

# 1. Introduction

- This file was trained on google colab using data from link bellow:  
<https://www.kaggle.com/datasets/alessiocrrado99/animals10>
- It contains about 28K medium quality animal images belonging to 10 categories: dog, cat, horse, spyder, butterfly, chicken, sheep, cow, squirrel, elephant.
- Leverage VGG16 model to classify.

## 2. Prepare data.

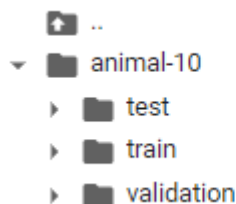
- Install kaggle lib and upload file kaggle.json (Link references:  
<https://www.kaggle.com/general/74235>)
- Download data file
- Unzip Data

```
1 !pip install -q kaggle
2 from google.colab import files
3 files.upload()
```

```
1 !mkdir ~/.kaggle
2 !cp kaggle.json ~/.kaggle/
3 !chmod 600 ~/.kaggle/kaggle.json
4 !kaggle datasets download -d alessiocrrado99/animals10
5 !unzip -qq animals10.zip
```

```
Downloading animals10.zip to /content
100% 584M/586M [00:26<00:00, 22.6MB/s]
100% 586M/586M [00:26<00:00, 22.8MB/s]
```

- Make directory follow contruction.



```
1 import os
2 list_animals_names = ['dog', 'horse', 'elephant', 'buterfly', 'chicken', 'cat', 'cow', 'sh
3 list_ori_folder = os.listdir('/content/raw-img')
```

```

4 # Make directory structure
5 path = "/content/animal-10"
6 os.mkdir(path)
7
8 def make_directory(name_catalogy):
9     sub_path = path + '/' + name_catalogy
10    os.mkdir(sub_path)
11    for animal_name in list_animals_names:
12        os.mkdir(sub_path + '/' + animal_name)
13    return
14
15 make_directory("train")
16 make_directory("validation")
17 make_directory("test")

```

- Make 10 sub directories in each "train", "validation", "test" and copy image from "raw-img" folder.
- We use 70% data for train, 15% for validation and 15% for test.

```

1 import shutil
2 import os
3 # ratio for train, val, and test set
4 ratio_train = 0.7
5 ratio_val = 0.15
6 ratio_test = ratio_train - ratio_val
7
8 ori_raw_folder = "/content/raw-img"
9 list_animals_names = ['dog', 'horse', 'elephant', 'butterfly', 'chicken', 'cat', 'cow', 'sheep']
10 list_ori_folder = sorted(os.listdir(ori_raw_folder))
11
12 for sub_ori_folder, sub_new_folder in zip(list_ori_folder, list_animals_names): # get a folder
13     sub_raw_img_name = ori_raw_folder + '/' + sub_ori_folder # address in "raw-img" folder
14     list_img_in_raw_img = os.listdir(sub_raw_img_name) # List image in sub "raw_img" folder
15     for i, img_org_name in enumerate(list_img_in_raw_img):
16         if i / len(list_img_in_raw_img) < ratio_train: # Copy train
17             shutil.copy(sub_raw_img_name + '/' + img_org_name,
18                         path + '/train/' + sub_new_folder + '/' +
19                         sub_new_folder + '.' + str(i) + '.jpeg' )
20         elif (i / len(list_img_in_raw_img) > ratio_train) and (i / len(list_img_in_raw_img) <
21                     ratio_train + ratio_val): # Copy validation
22             shutil.copy(sub_raw_img_name + '/' + img_org_name,
23                         path + '/validation/' + sub_new_folder + '/' +
24                         sub_new_folder + '.' + str(i) + '.jpeg' )
25         else: # Copy test
26             shutil.copy(sub_raw_img_name + '/' + img_org_name,
27                         path + '/test/' + sub_new_folder + '/' +
28                         sub_new_folder + '.' + str(i) + '.jpeg' )

```

- Using `image_dataset_from_directory` to read images. It will create and return a `tf.data.Dataset`

```

1 from tensorflow.keras.utils import image_dataset_from_directory
2 img_size = (180, 180)
3
4 train_dataset = image_dataset_from_directory(
5     path + "/train",
6     image_size=img_size,
7     batch_size=32)
8 validation_dataset = image_dataset_from_directory(
9     path + "/validation",
10    image_size=img_size,
11    batch_size=32)
12 test_dataset = image_dataset_from_directory(
13     path + "/test",
14     image_size=img_size,
15     batch_size=32)

```

```

Found 18331 files belonging to 10 classes.
Found 3925 files belonging to 10 classes.
Found 3923 files belonging to 10 classes.

```

### ▼ 3. Build the model.

- Model build base on VGG16 model with weight pretrain on imagenet image data
- After VGG16 convolution layer, we use GlobalAveragePooling2d to fatten it and add 10 neural for softmax
- Use callbacks to save the best train model.

```

1 from tensorflow import keras
2 from tensorflow.keras import layers
3
4 conv_base = keras.applications.vgg16.VGG16(
5     weights="imagenet",
6     include_top=False,
7     input_shape=(180, 180, 3))

```

```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/58889256/58889256 [=====] - 4s 0us/step

```



```

1
2 conv_base.trainable = False
3 """for layer in conv_base.layers[:-2]:
4     layer.trainable = False"""
5
6 data_augmentation = keras.Sequential(

```

```

7     [
8         layers.RandomRotation(0.1),
9         layers.RandomZoom(0.1)
10    ]
11 )
12
13 inputs = keras.Input(shape=(180, 180, 3))
14 #x = data_augmentation(inputs)
15 x = conv_base(inputs)
16 x = layers.GlobalAveragePooling2D()(x)
17 outputs = layers.Dense(10, activation="softmax")(x)
18 model = keras.Model(inputs, outputs)
19 model.compile(loss='sparse_categorical_crossentropy',
20               optimizer=keras.optimizers.Adam(learning_rate=0.001),
21               metrics=["accuracy"])
22
23 callbacks = [
24     keras.callbacks.ModelCheckpoint(
25         filepath="classify_animals_10.keras",
26         save_best_only=True,
27         monitor="val_loss"
28     )
29 ]

```

```

1 conv_base.summary()
2 model.summary()

```

Model: "vgg16"

Layer (type)	Output Shape	Param #
=====		
input_1 (InputLayer)	[(None, 180, 180, 3)]	0
block1_conv1 (Conv2D)	(None, 180, 180, 64)	1792
block1_conv2 (Conv2D)	(None, 180, 180, 64)	36928
block1_pool (MaxPooling2D)	(None, 90, 90, 64)	0
block2_conv1 (Conv2D)	(None, 90, 90, 128)	73856
block2_conv2 (Conv2D)	(None, 90, 90, 128)	147584
block2_pool (MaxPooling2D)	(None, 45, 45, 128)	0
block3_conv1 (Conv2D)	(None, 45, 45, 256)	295168
block3_conv2 (Conv2D)	(None, 45, 45, 256)	590080
block3_conv3 (Conv2D)	(None, 45, 45, 256)	590080
block3_pool (MaxPooling2D)	(None, 22, 22, 256)	0

```

block4_conv1 (Conv2D)      (None, 22, 22, 512)      1180160
block4_conv2 (Conv2D)      (None, 22, 22, 512)      2359808
block4_conv3 (Conv2D)      (None, 22, 22, 512)      2359808
block4_pool (MaxPooling2D) (None, 11, 11, 512)      0
block5_conv1 (Conv2D)      (None, 11, 11, 512)      2359808
block5_conv2 (Conv2D)      (None, 11, 11, 512)      2359808
block5_conv3 (Conv2D)      (None, 11, 11, 512)      2359808
block5_pool (MaxPooling2D) (None, 5, 5, 512)        0

```

=====  
Total params: 14,714,688

Trainable params: 0

Non-trainable params: 14,714,688

---

Model: "model\_4"

Layer (type)	Output Shape	Param #
input_5 (InputLayer)	[(None, 180, 180, 3)]	0
vgg16 (Functional)	(None, 5, 5, 512)	14714688
global_average_pooling2d_2 (GlobalAveragePooling2D)	(None, 512)	0

## ▼ 4. Train

- Now, we start train model for 10 epochs
- The model starting overfit after 8 epochs
- The accuracy nearly reach 90% on the validation data

```

1 history = model.fit(
2   train_dataset,
3   epochs=10,
4   validation_data=validation_dataset,
5   callbacks=callbacks)

```

Epoch 1/10

573/573 [=====] - 85s 148ms/step - loss: 1.8033 - accuracy: 0.7

Epoch 2/10

573/573 [=====] - 85s 147ms/step - loss: 0.6165 - accuracy: 0.8

Epoch 3/10

573/573 [=====] - 85s 148ms/step - loss: 0.4539 - accuracy: 0.9

Epoch 4/10

```

573/573 [=====] - 85s 147ms/step - loss: 0.3583 - accuracy: 0.9
Epoch 5/10
573/573 [=====] - 85s 148ms/step - loss: 0.3138 - accuracy: 0.9
Epoch 6/10
573/573 [=====] - 85s 148ms/step - loss: 0.2769 - accuracy: 0.9
Epoch 7/10
573/573 [=====] - 85s 148ms/step - loss: 0.2552 - accuracy: 0.9
Epoch 8/10
573/573 [=====] - 85s 148ms/step - loss: 0.2428 - accuracy: 0.9
Epoch 9/10
573/573 [=====] - 85s 147ms/step - loss: 0.2276 - accuracy: 0.9
Epoch 10/10
573/573 [=====] - 85s 148ms/step - loss: 0.2209 - accuracy: 0.9

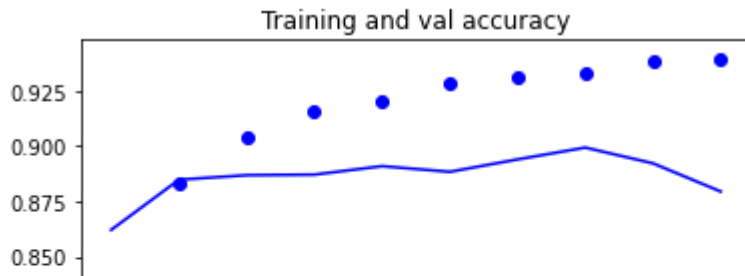
```



```

1 import matplotlib.pyplot as plt
2 accuracy = history.history["accuracy"]
3 val_accuracy = history.history["val_accuracy"]
4 loss = history.history["loss"]
5 val_loss = history.history["val_loss"]
6 epochs = range(1, len(accuracy) + 1)
7 plt.plot(epochs, accuracy, "bo", label="training accuracy")
8 plt.plot(epochs, val_accuracy, "b", label="Val_accuracy")
9 plt.title("Training and val accuracy")
10 plt.show()
11 plt.figure()
12 plt.plot(epochs, loss, "bo", label="training loss")
13 plt.plot(epochs, val_loss, "b", label="validation loss")
14 plt.title("Training and val loss")
15 plt.show()

```



## 5. Test model on validation data

```
1 # Evaluating the model on the test set
2 test_model = keras.models.load_model("classify_animals_10.keras")
3 test_loss, test_acc = test_model.evaluate(test_dataset)
4 print("Test accuracy: %.4f" %test_acc)
```

```
123/123 [=====] - 15s 120ms/step - loss: 0.5752 - accuracy: 0.8968
Test accuracy: 0.8968
```



## 6. Visualizing heatmaps of class activation (Grad-CAM)

- First, We must create a model that maps the input images to the activations of the last convolution layer.
- Then, We creast the model that maps the activation of the last convolutional to the final class predictions.
- After that, the `get_grads()` funtion will compute the gradient of top predict class with respect to the activaton of the last convolution layer

```
1 #Setting up a model return the last convolutinal output
2 last_conv_layer_name = "block5_conv3"
3 classifier_layer_names = ["global_average_pooling2d_2","dense_2"]
4 last_conv_layer = conv_base.get_layer(last_conv_layer_name)
5 last_conv_layer_model = keras.Model(conv_base.inputs, last_conv_layer.output)
```

```
1 # Reapplying the classifier on top of the last convolutional output
2 classifier_input = keras.Input(shape=last_conv_layer.output.shape[1:])
3 x = classifier_input
4 for layer_name in classifier_layer_names:
5     x = test_model.get_layer(layer_name)(x)
6 classifier_model = keras.Model(classifier_input, x)
```

Nhấp đúp (hoặc nhấn Enter) để chỉnh sửa

```

1 # Retrieving the gradients of the top predicted class
2 import tensorflow as tf
3
4 def get_grads(img_array):
5     with tf.GradientTape() as tape:
6         last_conv_layer_output = last_conv_layer_model(img_array) #TensorShape([1, 11, 11, 512
7         tape.watch(last_conv_layer_output)
8         preds = classifier_model(last_conv_layer_output) #TensorShape([1, 10])
9         top_pred_index = tf.argmax(preds[-1]) #TensorShape([10])
10        top_class_channel = preds[:, top_pred_index]
11        grads = tape.gradient(top_class_channel, last_conv_layer_output)
12        return grads, last_conv_layer_output

```

- I use 9 image downloaded from google search for CAM visuallazation

```

1 import matplotlib.cm as cm
2 import tensorflow as tf
3 import numpy as np
4
5 list_online_image =[
6     "https://www.eekwi.org/sites/default/files/2019-11/greysquirrel.jpg",
7     "https://www.dogsnsn.org.au/media/1007/breeding-dogs.jpg",
8     "https://encrypted-tbn0.gstatic.com/images?q=tbn:ANd9GcTJCujBD7aONJI7xzKl89nQqWww_Aub7
9     "https://www.treehugger.com/thmb/SShPLoEHvhEVIntPvs82-QcCPrQ=/2121x1193/smart/filters:
10    "https://morningchores.com/wp-content/uploads/2020/07/Sheep-Terms.jpg",
11    "https://www.jesmond.com/wp-content/uploads/2020/07/spider-436947_1920-CBen_Kerckx-102
12    "https://images.theconversation.com/files/472297/original/file-20220704-12-7zgqd5.jpg?
13    "https://i.natgeofe.com/n/548467d8-c5f1-4551-9f58-6817a8d2c45e/NationalGeographic_2572
14    "https://upload.wikimedia.org/wikipedia/commons/f/f7/Sparassidae_Palystes_castaneus_ma
15 ]
16 heatmaps = np.zeros((9, 11, 11))
17 img_arrays = np.zeros((9, 180, 180, 3))
18
19 #Convert link image to array
20 def get_img_array(img_path, target_size):
21     img_path = keras.utils.get_file(origin = img_path)
22     img = keras.utils.load_img(img_path, target_size=target_size)
23     array = keras.utils.img_to_array(img)
24     array = np.expand_dims(array, axis=0)
25     array = keras.applications.xception.preprocess_input(array)
26     return array
27
28
29 for k, link in enumerate(list_online_image):
30     img_array = get_img_array(link, target_size=(180, 180))
31     img_arrays[k, :, :, :] = img_array
32     grad, last_conv_layer_output = get_grads(img_array) #get grads funtion
33

```



```

34 pooled_grads = tf.reduce_mean(grad, axis=(0, 1, 2)).numpy()
35 last_conv_layer_output = last_conv_layer_output.numpy()[0]
36 for i in range(pooled_grads.shape[-1]):
37     last_conv_layer_output[:, :, i] *= pooled_grads[i]
38 heatmap = np.mean(last_conv_layer_output, axis=-1)
39
40 heatmap = np.maximum(heatmap, 0)
41 heatmap /= np.max(heatmap)
42 heatmaps[k, :, :] = heatmap

```

```

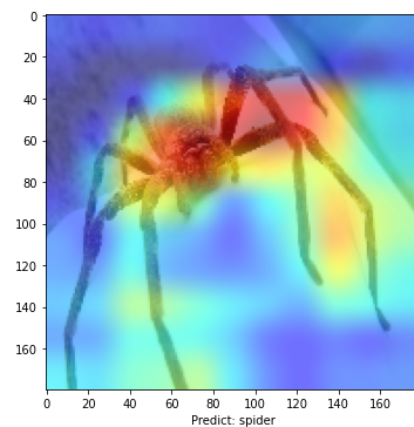
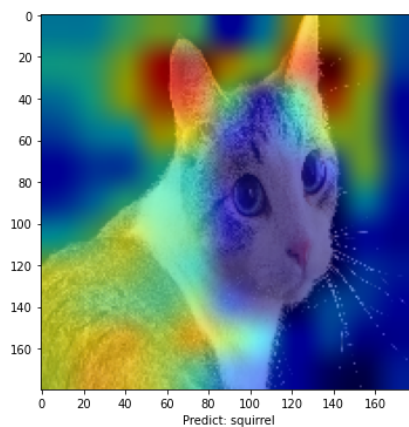
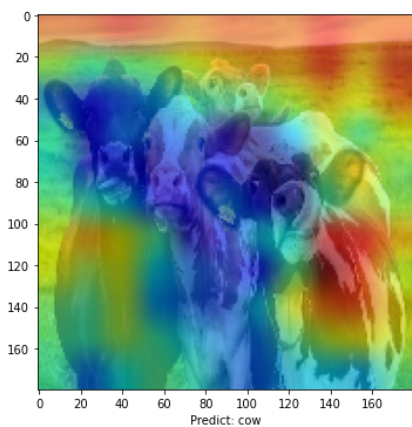
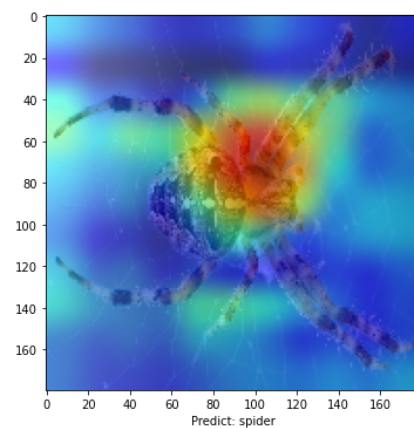
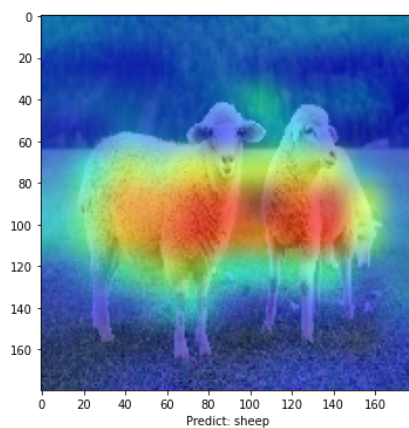
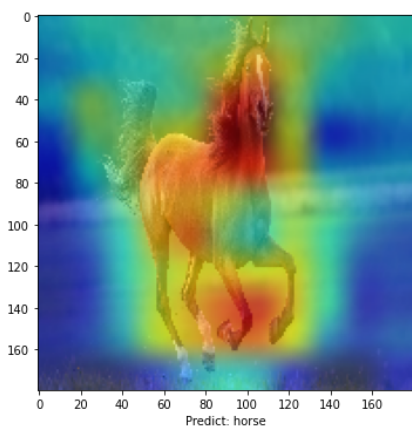
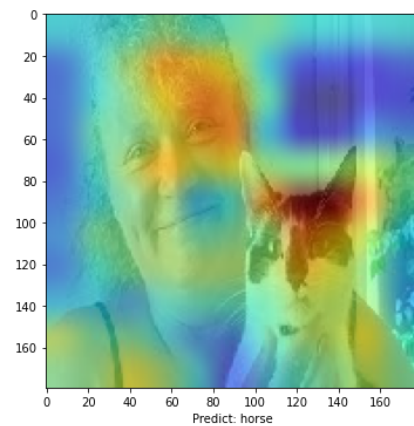
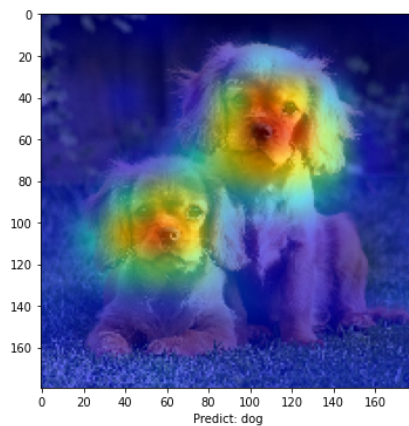
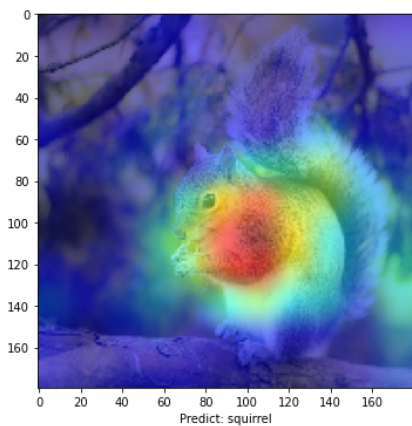
1 # Plot the result.
2 import matplotlib.cm as cm
3 import matplotlib.pyplot as plt
4
5 catalogi = sorted(os.listdir('/content/animal-10/train'))
6 predict_value = test_model.predict(img_arrays)
7 predict_value = np.argmax(predict_value, axis = -1)
8 print(predict_value)
9
10 plt.figure(figsize=(20,20))
11 for i in range(9):
12
13     plt.subplot(330 + 1 + i)
14     img = img_arrays[i]
15     heatmap = np.uint8(255 * heatmaps[i])
16     jet = cm.get_cmap("jet")
17     jet_colors = jet(np.arange(256))[:, :3]
18     jet_heatmap = jet_colors[heatmap]
19
20     jet_heatmap = keras.utils.array_to_img(jet_heatmap)
21     jet_heatmap = jet_heatmap.resize((img.shape[1], img.shape[0]))
22     jet_heatmap = keras.utils.img_to_array(jet_heatmap)
23
24     superimposed_img = jet_heatmap * 0.01 + img
25     superimposed_img = keras.utils.array_to_img(superimposed_img)
26     predict_index = predict_value[i]
27     name_of_predict = catalogi[predict_index]
28     plt.xlabel("Predict: %s" %name_of_predict)
29     plt.imshow(superimposed_img)
30 plt.show()
31

```



1/1 [=====] - 0s 21ms/step

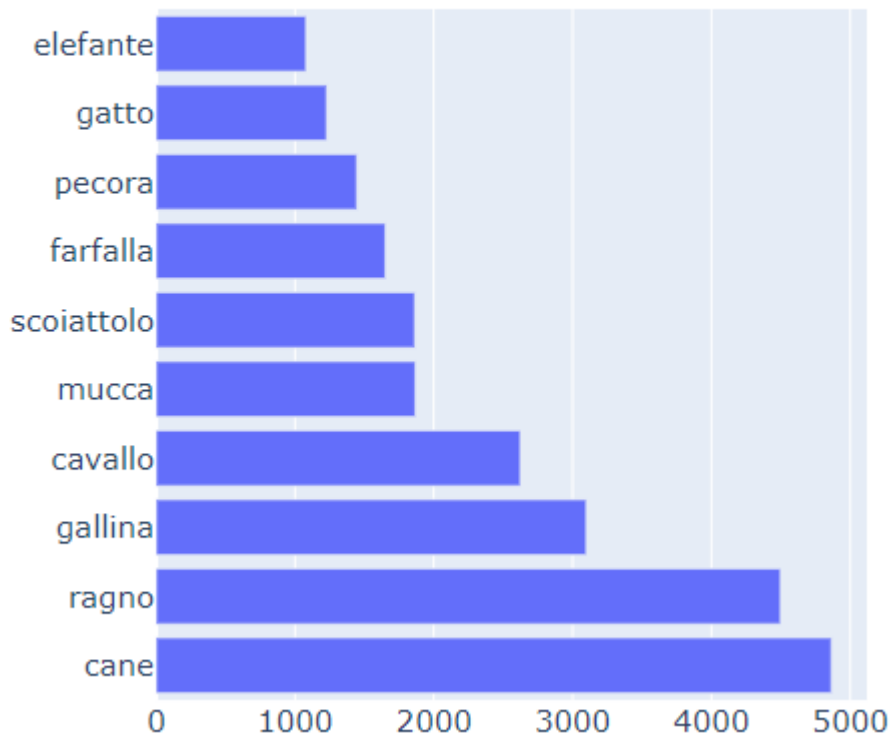
[9 4 6 6 7 8 3 9 8]



## 7. Conclusion

- After try with some random data, "elephant" class aren't classified well. Maybe due to the uneven distribution of the data.

Data Distribution in Bars



- Image at 3rd position with wrong classified, there is an error when there are people.
- CAM on 8th image wrong leading wrong classificaton.

Các sản phẩm có tính phí của Colab - [Huỷ hợp đồng tại đây](#)

✓ 10 giây    hoàn thành lúc 14:27

