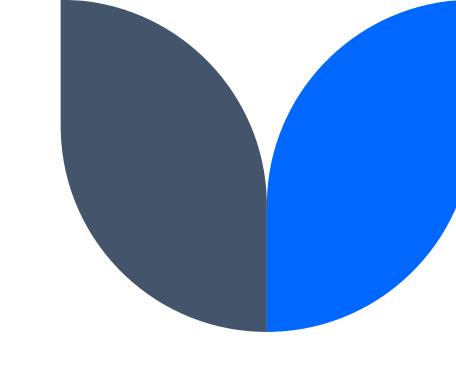
# Car classification

Deep learning 1000 ML



# Agenda

- Introduction
- Read Data
- Data Augmentation
- Resize all images
- Load images
- Splitting data
- Standardization
- Label encoding
- Creating pretrained model
- Testing the model



## Introduction

The problem is to identify the type of car through the camera, and this project can be used in many things, such as identifying the type of car in entering and exiting private places and other projects

The solution We used an DL model to identify the type of car

# **Reading Data**

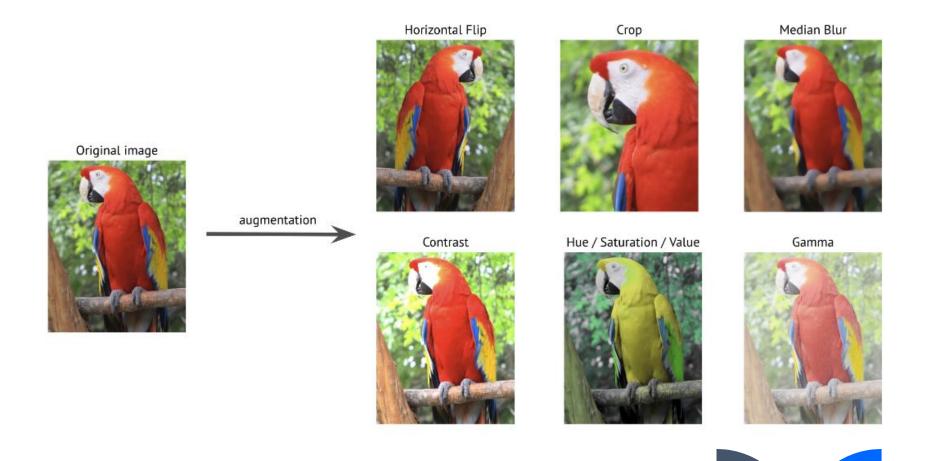
### **library**

```
1 # import necessary libraries
 2 import os
 3 import glob
 4 import cv2
 5 import random
 6 import albumentations as A
 7 import numpy as np
 8 from sklearn.preprocessing import LabelEncoder, StandardScaler
 9 import joblib
10 from sklearn.model_selection import train_test_split
11 from tensorflow.keras.utils import to_categorical
12 from tensorflow.keras.applications import MobileNet
13 from tensorflow.keras.models import Sequential
14 from tensorflow.keras.models import Model
15 from tensorflow.keras.layers import Dense, Dropout, GlobalAveragePooling2D
16 from tensorflow.keras.optimizers import Adam
17 from tensorflow.keras.applications.resnet50 import ResNet50
18 from keras.applications.mobilenet import MobileNet
19 from keras.layers import GlobalAveragePooling2D, Dropout, Dense
20 from tensorflow.keras.layers import Dense, Flatten
21 from keras.models import Sequential
22 from tensorflow.keras.optimizers import Adam
23 from keras.callbacks import EarlyStopping, ReduceLROnPlateau, ModelCheckpoint, CSVLogger
24 import time
```

### Read data

```
for subdir in os.listdir(root_dir):
    subdir = os.path.join(root_dir, subdir)
    if os.path.isdir(subdir):
        images = glob.glob(os.path.join(subdir, '*.jpg'))
        for image in images:
            # Load the image
            img = cv2.imread(image)
            # Resize the image
            img_resized = cv2.resize(img, (width, height))
            # Save the resized image
            cv2.imwrite(image, img_resized)
duration = time.time() - start_time
```

# **Data Augmentation**



# Data Augmentation and resizing all images

### **Data Augmentation Resize all images**

```
def augment_images(root_dir, width, height)
    Augment the images in the given directory and write the augmented images to the same directory.
    print(f"\nAugmenting images started..")
    start_time = time.time()
    max_images = 0
    for subdir in os.listdir(root_dir)
        subdir = os.path.join(root_dir, subdir)
        if os.path.isdir(subdir):
            images = glob.glob(os.path.join(subdir, '*.jpg'))
            max_images = max(max_images, len(images))
    for subdir in os.listdir(root_dir)
        subdir = os.path.join(root_dir, subdir)
        if os.path.isdir(subdir):
            images = glob.glob(os.path.join(subdir, '*.jpg'))
            while len(images) < max_images:
                label = os.path.basename(subdir)
                img = cv2.imread(random.sample(images,1)[0])
                if img is None
                    continue
                 else
                    img_resized = cv2.resize(img, (width, height))
                    img_aug = apply_augmentations()(image=img_resized)['image']
                    cv2.imwrite(f'{subdir}/augmented_{len(images)}.jpg',img_aug)
                    images.append(f'{subdir}/augmented_{len(images)}.jpg')
    for subdir in os.listdir(root_dir):
        subdir = os.path.join(root dir. subdir)
        if os.path.isdir(subdir):
            images = glob.glob(os.path.join(subdir, '*.jpg'))
            for image in images:
                # Load the image
                img = cv2.imread(image)
                # Resize the image
                img_resized = cv2.resize(img, (width, height))
                # Save the resized image
                cv2.imwrite(image, img_resized)
    duration = time.time() - start_time
    print(f"apply_augmentations() took {round(duration, 2)} seconds")
```

### **Augmentation function**



# Loading and splitting data

### **Loading data**

```
def load_data(root_dir):
   Load the images from the given directory and return them as numpy arrays of images and labels.
   print(f"\nLoading data started..")
   start_time = time.time()
   for subdir in os.listdir(root_dir):
       subdir = os.path.join(root_dir, subdir)
       if os.path.isdir(subdir):
            images = glob.glob(os.path.join(subdir, '*.jpg'))
           for image in images:
               img = cv2.imread(image)
               X.append(img)
               label = os.path.basename(subdir)
               y.append(label)
   X = np.array(X)
   y = np.array(y)
   duration = time.time() - start_time
   print(f"load_data() took {round(duration, 2)} seconds")
```

### **Splitting data**

```
def split_data(X, y):
    # Split the data into training and testing sets
    print(f"\nSplitting data started..")
    start_time = time.time()
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42, stratify=y)
    duration = time.time() - start_time
    print(f"split_data() took {round(duration, 2)} seconds")
    return X_train, X_test, y_train, y_test
```



# Standardization

### Standard scaler

```
def scale_data(X_train, X_test):
    # Reshape the data to 2D for scaling
    print(f"\nScaling data started..")
    start_time = time.time()
    X_train = np.reshape(X_train, (X_train.shape[0], -1))
    X_{\text{test}} = \text{np.reshape}(X_{\text{test}}, (X_{\text{test.shape}}[0], -1))
    # Scale the data using StandardScaler
    scaler = StandardScaler()
    X_train = scaler.fit_transform(X_train)
    # Save the scaler to file
    scaler_filename = "scaler.save"
    joblib.dump(scaler, scaler_filename)
    # Scale the test data using the fitted scaler
    X_test = scaler.transform(X_test)
    # Reshape the data back to 3D
    X_{train} = np.reshape(X_{train}, (X_{train}.shape[0], 224, 224, 3))
    X_{\text{test}} = \text{np.reshape}(X_{\text{test}}, (X_{\text{test.shape}}[0], 224, 224, 3))
    duration = time.time() - start_time
    print(f"scale_data() took {round(duration, 2)} seconds")
    return X_train, X_test
```

# Label encoding (using one hot encoding)

```
def encode_labels(y):
    Encode the given labels and return the encoded labels.
    print(f"\nEncoding labels started..")
    start_time = time.time()
   le = LabelEncoder()
   y = le.fit_transform(y)
   y = to_categorical(y, num_classes=20)
   labelEncoder filename = "labelEncoder.save"
   joblib.dump(le, labelEncoder_filename)
    duration = time.time() - start_time
    print(f"encode_labels() took {round(duration, 2)} seconds")
    return y
def load label encode():
    # Load label encoder from file
   le = joblib.load("labelEncoder.save")
    return le
```

# **Model creation**

Car classification

### **Creating pretrained model**

```
def create_model():
   Load pre-trained MobileNet model and freeze its layers.
   Add additional layers to the model.
   Compile and return the model.
   print(f"\nCreating model started..")
   start_time = time.time()
   base_model = MobileNet(weights='imagenet', include_top=False, input_shape=(224,224,3))
   base model.trainable = False
#base_model = ResNet50(include_top=False, input_shape=(224, 224, 3), weights = 'imagenet')
#for layer in base_model.layers:
# laver.trainable = False
   #model = ResNet50(include_top=False, input_shape=(224, 224, 3), weights = 'imagenet')
# mark loaded layers as not trainable
   #for layer in model.layers[:50]:
       #layer.trainable = False
   #for layer in model.layers[50:]:
       #layer.trainable = True
   model = Sequential([
       base_model,
       GlobalAveragePooling2D(),
      Dropout(0.5),
       Dense(1024, activation='relu'),
       Dropout(0.5),
       Dense(1024, activation='relu'),
       Dropout(0.5),
       Dense(20, activation='softmax')
```

### Fitting to our data

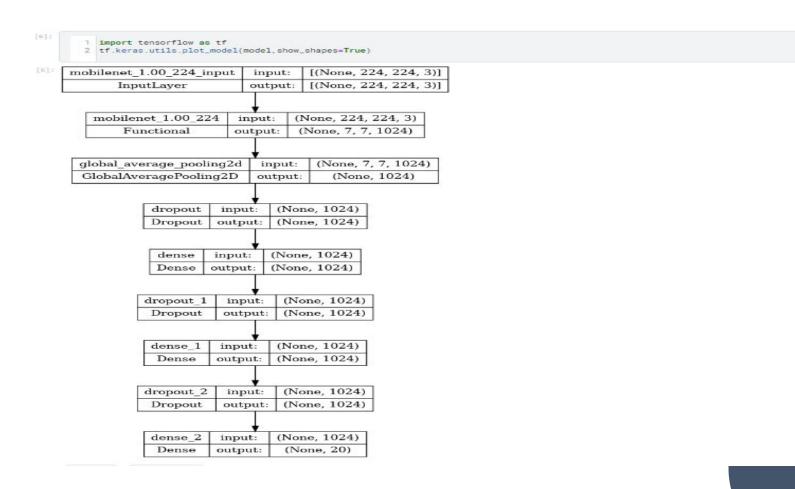
```
def train_model(model, X_train, y_train, X_test, y_test):
    """

    Train the model on the given data.
    """

    earlystop=EarlyStopping(patience=25)
    filepath = "model_class.h5"
    checkpoint = ModelCheckpoint(filepath, monitor='val_accuracy', verbose=1,save_best_only=True, mode='max')
    log_fname = 'model_log.csv'
    csv_logger = CSVLogger(filename=log_fname,separator=',',append=False)
    callbacks_list = [checkpoint, csv_logger,earlystop]
    print(f"\nTraining model started..")
    start_time = time.time()
    model.fit(X_train, y_train, epochs=220, batch_size=100, validation_data=(X_test, y_test),callbacks=[callbacks_list])
    duration = time.time() - start_time

print(f"train_model() took {round(duration, 2)} seconds")
```

# Model architecture



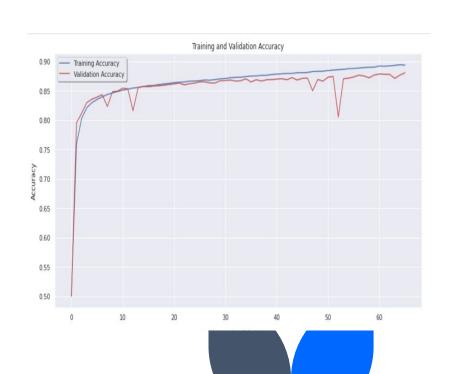
2/18/2023

# **Model Training**

# After 200 epochs, we reached validated accuracy of 92%

# loss: 0.1752 - accuracy: 0.9440 - val\_loss: 0.3435 - val\_accuracy: 0.9222 loss: 0.1980 - accuracy: 0.9371 - val\_loss: 0.3460 - val\_accuracy: 0.9208

# Model evaluation: train/val accuracy plot



# Testing the model

### Prediction using web app



### Prediction: seat ibiza



### **Prediction**









# Summary

Summary: our project is a multiple classification project to classify the type of the car, we used the transfer learning model (mobile net) and edit the classifier layers we get accuracy 92.5%, also we make a web bag to be as a user interface to take the car image from the user and classify it.

# Meet our team









Laila Ahmed Khalifa



**Mohammad Motasim** 



**Mahmoud Ali** 

# Thank you

Dr/ Hisham Assem

