snack_cakes
                                  chips_pretzels_snacks -
                  popcorn_peanuts_seeds_related_snacks
                                                                Predicted label
            Get predictions
In [ ]: def get_predictions_of_dir(mypath, dst):
                onlyfiles = [f for f in listdir(mypath) if isfile(join(mypath, f))]
                table = pd.DataFrame({'img':[], "prediction":[]})
                for file in onlyfiles:
                  img =cv2.imread(f"{mypath}/{file}")
                   resize = tf.image.resize(img, (160,160))
                   yhat = np.argmax(model.predict(np.expand_dims(resize/255, 0)))
                   table.loc[len(table.index)] = [int(file[:-4]), yhat]
                table.to_csv(dst, encoding = 'utf-8-sig', index =False)
                files.download(dst)
                return(table)
In [ ]: | test_table = get_predictions_of_dir('/content/drive/MyDrive/proj_DS/test/test_check', 'test.csv')
In [ ]: | popcorn_table = get_predictions_of_dir('/content/drive/MyDrive/proj_DS/train/popcorn_peanuts_seeds_related_snacks',
              'popcorn_peanuts_seeds_related_snacks.csv')
In [ ]: | cookies_biscuits_table = get_predictions_of_dir('/content/drive/MyDrive/proj_DS/train/cookies_biscuits', 'cookies_bi
             scuits.csv')
In [ ]: | chocolate_table = get_predictions_of_dir('/content/drive/MyDrive/proj_DS/train/chocolate', 'chocolate.csv')
In [ ]: | chips_table = get_predictions_of_dir('/content/drive/MyDrive/proj_DS/train/chips_pretzels_snacks', 'chips_pretzels_s
             nacks.csv')
In [ ]: candy_table = get_predictions_of_dir('/content/drive/MyDrive/proj_DS/train/candy', 'candy.csv')
In [ ]: cake_table = get_predictions_of_dir('/content/drive/MyDrive/proj_DS/train/cakes_cupcakes_snack_cakes', 'cakes_cupcakes_snack_cakes', 'cakes_cupcakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_snack_cakes_
             es_snack_cakes.csv')
```

ML Prediction - CNN for image classification

The model that we present here is thr last model we tried.

Prior to it we tried many different models with different hyperparemetrs ,such as :number of epochs, number of layers, with/without dropouts, different activation map and more.

All the other models gave us accuracy of 25%-40%. we had our 'breakthrough' when we tried to use weights from imagenet. Using those weights improved our prediction to around 70% accuracy, and we decided it's a good enough accuracy for us to use it in our main neural network.

summary