

# 1 Quizz

## 2 Image classification

```
import os
import shutil
import numpy as np
import matplotlib.pyplot as plt
import tensorflow as tf
from tensorflow.keras import layers
from tensorflow.keras import Model
from tensorflow.keras.optimizers import Adam, RMSprop
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.utils import img_to_array, load_img

import matplotlib.image as mpimg
from keras.callbacks import EarlyStopping, ModelCheckpoint,
ReduceLROnPlateau
import keras
```

## Creation des dataset Train et Validation

```
# Catégories à traiter
categories = ['Colonial', 'Modern', 'Prehispanic']
train_dir = 'MexCulture142/images_train'
validation_dir = 'MexCulture142/images_val'

# Créer les répertoires d'entraînement et de validation pour chaque catégorie
for category in categories:
    # Répertoires d'entraînement
    train_category_dir = os.path.join(train_dir, category)
    os.makedirs(train_category_dir, exist_ok=True)

    # Répertoires de validation
    validation_category_dir = os.path.join(validation_dir, category)
    os.makedirs(validation_category_dir, exist_ok=True)

# Répertoires de classes dans train_dir
train_colonial_dir = os.path.join(train_dir, 'Colonial')
train_modern_dir = os.path.join(train_dir, 'Modern')
train_prehispanic_dir = os.path.join(train_dir, 'Prehispanic')

# Répertoires de classes dans validation_dir
validation_colonial_dir = os.path.join(validation_dir, 'Colonial')
validation_modern_dir = os.path.join(validation_dir, 'Modern')
```

```

validation_prehispanic_dir = os.path.join(validation_dir,
'Prehispanic')

# Fonction pour déplacer les images
def move_images(directory, colonial_dir, modern_dir, prehispanic_dir):
    for filename in os.listdir(directory):
        if filename.endswith('.png') or filename.endswith('.jpg'):
            # Extraire le suffixe (la première partie avant
l'underscore '_')
            suffix = filename.split('_')[0]

            # Déplacer l'image dans le bon répertoire
            if suffix == 'Colonial':
                shutil.move(os.path.join(directory, filename),
os.path.join(colonial_dir, filename))
            elif suffix == 'Modern':
                shutil.move(os.path.join(directory, filename),
os.path.join(modern_dir, filename))
            elif suffix == 'Prehispanic':
                shutil.move(os.path.join(directory, filename),
os.path.join(prehispanic_dir, filename))

# Déplacer les images dans les sous-dossiers appropriés
move_images(train_dir, train_colonial_dir, train_modern_dir,
train_prehispanic_dir)
move_images(validation_dir, validation_colonial_dir,
validation_modern_dir, validation_prehispanic_dir)

# Vérifier le nombre d'images déplacées
print(f"There are {len(os.listdir(train_colonial_dir))} Colonial
images for training.")
print(f"There are {len(os.listdir(train_modern_dir))} Modern images
for training.")
print(f"There are {len(os.listdir(train_prehispanic_dir))} Prehispanic
images for training.\n")

print(f"There are {len(os.listdir(validation_colonial_dir))} Colonial
images for validation.")
print(f"There are {len(os.listdir(validation_modern_dir))} Modern
images for validation.")
print(f"There are {len(os.listdir(validation_prehispanic_dir))}
Prehispanic images for validation.")

There are 150 Colonial images for training.
There are 34 Modern images for training.
There are 52 Prehispanic images for training.

There are 16 Colonial images for validation.
There are 16 Modern images for validation.
There are 16 Prehispanic images for validation.

```

On peut voir que les classes sont très déséquilibrées, ce qui va mener le réseau de neurones à prédire Modern de façon aléatoire.

```
train_colonial_fnames = os.listdir(train_colonial_dir)
train_modern_fnames = os.listdir(train_modern_dir)
train_prehispanic_fnames = os.listdir(train_prehispanic_dir)

# Paramètres pour l'affichage
ncols = 4 # Nombre de colonnes
nrows = 2 # Nombre de lignes

# Créer la figure pour afficher les images
fig = plt.gcf()
fig.set_size_inches(ncols * 4, nrows * 4)

# Spécifiez le nombre d'images à afficher par classe
num_images_per_class = 4

# Afficher les images de chaque classe
for class_idx, (class_dir, class_fnames, title) in enumerate(zip(
    [train_colonial_dir, train_modern_dir, train_prehispanic_dir],
    [train_colonial_fnames, train_modern_fnames,
    train_prehispanic_fnames],
    ['Colonial', 'Modern', 'Prehispanic']
)):
    # Sélectionner les images à afficher
    selected_fnames = class_fnames[:num_images_per_class]

    for i, fname in enumerate(selected_fnames):
        img_path = os.path.join(class_dir, fname)

        # Set up subplot; subplot indices start at 1
        sp = plt.subplot(len(selected_fnames), ncols, class_idx *
num_images_per_class + i + 1)
        sp.axis('Off') # Ne pas afficher les axes (ou les lignes de
grille)

        img = mpimg.imread(img_path)
        plt.imshow(img)
        sp.set_title(title) # Ajouter le titre de la classe

plt.show()
```



```
def train_val_generators(TRAINING_DIR, VALIDATION_DIR):
    train_datagen = ImageDataGenerator(rescale=1./255
                                       )

    train_generator =
train_datagen.flow_from_directory(directory=TRAINING_DIR,
                                batch_size=32,
class_mode='categorical',
target_size=(224, 224))

    validation_datagen = ImageDataGenerator( rescale = 1.0/255. )

    validation_generator =
validation_datagen.flow_from_directory(directory=VALIDATION_DIR,
batch_size=32,
class_mode='categorical',
target_size=(224, 224))
    return train_generator, validation_generator

train_generator, validation_generator =
train_val_generators(train_dir, validation_dir)

Found 236 images belonging to 3 classes.
Found 48 images belonging to 3 classes.
```

## Création du Réseau de neurone

```
pre_trained_model = tf.keras.applications.ResNet50(  
    include_top=False,  
    weights='imagenet',  
    input_shape=(224, 224, 3),  
    classifier_activation='softmax'  
)  
for layer in pre_trained_model.layers:  
    layer.trainable = False
```

```
pre_trained_model.summary()
```

Model: "resnet50"

Layer (type)	Output Shape	Param #	Connected to
input_layer (InputLayer)	(None, 224, 224, 3)	0	-
conv1_pad input_layer[0][0] (ZeroPadding2D)	(None, 230, 230, 3)	0	
conv1_conv (Conv2D) [0]	(None, 112, 112, 64)	9,472	conv1_pad[0]
conv1_bn [0] (BatchNormalizatio...	(None, 112, 112, 64)	256	conv1_conv[0]
conv1_relu [0] (Activation)	(None, 112, 112, 64)	0	conv1_bn[0]

pool1_pad [0]	(None, 114, 114,	0	conv1_relu[0]
(ZeroPadding2D)	64)		
pool1_pool [0]	(None, 56, 56,	0	pool1_pad[0]
(MaxPooling2D)	64)		
conv2_block1_1_conv [0]	(None, 56, 56,	4,160	pool1_pool[0]
(Conv2D)	64)		
conv2_block1_1_bn conv2_block1_1_c...	(None, 56, 56,	256	
(BatchNormalizatio...	64)		
conv2_block1_1_relu conv2_block1_1_b...	(None, 56, 56,	0	
(Activation)	64)		
conv2_block1_2_conv conv2_block1_1_r...	(None, 56, 56,	36,928	
(Conv2D)	64)		
conv2_block1_2_bn conv2_block1_2_c...	(None, 56, 56,	256	
(BatchNormalizatio...	64)		
conv2_block1_2_relu conv2_block1_2_b...	(None, 56, 56,	0	
(Activation)	64)		

conv2_block1_0_conv [0]	(None, 56, 56,	16,640	pool1_pool[0]
(Conv2D)	256)		
conv2_block1_3_conv conv2_block1_2_r...	(None, 56, 56,	16,640	
(Conv2D)	256)		
conv2_block1_0_bn conv2_block1_0_c...	(None, 56, 56,	1,024	
(BatchNormalizatio...	256)		
conv2_block1_3_bn conv2_block1_3_c...	(None, 56, 56,	1,024	
(BatchNormalizatio...	256)		
conv2_block1_add conv2_block1_0_b...	(None, 56, 56,	0	
(Add)	256)		
conv2_block1_3_b...			
conv2_block1_out conv2_block1_add...	(None, 56, 56,	0	
(Activation)	256)		
conv2_block2_1_conv conv2_block1_out...	(None, 56, 56,	16,448	
(Conv2D)	64)		
conv2_block2_1_bn conv2_block2_1_c...	(None, 56, 56,	256	
(BatchNormalizatio...	64)		
conv2_block2_1_relu	(None, 56, 56,	0	

conv2_block2_1_b...	(Activation)	64	
conv2_block2_2_conv	(None, 56, 56,	36,928	
conv2_block2_1_r...	(Conv2D)	64	
conv2_block2_2_bn	(None, 56, 56,	256	
conv2_block2_2_c...	(BatchNormalizatio...	64	
conv2_block2_2_relu	(None, 56, 56,	0	
conv2_block2_2_b...	(Activation)	64	
conv2_block2_3_conv	(None, 56, 56,	16,640	
conv2_block2_2_r...	(Conv2D)	256	
conv2_block2_3_bn	(None, 56, 56,	1,024	
conv2_block2_3_c...	(BatchNormalizatio...	256	
conv2_block2_add	(None, 56, 56,	0	
conv2_block1_out...	(Add)	256	
conv2_block2_3_b...			
conv2_block2_out	(None, 56, 56,	0	
conv2_block2_add...	(Activation)	256	
conv2_block3_1_conv	(None, 56, 56,	16,448	
conv2_block2_out...			



(Conv2D)	64)		
conv2_block3_1_bn	(None, 56, 56,	256	
conv2_block3_1_c...	(BatchNormalizatio...	64)	
conv2_block3_1_relu	(None, 56, 56,	0	
conv2_block3_1_b...	(Activation)	64)	
conv2_block3_2_conv	(None, 56, 56,	36,928	
conv2_block3_1_r...	(Conv2D)	64)	
conv2_block3_2_bn	(None, 56, 56,	256	
conv2_block3_2_c...	(BatchNormalizatio...	64)	
conv2_block3_2_relu	(None, 56, 56,	0	
conv2_block3_2_b...	(Activation)	64)	
conv2_block3_3_conv	(None, 56, 56,	16,640	
conv2_block3_2_r...	(Conv2D)	256)	
conv2_block3_3_bn	(None, 56, 56,	1,024	
conv2_block3_3_c...	(BatchNormalizatio...	256)	
conv2_block3_add	(None, 56, 56,	0	
conv2_block2_out...	(Add)	256)	

conv2_block3_3_b...			
conv2_block3_out	(None, 56, 56,	0	
conv2_block3_add...	(Activation)	256)	
conv3_block1_1_conv	(None, 28, 28,	32,896	
conv2_block3_out...	(Conv2D)	128)	
conv3_block1_1_bn	(None, 28, 28,	512	
conv3_block1_1_c...	(BatchNormalizatio...	128)	
conv3_block1_1_relu	(None, 28, 28,	0	
conv3_block1_1_b...	(Activation)	128)	
conv3_block1_2_conv	(None, 28, 28,	147,584	
conv3_block1_1_r...	(Conv2D)	128)	
conv3_block1_2_bn	(None, 28, 28,	512	
conv3_block1_2_c...	(BatchNormalizatio...	128)	
conv3_block1_2_relu	(None, 28, 28,	0	
conv3_block1_2_b...	(Activation)	128)	
conv3_block1_0_conv	(None, 28, 28,	131,584	
conv2_block3_out...	(Conv2D)	512)	

conv3_block1_3_conv	(None, 28, 28,	66,048	
conv3_block1_2_r... (Conv2D)	512)		
conv3_block1_0_bn	(None, 28, 28,	2,048	
conv3_block1_0_c... (BatchNormalizatio...	512)		
conv3_block1_3_bn	(None, 28, 28,	2,048	
conv3_block1_3_c... (BatchNormalizatio...	512)		
conv3_block1_add	(None, 28, 28,	0	
conv3_block1_0_b... (Add)	512)		
conv3_block1_3_b...			
conv3_block1_out	(None, 28, 28,	0	
conv3_block1_add... (Activation)	512)		
conv3_block2_1_conv	(None, 28, 28,	65,664	
conv3_block1_out... (Conv2D)	128)		
conv3_block2_1_bn	(None, 28, 28,	512	
conv3_block2_1_c... (BatchNormalizatio...	128)		
conv3_block2_1_relu	(None, 28, 28,	0	
conv3_block2_1_b... (Activation)	128)		

conv3_block2_2_conv	(None, 28, 28,	147,584	
conv3_block2_1_r...	(Conv2D)	128)	
conv3_block2_2_bn	(None, 28, 28,	512	
conv3_block2_2_c...	(BatchNormalizatio...	128)	
conv3_block2_2_relu	(None, 28, 28,	0	
conv3_block2_2_b...	(Activation)	128)	
conv3_block2_3_conv	(None, 28, 28,	66,048	
conv3_block2_2_r...	(Conv2D)	512)	
conv3_block2_3_bn	(None, 28, 28,	2,048	
conv3_block2_3_c...	(BatchNormalizatio...	512)	
conv3_block2_add	(None, 28, 28,	0	
conv3_block1_out...	(Add)	512)	
conv3_block2_3_b...			
conv3_block2_out	(None, 28, 28,	0	
conv3_block2_add...	(Activation)	512)	
conv3_block3_1_conv	(None, 28, 28,	65,664	
conv3_block2_out...	(Conv2D)	128)	

conv3_block3_1_bn	(None, 28, 28,	512	
conv3_block3_1_c...	(BatchNormalizatio...	128)	
conv3_block3_1_relu	(None, 28, 28,	0	
conv3_block3_1_b...	(Activation)	128)	
conv3_block3_2_conv	(None, 28, 28,	147,584	
conv3_block3_1_r...	(Conv2D)	128)	
conv3_block3_2_bn	(None, 28, 28,	512	
conv3_block3_2_c...	(BatchNormalizatio...	128)	
conv3_block3_2_relu	(None, 28, 28,	0	
conv3_block3_2_b...	(Activation)	128)	
conv3_block3_3_conv	(None, 28, 28,	66,048	
conv3_block3_2_r...	(Conv2D)	512)	
conv3_block3_3_bn	(None, 28, 28,	2,048	
conv3_block3_3_c...	(BatchNormalizatio...	512)	
conv3_block3_add	(None, 28, 28,	0	
conv3_block2_out...	(Add)	512)	
conv3_block3_3_b...			
conv3_block3_out	(None, 28, 28,	0	

conv3_block3_add...	(Activation)	512)		
conv3_block4_1_conv	(None, 28, 28,	65,664		
conv3_block3_out...	(Conv2D)	128)		
conv3_block4_1_bn	(None, 28, 28,	512		
conv3_block4_1_c...	(BatchNormalizatio...	128)		
conv3_block4_1_relu	(None, 28, 28,	0		
conv3_block4_1_b...	(Activation)	128)		
conv3_block4_2_conv	(None, 28, 28,	147,584		
conv3_block4_1_r...	(Conv2D)	128)		
conv3_block4_2_bn	(None, 28, 28,	512		
conv3_block4_2_c...	(BatchNormalizatio...	128)		
conv3_block4_2_relu	(None, 28, 28,	0		
conv3_block4_2_b...	(Activation)	128)		
conv3_block4_3_conv	(None, 28, 28,	66,048		
conv3_block4_2_r...	(Conv2D)	512)		
conv3_block4_3_bn	(None, 28, 28,	2,048		
conv3_block4_3_c...				

(BatchNormalization)	512)		
conv3_block4_add conv3_block3_out...	(None, 28, 28,	0	
(Add)	512)		
conv3_block4_3_b...			
conv3_block4_out conv3_block4_add...	(None, 28, 28,	0	
(Activation)	512)		
conv4_block1_1_conv conv3_block4_out...	(None, 14, 14,	131,328	
(Conv2D)	256)		
conv4_block1_1_bn conv4_block1_1_c...	(None, 14, 14,	1,024	
(BatchNormalization)	256)		
conv4_block1_1_relu conv4_block1_1_b...	(None, 14, 14,	0	
(Activation)	256)		
conv4_block1_2_conv conv4_block1_1_r...	(None, 14, 14,	590,080	
(Conv2D)	256)		
conv4_block1_2_bn conv4_block1_2_c...	(None, 14, 14,	1,024	
(BatchNormalization)	256)		
conv4_block1_2_relu conv4_block1_2_b...	(None, 14, 14,	0	
(Activation)	256)		

conv4_block1_0_conv	(None, 14, 14,	525,312	
conv3_block4_out... (Conv2D)	1024)		
conv4_block1_3_conv	(None, 14, 14,	263,168	
conv4_block1_2_r... (Conv2D)	1024)		
conv4_block1_0_bn	(None, 14, 14,	4,096	
conv4_block1_0_c... (BatchNormalizatio...	1024)		
conv4_block1_3_bn	(None, 14, 14,	4,096	
conv4_block1_3_c... (BatchNormalizatio...	1024)		
conv4_block1_add	(None, 14, 14,	0	
conv4_block1_0_b... (Add)	1024)		
conv4_block1_3_b...			
conv4_block1_out	(None, 14, 14,	0	
conv4_block1_add... (Activation)	1024)		
conv4_block2_1_conv	(None, 14, 14,	262,400	
conv4_block1_out... (Conv2D)	256)		
conv4_block2_1_bn	(None, 14, 14,	1,024	
conv4_block2_1_c... (BatchNormalizatio...	256)		



conv4_block2_1_relu	(None, 14, 14,	0	
conv4_block2_1_bn	(Activation)	256	
conv4_block2_2_conv	(None, 14, 14,	590,080	
conv4_block2_1_relu	(Conv2D)	256	
conv4_block2_2_bn	(None, 14, 14,	1,024	
conv4_block2_2_conv	(BatchNormalizatio...	256	
conv4_block2_2_relu	(None, 14, 14,	0	
conv4_block2_2_bn	(Activation)	256	
conv4_block2_3_conv	(None, 14, 14,	263,168	
conv4_block2_2_relu	(Conv2D)	1024	
conv4_block2_3_bn	(None, 14, 14,	4,096	
conv4_block2_3_conv	(BatchNormalizatio...	1024	
conv4_block2_add	(None, 14, 14,	0	
conv4_block1_out	(Add)	1024	
conv4_block2_3_bn			
conv4_block2_out	(None, 14, 14,	0	
conv4_block2_add	(Activation)	1024	

conv4_block3_1_conv conv4_block2_out...	(None, 14, 14,	262,400	
(Conv2D)	256)		
<hr/>			
conv4_block3_1_bn conv4_block3_1_c...	(None, 14, 14,	1,024	
(BatchNormalizatio...	256)		
<hr/>			
conv4_block3_1_relu conv4_block3_1_b...	(None, 14, 14,	0	
(Activation)	256)		
<hr/>			
conv4_block3_2_conv conv4_block3_1_r...	(None, 14, 14,	590,080	
(Conv2D)	256)		
<hr/>			
conv4_block3_2_bn conv4_block3_2_c...	(None, 14, 14,	1,024	
(BatchNormalizatio...	256)		
<hr/>			
conv4_block3_2_relu conv4_block3_2_b...	(None, 14, 14,	0	
(Activation)	256)		
<hr/>			
conv4_block3_3_conv conv4_block3_2_r...	(None, 14, 14,	263,168	
(Conv2D)	1024)		
<hr/>			
conv4_block3_3_bn conv4_block3_3_c...	(None, 14, 14,	4,096	
(BatchNormalizatio...	1024)		
<hr/>			
conv4_block3_add	(None, 14, 14,	0	

conv4_block2_out...	(Add)	1024)		
conv4_block3_3_b...				
conv4_block3_out	(None, 14, 14,	0		
conv4_block3_add...	(Activation)	1024)		
conv4_block4_1_conv	(None, 14, 14,	262,400		
conv4_block3_out...	(Conv2D)	256)		
conv4_block4_1_bn	(None, 14, 14,	1,024		
conv4_block4_1_c...	(BatchNormalizatio...	256)		
conv4_block4_1_relu	(None, 14, 14,	0		
conv4_block4_1_b...	(Activation)	256)		
conv4_block4_2_conv	(None, 14, 14,	590,080		
conv4_block4_1_r...	(Conv2D)	256)		
conv4_block4_2_bn	(None, 14, 14,	1,024		
conv4_block4_2_c...	(BatchNormalizatio...	256)		
conv4_block4_2_relu	(None, 14, 14,	0		
conv4_block4_2_b...	(Activation)	256)		
conv4_block4_3_conv	(None, 14, 14,	263,168		
conv4_block4_2_r...				

(Conv2D)	1024)		
conv4_block4_3_bn	(None, 14, 14,	4,096	
conv4_block4_3_c...	(BatchNormalizatio...	1024)	
conv4_block4_add	(None, 14, 14,	0	
conv4_block3_out...	(Add)	1024)	
conv4_block4_3_b...			
conv4_block4_out	(None, 14, 14,	0	
conv4_block4_add...	(Activation)	1024)	
conv4_block5_1_conv	(None, 14, 14,	262,400	
conv4_block4_out...	(Conv2D)	256)	
conv4_block5_1_bn	(None, 14, 14,	1,024	
conv4_block5_1_c...	(BatchNormalizatio...	256)	
conv4_block5_1_relu	(None, 14, 14,	0	
conv4_block5_1_b...	(Activation)	256)	
conv4_block5_2_conv	(None, 14, 14,	590,080	
conv4_block5_1_r...	(Conv2D)	256)	
conv4_block5_2_bn	(None, 14, 14,	1,024	
conv4_block5_2_c...	(BatchNormalizatio...	256)	

conv4_block5_2_relu conv4_block5_2_b...	(None, 14, 14, (Activation)	0 256)	
conv4_block5_3_conv conv4_block5_2_r...	(None, 14, 14, (Conv2D)	263,168 1024)	
conv4_block5_3_bn conv4_block5_3_c...	(None, 14, 14, (BatchNormalizatio...	4,096 1024)	
conv4_block5_add conv4_block4_out...	(None, 14, 14, (Add)	0 1024)	
conv4_block5_3_b...			
conv4_block5_out conv4_block5_add...	(None, 14, 14, (Activation)	0 1024)	
conv4_block6_1_conv conv4_block5_out...	(None, 14, 14, (Conv2D)	262,400 256)	
conv4_block6_1_bn conv4_block6_1_c...	(None, 14, 14, (BatchNormalizatio...	1,024 256)	
conv4_block6_1_relu conv4_block6_1_b...	(None, 14, 14, (Activation)	0 256)	

conv4_block6_2_conv	(None, 14, 14,	590,080	
conv4_block6_1_r... (Conv2D)	256)		
conv4_block6_2_bn	(None, 14, 14,	1,024	
conv4_block6_2_c... (BatchNormalizatio...	256)		
conv4_block6_2_relu	(None, 14, 14,	0	
conv4_block6_2_b... (Activation)	256)		
conv4_block6_3_conv	(None, 14, 14,	263,168	
conv4_block6_2_r... (Conv2D)	1024)		
conv4_block6_3_bn	(None, 14, 14,	4,096	
conv4_block6_3_c... (BatchNormalizatio...	1024)		
conv4_block6_add	(None, 14, 14,	0	
conv4_block5_out... (Add)	1024)		
conv4_block6_3_b...			
conv4_block6_out	(None, 14, 14,	0	
conv4_block6_add... (Activation)	1024)		
conv5_block1_1_conv	(None, 7, 7, 512)	524,800	
conv4_block6_out... (Conv2D)			

conv5_block1_1_bn	(None, 7, 7, 512)	2,048	
conv5_block1_1_conv	(BatchNormalization)		
conv5_block1_1_relu	(None, 7, 7, 512)	0	
conv5_block1_1_bn	(Activation)		
conv5_block1_2_conv	(None, 7, 7, 512)	2,359,808	
conv5_block1_1_relu	(Conv2D)		
conv5_block1_2_bn	(None, 7, 7, 512)	2,048	
conv5_block1_2_conv	(BatchNormalization)		
conv5_block1_2_relu	(None, 7, 7, 512)	0	
conv5_block1_2_bn	(Activation)		
conv5_block1_0_conv	(None, 7, 7, 2048)	2,099,200	
conv4_block6_output	(Conv2D)		
conv5_block1_3_conv	(None, 7, 7, 2048)	1,050,624	
conv5_block1_2_relu	(Conv2D)		
conv5_block1_0_bn	(None, 7, 7, 2048)	8,192	
conv5_block1_0_conv	(BatchNormalization)		

conv5_block1_3_bn	(None, 7, 7,	8,192	
conv5_block1_3_c...	(BatchNormalizatio...	2048)	
conv5_block1_add	(None, 7, 7,	0	
conv5_block1_0_b...	(Add)	2048)	
conv5_block1_3_b...			
conv5_block1_out	(None, 7, 7,	0	
conv5_block1_add...	(Activation)	2048)	
conv5_block2_1_conv	(None, 7, 7, 512)	1,049,088	
conv5_block1_out...	(Conv2D)		
conv5_block2_1_bn	(None, 7, 7, 512)	2,048	
conv5_block2_1_c...	(BatchNormalizatio...		
conv5_block2_1_relu	(None, 7, 7, 512)	0	
conv5_block2_1_b...	(Activation)		
conv5_block2_2_conv	(None, 7, 7, 512)	2,359,808	
conv5_block2_1_r...	(Conv2D)		
conv5_block2_2_bn	(None, 7, 7, 512)	2,048	
conv5_block2_2_c...	(BatchNormalizatio...		
conv5_block2_2_relu	(None, 7, 7, 512)	0	



conv5_block2_2_bn (Activation)			
conv5_block2_3_conv conv5_block2_2_relu (Conv2D)	(None, 7, 7, 2048)	1,050,624	
conv5_block2_3_bn conv5_block2_3_conv (BatchNormalization)	(None, 7, 7, 2048)	8,192	
conv5_block2_add conv5_block1_output (Add)	(None, 7, 7, 2048)	0	
conv5_block2_3_bn conv5_block2_add (Activation)	(None, 7, 7, 2048)	0	
conv5_block3_1_conv conv5_block2_output (Conv2D)	(None, 7, 7, 512)	1,049,088	
conv5_block3_1_bn conv5_block3_1_conv (BatchNormalization)	(None, 7, 7, 512)	2,048	
conv5_block3_1_relu conv5_block3_1_bn (Activation)	(None, 7, 7, 512)	0	
conv5_block3_2_conv conv5_block3_1_relu (Conv2D)	(None, 7, 7, 512)	2,359,808	

(Conv2D)			
conv5_block3_2_bn	(None, 7, 7, 512)	2,048	
conv5_block3_2_c...	(BatchNormalizatio...		
conv5_block3_2_relu	(None, 7, 7, 512)	0	
conv5_block3_2_b...	(Activation)		
conv5_block3_3_conv	(None, 7, 7,	1,050,624	
conv5_block3_2_r...	(Conv2D)	2048)	
conv5_block3_3_bn	(None, 7, 7,	8,192	
conv5_block3_3_c...	(BatchNormalizatio...	2048)	
conv5_block3_add	(None, 7, 7,	0	
conv5_block2_out...	(Add)	2048)	
conv5_block3_3_b...			
conv5_block3_out	(None, 7, 7,	0	
conv5_block3_add...	(Activation)	2048)	
Total params: 23,587,712 (89.98 MB)			
Trainable params: 0 (0.00 B)			
Non-trainable params: 23,587,712 (89.98 MB)			

Nous avons récupéré le réseau ResNet50 car il est plus petit, donc plus adapté à notre petit dataset, et il a de très bonnes performances. Nous avons désactivé chaque couche de

convolution car nous ne possédons pas un dataset assez gros pour modifier les poids. Nous avons enlevé les couches fully connected pour qu'il s'entraîne sur nos données.

```
def create_final_model(pre_trained_model, last_output):
    # Aplatir la couche de sortie
    x = layers.GlobalAveragePooling2D()(pre_trained_model.layers[-
1].output)
    x = layers.Dense(2048, activation='relu')(x)
    x = layers.Dropout(0.2)(x)
    x = layers.Dense(256, activation='relu')(x)
    x = layers.Dropout(0.2)(x)
    x = layers.Dense(3, activation='softmax')(x)

    # Créer le modèle complet
    model = Model(inputs=pre_trained_model.input, outputs=x)

    model.compile(optimizer=Adam(learning_rate=0.0001),
                  loss='categorical_crossentropy',
                  metrics=['accuracy'])

    return model

model = create_final_model(pre_trained_model,
pre_trained_model.output)
# Inspect parameters
total_params = model.count_params()
num_trainable_params = sum([w.shape.num_elements() for w in
model.trainable_weights])

print(f"There are {total_params:,} total parameters in this model.")
print(f"There are {num_trainable_params:,} trainable parameters in
this model.")

There are 28,309,379 total parameters in this model.
There are 4,721,667 trainable parameters in this model.

#Vu que les classes sont déséquilibré, soit il faut generer des
images par classes,
#soit rajouter des poids au classe
class_weight = {
    0: 0.524, # Colonial # 236/(150*3)
    1: 2.313, # Modern # 236/(34*3)
    2: 1.512 # Prehispanic # 236/(52*3)
}

earlyStopping = EarlyStopping(monitor='val_loss', patience=10,
verbose=0, mode='min')
mcp_save = ModelCheckpoint('bestModel.keras', save_best_only=True,
monitor='val_loss', mode='min')
reduce_lr_loss = ReduceLROnPlateau(monitor='val_loss', factor=0.1,
```

```

patience=7, verbose=1, epsilon=1e-4, mode='min')

val_images, val_labels = next(iter(validation_generator))

history = model.fit(train_generator,
                    validation_data = validation_generator,
                    epochs = 60,
                    callbacks=[earlyStopping, mcp_save,
reduce_lr_loss],
                    verbose=2,
                    class_weight=class_weight)

```

Epoch 1/60

```

/opt/anaconda3/lib/python3.11/site-packages/keras/src/trainers/
data_adapters/py_dataset_adapter.py:120: UserWarning: Your `PyDataset`
class should call `super().__init__(**kwargs)` in its constructor.
`**kwargs` can include `workers`, `use_multiprocessing`,
`max_queue_size`. Do not pass these arguments to `fit()`, as they will
be ignored.

```

```

self._warn_if_super_not_called()

```

```

8/8 - 32s - 4s/step - accuracy: 0.3559 - loss: 1.1405 - val_accuracy:
0.3542 - val_loss: 1.0642 - learning_rate: 1.0000e-04

```

Epoch 2/60

```

8/8 - 24s - 3s/step - accuracy: 0.3390 - loss: 1.1398 - val_accuracy:
0.4167 - val_loss: 1.0614 - learning_rate: 1.0000e-04

```

Epoch 3/60

```

8/8 - 24s - 3s/step - accuracy: 0.4153 - loss: 1.1305 - val_accuracy:
0.5208 - val_loss: 1.0264 - learning_rate: 1.0000e-04

```

Epoch 4/60

```

8/8 - 24s - 3s/step - accuracy: 0.4322 - loss: 1.0738 - val_accuracy:
0.3333 - val_loss: 1.0400 - learning_rate: 1.0000e-04

```

Epoch 5/60

```

8/8 - 25s - 3s/step - accuracy: 0.3898 - loss: 1.0899 - val_accuracy:
0.4375 - val_loss: 1.0115 - learning_rate: 1.0000e-04

```

Epoch 6/60

```

8/8 - 24s - 3s/step - accuracy: 0.4873 - loss: 1.0523 - val_accuracy:
0.6042 - val_loss: 0.9791 - learning_rate: 1.0000e-04

```

Epoch 7/60

```

8/8 - 24s - 3s/step - accuracy: 0.4110 - loss: 1.0153 - val_accuracy:
0.5833 - val_loss: 0.9663 - learning_rate: 1.0000e-04

```

Epoch 8/60

```

8/8 - 24s - 3s/step - accuracy: 0.5975 - loss: 0.9775 - val_accuracy:
0.5625 - val_loss: 0.9617 - learning_rate: 1.0000e-04

```

Epoch 9/60

```

8/8 - 25s - 3s/step - accuracy: 0.3814 - loss: 0.9718 - val_accuracy:
0.5833 - val_loss: 0.9550 - learning_rate: 1.0000e-04

```

Epoch 10/60

8/8 - 24s - 3s/step - accuracy: 0.5847 - loss: 0.9124 - val\_accuracy: 0.5208 - val\_loss: 0.9771 - learning\_rate: 1.0000e-04  
Epoch 11/60  
8/8 - 24s - 3s/step - accuracy: 0.6610 - loss: 1.0048 - val\_accuracy: 0.4792 - val\_loss: 0.9626 - learning\_rate: 1.0000e-04  
Epoch 12/60  
8/8 - 24s - 3s/step - accuracy: 0.4237 - loss: 0.9305 - val\_accuracy: 0.5833 - val\_loss: 0.9373 - learning\_rate: 1.0000e-04  
Epoch 13/60  
8/8 - 24s - 3s/step - accuracy: 0.5551 - loss: 0.9447 - val\_accuracy: 0.5417 - val\_loss: 0.9410 - learning\_rate: 1.0000e-04  
Epoch 14/60  
8/8 - 24s - 3s/step - accuracy: 0.6441 - loss: 0.9411 - val\_accuracy: 0.6042 - val\_loss: 0.9209 - learning\_rate: 1.0000e-04  
Epoch 15/60  
8/8 - 24s - 3s/step - accuracy: 0.4576 - loss: 0.9230 - val\_accuracy: 0.5000 - val\_loss: 0.9387 - learning\_rate: 1.0000e-04  
Epoch 16/60  
8/8 - 24s - 3s/step - accuracy: 0.6186 - loss: 0.8997 - val\_accuracy: 0.5833 - val\_loss: 0.9098 - learning\_rate: 1.0000e-04  
Epoch 17/60  
8/8 - 24s - 3s/step - accuracy: 0.5805 - loss: 0.8740 - val\_accuracy: 0.5833 - val\_loss: 0.8928 - learning\_rate: 1.0000e-04  
Epoch 18/60  
8/8 - 24s - 3s/step - accuracy: 0.5932 - loss: 0.8803 - val\_accuracy: 0.6042 - val\_loss: 0.8924 - learning\_rate: 1.0000e-04  
Epoch 19/60  
8/8 - 24s - 3s/step - accuracy: 0.6186 - loss: 0.8306 - val\_accuracy: 0.5833 - val\_loss: 0.8975 - learning\_rate: 1.0000e-04  
Epoch 20/60  
8/8 - 24s - 3s/step - accuracy: 0.6017 - loss: 0.8403 - val\_accuracy: 0.5833 - val\_loss: 0.8769 - learning\_rate: 1.0000e-04  
Epoch 21/60  
8/8 - 24s - 3s/step - accuracy: 0.6271 - loss: 0.8213 - val\_accuracy: 0.5833 - val\_loss: 0.8784 - learning\_rate: 1.0000e-04  
Epoch 22/60  
8/8 - 24s - 3s/step - accuracy: 0.6356 - loss: 0.8392 - val\_accuracy: 0.5833 - val\_loss: 0.8681 - learning\_rate: 1.0000e-04  
Epoch 23/60  
8/8 - 25s - 3s/step - accuracy: 0.7203 - loss: 0.8510 - val\_accuracy: 0.5625 - val\_loss: 0.9016 - learning\_rate: 1.0000e-04  
Epoch 24/60  
8/8 - 24s - 3s/step - accuracy: 0.6102 - loss: 0.8055 - val\_accuracy: 0.6042 - val\_loss: 0.8702 - learning\_rate: 1.0000e-04  
Epoch 25/60  
8/8 - 26s - 3s/step - accuracy: 0.6695 - loss: 0.8241 - val\_accuracy: 0.6042 - val\_loss: 0.8825 - learning\_rate: 1.0000e-04  
Epoch 26/60  
8/8 - 24s - 3s/step - accuracy: 0.6059 - loss: 0.8157 - val\_accuracy:

0.5417 - val\_loss: 0.8749 - learning\_rate: 1.0000e-04  
Epoch 27/60  
8/8 - 24s - 3s/step - accuracy: 0.6186 - loss: 0.7755 - val\_accuracy:  
0.5625 - val\_loss: 0.8604 - learning\_rate: 1.0000e-04  
Epoch 28/60  
8/8 - 24s - 3s/step - accuracy: 0.6907 - loss: 0.8017 - val\_accuracy:  
0.6042 - val\_loss: 0.8510 - learning\_rate: 1.0000e-04  
Epoch 29/60  
8/8 - 24s - 3s/step - accuracy: 0.6441 - loss: 0.7785 - val\_accuracy:  
0.5833 - val\_loss: 0.8444 - learning\_rate: 1.0000e-04  
Epoch 30/60  
8/8 - 25s - 3s/step - accuracy: 0.7246 - loss: 0.7766 - val\_accuracy:  
0.5625 - val\_loss: 0.8627 - learning\_rate: 1.0000e-04  
Epoch 31/60  
8/8 - 25s - 3s/step - accuracy: 0.6780 - loss: 0.7242 - val\_accuracy:  
0.5417 - val\_loss: 0.8368 - learning\_rate: 1.0000e-04  
Epoch 32/60  
8/8 - 26s - 3s/step - accuracy: 0.6992 - loss: 0.7420 - val\_accuracy:  
0.5625 - val\_loss: 0.8935 - learning\_rate: 1.0000e-04  
Epoch 33/60  
8/8 - 26s - 3s/step - accuracy: 0.5932 - loss: 0.7839 - val\_accuracy:  
0.6250 - val\_loss: 0.8984 - learning\_rate: 1.0000e-04  
Epoch 34/60  
8/8 - 24s - 3s/step - accuracy: 0.6059 - loss: 0.8181 - val\_accuracy:  
0.5417 - val\_loss: 0.9123 - learning\_rate: 1.0000e-04  
Epoch 35/60  
8/8 - 24s - 3s/step - accuracy: 0.6568 - loss: 0.7552 - val\_accuracy:  
0.6042 - val\_loss: 0.8275 - learning\_rate: 1.0000e-04  
Epoch 36/60  
8/8 - 25s - 3s/step - accuracy: 0.6144 - loss: 0.7275 - val\_accuracy:  
0.6042 - val\_loss: 0.8279 - learning\_rate: 1.0000e-04  
Epoch 37/60  
8/8 - 24s - 3s/step - accuracy: 0.6780 - loss: 0.7351 - val\_accuracy:  
0.5833 - val\_loss: 0.8571 - learning\_rate: 1.0000e-04  
Epoch 38/60  
8/8 - 24s - 3s/step - accuracy: 0.6356 - loss: 0.7082 - val\_accuracy:  
0.5833 - val\_loss: 0.8424 - learning\_rate: 1.0000e-04  
Epoch 39/60  
8/8 - 24s - 3s/step - accuracy: 0.7203 - loss: 0.6994 - val\_accuracy:  
0.6250 - val\_loss: 0.8369 - learning\_rate: 1.0000e-04  
Epoch 40/60  
8/8 - 24s - 3s/step - accuracy: 0.6822 - loss: 0.7288 - val\_accuracy:  
0.5833 - val\_loss: 0.8594 - learning\_rate: 1.0000e-04  
Epoch 41/60  
8/8 - 24s - 3s/step - accuracy: 0.6737 - loss: 0.7144 - val\_accuracy:  
0.6042 - val\_loss: 0.8291 - learning\_rate: 1.0000e-04  
Epoch 42/60  
  
Epoch 42: ReduceLROnPlateau reducing learning rate to

9.999999747378752e-06.

8/8 - 24s - 3s/step - accuracy: 0.6610 - loss: 0.6882 - val\_accuracy: 0.6042 - val\_loss: 0.8276 - learning\_rate: 1.0000e-04

Epoch 43/60

8/8 - 24s - 3s/step - accuracy: 0.6822 - loss: 0.6515 - val\_accuracy: 0.6042 - val\_loss: 0.8226 - learning\_rate: 1.0000e-05

Epoch 44/60

8/8 - 24s - 3s/step - accuracy: 0.7331 - loss: 0.6247 - val\_accuracy: 0.5833 - val\_loss: 0.8223 - learning\_rate: 1.0000e-05

Epoch 45/60

8/8 - 24s - 3s/step - accuracy: 0.7415 - loss: 0.6742 - val\_accuracy: 0.6042 - val\_loss: 0.8243 - learning\_rate: 1.0000e-05

Epoch 46/60

8/8 - 23s - 3s/step - accuracy: 0.7288 - loss: 0.6771 - val\_accuracy: 0.6042 - val\_loss: 0.8268 - learning\_rate: 1.0000e-05

Epoch 47/60

8/8 - 15s - 2s/step - accuracy: 0.7542 - loss: 0.6260 - val\_accuracy: 0.6250 - val\_loss: 0.8309 - learning\_rate: 1.0000e-05

Epoch 48/60

8/8 - 13s - 2s/step - accuracy: 0.6737 - loss: 0.6602 - val\_accuracy: 0.5833 - val\_loss: 0.8235 - learning\_rate: 1.0000e-05

Epoch 49/60

8/8 - 14s - 2s/step - accuracy: 0.7076 - loss: 0.6265 - val\_accuracy: 0.5833 - val\_loss: 0.8210 - learning\_rate: 1.0000e-05

Epoch 50/60

8/8 - 14s - 2s/step - accuracy: 0.6992 - loss: 0.6843 - val\_accuracy: 0.5833 - val\_loss: 0.8223 - learning\_rate: 1.0000e-05

Epoch 51/60

8/8 - 14s - 2s/step - accuracy: 0.6737 - loss: 0.6797 - val\_accuracy: 0.5833 - val\_loss: 0.8242 - learning\_rate: 1.0000e-05

Epoch 52/60

8/8 - 14s - 2s/step - accuracy: 0.7034 - loss: 0.6733 - val\_accuracy: 0.6250 - val\_loss: 0.8319 - learning\_rate: 1.0000e-05

Epoch 53/60

8/8 - 14s - 2s/step - accuracy: 0.7119 - loss: 0.6822 - val\_accuracy: 0.6042 - val\_loss: 0.8264 - learning\_rate: 1.0000e-05

Epoch 54/60

8/8 - 14s - 2s/step - accuracy: 0.7161 - loss: 0.6545 - val\_accuracy: 0.5833 - val\_loss: 0.8245 - learning\_rate: 1.0000e-05

Epoch 55/60

8/8 - 14s - 2s/step - accuracy: 0.7161 - loss: 0.6636 - val\_accuracy: 0.5833 - val\_loss: 0.8224 - learning\_rate: 1.0000e-05

Epoch 56/60

Epoch 56: ReduceLROnPlateau reducing learning rate to 9.999999747378752e-07.

8/8 - 14s - 2s/step - accuracy: 0.7161 - loss: 0.6550 - val\_accuracy: 0.6042 - val\_loss: 0.8258 - learning\_rate: 1.0000e-05

Epoch 57/60

```
8/8 - 15s - 2s/step - accuracy: 0.6992 - loss: 0.6848 - val_accuracy: 0.6042 - val_loss: 0.8264 - learning_rate: 1.0000e-06
Epoch 58/60
8/8 - 14s - 2s/step - accuracy: 0.7246 - loss: 0.6164 - val_accuracy: 0.6042 - val_loss: 0.8266 - learning_rate: 1.0000e-06
Epoch 59/60
8/8 - 14s - 2s/step - accuracy: 0.7034 - loss: 0.6391 - val_accuracy: 0.6042 - val_loss: 0.8266 - learning_rate: 1.0000e-06
```

## Performance

```
# Plot the training and validation accuracies for each epoch
```

```
acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']

epochs = range(len(acc))

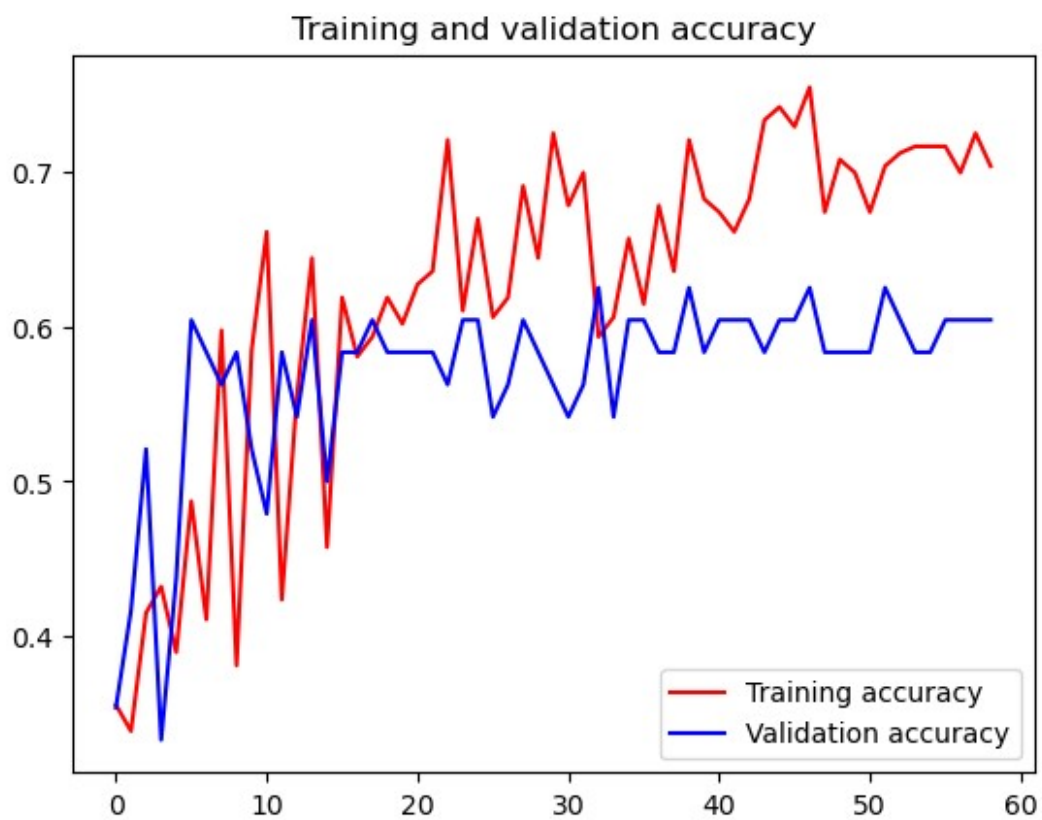
plt.plot(epochs, acc, 'r', label='Training accuracy')
plt.plot(epochs, val_acc, 'b', label='Validation accuracy')
plt.title('Training and validation accuracy')
plt.legend(loc=0)
plt.figure()

plt.show()

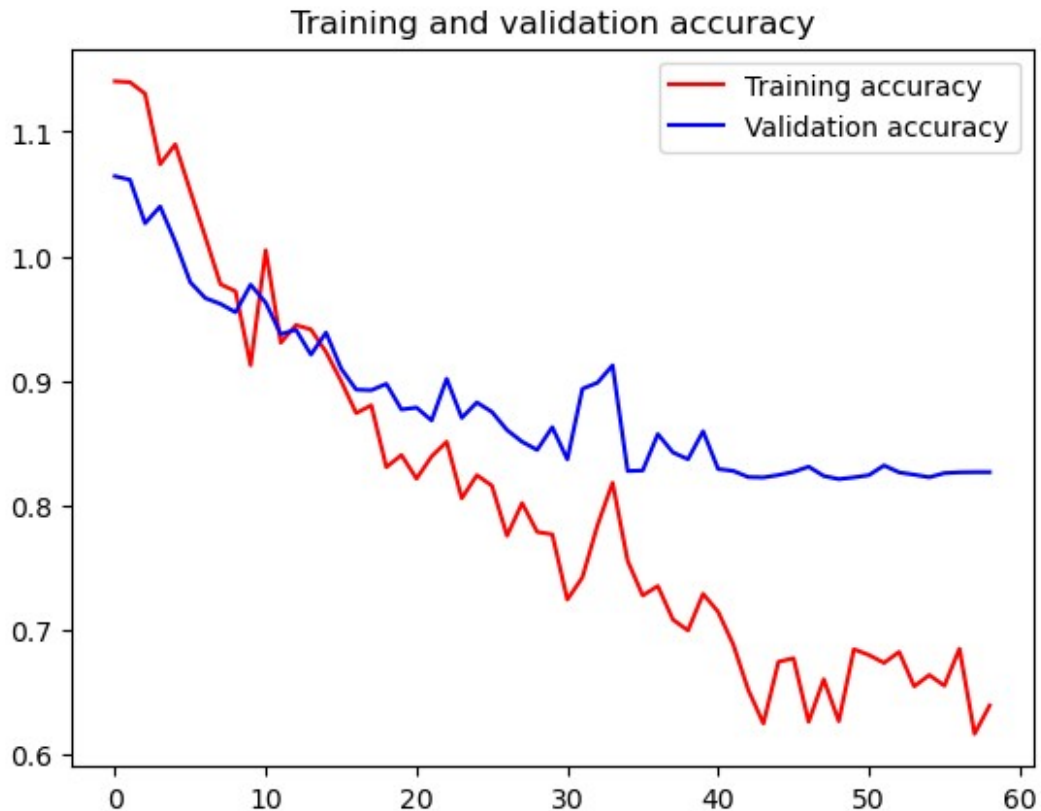
plt.plot(epochs, loss, 'r', label='Training accuracy')
plt.plot(epochs, val_loss, 'b', label='Validation accuracy')
plt.title('Training and validation accuracy')
plt.legend(loc=0)
plt.figure()

plt.show()
```





<Figure size 640x480 with 0 Axes>



<Figure size 640x480 with 0 Axes>

On peut constater qu'il faut baisser le learning rate car il est en dents de scie

## Chargement du meilleur model de validation

```
model = keras.saving.load_model('bestModel.keras')
```

## Sur les données de validation

```
import numpy as np
from sklearn.metrics import classification_report

# Supposons que validation_generator est déjà défini
val_images, val_labels = next(iter(validation_generator)) # Obtenir
un lot d'images et d'étiquettes de validation

# Faire des prédictions sur les données de validation
predictions = model.predict(val_images)

# Convertir les prédictions en classes
predicted_classes = np.argmax(predictions, axis=1)
true_classes = np.argmax(val_labels, axis=1)

report = classification_report(true_classes, predicted_classes,
```

```

output_dict=True)

# Afficher l'accuracy pour chaque classe
for class_label, metrics in report.items():
    if isinstance(metrics, dict): # Pour éviter les entrées non-
    classe comme 'accuracy' globale
        print(f"Classe {class_label}: Accuracy =
{metrics['precision']:.2f}")

1/1 ————— 1s 1s/step
Classe 0: Accuracy = 0.33
Classe 1: Accuracy = 0.29
Classe 2: Accuracy = 0.67
Classe macro avg: Accuracy = 0.43
Classe weighted avg: Accuracy = 0.46

```

On peut voir que la classe Modern a toujours du mal malgré les poids sur les classes ; ainsi, il aurait quand même fallu augmenter le dataset

```

def is_image_in_list(image, image_list):
    for img in image_list:
        if np.array_equal(image, img[0]):
            return True
    return False

val_images, val_labels = next(iter(validation_generator))

# Faire des prédictions
predictions = model.predict(val_images)
predicted_classes = np.argmax(predictions, axis=1)
true_classes = np.argmax(val_labels, axis=1)

# Identifier les prédictions incorrectes
wrong_predictions = []

def save_image(predictions_list, condition, comparator):
    for i in range(len(val_images)):
        if condition(predicted_classes[i], true_classes[i]):
            # Enregistrer l'image, la prédiction et le score
            d'exactitude
            accuracy = comparator(predictions[i]) # Utiliser les
            prédictions actuelles
            if not is_image_in_list(val_images[i], predictions_list):
                predictions_list.append((val_images[i],
                predicted_classes[i], accuracy))
    return predictions_list # Correction pour retourner la liste des
    prédictions

# Utilisation des opérateurs pour les conditions

```

```
wrong_predictions = save_image(wrong_predictions, lambda pred, true:
pred != true, np.max)
```

1/1 ————— 2s 2s/step

```
def plot_top_predictions(predictions, top_n=10):
    # Trier les prédictions par score
    predictions.sort(key=lambda x: x[2], reverse=True) # Trier par
    précision

    # Afficher les images avec les scores les plus élevés
    plt.figure(figsize=(15, 10))
    for i in range(min(top_n, len(predictions))):
        img, pred_class, accuracy = predictions[i]
        plt.subplot(2, 5, i + 1)
        plt.imshow(img)
        plt.title(f'Pred: {pred_class}, Acc: {accuracy:.2f}')
        plt.axis('off')
    plt.show()

print("____Wrong Predictions____")
plot_top_predictions(wrong_predictions, top_n=10)
```

\_\_\_\_Wrong Predictions\_\_\_\_

Pred: 0, Acc: 0.87



Pred: 0, Acc: 0.79



Pred: 0, Acc: 0.78



Pred: 0, Acc: 0.74



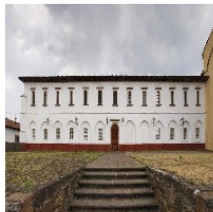
Pred: 2, Acc: 0.67



Pred: 1, Acc: 0.62



Pred: 1, Acc: 0.61



Pred: 2, Acc: 0.58



Pred: 0, Acc: 0.50



Pred: 1, Acc: 0.45

