Redes Neuronales y Deep Learning: Proyecto de programación "Deep Vision in classification tasks"

ID Group: 07MIAR06

Dataset: Plant Seedlings Classification

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El código de la práctica se encuentra en el siguiente repositorio de GitHub: https://github.com/daniabellan/Plant-Seedlings-Classification

Disclaimer:

Dado que el dataset de Kaggle no presentan etiquetas correspondientes a las imágenes, hemos tomado la decisión de evaluar los modelos en el dataset de validación. Somos conscientes de que no debería ser así en un entorno real, pero no tenemos otra manera de aplicar el performance de los algoritmos y modelos.

Imports Librerías

```
import tensorflow as tf
import os
import pandas as pd
from PIL import Image
import matplotlib.pyplot as plt
from typing import Tuple, Dict
import numpy as np
import math

from sklearn.metrics import classification_report, confusion_matrix
from sklearn.utils.class_weight import compute_class_weight
import tensorflow.keras.applications.efficientnet_v2 as effnV2
from tensorflow.keras.layers import Dense, Concatenate, GlobalMaxPooling2D, GlobalAveragePooling2D, Dropout
```

Descarga de datos desde la plataforma Kaggle

Para este proyecto se va a hacer uso del dataset de kaggle https://www.kaggle.com/c/plant-seedlings-classification/data

En esta ocasión, por dificultades en la descarga directa del dataset desde kaggle, hemos obtado por descargarlo y hacer uso de éste en local.

Dentro del repositorio de Github con el código, en la carpeta de *dataset/* disponemo de dos subcarpetas llamadas *test/* y *train/* con las muestras pertinentes a cada partición. Todo esto es respetando la jerarquía original del dataset.

1. Carga del conjunto de datos

Definimos las funciones necesarias para la carga de los datasets

```
In [2]: # Definimos un diccionario para mapear los nombres de las clases con sus indices
dict_map_class = {
    'Loose Silky-bent': 0,
    'Common Chickweed': 1,
    'Scentless Mayweed': 2,
    'Small-flowered Cranesbill': 3,
    'Fat Hen': 4,
    'Charlock': 5,
    'Sugar beet': 6,
```

```
'Black-grass': 8,
    'Shepherds Purse': 9,
    'Common wheat': 10,
    'Maize': 11.
}
# Invertimos el diccioanrio del mapeo de clases
dict_map_class_inverted = {v: k for k, v in dict_map_class.items()}
def get dict dataset(
    dataset_path: str = 'dataset'
    Crea un diccionario que asigna las rutas de las imágenes a tuplas que contienen la clase y la partición.
        dataset_path (str): Ruta al directorio del conjunto de datos. Por defecto, es 'dataset'.
       dict: Un diccionario donde las claves son las rutas de las imágenes y los valores son tuplas (clase, pa
    dict_dataset = {}
    train_classes = os.listdir(os.path.join(dataset_path, 'train'))
    for train class in train classes:
        class_path = os.path.join(dataset_path, 'train', train_class)
        train_imgs = os.listdir(class_path)
        # Split Train images to a 80% for a Train Split for each class
        for train img in train imgs[:int(len(train imgs)*0.8)]:
            train img path = os.path.join(class path, train img)
            dict_dataset[train_img_path] = (dict_map_class[train_class], 'Train')
        # Assign the rest 20% to Valid Split for each class
        for valid img in train imgs[int(len(train imgs)*0.8):]:
            valid_img_path = os.path.join(class_path, valid_img)
            dict_dataset[valid_img_path] = (dict_map_class[train_class], 'Valid')
    # Geting Test Images
    test_path = os.path.join(dataset_path, 'test')
    test_imgs = os.listdir(test_path)
    for test img in test imgs:
        test img path = os.path.join(test path, test img)
        dict dataset[test img path] = ("Unkown", 'Test')
    return dict dataset
def dict2dataframe(
    input_dict: Dict[str, Tuple]
    Convierte un diccionario a un DataFrame de pandas con columnas para 'path', 'label', y 'split'.
        input dict (dict): Un diccionario donde las claves son las rutas de las imágenes y los valores son tupl
    pd.DataFrame: Un DataFrame con columnas 'path', 'label' y 'split'.
    df = pd.DataFrame([(key, values[0], values[1]) for key, values in input_dict.items()], columns=['path', 'la
    # Returns shuffled datasets
    return df.sample(frac=1, random state=42).reset index(drop=True)
```

Definimos las funciones para la creación del Datset de TensorFlow

'Cleavers': 7,

```
image = output[0]
        # De manera probabilística, con 50% de que ocurra, se aplicará
        # una de estas transformaciones
        if tf.random.uniform([1], minval=0, maxval=1) > 0.5:
            image = tf.image.random_flip_up_down(image)
        if tf.random.uniform([1], minval=0, maxval=1) > 0.5:
            image = tf.image.random_flip_left_right(image)
        if tf.random.uniform([1], minval=0, maxval=1) > 0.5:
            image = tf.image.random_brightness(image, 0.15)
        if tf.random.uniform([1], minval=0, maxval=1) > 0.7:
            image = tf.image.random_contrast(image, 0.6, 1.4)
        output[0] = image
        return output
    return f
@tf.function
def load_image(*inputs):
    Función de TensorFlow para cargar una imagen utilizando una función de NumPy.
        *inputs: Número variable de tensores de entrada.
    Returns:
    list: Una lista de tensores de salida con la imagen cargada como primer elemento.
    outputs = list(inputs)
    image = tf.numpy_function(load_image_np, [inputs[0]], tf.float32)
image.set_shape([None, None, 3])
    outputs[0] = image
    return outputs
def load image np(path):
    Carga una imagen con PIL desde la ruta especificada y la convierte en un array de NumPy.
       path (str): La ruta al archivo de la imagen.
    np.ndarray: Un array de NumPy que representa la imagen cargada en formato RGB.
    return np.array(Image.open(path).convert('RGB')).astype(np.float32)
def resize(index=0, resize_to=None):
    Devuelve una función de TensorFlow para cambiar el tamaño de una imagen en un conjunto de datos.
        index (int): Índice del tensor de la imagen en el conjunto de datos. Por defecto, es 0.
        resize_to (tuple, list o None): Tamaño objetivo para cambiar el tamaño. Si es None, no se realiza ningú
    callable: Una función de TensorFlow que cambia el tamaño de la imagen en el conjunto de datos.
    def f(*dataset):
        output = list(dataset)
        # Hacemos resize al input size deseado
        resized_image = tf.image.resize(dataset[index], resize_to)
        # Casteamos de nuevo a uint8 para dejar la imagen entre 0 y 255
        resized_image = tf.cast(resized_image, tf.uint8)
output[index] = resized_image
        return output
    return f
def preprocess input(index):
    Devuelve una función de TensorFlow para preprocesar una imagen en un conjunto de datos.
    Aras:
        index (int): Índice del tensor de la imagen en el conjunto de datos.
    callable: Una función de TensorFlow que preprocesa la imagen en el conjunto de datos.
    @tf.function
    def f(*dataset):
        output = list(dataset)
        image = dataset[index]
        # Convertimos a tf.float32 la imagen
        image = tf.cast(image, tf.float32)
```

```
# Normalizamos la imagen entre 0 y 1
        image = image / 255.
       output[index] = image
        return output
    return f
def get_dataset(
   df: pd.DataFrame,
   input_size: Tuple[int, int],
    shuffle: bool = False,
    batch_size: int = None,
   gray_scale: bool = False,
   augmenter: bool = False,
   num_aug: int = None,
    test set: bool = False
)->tf.data.Dataset:
    Crea un dataset de TensorFlow a partir de un DataFrame.
   Aras:
        df (pd.DataFrame): DataFrame que contiene información sobre el conjunto de datos.
        input size (Tuple[int, int]): Tupla que representa el tamaño objetivo para cambiar el tamaño de las imá
        shuffle (bool): Si se debe barajar el conjunto de datos. Por defecto, es False.
        batch_size (int): Tamaño del lote para el conjunto de datos. Si es None, no se realiza agrupamiento. Po
        gray_scale (bool): Si se deben convertir las imágenes a escala de grises. Por defecto, es False.
       augmenter (bool): Si se debe aplicar aumentación de datos. Por defecto, es False.
num_aug (int): Número de aumentaciones a aplicar si augmenter es True. Por defecto, es None.
        test_set (bool): Si el conjunto de datos es un conjunto de prueba. Por defecto, es False.
   Returns:
    tf.data.Dataset: Un conjunto de datos de TensorFlow preparado según las opciones proporcionadas.
    # Imprimir información sobre la distribución de las etiquetas
   print('Number of instances per label: '
          pd.Series(df['label']).value_counts(), sep='\n')
    print('\nPercentaje of instances per label:
          pd.Series(df['label']).value_counts().div(pd.Series(df['label']).shape[0]),
          sep='\n')
   names = np.array(df['path'], dtype=str)
    if not test set:
        labels = np.array(tf.keras.utils.to categorical(df['label'], num classes=12))
   else:
        labels = np.ones(len(names))
   data = names, labels
    # Creamos un dataet de TensorFlow a partir de las rutas y etiquetas
   dataset = tf.data.Dataset.from tensor slices(data)
    # Barajamos el dataset
   if shuffle:
        print(' > Shuffle')
        dataset = dataset.shuffle(len(names))
    # Cargamos las imagenes
   dataset = dataset.map(load_image, num_parallel_calls=tf.data.experimental.AUTOTUNE)
    # Aplicamos el resize
   dataset = dataset.map(resize(resize to=input size), num parallel calls=tf.data.experimental.AUTOTUNE)
    # Aumentamos los datos
    if augmenter:
        print(f' > Augmentamos datos numero {num aug}')
        if num_aug == 1:
            dataset = dataset.map(tf augmenter(), num parallel calls=tf.data.experimental.AUTOTUNE)
   # Preprocesamos las entradas
   dataset = dataset.map(preprocess input(0), num parallel calls=tf.data.experimental.AUTOTUNE)
    # Convertimos a escala de grises
   if gray scale:
        print(' > Escala de grises')
        dataset = dataset.map(lambda *args: (tf.image.rgb_to_grayscale(args[0]), *args[1:]))
    # Definimos el batch size
   if batch size is not None:
        print(' > Establecemos el batchsize')
        dataset = dataset.batch(batch_size)
    # Prefetch para paralelizar los siguientes batches mientras se evalua un batch
   dataset = dataset.prefetch(tf.data.experimental.AUTOTUNE)
    return dataset
```

Cargamos el dataset en un pd.DataFrame y visualizamos los datos

```
In [4]: dict_dataset = get_dict_dataset(dataset path="dataset/")
          # Creamos el DataFrame
          df dataset = dict2dataframe(dict dataset)
          df dataset
                                                      path
                                                               label split
Out[4]:
             0 dataset/train/Scentless Mayweed/8c496e84a.png
                                                                  2 Train
                     dataset/train/Common wheat/4a56f32c6.png
                                                                     Train
             2
                                  dataset/test/99569b224.png
                                                                     Test
                                                           Unkown
             3
                        dataset/train/Black-grass/e0380dff9.png
                                                                  8 Valid
             4
                           dataset/train/Fat Hen/9708f9c0e.png
                                                                  4 Train
          5539
                   dataset/train/Shepherds Purse/953ced7c6.png
                                                                  9 Train
                                   dataset/test/8cf909eb3.png
                                                                     Test
                                   dataset/test/98062cd87.png
          5541
                                                           Unkown Test
          5542
                                   dataset/test/cc3d2a59a.png
                                                                     Test
                          dataset/train/Cleavers/b6220c08e.png
                                                                     Train
```

En el DataFrame anterior podemos visualizar por cada índice, el path de la imagen junto a su etiqueta asociada y al split que perteneces (Train, Valid o Test).

Aquellas imagenes pertenecientes a Test presentan la etiqueta de Unkown. Esto se debe a que el dataset de kaggle no tiene el dataset de test etiquetado, por ello para diferenciarlo aún más le ponemos esta etiqueta.

2. Inspección del conjunto de datos

5544 rows × 3 columns

Dado que ya disponemos del DataFrame con todas las muestras, vamos a crear cada partición y a su vez, visualizar las distribuciones de las etiquetas por cada uno de los splits.

```
In [5]: # Creamos dataset de Train
          df_dataset_train = df_dataset[df_dataset['split'] == 'Train'].drop(columns=['split'])
          df_dataset_train
                                                     path label
             0 dataset/train/Scentless Mayweed/8c496e84a.png
                    dataset/train/Common wheat/4a56f32c6.png
                                                             10
             4
                           dataset/train/Fat Hen/9708f9c0e.png
                       dataset/train/Sugar beet/6d623072a.png
                dataset/train/Small-flowered Cranesbill/869252
             8
                                                              3
          5534
                dataset/train/Scentless Mayweed/948251df3.png
          5535
                        dataset/train/Sugar beet/29a0e6bf9.png
          5537
                         dataset/train/Charlock/8b35222d0.png
          5539
                  dataset/train/Shepherds Purse/953ced7c6.png
          5543
                         dataset/train/Cleavers/b6220c08e.png
         3794 rows × 2 columns
          # Visualizamos distribuciones por clase
          df_dataset_train['label'].value_counts()
```

```
Out[6]: label
                523
         1
                488
         2
                412
         3
                396
         4
                380
                312
         6
                308
         7
                229
         8
                210
         9
                184
         10
                176
         11
                176
         Name: count, dtype: int64
```

Si nos fijamos en la distribución de muestras por clases del splir de train podemos ver como se presenta un desbalanceo de clases notable. Podemos ver como de la clase 0 disponemos de 523 muestras, mientras que de la clase 11 disponemos nomás de 176 muestras.

Dado este desbalanceo, se va a obtar por entrenar los modelos pasándoles los pesos de cada una de las clases mediante el netodo compute_class_weight() de la librería de sklearn.

Otra opción era duplicar las muestras de las clases minoritarias e igualar las muestras de la mayoritaria, pero esto supondría la repetición de muchas imágenes iguales y se ha descartado

```
In [7]:
        # Calculamos los pesos de cada clase debido al desbalanceo de las clases
        class_weights = compute_class_weight(class_weight='balanced'
                                              classes=np.unique(df dataset train['label'].values),
                                              y=df_dataset_train['label'].values
        )
        class weight dict = dict(enumerate(class_weights))
        class_weight_dict
        {0: 0.6045251752708731,
         1: 0.6478825136612022,
         2: 0.7673948220064725,
         3: 0.7984006734006734,
         4: 0.8320175438596491.
         5: 1.0133547008547008,
         6: 1.0265151515151516,
         7: 1.3806404657933042.
         8: 1.50555555555555,
         9: 1.7182971014492754,
         10: 1.7964015151515151,
         11: 1.7964015151515151}
```

Por lo que para calcular los pesos de las clases, que posteriormente se les pasarán en el fit del modelo, se hace uso del siguiente código.

Del resultado podemos ver como las clases minoritarias presentan mayor peso que las que tienen mas presencia en nuestro dataset. Por ello, de las clases con mayor peso, la red ponderará más.

Este cálculo de los pesos sólo lo haremos para el set de Train, para los otros no es necesario

```
In [8]: # Creamos dataset de Validación
    df_dataset_valid = df_dataset[df_dataset['split'] == 'Valid'].drop(columns=['split'])
    df_dataset_valid
```

```
Out[8]:
               3
                           dataset/train/Black-grass/e0380dff9.png
              10
                  dataset/train/Scentless Mayweed/d748c7307.png
              15
                             dataset/train/Fat Hen/e6b756e98.png
             22
                    dataset/train/Small-flowered Cranesbill/ecf58a...
                                                                      3
              24
                             dataset/train/Charlock/d1b362c43.png
           5500
                        dataset/train/Loose Silky-bent/fc2b27fff.png
                                                                      0
           5501
                             dataset/train/Charlock/fc3e15a2e.png
           5503 dataset/train/Scentless Mayweed/d1e775b97.png
                                                                      2
           5505 dataset/train/Common Chickweed/f50c8181a.png
                                                                      1
           5538
                                dataset/train/Maize/e30accd2f.png
```

956 rows × 2 columns

```
In [9]: # Visualizamos distribuciones por clase
df_dataset_valid['label'].value_counts()
```

```
Out[9]:
           0
                  131
           1
                  123
                  104
           3
                  100
           4
                   95
                  78
           6
                   77
           7
                   58
           8
                   53
                   47
           11
                   45
           10
                   45
           Name: count, dtype: int64
In [10]: # Creamos dataset de Test
           df_dataset_test = df_dataset[df_dataset['split'] == 'Test'].drop(columns=['split'])
           df_dataset_test
                                   path
                                           label
Out[10]:
              2 dataset/test/99569b224.png Unkown
              6 dataset/test/d14aa43f3.png Unkown
              7 dataset/test/b47691c08.png Unkown
                  dataset/test/19fdf19fb.png Unkown
             14 dataset/test/7d4cd07ad.png Unkown
           5526 dataset/test/b30ab4659.png Unkown
           5536 dataset/test/bb1c84bbc.png Unkown
           5540 dataset/test/8cf909eb3.png Unkown
           5541 dataset/test/98062cd87.png Unkown
           5542 dataset/test/cc3d2a59a.png Unkown
          794 rows × 2 columns
In [11]: # Visualizamos distribuciones por clase
           df_dataset_test['label'].value_counts()
           Unkown
                      794
           Name: count, dtype: int64
           Vistas las distribuciones de muestras por Split, vamos a visualizar unas muestras del Entrenamiento.
In [12]: fig = plt.figure(figsize = (20, 5))
           for i in range(20):
                img = Image.open(df_dataset_train.iloc[i]['path'])
               ax = fig.add_subplot(2, 10, i + 1)
                ax.imshow(img.convert('RGB'))
                title = dict_map class inverted[df_dataset_train.iloc[i]['label']]
               ax.set_title(title, fontsize=10)
               ax.axis('off')
           Scentless Mayweed
                                                                               Charlock
                                                                                       Small-flowered Cranesbill Scentless Mayv
                                                                                                                                   Charlock
              Fat Hen
                           Cleavers
                                      Shepherds Purse
                                                     Sugar beet
                                                                 Black-grass
                                                                                Maize
                                                                                                      Common Chickweed
                                                                                                                    Loose Silky-bent
```

3. Acondicionamiento del conjunto de datos

Creados los Splits para el Dataset original, podemos pasar a preparar un ejemplo de los TensorFlow Datasets pertinentes para cada uno de los Splits (Train, Val y Test) que se van a usar para las siguientes pruebas. A su vez, también los visualizaremos para ver qué imagenes entrarían a la red

label

```
TH [T9]:
         train_tfdataset = get_dataset(
             df=df_dataset_train,
             input size=(224,224),
             batch_size=16,
             shuffle=True,
             gray scale=False,
             augmenter=True,
             num aug=1,
         train_tfdataset
         Number of instances per label:
         label
         0
               523
               488
         1
         2
               412
         3
               396
         4
               380
         5
               312
         6
               308
         7
               229
         8
               210
         9
               184
         10
               176
         11
               176
         Name: count, dtype: int64
         Percentaje of instances per label:
         label
               0.137849
         0
         1
               0.128624
               0.108593
         3
               0.104375
         4
               0.100158
         5
               0.082235
         6
               0.081181
         7
               0.060358
         8
               0.055351
               0.048498
         10
               0.046389
         11
               0.046389
         Name: count, dtype: float64
         2023-11-26 18:42:44.832637: I tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa_node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:44.837225: I tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa_node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:44.837457: I tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:44.840091: I tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:44.840385: I tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:44.840636: I tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa_node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:45.186309: I tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa_node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:45.186661: I tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa_node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:45.186697: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1977] Could not identify NUMA no
         de of platform GPU id 0, defaulting to 0. Your kernel may not have been built with NUMA support.
         2023-11-26 18:42:45.187050: I tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:880] could not
         open file to read NUMA node: /sys/bus/pci/devices/0000:06:00.0/numa node
         Your kernel may have been built without NUMA support.
         2023-11-26 18:42:45.187096: I tensorflow/core/common_runtime/gpu/gpu_device.cc:1886] Created device /job:localh
         ost/replica:0/task:0/device:GPU:0 with 7551 MB memory: -> device: 0, name: NVIDIA GeForce RTX 3080, pci bus id
         : 0000:06:00.0, compute capability: 8.6
          > Shuffle
          > Augmentamos datos numero 1
          > Establecemos el batchsize
         <_PrefetchDataset element_spec=(TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name=None), TensorSpec(</pre>
Out[13]:
         shape=(None, 12), dtype=tf.float32, name=None))>
In [14]: # Visualizamos dataset de Train
         batch = next(iter(train tfdataset))
         batches card np = batch[0].numpy()
         y = batch[1].numpy()
         images = list(iter(batches_card_np))
```

```
fig, axs = plt.subplots(math.ceil(len(images)/4), 4, figsize=(15, math.ceil(len(images)/4)*4))
axs = axs.ravel()
for i, image in enumerate(images):
     axs[i].imshow(image, cmap='gray')
title = f"{dict_map_class_inverted[np.argmax(y[i])]}"
     axs[i].set_title(title)
           Loose Silky-bent
                                             Common Chickweed
                                                                                 Scentless Mayweed
                                                                                                                      Loose Silky-bent
  0
                                      0
                                                                         0
                                                                                                             0
 25
                                                                                                            25
                                     25
                                                                        25
 50
                                     50
                                                                        50
                                                                                                            50
 75
                                     75
                                                                        75
                                                                                                            75
                                                                        100
100
                                    100
                                                                                                           100
125
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# Creamos un ejemplo de datastet de Validacion
valid_tfdataset = get_dataset(
     df=df dataset valid,
     input size=(224,224),
```

```
Number of instances per label:
          label
          0
                131
                123
          1
          2
                104
          3
                100
          4
                 95
          5
                 78
          6
                 77
          7
                 58
          8
                 53
          9
                 47
          11
                 45
          10
                 45
          Name: count, dtype: int64
          Percentaje of instances per label:
          label
          0
                0.137029
                0.128661
          1
          2
                0.108787
          3
                0.104603
                0.099372
          4
          5
                0.081590
          6
                0.080544
                0.060669
          7
                0.055439
          8
          9
                0.049163
               0.047071
          11
               0.047071
          10
          Name: count, dtype: float64
          > Shuffle
          > Establecemos el batchsize
Out[15]: < PrefetchDataset element spec=(TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name=None), TensorSpec(shape=(None, 12), dtype=tf.float32, name=None))>
In [16]: # Visualizamos dataset de Train
          batch = next(iter(valid_tfdataset))
          batches_card_np = batch[0].numpy()
          y = batch[1].numpy()
          images = list(iter(batches card np))
          fig, axs = plt.subplots(math.ceil(len(images)/4), 4, figsize=(15, math.ceil(len(images)/4)*4))
          axs = axs.ravel()
          for i, image in enumerate(images):
              axs[i].imshow(image, cmap='gray')
              title = f"{dict_map_class_inverted[np.argmax(y[i])]}"
              axs[i].set_title(title)
```



```
images = list(iter(batches_card_np))
fig, axs = plt.subplots(math.ceil(len(images)/4), 4, figsize=(15, math.ceil(len(images)/4)*4))
axs = axs.ravel()
for i, image in enumerate(images):
     axs[i].imshow(image, cmap='gray')
     title = f"{dict_map_class_inverted[np.argmax(y[i])]}"
     axs[i].set_title(title)
           Loose Silky-bent
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                                    125
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                                    150
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175
                                    175
                                                                       175
                                                                                                           175
200
                                                    100
                                                          150
```

4.1 Desarrollo de la arquitectura de red neuronal y entrenamiento de la solución (from scratch)

Creamos los callbacks

```
In [19]:
    from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping
    from tensorflow.keras.callbacks import Callback
    import matplotlib
    import numpy as np
    import matplotlib.pyplot as plt
    import pandas as pd

class HistoryAndPlotCallback(Callback):
    def __init__(self, model_name):
        super(HistoryAndPlotCallback, self).__init__()
```

```
self.model name = model name
    def on train end(self, logs=None):
         # Guardar el historial del modelo
         history = self.model.history.history
         # Calcular límite eje Y
         loss_av = np.average(history["loss"][-1])
         val_loss_av = np.average(history["val_loss"][-1])
         acc_av = np.average(history["acc"][-1])
         val_acc_av = np.average(history["val_acc"][-1])
         y_upper_limit = np.max([loss_av, val_loss_av, acc_av, val_acc_av])
         y_upper_limit *= 2
         np.save(f'models/{self.model name}/training history.npy', history)
         # Guardar métricas y pérdidas en una sola gráfica
         epochs = len(history["loss"])
         plt.style.use("ggplot")
         plt.figure(figsize=(12, 8))
        plt.plot(np.arange(0, epochs), history["loss"], label="train_loss")
plt.plot(np.arange(0, epochs), history["val_loss"], label="val_loss")
plt.plot(np.arange(0, epochs), history["acc"], label="train_acc")
         plt.plot(np.arange(0, epochs), history["val acc"], label="val acc")
         plt.title(f"Training Loss and Accuracy - {self.model_name}")
         plt.xlabel("Epoch #")
         plt.ylabel("Loss/Accuracy")
         plt.legend(loc='upper left')
         plt.ylim(0, int(y upper limit))
         plt.grid(True)
         # Guardar la gráfica
         plt.savefig(f'models/{self.model name}/training plot.png')
         plt.close()
         # Guardar history para revisar en el futuro por si se activa un earlystopping o cualquier problema
         pd.DataFrame.from_dict(history).to_csv(f'models/{self.model_name}/history.csv', index=False)
def get callbacks(model name):
    callbacks = [
         EarlyStopping(
    monitor = 'val_loss',
             mode = 'min'
             patience = 20,
             verbose=1,
             restore best weights=False),
         ModelCheckpoint(
             filepath = f'models/{model_name}',
monitor = 'val_loss',
             mode = 'min'
             save_best_only = True,
             verbose = 1)
         HistoryAndPlotCallback(model name),
    ]
    return callbacks
```

Hemos creado Callbacks para almacenar los mejores checkpoints de los modelos a medida que se iban entrenando las redes.

Hemos decidido incluir el EarlyStopping para evitar un sobreentrenamiento y gastar recursos computacionales.

Las métricas a monitorear es el val_loss. Se intenta minimizar en cada época.

También hemos almacenado el historial de entrenamiento para, posteriormente, poder obtener las gráficas. También está añadido el callback para guardar las gráficas después de cada época.

Creamos los datasets para train, val y test

```
Dataset Train (TensorFlow):
         Number of instances per label:
         label
         0
                488
         1
         2
                412
         3
                396
         4
                380
         5
                312
         6
                308
                229
         8
                210
         9
                184
         10
                176
                176
         11
         Name: count, dtype: int64
         Percentaje of instances per label:
         label
                0.137849
         0
                0.128624
         2
                0.108593
         3
                0.104375
         4
                0.100158
                0.082235
         6
                0.081181
         7
                0.060358
         8
                0.055351
         9
                0.048498
                0.046389
         10
         11
                0.046389
         Name: count, dtype: float64
          > Shuffle
          > Augmentamos datos numero 1
          > Establecemos el batchsize
         <_PrefetchDataset element_spec=(TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name=None), TensorSpec(</pre>
         shape=(None, 12), dtype=tf.float32, name=None))>
In [21]: print("Dataset Valid (TensorFlow):")
         valid_tfdataset = get_dataset(
              df=df dataset valid,
              input size=(224,224),
              batch_size=32,
              shuffle=False,
              gray scale=False,
              augmenter=False,
         valid tfdataset
         Dataset Valid (TensorFlow):
         Number of instances per label:
         label
                131
         0
         1
                123
         2
                104
                100
         3
         4
                 95
         5
                 78
         6
                 77
         7
                 58
         8
                 53
         9
                 47
         11
                 45
         10
                 45
         Name: count, dtype: int64
         Percentaje of instances per label:
         label
         0
                0.137029
                0.128661
         1
         2
                0.108787
         3
                0.104603
                0.099372
         5
                0.081590
         6
                0.080544
                0.060669
         8
                0.055439
         9
                0.049163
         11
                0.047071
         10
                0.047071
         Name: count, dtype: float64
          > Establecemos el batchsize
         <_PrefetchDataset element_spec=(TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name=None), TensorSpec(</pre>
Out[21]:
         shape=(None, 12), dtype=tf.float32, name=None))>
In [22]: print("Dataset Test (TensorFlow):")
```

```
test tfdataset = get dataset(
              df=df_dataset_test,
              input_size=(2\overline{24},224),
              batch size=32,
              gray_scale=False,
              augmenter=False,
              test set=True,
          test tfdataset
          Dataset Test (TensorFlow):
          Number of instances per label:
                    794
          Unkown
          Name: count, dtype: int64
          Percentaje of instances per label:
          label
          Unkown
                    1.0
          Name: count, dtype: float64
          > Establecemos el batchsize
          <_PrefetchDataset element_spec=(TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name=None), TensorSpec(</pre>
Out[22]:
          shape=(None,), dtype=tf.float64, name=None))>
```

Función para lanzar factoría de modelos from scratch

En este bucle se carga la factoría de modelos from scratch para ir probando cada uno y guardando sus resultados con los callbacks personalizados. Las configuraciones de los modelos vienen desde el archivo "scratch_configs/scratch_experiments.json". Estas configuraciones se pasan a la factoría y se crean los modelos.

```
In [23]: # model factory
         from tensorflow.keras import models, layers
         class ConvBlockConfig:
             def __init__(self, filters, kernel_size, use_maxpooling=True, use_batchnorm=True, dropout_value=0, name=Non
    """Configuracion de cada bloque con los parametros que vienen del json
                 Aras:
                      filters ( type ): Numero de filtros del bloque (una capa convolucional por bloque)
                      kernel size ( type ): Tamaño del kernel de la capa convolucional
                      use_maxpooling (bool, optional): Usar regularizacion MaxPooling. Defaults to True.
                      use_batchnorm (bool, optional): Usar regularizacion BatchNorm. Defaults to True.
                      dropout_value (int, optional): Usar regularizacion Dropout. Defaults to 0.
                     name (_type_, optional): Nombre del bloque. Defaults to None.
                  self.filters = filters
                  self.kernel size = kernel size
                  self.use_maxpooling = use_maxpooling
                  self.use_batchnorm = use_batchnorm
                  self.dropout_value = dropout_value
                  self.name = name
         def create conv block(config, input shape):
               ""Crea los bloques convolucionales con sus regularizaciones (MaxPooling, BatchNorm y Dropout)
                  config ( type ): Configuracion de cada bloque (viene del json de experimentos)
                  input shape (type): Tamaño de entrada del bloque
             Returns:
             _type_: Bloque convolucional
             # Creacion del modelo
             block = models.Sequential(name=config.name)
             # CNN input
             block.add(layers.Conv2D(config.filters, config.kernel size, activation='relu', padding='same', input shape=
             # Regularizacion
             if config.use_batchnorm:
                 block.add(layers.BatchNormalization(name=f'{config.name}_batchnorm'))
             if config.use_maxpooling:
                  block.add(layers.MaxPooling2D((2, 2), name=f'{config.name} maxpooling'))
             if config.dropout value > 0:
                 block.add(layers.Dropout(config.dropout value))
              return block
         def create model with configurations(conv block configs, input shape, num classes):
              """Crea el modelo en base a los bloques que le hemos pasado con la configuracion de los experimentos
             Aras:
                  conv block configs ( type ): Configuracion de los bloques
                  input_shape (_type_): Dimensiones de entrada
```

```
Returns:
              __type_: Modelo
              model = models.Sequential()
              # Incluimos la capa de entrada directamente en el modelo principal
              model.add(layers.InputLayer(input_shape=input_shape, name='input'))
              # Creacion de los bloques para la extraccion de caracteristicas
              for i, config in enumerate(conv_block_configs, start=1):
                  block = create_conv_block(config, input_shape)
                   model.add(block)
                   input shape = (input shape[0]//2, input shape[1]//2, config.filters) # Ajustamos input shape después d
              # Creacion del top model para obtener la clase
              model.add(layers.Flatten(name='top model flatten'))
              model.add(layers.Dense(128, activation='relu', name='top_model_dense_1'))
#model.add(layers.Dropout(0.5, name='top_model_dropout')) # Opcional
              model.add(layers.Dense(num classes, activation='softmax', name='top model output'))
              return model
In [24]: def scratch train(input shape, experiment config:dict):
              # Crear modelo, compilarlo y entrenarlo
              for experiment_name, config in experiment_config.items():
                  print(f"Running Experiment: {experiment_name}")
                   # Configurar el modelo utilizando las configuraciones
                   block configs = []
                   for block_conf in config["model"]["conv_block_configs"]:
    block = ConvBlockConfig(block_conf["filters"],
                                                 block_conf["kernel_size"],
                                                 block_conf["use_maxpooling"],
block_conf["use_batchnorm"],
                                                 block_conf["dropout_value"],
                                                 block conf["name"])
                       block configs.append(block)
                  model = create model with configurations(block configs, input shape, num classes)
                   # Compilar modelo
                   model.compile(
                       loss = tf.keras.losses.categorical crossentropy,
                       optimizer = tf.keras.optimizers.Adam(0.0001),
                       metrics = ['acc'])
                   # Crear callbacks
                   callbacks = get_callbacks(experiment_name)
                   # Mostrar arquitectura
                  print(model.summary())
                   # Entrenar modelo
                   history = model.fit(train tfdataset.repeat(),
                                         validation data=valid tfdataset,
                                         class weight = class weight dict,
                                         epochs = config["training"]["epochs"],
                                         steps_per_epoch=config["training"]["steps_per_epoch"],
                                         callbacks = callbacks,
                                         verbose=1)
```

num classes (type): Numeros de clases para la capa de salida

5.1 Monitorización del proceso de entrenamiento para la toma de decisiones (from scratch)

Lanzar entrenamiento from scratch

Usando la configuración de los modelos de la factoría, se lanzan todos los experimentos. De esta manera se ejecutan uno detrás de otro sin necesidad de estar pendiente de lanzarlos manualmente. La configuración de todos los experimentos se detalla a continuación:

```
"name": "block1"
                  },
                  {
                        "filters": 128,
                        "kernel_size": [3, 3],
                        "use_maxpooling": True,
                        "use_batchnorm": True,
"dropout_value": 0.1,
"name": "block2"
                  },
                        "filters": 256,
                       "kernel_size": [3, 3],
"use_maxpooling": True,
"use_batchnorm": True,
"dropout_value": 0.1,
                        "name": "block3"
           ]
     "steps_per_epoch": 32,
"epochs": 50
     }
},
"model1 nobatchnorm": {
      "model": {
            "conv block configs": [
                 {
                        "filters": 64,
                        "kernel_size": [3, 3],
"use_maxpooling": True,
                        "use_batchnorm": False,
"dropout_value": 0.1,
"name": "block1"
                 },
{
                        "filters": 128,
                        "kernel size": [3, 3],
                        "use_maxpooling": True,
"use_batchnorm": False,
"dropout_value": 0.1,
                        "name": "block2"},
                        "filters": 256,
                       "kernel_size": [3, 3],
"use_maxpooling": True,
"use_batchnorm": False,
                        "dropout_value": 0.1,
                        "name": "block3"
                 }
           ]
      },
"training": {
            "steps_per_epoch": 32,
"epochs": 50
},
"model2": {
      "model": {
            "conv_block_configs": [
                 {
                        "filters": 64,
                        "kernel_size": [5, 5],
"use_maxpooling": True,
                        "use_batchnorm": True,
"dropout_value": 0.1,
"name": "block1"},
                        "filters": 128,
                        "kernel_size": [5, 5],
"use_maxpooling": True,
                        "use_batchnorm": True,
                        "dropout_value": 0.1,
"name": "block2"
                 }
            1
       "training": {
            "steps_per_epoch": 32,
            "epochs": 50
"model2 nobatchnorm": {
```

```
"model": {
              "conv_block_configs": [
                   {
                           "filters": 64,
                           "kernel_size": [5, 5],
"use_maxpooling": True,
                           "use batchnorm": False,
                           "dropout_value": 0.1,
"name": "block1"},
                           "filters": 128,
                           "kernel_size": [5, 5],
"use_maxpooling": True,
                          "use_batchnorm": False,
"dropout_value": 0.1,
"name": "block2"
                    }
             1
      },
"training": {
   "tans per
              "steps_per_epoch": 32,
              "epochs": 50
      }
},
"model3": {
    "model": {
              "conv block configs": [
                   {
                           "filters": 64,
                           "kernel_size": [3, 3],
"use_maxpooling": True,
                           "use_batchnorm": True,
"dropout_value": 0.1,
                           "name": "block1"
                   },
{
                          "filters": 128,
                          "kernel_size": [3, 3],
"use_maxpooling": True,
"use_batchnorm": True,
"dropout_value": 0.1,
"name": "block2"
                   },
{
                           "filters": 256,
                           "kernel_size": [3, 3],
"use_maxpooling": True,
                          "use_batchnorm": True,
"dropout_value": 0.1,
"name": "block3"
                    },
                           "filters": 512,
                           "kernel_size": [3, 3],
                           "use_maxpooling": True,
"use_batchnorm": True,
                           "dropout_value": 0.1,
"name": "block4"
                   }
             ]
       "training": {
             "steps_per_epoch": 32,
"epochs": 50
      }
},
"model3 nobatchnorm": {
       "model": {
             "conv_block_configs": [
                    {
                           "filters": 64,
                           "kernel_size": [3, 3],
"use_maxpooling": True,
                           "use_batchnorm": False,
"dropout_value": 0.1,
                           "name": "block1"
                          "filters": 128,

"kernel_size": [3, 3],

"use_maxpooling": True,

"use_batchnorm": False,

"dropout_value": 0.1,
                           "name": "block2"
                           "filters": 256,
```

```
"kernel size": [3, 3],
                     "use_maxpooling": True,
"use_batchnorm": False,
                     "dropout value": 0.1,
                     "name": "block3"
                     "filters": 512,
                     "kernel_size": [3, 3],
                     "use_maxpooling": True,
                     "use batchnorm": False,
                     "dropout_value": 0.1,
"name": "block4"
           1
       "training": {
            "steps per epoch": 32,
            "epochs": 50
      }
 }
}
```

```
In [26]:
         # Dimensiones de entrada
         input\_shape = (224, 224, 3)
         # Número de clases
         num classes = df dataset train["label"].max() + 1
         # Lanzar entrenamiento
         scratch train(input shape, scratch experiments)
```

Running Experiment: model1 Model: "sequential"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	2048
block2 (Sequential)	(None, 56, 56, 128)	74368
block3 (Sequential)	(None, 28, 28, 256)	296192
<pre>top_model_flatten (Flatten)</pre>	(None, 200704)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	25690240
<pre>top_model_output (Dense)</pre>	(None, 12)	1548
=======================================		========

INFO:tensorflow:Assets written to: models/model1/assets

Total params: 26064396 (99.43 MB) Trainable params: 26063500 (99.42 MB) Non-trainable params: 896 (3.50 KB)

None

```
Epoch 1/50
2023-11-26 18:42:56.542931: E tensorflow/core/grappler/optimizers/meta optimizer.cc:961] layout failed: INVALID
_ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential/block1/dropout/drop
out/SelectV2-2-TransposeNHWCToNCHW-LayoutOptimizer
2023-11-26 18:42:57.095623: I tensorflow/compiler/xla/stream executor/cuda/cuda dnn.cc:442] Loaded cuDNN versio
n 8700
2023-11-26 18:42:57.397182: I tensorflow/tsl/platform/default/subprocess.cc:304] Start cannot spawn child proce
ss: No such file or directory
2023-11-26 18:42:57.927327: I tensorflow/tsl/platform/default/subprocess.cc:304] Start cannot spawn child proce
ss: No such file or directory
2023-11-26 18:42:58.734794: I tensorflow/compiler/xla/service/service.cc:168] XLA service 0x7f8ca803a590 initia
lized for platform CUDA (this does not guarantee that XLA will be used). Devices: 2023-11-26 18:42:58.734834: I tensorflow/compiler/xla/service/service.cc:176] StreamExecutor device (0): NVID
IA GeForce RTX 3080, Compute Capability 8.6
2023-11-26 18:42:58.739453: I tensorflow/compiler/mlir/tensorflow/utils/dump_mlir_util.cc:269] disabling MLIR c
rash reproducer, set env var `MLIR_CRASH_REPRODUCER_DIRECTORY` to enable. 2023-11-26 18:42:58.866988: I ./tensorflow/compiler/jit/device_compiler.h:186] Compiled cluster using XLA! Thi
s line is logged at most once for the lifetime of the process.
 6/32 [====>.....] - ETA: 2s - loss: 8.9099 - acc: 0.1146WARNING:tensorflow:Callback method
`on train batch end` is slow compared to the batch time (batch time: 0.0357s vs `on train batch end` time: 0.05
94s). Check your callbacks.
32/32 [=====
                        ========] - ETA: Os - loss: 4.8960 - acc: 0.1680
Epoch 1: val loss improved from inf to 23.67860, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
```

```
acc: 0.1287
Epoch 2/50
Epoch 2: val loss did not improve from 23.67860
32/32 [===========] - 9s 291ms/step - loss: 2.2053 - acc: 0.2080 - val loss: 40.6756 - val a
cc: 0.1287
Epoch 3/50
Epoch 3: val loss did not improve from 23.67860
32/32 [========] - 10s 321ms/step - loss: 2.1161 - acc: 0.2500 - val loss: 50.5102 - val
acc: 0.1287
Epoch 4/50
Epoch 4: val loss did not improve from 23.67860
acc: 0.1287
Epoch 5/50
Epoch 5: val loss did not improve from 23.67860
32/32 [============] - 11s 349ms/step - loss: 1.9432 - acc: 0.3125 - val loss: 57.2173 - val
acc: 0.1287
Epoch 6/50
32/32 [====
        Epoch 6: val loss did not improve from 23.67860
cc: 0.1287
Epoch 7/50
Epoch 7: val loss did not improve from 23.67860
32/32 [=========] - 10s 323ms/step - loss: 1.8529 - acc: 0.3701 - val loss: 51.2696 - val
acc: 0.1287
Epoch 8/50
Epoch 8: val loss did not improve from 23.67860
32/32 [=========] - 10s 304ms/step - loss: 1.8377 - acc: 0.3554 - val loss: 41.3990 - val
acc: 0.1287
Epoch 9/50
Epoch 9: val loss did not improve from 23.67860
cc: 0.1287
Epoch 10/50
Epoch 10: val loss improved from 23.67860 to 20.57506, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [=========] - 12s 388ms/step - loss: 1.6856 - acc: 0.4561 - val_loss: 20.5751 - val_
acc: 0.1308
Epoch 11/50
32/32 [========] - ETA: 0s - loss: 1.7020 - acc: 0.4141
Epoch 11: val loss improved from 20.57506 to 11.84752, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [=========] - 12s 371ms/step - loss: 1.7020 - acc: 0.4141 - val loss: 11.8475 - val
acc: 0.1444
Epoch 12/50
32/32 [============] - ETA: 0s - loss: 1.4994 - acc: 0.4762
Epoch 12: val loss improved from 11.84752 to 3.79100, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [============] - 13s 397ms/step - loss: 1.4994 - acc: 0.4762 - val loss: 3.7910 - val a
cc: 0.3159
Epoch 13/50
Epoch 13: val loss improved from 3.79100 to 3.02257, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [====
          cc: 0.3096
Epoch 14/50
Epoch 14: val loss improved from 3.02257 to 2.89093, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [====
          cc: 0.2542
Epoch 15/50
Epoch 15: val loss did not improve from 2.89093
32/32 [=======] - 10s 303ms/step - loss: 1.4071 - acc: 0.5089 - val loss: 3.4234 - val a
cc: 0.2113
Epoch 16/50
32/32 [=====
        Epoch 16: val loss improved from 2.89093 to 2.12441, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
```

```
INFO:tensorflow:Assets written to: models/model1/assets
cc: 0.3023
Epoch 17/50
Epoch 17: val loss improved from 2.12441 to 1.81835, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [========] - 12s 382ms/step - loss: 1.4387 - acc: 0.5059 - val_loss: 1.8183 - val a
cc: 0.3421
Epoch 18/50
                   ======] - ETA: 0s - loss: 1.4313 - acc: 0.5225
32/32 [==
Epoch 18: val loss improved from 1.81835 to 1.68183, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [=========] - 11s 359ms/step - loss: 1.4313 - acc: 0.5225 - val_loss: 1.6818 - val a
cc: 0.3860
Epoch 19/50
                 =======] - ETA: 0s - loss: 1.4047 - acc: 0.5099
32/32 [===
Epoch 19: val loss did not improve from 1.68183
cc: 0.4205
Epoch 20/50
Epoch 20: val loss improved from 1.68183 to 1.62373, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [==========] - 11s 358ms/step - loss: 1.2930 - acc: 0.5566 - val loss: 1.6237 - val a
cc: 0.4623
Epoch 21/50
32/32 [============== ] - ETA: 0s - loss: 1.3711 - acc: 0.5273
Epoch 21: val loss improved from 1.62373 to 1.46950, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
cc: 0.5126
Epoch 22/50
Epoch 22: val loss did not improve from 1.46950
32/32 [====
        c: 0.4027
Epoch 23/50
          32/32 [=====
Epoch 23: val_loss did not improve from 1.46950
cc: 0.4833
Epoch 24/50
32/32 [==============] - ETA: 0s - loss: 1.2066 - acc: 0.5908
Epoch 24: val loss improved from 1.46950 to 1.44475, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [===========] - 12s 381ms/step - loss: 1.2066 - acc: 0.5908 - val_loss: 1.4447 - val a
cc: 0.5356
Epoch 25/50
32/32 [==============] - ETA: 0s - loss: 1.1184 - acc: 0.6162
Epoch 25: val loss improved from 1.44475 to 1.30158, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
cc: 0.5628
Epoch 26/50
Epoch 26: val_loss did not improve from 1.30158
               =======] - 8s 266ms/step - loss: 1.1335 - acc: 0.6104 - val loss: 1.3503 - val ac
32/32 [===
c: 0.5764
Epoch 27/50
32/32 [=====
          Epoch 27: val_loss did not improve from 1.30158
cc: 0.5523
Epoch 28/50
Epoch 28: val loss improved from 1.30158 to 1.16862, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
```

```
32/32 [==========] - 12s 378ms/step - loss: 1.0763 - acc: 0.6299 - val loss: 1.1686 - val a
cc: 0.6077
Epoch 29/50
Epoch 29: val loss did not improve from 1.16862
32/32 [============] - 10s 323ms/step - loss: 0.9967 - acc: 0.6396 - val loss: 1.3106 - val a
cc: 0.5994
Epoch 30/50
         32/32 [=====
Epoch 30: val_loss did not improve from 1.16862
32/32 [=========] - 9s 294ms/step - loss: 0.9920 - acc: 0.6604 - val loss: 1.1948 - val ac
c: 0.6297
Epoch 31/50
                 =======] - ETA: 0s - loss: 0.9455 - acc: 0.6621
32/32 [===
Epoch 31: val loss did not improve from 1.16862
32/32 [==========] - 10s 309ms/step - loss: 0.9455 - acc: 0.6621 - val loss: 1.1806 - val a
cc: 0.6077
Epoch 32/50
Epoch 32: val loss improved from 1.16862 to 1.10771, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [==========] - 12s 369ms/step - loss: 0.8997 - acc: 0.6885 - val loss: 1.1077 - val a
cc: 0.6391
Epoch 33/50
Epoch 33: val loss did not improve from 1.10771
32/32 [==
               ========] - 9s 286ms/step - loss: 0.9859 - acc: 0.6797 - val loss: 1.2130 - val ac
c: 0.6130
Epoch 34/50
Epoch 34: val loss improved from 1.10771 to 1.06737, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [===
              :========] - 12s 394ms/step - loss: 0.8005 - acc: 0.7228 - val loss: 1.0674 - val a
cc: 0.6642
Epoch 35/50
Epoch 35: val_loss improved from 1.06737 to 0.98820, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
cc: 0.6987
Epoch 36/50
Epoch 36: val loss did not improve from 0.98820
cc: 0.5523
Epoch 37/50
Epoch 37: val loss did not improve from 0.98820
32/32 [============] - 8s 261ms/step - loss: 0.9203 - acc: 0.6826 - val loss: 1.1190 - val ac
c: 0.6569
Epoch 38/50
Epoch 38: val_loss did not improve from 0.98820
cc: 0.6506
Epoch 39/50
Epoch 39: val loss did not improve from 0.98820
cc: 0.6538
Epoch 40/50
Epoch 40: val loss did not improve from 0.98820
32/32 [=============] - 10s 311ms/step - loss: 0.7568 - acc: 0.7197 - val loss: 1.1238 - val a
cc: 0.6360
Epoch 41/50
32/32 [===
               ======== ] - ETA: Os - loss: 0.7979 - acc: 0.7396
Epoch 41: val loss improved from 0.98820 to 0.97991, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
```

```
32/32 [============= ] - 10s 317ms/step - loss: 0.7979 - acc: 0.7396 - val loss: 0.9799 - val a
cc: 0.6946
Epoch 42/50
Epoch 42: val loss did not improve from 0.97991
32/32 [=============] - 10s 301ms/step - loss: 0.6619 - acc: 0.7773 - val loss: 1.0448 - val a
cc: 0.6789
Epoch 43/50
          32/32 [===
Epoch 43: val loss did not improve from 0.97991
cc: 0.6778
Epoch 44/50
Epoch 44: val_loss did not improve from 0.97991
c: 0.6778
Epoch 45/50
Epoch 45: val loss improved from 0.97991 to 0.92499, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [=============] - 10s 324ms/step - loss: 0.6078 - acc: 0.8020 - val loss: 0.9250 - val a
cc: 0.7312
Epoch 46/50
Epoch 46: val loss did not improve from 0.92499
cc: 0.6433
Epoch 47/50
32/32 [====
           Epoch 47: val loss did not improve from 0.92499
32/32 [=========] - 10s 302ms/step - loss: 0.6333 - acc: 0.8027 - val loss: 0.9724 - val a
cc: 0.7123
Epoch 48/50
32/32 [============= ] - ETA: 0s - loss: 0.5638 - acc: 0.8047
Epoch 48: val loss did not improve from 0.92499
               32/32 [==
c: 0.7134
Epoch 49/50
32/32 [=============] - ETA: 0s - loss: 0.5466 - acc: 0.8218
Epoch 49: val loss improved from 0.92499 to 0.89157, saving model to models/model1
INFO:tensorflow:Assets written to: models/model1/assets
INFO:tensorflow:Assets written to: models/model1/assets
32/32 [===
                   =======] - 11s 355ms/step - loss: 0.5466 - acc: 0.8218 - val loss: 0.8916 - val a
cc: 0.7103
Epoch 50/50
Epoch 50: val_loss did not improve from 0.89157
c: 0.7186
Running Experiment: model1_nobatchnorm
Model: "sequential_1"
Layer (type)
                    Output Shape
                                       Param #
block1 (Sequential)
                    (None, 112, 112, 64)
                                       1792
                                       73856
block2 (Seguential)
                    (None. 56. 56. 128)
block3 (Sequential)
                    (None, 28, 28, 256)
                                       295168
top_model_flatten (Flatten (None, 200704)
top model dense 1 (Dense)
                                       25690240
                    (None, 128)
top model output (Dense)
                    (None, 12)
                                       1548
______
Total params: 26062604 (99.42 MB)
Trainable params: 26062604 (99.42 MB)
Non-trainable params: 0 (0.00 Byte)
None
Epoch 1/50
2023-11-26 18:51:47.334375: E tensorflow/core/grappler/optimizers/meta_optimizer.cc:961] layout failed: INVALID
ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential 1/block1/dropout 3/
dropout/SelectV2-2-TransposeNHWCToNCHW-LayoutOptimizer
6/32 [====>.....] - ETA: 2s - loss: 2.6361 - acc: 0.0521WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0357s vs `on_train_batch_end` time: 0.05
14s). Check your callbacks.
WARNING:tensorflow:Callback method `on train batch end` is slow compared to the batch time (batch time: 0.0357s
```

vs `on_train_batch_end` time: 0.0514s). Check your callbacks.

INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets

Epoch 1: val loss improved from inf to 2.47914, saving model to models/model1 nobatchnorm

```
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
32/32 [=========] - 12s 344ms/step - loss: 2.5038 - acc: 0.0703 - val loss: 2.4791 - val a
cc: 0.1255
Epoch 2/50
32/32 [=====
          Epoch 2: val loss improved from 2.47914 to 2.46177, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
                 cc: 0.0910
Epoch 3/50
32/32 [============] - ETA: 0s - loss: 2.3208 - acc: 0.1533
Epoch 3: val loss improved from 2.46177 to 2.26747, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
                cc: 0.2427
Fnoch 4/50
32/32 [=========] - ETA: 0s - loss: 2.1349 - acc: 0.2703
Epoch 4: val loss improved from 2.26747 to 2.04743, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
cc: 0.3410
Epoch 5/50
32/32 [===========] - ETA: 0s - loss: 1.8888 - acc: 0.3135
Epoch 5: val loss improved from 2.04743 to 1.83637, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
32/32 [==========] - 11s 338ms/step - loss: 1.8888 - acc: 0.3135 - val_loss: 1.8364 - val_a
cc: 0.3902
Epoch 6/50
32/32 [====
           Epoch 6: val loss improved from 1.83637 to 1.75534, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
32/32 [=========] - 11s 363ms/step - loss: 1.8411 - acc: 0.3652 - val loss: 1.7553 - val a
cc: 0.4006
Epoch 7/50
32/32 [=====
          Epoch 7: val loss improved from 1.75534 to 1.56954, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
32/32 [=========] - 10s 326ms/step - loss: 1.6808 - acc: 0.3809 - val loss: 1.5695 - val a
cc: 0.4728
Epoch 8/50
32/32 [============== ] - ETA: 0s - loss: 1.5765 - acc: 0.4653
Epoch 8: val loss improved from 1.56954 to 1.49639, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
32/32 [=============] - 12s 373ms/step - loss: 1.5765 - acc: 0.4653 - val loss: 1.4964 - val a
cc: 0.5031
Epoch 9/50
Epoch 9: val loss improved from 1.49639 to 1.43752, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
32/32 [=========================== ] - 12s 366ms/step - loss: 1.4623 - acc: 0.4902 - val loss: 1.4375 - val a
cc: 0.5209
Epoch 10/50
Epoch 10: val loss did not improve from 1.43752
c: 0.4833
Epoch 11/50
32/32 [====
                   ========] - ETA: Os - loss: 1.3917 - acc: 0.5186
Epoch 11: val loss did not improve from 1.43752
c: 0.4906
Epoch 12/50
Epoch 12: val loss improved from 1.43752 to 1.25642, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
```

```
32/32 [============] - 11s 364ms/step - loss: 1.3857 - acc: 0.5139 - val loss: 1.2564 - val a
cc: 0.5764
Epoch 13/50
Epoch 13: val loss did not improve from 1.25642
c: 0.5377
Epoch 14/50
Epoch 14: val loss did not improve from 1.25642
cc: 0.5481
Epoch 15/50
Epoch 15: val loss improved from 1.25642 to 1.18652, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
32/32 [=========] - 12s 387ms/step - loss: 1.2128 - acc: 0.5663 - val loss: 1.1865 - val a
cc: 0.6151
Epoch 16/50
32/32 [=====
          Epoch 16: val loss improved from 1.18652 to 1.16294, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
32/32 [==========] - 11s 360ms/step - loss: 1.1516 - acc: 0.6035 - val_loss: 1.1629 - val_a
cc: 0.6098
Epoch 17/50
          32/32 [=====
Epoch 17: val loss did not improve from 1.16294
32/32 [============] - 10s 300ms/step - loss: 1.1295 - acc: 0.6201 - val loss: 1.1732 - val a
cc: 0.6088
Epoch 18/50
Epoch 18: val_loss did not improve from 1.16294
                :========] - 10s 308ms/step - loss: 1.1216 - acc: 0.6123 - val loss: 1.2293 - val a
32/32 [==
cc: 0.5889
Epoch 19/50
32/32 [=============] - ETA: 0s - loss: 1.0590 - acc: 0.6248
Epoch 19: val loss improved from 1.16294 to 1.03423, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
32/32 [===
                  cc: 0.6517
Epoch 20/50
Epoch 20: val_loss did not improve from 1.03423
32/32 [============] - 11s 348ms/step - loss: 1.0669 - acc: 0.6338 - val loss: 1.0654 - val a
cc: 0.6308
Epoch 21/50
Epoch 21: val loss did not improve from 1.03423
32/32 [==========] - 10s 316ms/step - loss: 1.0440 - acc: 0.6279 - val loss: 1.0716 - val a
cc: 0.6423
Epoch 22/50
         32/32 [===
Epoch 22: val loss improved from 1.03423 to 0.98048, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
32/32 [====
        cc: 0.6778
Epoch 23/50
32/32 [=====
         Epoch 23: val loss improved from 0.98048 to 0.96465, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
```

```
32/32 [=============] - 13s 405ms/step - loss: 0.9751 - acc: 0.6584 - val loss: 0.9646 - val a
cc: 0.6820
Epoch 24/50
Epoch 24: val loss did not improve from 0.96465
32/32 [=======] - 11s 347ms/step - loss: 0.9801 - acc: 0.6543 - val loss: 0.9801 - val a
cc: 0.6799
Epoch 25/50
32/32 [=============] - ETA: 0s - loss: 0.9979 - acc: 0.6484
Epoch 25: val loss did not improve from 0.96465
32/32 [=============] - 10s 311ms/step - loss: 0.9979 - acc: 0.6484 - val loss: 1.0371 - val a
cc: 0.6684
Epoch 26/50
Epoch 26: val loss did not improve from 0.96465
         32/32 [====
cc: 0.6391
Epoch 27/50
32/32 [============] - ETA: 0s - loss: 0.9230 - acc: 0.6772
Epoch 27: val loss did not improve from 0.96465
cc: 0.6370
Epoch 28/50
Epoch 28: val loss improved from 0.96465 to 0.93269, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
32/32 [=========] - 13s 408ms/step - loss: 0.8002 - acc: 0.7197 - val loss: 0.9327 - val a
cc: 0.6893
Epoch 29/50
Epoch 29: val loss improved from 0.93269 to 0.91702, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
cc: 0.6841
Epoch 30/50
Epoch 30: val loss did not improve from 0.91702
c: 0.6935
Epoch 31/50
                 =======] - ETA: Os - loss: 0.8090 - acc: 0.7158
32/32 [==
Epoch 31: val loss improved from 0.91702 to 0.89510, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
32/32 [==========] - 12s 371ms/step - loss: 0.8090 - acc: 0.7158 - val loss: 0.8951 - val a
cc: 0.6893
Epoch 32/50
32/32 [============] - ETA: 0s - loss: 0.8965 - acc: 0.6699
Epoch 32: val_loss did not improve from 0.89510
c: 0.6946
Epoch 33/50
Epoch 33: val loss did not improve from 0.89510
32/32 [===
        c: 0.6559
Epoch 34/50
32/32 [=====
         Epoch 34: val loss did not improve from 0.89510
c: 0.6757
Epoch 35/50
Epoch 35: val loss improved from 0.89510 to 0.89199, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
32/32 [===========] - 12s 375ms/step - loss: 0.7849 - acc: 0.7090 - val loss: 0.8920 - val a
cc: 0.6977
Epoch 36/50
Epoch 36: val loss improved from 0.89199 to 0.86073, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
32/32 [===========] - 12s 375ms/step - loss: 0.7142 - acc: 0.7617 - val loss: 0.8607 - val a
cc: 0.6726
Epoch 37/50
Epoch 37: val loss improved from 0.86073 to 0.83864, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
```

INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets

```
32/32 [============] - 11s 342ms/step - loss: 0.7893 - acc: 0.7275 - val loss: 0.8386 - val a
cc: 0.6977
Epoch 38/50
Epoch 38: val loss did not improve from 0.83864
cc: 0.6956
Epoch 39/50
32/32 [===========] - ETA: 0s - loss: 0.7393 - acc: 0.7295
Epoch 39: val_loss did not improve from 0.83864
cc: 0.7165
Epoch 40/50
Epoch 40: val loss improved from 0.83864 to 0.79920, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
     32/32 [====
cc: 0.7280
Epoch 41/50
Epoch 41: val loss did not improve from 0.79920
c: 0.7144
Epoch 42/50
32/32 [===
          Epoch 42: val loss did not improve from 0.79920
c: 0.7301
Epoch 43/50
Epoch 43: val loss improved from 0.79920 to 0.75420, saving model to models/model1 nobatchnorm
INFO:tensorflow:Assets written to: models/model1 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model1_nobatchnorm/assets
```

```
32/32 [===========] - 12s 389ms/step - loss: 0.6849 - acc: 0.7832 - val loss: 0.7542 - val a
cc: 0.7395
Epoch 44/50
Epoch 44: val loss did not improve from 0.75420
cc: 0.7301
Epoch 45/50
        32/32 [==
Epoch 45: val loss did not improve from 0.75420
c: 0.7249
Epoch 46/50
Epoch 46: val loss did not improve from 0.75420
                    32/32 [==
c: 0.7479
Epoch 47/50
        32/32 [=====
Epoch 47: val loss did not improve from 0.75420
c: 0.7186
Fnoch 48/50
Epoch 48: val_loss did not improve from 0.75420
32/32 [===
      c: 0.7218
Epoch 49/50
32/32 [===
              ========] - ETA: 0s - loss: 0.6076 - acc: 0.8020
Epoch 49: val loss did not improve from 0.75420
c: 0.7228
Epoch 50/50
Epoch 50: val loss did not improve from 0.75420
32/32 [===
        cc: 0.7019
Running Experiment: model2
Model: "sequential_2'
                               Param #
Layer (type)
                Output Shape
block1 (Sequential)
                (None, 112, 112, 64)
                               5120
block2 (Sequential)
                (None, 56, 56, 128)
                               205440
top model flatten (Flatten (None, 401408)
                               51380352
top model dense 1 (Dense)
                (None, 128)
top_model_output (Dense)
                (None, 12)
                               1548
Total params: 51592460 (196.81 MB)
Trainable params: 51592076 (196.81 MB)
Non-trainable params: 384 (1.50 KB)
None
Epoch 1/50
2023-11-26 19:00:28.375089: E tensorflow/core/grappler/optimizers/meta optimizer.cc:961] layout failed: INVALID
ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential_2/block1/dropout_6/
dropout/SelectV2-2-TransposeNHWCToNCHW-LayoutOptimizer
                .....] - ETA: 2s - loss: 12.5007 - acc: 0.1198WARNING:tensorflow:Callback metho
d 'on train batch end' is slow compared to the batch time (batch time: 0.0383s vs 'on train batch end' time: 0.
0636s). Check your callbacks.
WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0383s
vs `on_train_batch_end` time: 0.0636s). Check your callbacks.
Epoch 1: val_loss improved from inf to 5.57673, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
             :========] - 17s 372ms/step - loss: 7.1232 - acc: 0.1562 - val_loss: 5.5767 - val_a
32/32 [===
cc: 0.0533
Epoch 2/50
Epoch 2: val_loss did not improve from 5.57673
acc: 0.0492
Epoch 3/50
       Epoch 3: val_loss did not improve from 5.57673
cc: 0.0492
```

Epoch 4: val_loss did not improve from 5.57673

```
acc: 0.0492
Epoch 5/50
Epoch 5: val loss did not improve from 5.57673
acc: 0.0492
Epoch 6/50
Epoch 6: val loss did not improve from 5.57673
32/32 [==========] - 10s 320ms/step - loss: 2.1541 - acc: 0.2598 - val loss: 71.6949 - val
acc: 0.0492
Epoch 7/50
Epoch 7: val loss did not improve from 5.57673
cc: 0.0492
Epoch 8/50
Epoch 8: val_loss did not improve from 5.57673
acc: 0.0492
Epoch 9/50
Epoch 9: val_loss did not improve from 5.57673
cc: 0.0492
Epoch 10/50
32/32 [=====
       Epoch 10: val loss did not improve from 5.57673
32/32 [=========] - 9s 293ms/step - loss: 1.8700 - acc: 0.3672 - val loss: 74.8372 - val a
cc: 0.0492
Epoch 11/50
Epoch 11: val loss did not improve from 5.57673
       32/32 [==
cc: 0.0492
Epoch 12/50
Epoch 12: val loss did not improve from 5.57673
32/32 [=========] - 10s 313ms/step - loss: 1.9599 - acc: 0.3663 - val_loss: 60.8676 - val_
acc: 0.0523
Epoch 13/50
32/32 [============ ] - ETA: 0s - loss: 1.9201 - acc: 0.3506
Epoch 13: val loss did not improve from 5.57673
32/32 [============= ] - 9s 300ms/step - loss: 1.9201 - acc: 0.3506 - val loss: 56.8206 - val a
cc: 0.0565
Epoch 14/50
Epoch 14: val loss did not improve from 5.57673
32/32 [========] - 9s 279ms/step - loss: 1.8320 - acc: 0.3945 - val loss: 46.8248 - val a
cc: 0.0994
Epoch 15/50
Epoch 15: val loss did not improve from 5.57673
cc: 0.1820
Epoch 16/50
Epoch 16: val loss did not improve from 5.57673
32/32 [==========] - 9s 298ms/step - loss: 1.7437 - acc: 0.4180 - val loss: 14.7768 - val a
cc: 0.2626
Epoch 17/50
Epoch 17: val_loss did not improve from 5.57673
     32/32 [====
c: 0.3201
Fnoch 18/50
Epoch 18: val_loss did not improve from 5.57673
acc: 0.2636
Epoch 19/50
            =======] - ETA: Os - loss: 1.7201 - acc: 0.4178
32/32 [===
Epoch 19: val loss improved from 5.57673 to 4.40921, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
cc: 0.3619
Epoch 20/50
          32/32 [====
Epoch 20: val loss improved from 4.40921 to 3.08332, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
```

INFO:tensorflow:Assets written to: models/model2/assets

```
32/32 [==========] - 12s 390ms/step - loss: 1.6566 - acc: 0.4639 - val loss: 3.0833 - val a
cc: 0.3180
Epoch 21/50
Epoch 21: val loss improved from 3.08332 to 2.98711, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
            :========] - 13s 402ms/step - loss: 1.7541 - acc: 0.4219 - val loss: 2.9871 - val a
cc: 0.4048
Epoch 22/50
Epoch 22: val loss improved from 2.98711 to 2.98685, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
cc: 0.3985
Epoch 23/50
Epoch 23: val_loss improved from 2.98685 to 1.91801, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
cc: 0.4477
Fnoch 24/50
Epoch 24: val_loss improved from 1.91801 to 1.71955, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
cc: 0.4372
Epoch 25/50
Epoch 25: val loss did not improve from 1.71955
32/32 [=======] - 10s 310ms/step - loss: 1.6197 - acc: 0.4512 - val loss: 1.8349 - val a
cc: 0.4508
Epoch 26/50
Epoch 26: val_loss did not improve from 1.71955
c: 0.4644
Epoch 27/50
Epoch 27: val loss did not improve from 1.71955
cc: 0.4435
Epoch 28/50
Epoch 28: val loss improved from 1.71955 to 1.61978, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
cc: 0.4843
Epoch 29/50
       32/32 [=====
Epoch 29: val loss did not improve from 1.61978
cc: 0.4446
Epoch 30/50
Epoch 30: val loss did not improve from 1.61978
c: 0.3400
Epoch 31/50
32/32 [===
       Epoch 31: val loss did not improve from 1.61978
32/32 [============] - 11s 335ms/step - loss: 1.4866 - acc: 0.5205 - val loss: 2.3023 - val a
cc: 0.4435
Epoch 32/50
Epoch 32: val loss did not improve from 1.61978
32/32 [===
       c: 0.5042
Epoch 33/50
32/32 [============] - ETA: 0s - loss: 1.4685 - acc: 0.5098
Epoch 33: val loss did not improve from 1.61978
32/32 [==========] - 9s 276ms/step - loss: 1.4685 - acc: 0.5098 - val loss: 1.9977 - val ac
c: 0.4644
Epoch 34/50
Epoch 34: val loss did not improve from 1.61978
c: 0.4812
Epoch 35/50
       32/32 [===
Epoch 35: val loss did not improve from 1.61978
32/32 [=========] - 10s 310ms/step - loss: 1.4386 - acc: 0.5498 - val loss: 4.5071 - val a
```

cc: 0.3211

```
Epoch 36/50
         32/32 [==
Epoch 36: val loss did not improve from 1.61978
32/32 [==========] - 9s 294ms/step - loss: 1.3399 - acc: 0.5791 - val loss: 5.0827 - val ac
c: 0.3588
Epoch 37/50
Epoch 37: val loss did not improve from 1.61978
32/32 [===
        c: 0.4195
Epoch 38/50
Epoch 38: val_loss did not improve from 1.61978
c: 0.5324
Epoch 39/50
Epoch 39: val loss did not improve from 1.61978
32/32 [=============] - 10s 308ms/step - loss: 1.2683 - acc: 0.5791 - val loss: 2.0091 - val a
cc: 0.5031
Epoch 40/50
                =======] - ETA: Os - loss: 1.3099 - acc: 0.6016
32/32 [=====
Epoch 40: val loss did not improve from 1.61978
32/32 [===========] - 9s 296ms/step - loss: 1.3099 - acc: 0.6016 - val loss: 1.7055 - val ac
c: 0.5251
Epoch 41/50
Epoch 41: val loss did not improve from 1.61978
32/32 [==
              =========] - 8s 264ms/step - loss: 1.2564 - acc: 0.5921 - val loss: 2.1295 - val ac
c: 0.4833
Epoch 42/50
Epoch 42: val loss did not improve from 1.61978
32/32 [==========] - 10s 325ms/step - loss: 1.2293 - acc: 0.6152 - val_loss: 1.8648 - val_a
cc: 0.5031
Epoch 43/50
Epoch 43: val loss did not improve from 1.61978
c: 0.5042
Epoch 44/50
Epoch 44: val loss did not improve from 1.61978
c: 0.4404
Epoch 45/50
                =======] - ETA: 0s - loss: 1.2669 - acc: 0.5772
32/32 [==
Epoch 45: val_loss did not improve from 1.61978
c: 0.5010
Epoch 46/50
Epoch 46: val loss did not improve from 1.61978
32/32 [=========] - 10s 312ms/step - loss: 1.2712 - acc: 0.5977 - val loss: 1.6315 - val a
cc: 0.5272
Epoch 47/50
32/32 [=============] - ETA: 0s - loss: 1.2668 - acc: 0.5977
Epoch 47: val loss improved from 1.61978 to 1.53502, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
              :========] - 12s 391ms/step - loss: 1.2668 - acc: 0.5977 - val loss: 1.5350 - val a
32/32 [===
cc: 0.5826
Epoch 48/50
32/32 [===
              ======== ] - ETA: Os - loss: 1.2746 - acc: 0.5928
Epoch 48: val loss improved from 1.53502 to 1.41117, saving model to models/model2
INFO:tensorflow:Assets written to: models/model2/assets
INFO:tensorflow:Assets written to: models/model2/assets
```

```
32/32 [============] - 11s 361ms/step - loss: 1.2746 - acc: 0.5928 - val loss: 1.4112 - val a
cc: 0.5513
Epoch 49/50
Epoch 49: val loss did not improve from 1.41117
cc: 0.5701
Epoch 50/50
           32/32 [=====
Epoch 50: val loss did not improve from 1.41117
c: 0.5596
{\tt Running\ Experiment:\ model2\_nobatchnorm}
Model: "sequential_3"
Layer (type)
                       Output Shape
                                            Param #
                       -----
block1 (Sequential)
                       (None, 112, 112, 64)
block2 (Sequential)
                       (None, 56, 56, 128)
                                            204928
top_model_flatten (Flatten (None, 401408)
top_model_dense_1 (Dense)
                      (None, 128)
                                            51380352
top model output (Dense)
                       (None, 12)
                                            1548
Total params: 51591692 (196.81 MB)
Trainable params: 51591692 (196.81 MB)
Non-trainable params: 0 (0.00 Byte)
None
Epoch 1/50
2023-11-26 19:08:45.971982: E tensorflow/core/grappler/optimizers/meta_optimizer.cc:961] layout failed: INVALID
ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential 3/block1/dropout 8/
dropout/SelectV2-2-TransposeNHWCToNCHW-LayoutOptimizer
6/32 [====>......] - ETA: 3s - loss: 2.8470 - acc: 0.0469WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0443s vs `on_train_batch_end` time: 0.05
98s). Check your callbacks.
WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0443s
vs `on_train_batch_end` time: 0.0598s). Check your callbacks.
Epoch 1: val_loss improved from inf to 2.47979, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.1046
Epoch 2/50
32/32 [============== ] - ETA: 0s - loss: 2.5112 - acc: 0.0986
Epoch 2: val loss improved from 2.47979 to 2.44516, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.1067
Epoch 3/50
32/32 [====
                       ======] - ETA: Os - loss: 2.3284 - acc: 0.1396
Epoch 3: val loss improved from 2.44516 to 2.35855, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
32/32 [==========] - 12s 386ms/step - loss: 2.3284 - acc: 0.1396 - val_loss: 2.3586 - val_a
cc: 0.1433
Epoch 4/50
32/32 [===
                      :=======] - ETA: 0s - loss: 2.3528 - acc: 0.1238
Epoch 4: val loss improved from 2.35855 to 2.27866, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
32/32 [======
            cc: 0.1412
Epoch 5/50
32/32 [====
                   =========] - ETA: Os - loss: 2.2192 - acc: 0.1729
Epoch 5: val loss improved from 2.27866 to 2.11659, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
                         =====] - 12s 383ms/step - loss: 2.2192 - acc: 0.1729 - val loss: 2.1166 - val a
32/32 [==
cc: 0.2563
Epoch 6/50
32/32 [=====
          Epoch 6: val loss improved from 2.11659 to 1.92883, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
```

```
32/32 [===========] - 12s 376ms/step - loss: 1.9854 - acc: 0.2686 - val loss: 1.9288 - val a
cc: 0.3358
Epoch 7/50
Epoch 7: val loss improved from 1.92883 to 1.87172, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
32/32 [=========] - 11s 359ms/step - loss: 1.9476 - acc: 0.3193 - val loss: 1.8717 - val a
cc: 0.3609
Epoch 8/50
32/32 [=====
        Epoch 8: val loss improved from 1.87172 to 1.72532, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
          32/32 [====
cc: 0.3724
Epoch 9/50
32/32 [=====
        Epoch 9: val_loss improved from 1.72532 to 1.65652, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
32/32 [===
               =========] - 11s 357ms/step - loss: 1.7319 - acc: 0.3965 - val loss: 1.6565 - val a
cc: 0.3912
Epoch 10/50
Epoch 10: val_loss improved from 1.65652 to 1.58150, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
cc: 0.4906
Fnoch 11/50
Epoch 11: val_loss improved from 1.58150 to 1.50213, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.4791
Fnoch 12/50
Epoch 12: val loss improved from 1.50213 to 1.45472, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
cc: 0.5481
Epoch 13/50
Epoch 13: val loss improved from 1.45472 to 1.38676, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.5324
Epoch 14/50
Epoch 14: val loss improved from 1.38676 to 1.38558, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
32/32 [===========] - 12s 385ms/step - loss: 1.3642 - acc: 0.5117 - val loss: 1.3856 - val a
cc: 0.5523
Epoch 15/50
Epoch 15: val loss improved from 1.38558 to 1.33885, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.5649
Epoch 16/50
Epoch 16: val loss improved from 1.33885 to 1.27741, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.5701
Epoch 17/50
32/32 [========] - ETA: 0s - loss: 1.3034 - acc: 0.5684
Epoch 17: val loss improved from 1.27741 to 1.19331, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
```

INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets

```
32/32 [==========] - 12s 393ms/step - loss: 1.3034 - acc: 0.5684 - val loss: 1.1933 - val a
cc: 0.6140
Epoch 18/50
Epoch 18: val loss did not improve from 1.19331
cc: 0.5785
Epoch 19/50
Epoch 19: val loss did not improve from 1.19331
c: 0.5774
Epoch 20/50
Epoch 20: val_loss improved from 1.19331 to 1.14081, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
32/32 [==========] - 11s 362ms/step - loss: 1.2346 - acc: 0.5918 - val loss: 1.1408 - val a
cc: 0.6339
Epoch 21/50
Epoch 21: val loss improved from 1.14081 to 1.13441, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.6245
Epoch 22/50
Epoch 22: val loss did not improve from 1.13441
        32/32 [==
c: 0.5868
Epoch 23/50
32/32 [=====
        Epoch 23: val loss did not improve from 1.13441
cc: 0.6046
Epoch 24/50
32/32 [========] - ETA: 0s - loss: 1.0908 - acc: 0.6348
Epoch 24: val loss improved from 1.13441 to 1.11488, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.6172
Epoch 25/50
Epoch 25: val loss did not improve from 1.11488
32/32 [============] - 10s 308ms/step - loss: 1.1253 - acc: 0.6074 - val loss: 1.2634 - val a
cc: 0.5492
Epoch 26/50
32/32 [===
            =========] - ETA: Os - loss: 1.0836 - acc: 0.6455
Epoch 26: val loss improved from 1.11488 to 1.09459, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
32/32 [==========] - 10s 334ms/step - loss: 1.0836 - acc: 0.6455 - val_loss: 1.0946 - val_a
cc: 0.6349
Epoch 27/50
Epoch 27: val loss did not improve from 1.09459
32/32 [========] - 13s 409ms/step - loss: 0.9800 - acc: 0.6822 - val_loss: 1.0979 - val a
cc: 0.6213
Epoch 28/50
     32/32 [===
Epoch 28: val_loss did not improve from 1.09459
32/32 [===
       c: 0.6370
Epoch 29/50
Epoch 29: val loss did not improve from 1.09459
c: 0.5994
Epoch 30/50
32/32 [==
             ========= ] - ETA: Os - loss: 1.0368 - acc: 0.6663
Epoch 30: val loss did not improve from 1.09459
c: 0.6192
Epoch 31/50
Epoch 31: val loss did not improve from 1.09459
32/32 [===
      c: 0.6433
Epoch 32/50
32/32 [===
            =========] - ETA: Os - loss: 0.9685 - acc: 0.6973
Epoch 32: val loss improved from 1.09459 to 1.03345, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
```

INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets

```
32/32 [==========] - 13s 409ms/step - loss: 0.9685 - acc: 0.6973 - val loss: 1.0335 - val a
cc: 0.6579
Epoch 33/50
Epoch 33: val loss improved from 1.03345 to 1.01040, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
32/32 [=========] - 12s 373ms/step - loss: 0.9271 - acc: 0.7012 - val loss: 1.0104 - val a
cc: 0.6841
Epoch 34/50
Epoch 34: val loss did not improve from 1.01040
c: 0.6454
Epoch 35/50
       32/32 [=====
Epoch 35: val loss did not improve from 1.01040
32/32 [==========] - 9s 297ms/step - loss: 0.9857 - acc: 0.6836 - val loss: 1.0220 - val ac
c: 0.6705
Epoch 36/50
Epoch 36: val loss did not improve from 1.01040
       32/32 [====
cc: 0.6318
Epoch 37/50
Epoch 37: val_loss did not improve from 1.01040
c: 0.6255
Epoch 38/50
Epoch 38: val loss did not improve from 1.01040
32/32 [========] - 9s 298ms/step - loss: 0.8340 - acc: 0.7119 - val loss: 1.0240 - val ac
c: 0.6444
Epoch 39/50
32/32 [===
        Epoch 39: val_loss improved from 1.01040 to 1.00646, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
       32/32 [====
cc: 0.6789
Epoch 40/50
Epoch 40: val_loss did not improve from 1.00646
32/32 [===========] - 10s 298ms/step - loss: 0.7536 - acc: 0.7432 - val loss: 1.0395 - val a
cc: 0.6674
Epoch 41/50
Epoch 41: val loss did not improve from 1.00646
32/32 [===
        c: 0.6726
Epoch 42/50
Epoch 42: val_loss did not improve from 1.00646
c: 0.6768
Epoch 43/50
32/32 [=============] - ETA: 0s - loss: 0.8051 - acc: 0.7559
Epoch 43: val loss improved from 1.00646 to 1.00357, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
32/32 [==========] - 12s 379ms/step - loss: 0.8051 - acc: 0.7559 - val_loss: 1.0036 - val_a
cc: 0.6799
Epoch 44/50
32/32 [===
        Epoch 44: val loss did not improve from 1.00357
c: 0.6684
Epoch 45/50
Epoch 45: val_loss improved from 1.00357 to 0.97444, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2 nobatchnorm/assets
cc: 0.6935
Epoch 46/50
Epoch 46: val loss improved from 0.97444 to 0.95030, saving model to models/model2_nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
cc: 0.7029
Epoch 47/50
Epoch 47: val loss improved from 0.95030 to 0.93477, saving model to models/model2 nobatchnorm
INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets
```

INFO:tensorflow:Assets written to: models/model2_nobatchnorm/assets

```
32/32 [===========] - 11s 348ms/step - loss: 0.7385 - acc: 0.7578 - val_loss: 0.9348 - val_a
cc: 0.7061
Epoch 48/50
Epoch 48: val loss did not improve from 0.93477
c: 0.6946
Epoch 49/50
Epoch 49: val loss did not improve from 0.93477
32/32 [====
     c: 0.6412
Epoch 50/50
Epoch 50: val_loss did not improve from 0.93477
32/32 [==========] - 9s 280ms/step - loss: 0.6637 - acc: 0.7783 - val loss: 1.0084 - val ac
c: 0.6705
Running Experiment: model3
Model: "sequential 4"
```

Output Shape	Param #
(None, 112, 112, 64)	2048
(None, 56, 56, 128)	74368
(None, 28, 28, 256)	296192
(None, 14, 14, 512)	1182208
(None, 100352)	0
(None, 128)	12845184
(None, 12)	1548
	(None, 112, 112, 64) (None, 56, 56, 128) (None, 28, 28, 256) (None, 14, 14, 512) (None, 100352)

Total params: 14401548 (54.94 MB) Trainable params: 14399628 (54.93 MB) Non-trainable params: 1920 (7.50 KB)

None

Epoch 1/50

2023-11-26 19:17:37.620924: E tensorflow/core/grappler/optimizers/meta optimizer.cc:961] layout failed: INVALID _ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential_4/block1/dropout_10 /dropout/SelectV2-2-TransposeNHWCToNCHW-LayoutOptimizer

6/32 [====>.....] - ETA: 2s - loss: 7.0151 - acc: 0.1094WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0375s vs `on_train_batch_end` time: 0.05 39s). Check your callbacks.

WARNING:tensorflow:Callback method `on train batch end` is slow compared to the batch time (batch time: 0.0375s vs `on_train_batch_end` time: 0.0539s). Check your callbacks.

32/32 [==============] - ETA: 0s - loss: 4.0761 - acc: 0.1533

Epoch 1: val loss improved from inf to 11.34777, saving model to models/model3

INFO:tensorflow:Assets written to: models/model3/assets

INFO:tensorflow:Assets written to: models/model3/assets

```
acc: 0.1370
Epoch 2/50
Epoch 2: val loss did not improve from 11.34777
32/32 [========] - 10s 306ms/step - loss: 2.2385 - acc: 0.1885 - val loss: 19.6838 - val
acc: 0.1370
Epoch 3/50
        32/32 [====
Epoch 3: val loss did not improve from 11.34777
32/32 [=========] - 10s 319ms/step - loss: 2.0913 - acc: 0.2412 - val loss: 22.0937 - val
acc: 0.1370
Epoch 4/50
Epoch 4: val loss did not improve from 11.34777
                    =====] - 8s 261ms/step - loss: 1.9802 - acc: 0.2901 - val loss: 26.7574 - val a
32/32 [===
cc: 0.1370
Epoch 5/50
        32/32 [=====
Epoch 5: val loss did not improve from 11.34777
acc: 0.1370
Fnoch 6/50
Epoch 6: val_loss did not improve from 11.34777
32/32 [============] - 9s 298ms/step - loss: 1.5902 - acc: 0.4443 - val loss: 26.7227 - val a
cc: 0.1370
Epoch 7/50
32/32 [===
         Epoch 7: val loss did not improve from 11.34777
cc: 0.1370
Epoch 8/50
32/32 [===
                 =======] - ETA: Os - loss: 1.3746 - acc: 0.5228
Epoch 8: val loss did not improve from 11.34777
32/32 [==========] - 9s 272ms/step - loss: 1.3746 - acc: 0.5228 - val loss: 24.6361 - val a
cc: 0.1370
Epoch 9/50
Epoch 9: val loss did not improve from 11.34777
cc: 0.1370
Epoch 10/50
32/32 [==============] - ETA: 0s - loss: 1.3731 - acc: 0.5605
Epoch 10: val_loss did not improve from 11.34777
32/32 [==========] - 9s 281ms/step - loss: 1.3731 - acc: 0.5605 - val loss: 11.6116 - val a
cc: 0.1454
Epoch 11/50
Epoch 11: val loss improved from 11.34777 to 6.09095, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
cc: 0.1789
Epoch 12/50
32/32 [=====
                  ======] - ETA: 0s - loss: 1.0585 - acc: 0.6426
Epoch 12: val loss improved from 6.09095 to 5.28668, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [=========] - 12s 371ms/step - loss: 1.0585 - acc: 0.6426 - val_loss: 5.2867 - val_a
cc: 0.1946
Epoch 13/50
32/32 [===
                  ======] - ETA: 0s - loss: 1.0073 - acc: 0.6494
Epoch 13: val loss improved from 5.28668 to 3.72143, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [=========] - 12s 387ms/step - loss: 1.0073 - acc: 0.6494 - val_loss: 3.7214 - val_a
cc: 0.2782
Epoch 14/50
32/32 [=====
                ========] - ETA: Os - loss: 1.0296 - acc: 0.6416
Epoch 14: val loss improved from 3.72143 to 3.09427, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [==
                  =======] - 11s 361ms/step - loss: 1.0296 - acc: 0.6416 - val loss: 3.0943 - val a
cc: 0.2709
Epoch 15/50
Epoch 15: val_loss did not improve from 3.09427
c: 0.3787
Epoch 16/50
Epoch 16: val_loss improved from 3.09427 to 2.09064, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
```

INFO:tensorflow:Assets written to: models/model3/assets

```
32/32 [==========] - 12s 389ms/step - loss: 0.9386 - acc: 0.6768 - val loss: 2.0906 - val a
cc: 0.4257
Epoch 17/50
Epoch 17: val loss improved from 2.09064 to 1.90015, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [==========] - 11s 351ms/step - loss: 0.7969 - acc: 0.7207 - val loss: 1.9001 - val a
cc: 0.4215
Epoch 18/50
Epoch 18: val loss improved from 1.90015 to 1.65517, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [=========] - 11s 338ms/step - loss: 0.8099 - acc: 0.7305 - val loss: 1.6552 - val a
cc: 0.4770
Epoch 19/50
Epoch 19: val loss did not improve from 1.65517
c: 0.5230
Epoch 20/50
Epoch 20: val loss improved from 1.65517 to 1.32886, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [=========] - 11s 347ms/step - loss: 0.8046 - acc: 0.7383 - val_loss: 1.3289 - val_a
cc: 0.5921
Epoch 21/50
Epoch 21: val loss improved from 1.32886 to 0.97684, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [========] - 12s 379ms/step - loss: 0.6893 - acc: 0.7578 - val_loss: 0.9768 - val_a
cc: 0.6663
Epoch 22/50
Epoch 22: val loss improved from 0.97684 to 0.86301, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [==========] - 11s 344ms/step - loss: 0.7622 - acc: 0.7490 - val loss: 0.8630 - val a
cc: 0.7186
Epoch 23/50
Epoch 23: val_loss did not improve from 0.86301
32/32 [============] - 10s 312ms/step - loss: 0.7323 - acc: 0.7455 - val loss: 0.8913 - val a
cc: 0.7238
Epoch 24/50
Epoch 24: val loss improved from 0.86301 to 0.81027, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
cc: 0.7479
Epoch 25/50
           32/32 [=====
Epoch 25: val loss did not improve from 0.81027
32/32 [=======] - 10s 298ms/step - loss: 0.6319 - acc: 0.7676 - val loss: 0.8207 - val a
cc: 0.7270
Epoch 26/50
Epoch 26: val loss improved from 0.81027 to 0.73432, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [============== ] - 10s 324ms/step - loss: 0.6516 - acc: 0.7842 - val loss: 0.7343 - val a
cc: 0.7469
Epoch 27/50
Epoch 27: val_loss did not improve from 0.73432
         32/32 [======
cc: 0.7458
Epoch 28/50
32/32 [============] - ETA: 0s - loss: 0.6028 - acc: 0.8105
Epoch 28: val loss improved from 0.73432 to 0.64487, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
```

INFO:tensorflow:Assets written to: models/model3/assets

```
32/32 [===========] - 10s 329ms/step - loss: 0.6028 - acc: 0.8105 - val loss: 0.6449 - val a
cc: 0.7793
Epoch 29/50
Epoch 29: val loss did not improve from 0.64487
cc: 0.7584
Epoch 30/50
       32/32 [===
Epoch 30: val loss did not improve from 0.64487
c: 0.7542
Epoch 31/50
Epoch 31: val loss improved from 0.64487 to 0.64376, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [=========] - 13s 407ms/step - loss: 0.5196 - acc: 0.8125 - val loss: 0.6438 - val a
cc: 0.7929
Epoch 32/50
Epoch 32: val loss did not improve from 0.64376
      32/32 [====
c: 0.7824
Epoch 33/50
Epoch 33: val_loss did not improve from 0.64376
32/32 [==========] - 10s 305ms/step - loss: 0.5827 - acc: 0.8105 - val loss: 0.7436 - val a
cc: 0.7615
Epoch 34/50
Epoch 34: val loss did not improve from 0.64376
c: 0.7333
Epoch 35/50
32/32 [=====
       Epoch 35: val_loss did not improve from 0.64376
32/32 [=========] - 10s 319ms/step - loss: 0.5140 - acc: 0.8271 - val loss: 0.8938 - val a
cc: 0.7228
Epoch 36/50
Epoch 36: val loss did not improve from 0.64376
          :=========] - 10s 302ms/step - loss: 0.4562 - acc: 0.8545 - val loss: 0.7480 - val a
32/32 [==
cc: 0.7699
Epoch 37/50
Epoch 37: val loss did not improve from 0.64376
c: 0.7573
Epoch 38/50
Epoch 38: val loss improved from 0.64376 to 0.64312, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
32/32 [==========] - 13s 401ms/step - loss: 0.4100 - acc: 0.8505 - val loss: 0.6431 - val a
cc: 0.7845
Epoch 39/50
Epoch 39: val loss did not improve from 0.64312
32/32 [=====
       cc: 0.7897
Epoch 40/50
Epoch 40: val loss improved from 0.64312 to 0.51126, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
```

```
32/32 [===========] - 12s 394ms/step - loss: 0.5067 - acc: 0.8369 - val loss: 0.5113 - val a
cc: 0.8190
Epoch 41/50
Epoch 41: val loss did not improve from 0.51126
32/32 [=============] - 10s 315ms/step - loss: 0.4157 - acc: 0.8416 - val loss: 0.6337 - val a
cc: 0.8002
Epoch 42/50
       32/32 [=====
Epoch 42: val loss did not improve from 0.51126
32/32 [=============] - 10s 303ms/step - loss: 0.3793 - acc: 0.8613 - val loss: 0.6934 - val a
cc: 0.7835
Epoch 43/50
Epoch 43: val loss did not improve from 0.51126
                  =====] - 10s 312ms/step - loss: 0.4491 - acc: 0.8604 - val loss: 0.6739 - val a
32/32 [==:
cc: 0.7981
Epoch 44/50
32/32 [=============] - ETA: 0s - loss: 0.4175 - acc: 0.8594
Epoch 44: val loss did not improve from 0.51126
cc: 0.7510
Epoch 45/50
32/32 [=====
             ======== ] - ETA: Os - loss: 0.3934 - acc: 0.8525
Epoch 45: val_loss did not improve from 0.51126
32/32 [============] - 10s 332ms/step - loss: 0.3934 - acc: 0.8525 - val loss: 0.6475 - val a
cc: 0.7939
Epoch 46/50
Epoch 46: val loss did not improve from 0.51126
32/32 [===
       cc: 0.8159
Epoch 47/50
32/32 [=====
        Epoch 47: val loss did not improve from 0.51126
c: 0.7573
Epoch 48/50
Epoch 48: val loss improved from 0.51126 to 0.50210, saving model to models/model3
INFO:tensorflow:Assets written to: models/model3/assets
INFO:tensorflow:Assets written to: models/model3/assets
cc: 0.8358
Epoch 49/50
Epoch 49: val loss did not improve from 0.50210
c: 0.8379
Epoch 50/50
32/32 [===
            ========] - ETA: Os - loss: 0.3898 - acc: 0.8818
Epoch 50: val loss did not improve from 0.50210
c: 0.8159
Running Experiment: model3 nobatchnorm
Model: "sequential 5"
```

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	1792
block2 (Sequential)	(None, 56, 56, 128)	73856
block3 (Sequential)	(None, 28, 28, 256)	295168
block4 (Sequential)	(None, 14, 14, 512)	1180160
<pre>top_model_flatten (Flatten)</pre>	(None, 100352)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	12845184
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

Total params: 14397708 (54.92 MB) Trainable params: 14397708 (54.92 MB) Non-trainable params: 0 (0.00 Byte)

None

Epoch 1/50

2023-11-26 19:26:09.154955: E tensorflow/core/grappler/optimizers/meta_optimizer.cc:961] layout failed: INVALID _ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential_5/block1/dropout_14/dropout/SelectV2-2-TransposeNHWCToNCHW-LayoutOptimizer

6/32 [====>......] - ETA: 3s - loss: 2.5770 - acc: 0.0833WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0415s vs `on_train_batch_end` time: 0.06 31s). Check your callbacks.

```
WARNING:tensorflow:Callback method `on train batch end` is slow compared to the batch time (batch time: 0.0415s
vs `on_train_batch_end` time: 0.0631s). Check your callbacks.
Epoch 1: val loss improved from inf to 2.48293, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
           cc: 0.0554
Epoch 2/50
Epoch 2: val loss improved from 2.48293 to 2.45301, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [===
              =========] - 11s 345ms/step - loss: 2.4596 - acc: 0.0771 - val loss: 2.4530 - val a
cc: 0.0826
Epoch 3/50
Epoch 3: val loss improved from 2.45301 to 2.24443, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [====
           cc: 0.2092
Epoch 4/50
Epoch 4: val loss improved from 2.24443 to 2.01219, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [==========] - 10s 326ms/step - loss: 2.0726 - acc: 0.2109 - val_loss: 2.0122 - val_a
cc: 0.2678
Epoch 5/50
Epoch 5: val loss improved from 2.01219 to 1.86625, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [===========] - 10s 330ms/step - loss: 1.9160 - acc: 0.2451 - val loss: 1.8662 - val a
cc: 0.3065
Epoch 6/50
Epoch 6: val loss improved from 1.86625 to 1.80241, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [==========] - 12s 384ms/step - loss: 1.8128 - acc: 0.2852 - val loss: 1.8024 - val a
cc: 0.3232
Epoch 7/50
Epoch 7: val loss improved from 1.80241 to 1.78116, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
32/32 [============= ] - 12s 394ms/step - loss: 1.7701 - acc: 0.3125 - val loss: 1.7812 - val a
cc: 0.2782
Epoch 8/50
Epoch 8: val loss improved from 1.78116 to 1.69187, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
cc: 0.3169
Epoch 9/50
Epoch 9: val loss improved from 1.69187 to 1.59509, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
cc: 0.4613
Epoch 10/50
Epoch 10: val loss improved from 1.59509 to 1.50615, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
32/32 [=========] - 12s 392ms/step - loss: 1.5784 - acc: 0.4170 - val loss: 1.5061 - val a
cc: 0.4278
Epoch 11/50
32/32 [===
                  ========] - ETA: Os - loss: 1.5440 - acc: 0.4346
Epoch 11: val loss improved from 1.50615 to 1.48567, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
         32/32 [====
cc: 0.5042
Epoch 12/50
32/32 [====
                =========] - ETA: Os - loss: 1.4612 - acc: 0.4861
Epoch 12: val loss improved from 1.48567 to 1.42787, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
```

INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets

```
32/32 [===========] - 11s 352ms/step - loss: 1.4612 - acc: 0.4861 - val loss: 1.4279 - val a
cc: 0.5178
Epoch 13/50
Epoch 13: val loss improved from 1.42787 to 1.42226, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [=========] - 12s 366ms/step - loss: 1.3907 - acc: 0.5352 - val loss: 1.4223 - val a
cc: 0.4812
Epoch 14/50
Epoch 14: val loss improved from 1.42226 to 1.36990, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [=========] - 12s 378ms/step - loss: 1.4972 - acc: 0.4775 - val loss: 1.3699 - val a
cc: 0.5701
Epoch 15/50
Epoch 15: val loss improved from 1.36990 to 1.25837, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [=========] - 11s 346ms/step - loss: 1.2982 - acc: 0.5446 - val loss: 1.2584 - val a
cc: 0.5962
Epoch 16/50
Epoch 16: val_loss did not improve from 1.25837
          32/32 [=====
cc: 0.5450
Epoch 17/50
Epoch 17: val_loss improved from 1.25837 to 1.25134, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
cc: 0.5973
Epoch 18/50
Epoch 18: val loss improved from 1.25134 to 1.20503, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
cc: 0.5952
Epoch 19/50
Epoch 19: val loss improved from 1.20503 to 1.09078, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [==========] - 11s 360ms/step - loss: 1.0758 - acc: 0.6218 - val loss: 1.0908 - val a
cc: 0.6485
Epoch 20/50
32/32 [=============] - ETA: 0s - loss: 1.0611 - acc: 0.6494
Epoch 20: val loss improved from 1.09078 to 1.02915, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
32/32 [==========] - 11s 354ms/step - loss: 1.0611 - acc: 0.6494 - val loss: 1.0291 - val a
cc: 0.6663
Epoch 21/50
Epoch 21: val loss improved from 1.02915 to 1.01688, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
32/32 [=============] - 11s 364ms/step - loss: 1.0461 - acc: 0.6367 - val loss: 1.0169 - val a
cc: 0.6820
Epoch 22/50
32/32 [=============] - ETA: 0s - loss: 1.0695 - acc: 0.6367
Epoch 22: val loss improved from 1.01688 to 0.98831, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
32/32 [==========] - 10s 332ms/step - loss: 1.0695 - acc: 0.6367 - val loss: 0.9883 - val a
cc: 0.6956
Epoch 23/50
32/32 [==============] - ETA: 0s - loss: 0.8842 - acc: 0.7228
Epoch 23: val loss improved from 0.98831 to 0.97229, saving model to models/model3 nobatchnorm
INFO: tensorflow: Assets \ written \ to: \ models/model3\_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
32/32 [=============] - 11s 359ms/step - loss: 0.8842 - acc: 0.7228 - val loss: 0.9723 - val a
cc: 0.6663
Epoch 24/50
32/32 [============] - ETA: 0s - loss: 0.9008 - acc: 0.6729
Epoch 24: val loss improved from 0.97229 to 0.87410, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
```

INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets

```
32/32 [==========] - 11s 347ms/step - loss: 0.9008 - acc: 0.6729 - val loss: 0.8741 - val a
cc: 0.7071
Epoch 25/50
Epoch 25: val loss did not improve from 0.87410
32/32 [==========] - 10s 316ms/step - loss: 0.9123 - acc: 0.6855 - val_loss: 0.9550 - val_a
cc: 0.6841
Epoch 26/50
Epoch 26: val_loss improved from 0.87410 to 0.84050, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
cc: 0.7312
Epoch 27/50
Epoch 27: val loss improved from 0.84050 to 0.82775, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
32/32 [===========] - 12s 374ms/step - loss: 0.8056 - acc: 0.7208 - val loss: 0.8277 - val a
cc: 0.7322
Epoch 28/50
Epoch 28: val loss did not improve from 0.82775
cc: 0.7113
Epoch 29/50
32/32 [=====
        Epoch 29: val loss did not improve from 0.82775
c: 0.7092
Epoch 30/50
Epoch 30: val_loss did not improve from 0.82775
        32/32 [===
c: 0.6925
Epoch 31/50
Epoch 31: val loss did not improve from 0.82775
c: 0.7207
Epoch 32/50
Epoch 32: val loss improved from 0.82775 to 0.74871, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [=========] - 12s 391ms/step - loss: 0.7282 - acc: 0.7480 - val loss: 0.7487 - val a
cc: 0.7563
Epoch 33/50
Epoch 33: val loss did not improve from 0.74871
c: 0.7458
Epoch 34/50
32/32 [===
       Epoch 34: val loss improved from 0.74871 to 0.74466, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
```

```
32/32 [=============] - 12s 370ms/step - loss: 0.6231 - acc: 0.7851 - val loss: 0.7447 - val a
cc: 0.7479
Epoch 35/50
Epoch 35: val loss did not improve from 0.74466
32/32 [==========] - 11s 351ms/step - loss: 0.7369 - acc: 0.7305 - val loss: 0.8300 - val a
cc: 0.7155
Epoch 36/50
       32/32 [==
Epoch 36: val loss did not improve from 0.74466
c: 0.7238
Epoch 37/50
Epoch 37: val loss did not improve from 0.74466
                  ====] - 8s 240ms/step - loss: 0.7340 - acc: 0.7363 - val loss: 0.8206 - val ac
32/32 [==
c: 0.7259
Epoch 38/50
        32/32 [=====
Epoch 38: val loss did not improve from 0.74466
c: 0.7155
Fnoch 39/50
Epoch 39: val_loss did not improve from 0.74466
32/32 [===
      c: 0.7490
Epoch 40/50
32/32 [===
            =========] - ETA: Os - loss: 0.6325 - acc: 0.7812
Epoch 40: val loss did not improve from 0.74466
cc: 0.7458
Epoch 41/50
Epoch 41: val_loss did not improve from 0.74466
32/32 [==
             =======] - 9s 299ms/step - loss: 0.6104 - acc: 0.7891 - val loss: 0.8864 - val ac
c: 0.6872
Epoch 42/50
Epoch 42: val loss did not improve from 0.74466
32/32 [========] - 10s 331ms/step - loss: 0.6650 - acc: 0.7617 - val loss: 0.7743 - val a
cc: 0.7385
Epoch 43/50
Epoch 43: val loss improved from 0.74466 to 0.71436, saving model to models/model3 nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [=========] - 11s 348ms/step - loss: 0.5441 - acc: 0.8223 - val loss: 0.7144 - val a
cc: 0.7667
Epoch 44/50
Epoch 44: val loss did not improve from 0.71436
32/32 [============= ] - 11s 331ms/step - loss: 0.5878 - acc: 0.8037 - val loss: 0.7323 - val a
cc: 0.7490
Epoch 45/50
       32/32 [==
Epoch 45: val loss did not improve from 0.71436
c: 0.7552
Epoch 46/50
Epoch 46: val_loss improved from 0.71436 to 0.66713, saving model to models/model3_nobatchnorm
INFO:tensorflow:Assets written to: models/model3_nobatchnorm/assets
INFO:tensorflow:Assets written to: models/model3 nobatchnorm/assets
32/32 [========] - 11s 350ms/step - loss: 0.5395 - acc: 0.8320 - val loss: 0.6671 - val a
cc: 0.7741
Epoch 47/50
Epoch 47: val_loss did not improve from 0.66713
32/32 [====
       cc: 0.7762
Epoch 48/50
Epoch 48: val_loss did not improve from 0.66713
c: 0.7615
Epoch 49/50
Epoch 49: val loss did not improve from 0.66713
32/32 [=========] - 10s 308ms/step - loss: 0.4937 - acc: 0.8307 - val loss: 0.7318 - val a
cc: 0.7469
Epoch 50/50
           32/32 [====
Epoch 50: val_loss did not improve from 0.66713
c: 0.7636
```

```
def print_classification_report(true_labels, predictions):
    predicted_classes = np.argmax(predictions, axis=1)
    report = classification_report(true_labels.tolist(), predicted_classes.tolist())
    print("Classification Report:\n", report)
```

```
In [28]: # Cargar arquitecturas de los modelos para cargar pesos en memoria
         models list = []
         for experiment_name, config in scratch_experiments.items():
             print(f"Running Experiment: {experiment_name}")
             # Configurar el modelo utilizando las configuraciones
             block configs = []
             for block conf in config["model"]["conv block configs"]:
                 block_conf["use_batchnorm"],
block_conf["dropout_value"],
                                         block conf["name"])
                 block_configs.append(block)
             # Crear modelo
             model = create model with configurations(block configs, input shape, num classes)
             # Compilar modelo
             model.compile(
                 loss = tf.keras.losses.categorical crossentropy,
                 optimizer = tf.keras.optimizers.Adam(0.0001),
metrics = ['acc'])
             # Crear callbacks
             callbacks = get callbacks(experiment name)
             # Mostrar arquitectura
             print(model.summary())
             models list.append(model)
```

Running Experiment: model1 Model: "sequential 6"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	2048
block2 (Sequential)	(None, 56, 56, 128)	74368
block3 (Sequential)	(None, 28, 28, 256)	296192
<pre>top_model_flatten (Flatten)</pre>	(None, 200704)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	25690240
<pre>top_model_output (Dense)</pre>	(None, 12)	1548
	=======================================	

Total params: 26064396 (99.43 MB) Trainable params: 26063500 (99.42 MB) Non-trainable params: 896 (3.50 KB)

Running Experiment: $model1_nobatchnorm$

Model: "sequential_7"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	1792
block2 (Sequential)	(None, 56, 56, 128)	73856
block3 (Sequential)	(None, 28, 28, 256)	295168
<pre>top_model_flatten (Flatten)</pre>	(None, 200704)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	25690240
top_model_output (Dense)	(None, 12)	1548

Total params: 26062604 (99.42 MB) Trainable params: 26062604 (99.42 MB) Non-trainable params: 0 (0.00 Byte)

None

Running Experiment: model2

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	5120
block2 (Sequential)	(None, 56, 56, 128)	205440
<pre>top_model_flatten (Flatten)</pre>	(None, 401408)	Θ
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	51380352
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

Total params: 51592460 (196.81 MB)

Trainable params: 51592076 (196.81 MB) Non-trainable params: 384 (1.50 KB)

Running Experiment: model2_nobatchnorm Model: "sequential_9"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	4864
block2 (Sequential)	(None, 56, 56, 128)	204928
<pre>top_model_flatten (Flatten)</pre>	(None, 401408)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	51380352
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

_____ Total params: 51591692 (196.81 MB)

Trainable params: 51591692 (196.81 MB) Non-trainable params: 0 (0.00 Byte)

Running Experiment: model3 Model: "sequential_10"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	2048
block2 (Sequential)	(None, 56, 56, 128)	74368
block3 (Sequential)	(None, 28, 28, 256)	296192
block4 (Sequential)	(None, 14, 14, 512)	1182208
<pre>top_model_flatten (Flatten)</pre>	(None, 100352)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	12845184
top_model_output (Dense)	(None, 12)	1548

Total params: 14401548 (54.94 MB) Trainable params: 14399628 (54.93 MB) Non-trainable params: 1920 (7.50 KB)

None

Running Experiment: model3_nobatchnorm

Model: "sequential_11"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	1792
block2 (Sequential)	(None, 56, 56, 128)	73856
block3 (Sequential)	(None, 28, 28, 256)	295168
block4 (Sequential)	(None, 14, 14, 512)	1180160
<pre>top_model_flatten (Flatten)</pre>	(None, 100352)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	12845184
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

Total params: 14397708 (54.92 MB) Trainable params: 14397708 (54.92 MB) Non-trainable params: 0 (0.00 Byte)

None

Model: "sequential_6"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	2048
block2 (Sequential)	(None, 56, 56, 128)	74368
block3 (Sequential)	(None, 28, 28, 256)	296192
<pre>top_model_flatten (Flatten)</pre>	(None, 200704)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	25690240
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

2023-11-26 19:34:55.980413: W tensorflow/core/util/tensor_slice_reader.cc:98] Could not open models/model1: FAI LED_PRECONDITION: models/model1; Is a directory: perhaps your file is in a different file format and you need to use a different restore operator?

Total params: 26064396 (99.43 MB) Trainable params: 26063500 (99.42 MB) Non-trainable params: 896 (3.50 KB)

30/30 [======] - 5s 159ms/step

Model: "sequential_7"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	1792
block2 (Sequential)	(None, 56, 56, 128)	73856
block3 (Sequential)	(None, 28, 28, 256)	295168
<pre>top_model_flatten (Flatten)</pre>	(None, 200704)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	25690240
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

Total params: 26062604 (99.42 MB) Trainable params: 26062604 (99.42 MB) Non-trainable params: 0 (0.00 Byte)

2023-11-26 19:35:01.114545: W tensorflow/core/util/tensor_slice_reader.cc:98] Could not open models/model1_nobatchnorm: FAILED_PRECONDITION: models/model1_nobatchnorm; Is a directory: perhaps your file is in a different file format and you need to use a different restore operator?

30/30 [==========] - 5s 148ms/step

2023-11-26 19:35:05.855373: W tensorflow/core/util/tensor_slice_reader.cc:98] Could not open models/model2: FAI LED_PRECONDITION: models/model2; Is a directory: perhaps your file is in a different file format and you need to use a different restore operator?

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	5120
block2 (Sequential)	(None, 56, 56, 128)	205440
<pre>top_model_flatten (Flatten)</pre>	(None, 401408)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	51380352
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

Total params: 51592460 (196.81 MB) Trainable params: 51592076 (196.81 MB) Non-trainable params: 384 (1.50 KB)

30/30 [======] - 5s 148ms/step

2023-11-26 19:35:10.757952: W tensorflow/core/util/tensor_slice_reader.cc:98] Could not open models/model2_nobatchnorm: FAILED_PRECONDITION: models/model2_nobatchnorm; Is a directory: perhaps your file is in a different fi le format and you need to use a different restore operator?

Model: "sequential_9"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	4864
block2 (Sequential)	(None, 56, 56, 128)	204928
<pre>top_model_flatten (Flatten)</pre>	(None, 401408)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	51380352
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

Total params: 51591692 (196.81 MB) Trainable params: 51591692 (196.81 MB) Non-trainable params: 0 (0.00 Byte)

30/30 [=======] - 5s 148ms/step

Model: "sequential_10"

Layer (type)	Output Shape	Param #
block1 (Sequential)	(None, 112, 112, 64)	2048
block2 (Sequential)	(None, 56, 56, 128)	74368
block3 (Sequential)	(None, 28, 28, 256)	296192
block4 (Sequential)	(None, 14, 14, 512)	1182208
<pre>top_model_flatten (Flatten)</pre>	(None, 100352)	0
<pre>top_model_dense_1 (Dense)</pre>	(None, 128)	12845184
<pre>top_model_output (Dense)</pre>	(None, 12)	1548

Total params: 14401548 (54.94 MB) Trainable params: 14399628 (54.93 MB) Non-trainable params: 1920 (7.50 KB)

2023-11-26 19:35:15.647254: W tensorflow/core/util/tensor slice reader.cc:98] Could not open models/model3: FAI LED_PRECONDITION: models/model3; Is a directory: perhaps your file is in a different file format and you need t o use a different restore operator?

30/30 [=======] - 5s 149ms/step Model: "sequential_11"

Output Shape	Param #
(None, 112, 112, 64)	1792
(None, 56, 56, 128)	73856
(None, 28, 28, 256)	295168
(None, 14, 14, 512)	1180160
(None, 100352)	0
(None, 128)	12845184
(None, 12)	1548
	(None, 112, 112, 64) (None, 56, 56, 128) (None, 28, 28, 256) (None, 14, 14, 512) (None, 100352) (None, 128)

Total params: 14397708 (54.92 MB) Trainable params: 14397708 (54.92 MB) Non-trainable params: 0 (0.00 Byte)

2023-11-26 19:35:20.731592: W tensorflow/core/util/tensor_slice_reader.cc:98] Could not open models/model3_noba tchnorm: FAILED_PRECONDITION: models/model3_nobatchnorm; Is a directory: perhaps your file is in a different fi le format and you need to use a different restore operator? 30/30 [======] - 4s 141ms/step

```
In [30]: # Mostrar classification report
           for prediction, checkpoint_filepath in zip(predictions, checkpoint_paths):
    print(f"\nModel: {checkpoint_filepath}")
                print_classification_report(df_dataset_valid["label"].to_numpy(), prediction)
```

Model: models/model1 Classification Report:

	precision	recall	f1-score	support
0	0.75	0.55	0.63	131
1	0.89	0.76	0.82	123
2	0.88	0.49	0.63	104
3	0.90	0.84	0.87	100
4	0.69	0.78	0.73	95
5	0.81	0.77	0.79	78
6	0.76	0.79	0.78	77
7	0.79	0.93	0.86	58
8	0.33	0.32	0.32	53
9	0.45	0.83	0.58	47
10	0.40	0.67	0.50	45
11	0.72	0.96	0.82	45
accuracy			0.71	956
macro avg	0.70	0.72	0.69	956
weighted avg	0.74	0.71	0.71	956

Model: models/model1_nobatchnorm

Classification Report:

	precision	recall	f1-score	support
0	0.77	0.69	0.73	131
1	0.76	0.86	0.81	123
2	0.91	0.75	0.82	104
3	0.79	0.87	0.83	100
4	0.61	0.68	0.64	95
5	0.89	0.82	0.85	78
6	0.83	0.70	0.76	77
7	0.74	0.88	0.80	58
8	0.52	0.42	0.46	53
9	0.60	0.55	0.58	47
10	0.48	0.64	0.55	45
11	0.76	0.76	0.76	45
accuracy			0.74	956
macro avg	0.72	0.72	0.72	956
weighted avg	0.75	0.74	0.74	956

Model: models/model2 Classification Report

ation	Report: precision	recall	f1-score	support
0	0.57	0.74	0.64	131
1	0.84	0.52	0.64	123
2	0.55	0.33	0.41	104
3	0.80	0.81	0.81	100
4	0.22	0.39	0.28	95

5	0.69	0.62	0.65	78
6	0.81	0.45	0.58	77
7	0.64	0.74	0.69	58
8 9	0.45 0.40	0.38 0.17	0.41 0.24	53 47
10	0.43	0.17	0.49	45
11	0.44	0.76	0.56	45
accuracy			0.55	956
macro avg	0.57	0.54	0.53	956
weighted avg	0.60	0.55	0.55	956
	model2_nobate	chnorm		
Classification	n Report: precision	recall	f1-score	cupport
	precision	recatt	11-50016	support
0	0.72	0.75	0.73	131
1 2	0.90	0.77	0.83	123
3	0.56 0.85	0.81 0.87	0.66 0.86	104 100
4	0.72	0.53	0.61	95
5	0.81	0.86	0.83	78
6	0.73	0.64	0.68	77
7	0.80	0.84	0.82	58
8 9	0.45	0.28	0.35	53 47
10	0.61 0.37	0.47 0.56	0.53 0.45	47
11	0.72	0.76	0.74	45
accuracy	0.60	0.68	0.71 0.67	956 956
macro avg weighted avg	0.69 0.72	0.71	0.07	956
orgca arg	01.72	0.7.2	0.70	
Madaldala	/ d = 1.2			
Model: models, Classification				
Ctassification	precision	recall	f1-score	support
0	0.86	0.68	0.76	131
1 2	0.95 0.84	0.93 0.86	0.94 0.85	123 104
3	0.92	0.91	0.03	100
4	0.89	0.89	0.89	95
	0.05			
5	0.93	0.86	0.89	78
6	0.93 0.93	0.86 0.83	0.88	78 77
6 7	0.93 0.93 0.86	0.86 0.83 0.88	0.88 0.87	78 77 58
6 7 8	0.93 0.93 0.86 0.47	0.86 0.83 0.88 0.53	0.88 0.87 0.50	78 77 58 53
6 7	0.93 0.93 0.86	0.86 0.83 0.88	0.88 0.87	78 77 58
6 7 8 9	0.93 0.93 0.86 0.47 0.65	0.86 0.83 0.88 0.53 0.87	0.88 0.87 0.50 0.75	78 77 58 53 47
6 7 8 9 10 11	0.93 0.93 0.86 0.47 0.65 0.63	0.86 0.83 0.88 0.53 0.87 0.84	0.88 0.87 0.50 0.75 0.72 0.88	78 77 58 53 47 45 45
6 7 8 9 10	0.93 0.93 0.86 0.47 0.65 0.63	0.86 0.83 0.88 0.53 0.87 0.84	0.88 0.87 0.50 0.75 0.72 0.88	78 77 58 53 47 45
6 7 8 9 10 11	0.93 0.93 0.86 0.47 0.65 0.63	0.86 0.83 0.88 0.53 0.87 0.84	0.88 0.87 0.50 0.75 0.72 0.88	78 77 58 53 47 45 45
6 7 8 9 10 11 accuracy macro avg	0.93 0.93 0.86 0.47 0.65 0.63 0.84	0.86 0.83 0.88 0.53 0.87 0.84 0.93	0.88 0.87 0.50 0.75 0.72 0.88	78 77 58 53 47 45 45 956
6 7 8 9 10 11 accuracy macro avg weighted avg	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85	0.86 0.83 0.88 0.53 0.87 0.84 0.93	0.88 0.87 0.50 0.75 0.72 0.88	78 77 58 53 47 45 45 956
6 7 8 9 10 11 accuracy macro avg weighted avg	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85	0.86 0.83 0.88 0.53 0.87 0.84 0.93	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956
6 7 8 9 10 11 accuracy macro avg weighted avg	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85	0.86 0.83 0.88 0.53 0.87 0.84 0.93	0.88 0.87 0.50 0.75 0.72 0.88	78 77 58 53 47 45 45 956
6 7 8 9 10 11 accuracy macro avg weighted avg	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) 1 Report: precision 0.82 0.82 0.86	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956 956 956
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) Report: precision 0.82 0.86 0.73	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956 956 956
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate n Report: precision 0.82 0.86 0.73 0.95	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956 956 956 131 123 104 100
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) Report: precision 0.82 0.86 0.73	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956 956 956
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) Report: precision 0.82 0.86 0.73 0.95 0.68	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.98 0.76	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956 956 956 131 123 104 100 95 78 77
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6 7	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate 0.85 (model3_nobate 0.86 0.73 0.95 0.68 0.93 0.86 0.76	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.98 0.76 0.81 0.77	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84 0.84 0.82 0.84	78 77 58 53 47 45 45 956 956 956 956 131 123 104 100 95 78 77 58
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6 7 8	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) 1 Report: precision 0.82 0.86 0.73 0.95 0.68 0.93 0.86 0.76 0.40	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.98 0.76 0.81 0.77 0.91	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84 f1-score 0.68 0.88 0.81 0.91 0.72 0.86 0.81 0.83 0.40	78 77 58 53 47 45 45 956 956 956 956 956 956 77 78 77 58 53
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6 7 8 9	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) 0.82 0.85 (model3_nobate) 0.82 0.86 0.73 0.95 0.68 0.93 0.86 0.76 0.40 0.76	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.88 0.76 0.81 0.77 0.91 0.40	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84 f1-score 0.68 0.88 0.81 0.91 0.72 0.86 0.81 0.83 0.40	78 77 58 53 47 45 45 956 956 956 956 956 78 77 58 53 47
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6 7 8	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) 1 Report: precision 0.82 0.86 0.73 0.95 0.68 0.93 0.86 0.76 0.40	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.98 0.76 0.81 0.77 0.91	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84 f1-score 0.68 0.88 0.81 0.91 0.72 0.86 0.81 0.83 0.40	78 77 58 53 47 45 45 956 956 956 956 956 956 77 78 77 58 53
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6 7 8 9 10 11	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) 0.82 0.85 (model3_nobate) 0.82 0.86 0.73 0.95 0.68 0.93 0.86 0.73 0.95 0.69 0.76 0.76 0.76 0.76 0.76	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.90 0.88 0.76 0.81 0.77 0.91 0.40 0.68 0.71	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84 0.83 0.81 0.91 0.72 0.86 0.81 0.83 0.40 0.72 0.63 0.82	78 77 58 53 47 45 45 956 956 956 956 956 956 956 956 956 45 45 45 45
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6 7 8 9 10 11 accuracy	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) 0.82 0.86 0.73 0.95 0.68 0.93 0.86 0.76 0.40 0.76 0.57 0.79	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.98 0.76 0.81 0.77 0.91 0.68 0.71	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84 0.83 0.81 0.91 0.72 0.86 0.81 0.83 0.40 0.72 0.63 0.82	78 77 58 53 47 45 45 956 956 956 956 31 104 100 95 78 77 58 53 47 45 45
6 7 8 9 10 11 accuracy macro avg weighted avg Model: models, Classification 0 1 2 3 4 5 6 7 8 9 10 11	0.93 0.93 0.86 0.47 0.65 0.63 0.84 0.82 0.85 (model3_nobate) 0.82 0.85 (model3_nobate) 0.82 0.86 0.73 0.95 0.68 0.93 0.86 0.73 0.95 0.69 0.76 0.76 0.76 0.76 0.76	0.86 0.83 0.88 0.53 0.87 0.84 0.93 0.83 0.84 chnorm recall 0.59 0.90 0.90 0.90 0.88 0.76 0.81 0.77 0.91 0.40 0.68 0.71	0.88 0.87 0.50 0.75 0.72 0.88 0.84 0.82 0.84 0.83 0.81 0.91 0.72 0.86 0.81 0.83 0.40 0.72 0.63 0.82	78 77 58 53 47 45 45 956 956 956 956 956 956 956 956 956 45 45 45 45

5.2 Monitorización del proceso de entrenamiento para la toma de decisiones (pre-entrenada)

DenseNet

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten
# Tamaño entrada
input\_shape = (224, 224, 3)
# Callbacks y guardado de grádicas
callbacks = get_callbacks("DenseNet121")
# Cargar el modelo preentrenado de DenseNet121 sin incluir las capas densas (fully connected)
base model = DenseNet121(weights='imagenet', include top=False, input shape=input shape)
# Congelar las capas preentrenadas
for layer in base model.layers:
   layer.trainable = False
# Crear un nuevo modelo Sequential
model = Sequential()
# Agregar la base preentrenada de DenseNet121
model.add(base model)
# Añadir capas adicionales para la clasificación
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dense(12, activation='softmax'))
# Compilar el modelo
model.compile(optimizer='adam', loss='categorical crossentropy', metrics=['acc'])
# Mostrar la arquitectura del modelo
model.summary()
history = model.fit(
   train_tfdataset.repeat(),
   steps per epoch = 32,
   epochs = 50,
   validation_data = valid_tfdataset,
   callbacks = callbacks
Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/densenet/densenet121_weights
tf dim ordering tf kernels notop.h5
Model: "sequential 12"
                         Output Shape
Layer (type)
                                               Param #
densenet121 (Functional)
                       (None, 7, 7, 1024)
                                               7037504
flatten (Flatten)
                         (None, 50176)
                                                6422656
dense (Dense)
                         (None, 128)
dense_1 (Dense)
                         (None, 12)
                                               1548
Total params: 13461708 (51.35 MB)
Trainable params: 6424204 (24.51 MB)
Non-trainable params: 7037504 (26.85 MB)
Epoch 1/50
Epoch 1: val loss improved from inf to 1.84584, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
32/32 [==========] - 43s 1s/step - loss: 5.0648 - acc: 0.2617 - val loss: 1.8458 - val acc:
0.3944
Epoch 2/50
32/32 [====
             Epoch 2: val loss improved from 1.84584 to 1.34165, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
                    ========] - 39s 1s/step - loss: 1.6440 - acc: 0.4434 - val loss: 1.3416 - val acc:
32/32 [====
0.5429
Epoch 3/50
Epoch 3: val loss improved from 1.34165 to 1.31440, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
             32/32 [=====
0.5659
Epoch 4/50
32/32 [===========] - ETA: 0s - loss: 1.0090 - acc: 0.6525
Epoch 4: val loss improved from 1.31440 to 1.08237, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
```

```
INFO:tensorflow:Assets written to: models/DenseNet121/assets
32/32 [========] - 35s 1s/step - loss: 1.0090 - acc: 0.6525 - val_loss: 1.0824 - val_acc:
0.6151
Epoch 5/50
Epoch 5: val loss improved from 1.08237 to 0.78110, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
32/32 [==========] - 36s 1s/step - loss: 0.8035 - acc: 0.7266 - val loss: 0.7811 - val acc:
0.7249
Epoch 6/50
32/32 [========] - ETA: 0s - loss: 0.7419 - acc: 0.7500
Epoch 6: val loss improved from 0.78110 to 0.75672, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
0.7500
Epoch 7/50
Epoch 7: val loss improved from 0.75672 to 0.71590, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
0.7584
Epoch 8/50
Epoch 8: val_loss did not improve from 0.71590
32/32 [============= ] - 8s 263ms/step - loss: 0.6959 - acc: 0.7733 - val loss: 0.7267 - val ac
c: 0.7469
Epoch 9/50
Epoch 9: val loss did not improve from 0.71590
c: 0.7584
Epoch 10/50
32/32 [============== ] - ETA: 0s - loss: 0.5088 - acc: 0.8213
Epoch 10: val loss improved from 0.71590 to 0.66193, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
0.7573
Epoch 11/50
                  ======] - ETA: 0s - loss: 0.6131 - acc: 0.8008
32/32 [==
Epoch 11: val_loss did not improve from 0.66193
c: 0.7699
Epoch 12/50
Epoch 12: val loss improved from 0.66193 to 0.62316, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
0.7730
Epoch 13/50
Epoch 13: val loss did not improve from 0.62316
                 :=======] - 9s 287ms/step - loss: 0.4282 - acc: 0.8457 - val loss: 0.6300 - val ac
32/32 [==
c: 0.7929
Epoch 14/50
32/32 [=====
              ========= ] - ETA: Os - loss: 0.4668 - acc: 0.8340
Epoch 14: val_loss did not improve from 0.62316
32/32 [========] - 8s 242ms/step - loss: 0.4668 - acc: 0.8340 - val loss: 0.6337 - val ac
c: 0.8065
Epoch 15/50
Epoch 15: val loss did not improve from 0.62316
32/32 [=============] - 10s 326ms/step - loss: 0.4470 - acc: 0.8287 - val loss: 0.6533 - val a
cc: 0.7845
Epoch 16/50
32/32 [=====
          Epoch 16: val loss improved from 0.62316 to 0.54595, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
```

```
32/32 [==========] - 36s 1s/step - loss: 0.3566 - acc: 0.8730 - val loss: 0.5459 - val acc:
0.8086
Epoch 17/50
Epoch 17: val loss did not improve from 0.54595
c: 0.7877
Epoch 18/50
       32/32 [===
Epoch 18: val loss did not improve from 0.54595
c: 0.8232
Epoch 19/50
            ========>.] - ETA: Os - loss: 0.2981 - acc: 0.8906
Epoch 19: val loss did not improve from 0.54595
c: 0.8264
Epoch 20/50
Epoch 20: val_loss improved from 0.54595 to 0.53457, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
0.8128
Epoch 21/50
Epoch 21: val_loss improved from 0.53457 to 0.51804, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
0.8285
Epoch 22/50
Epoch 22: val_loss did not improve from 0.51804
       c: 0.8243
Epoch 23/50
32/32 [============] - ETA: 0s - loss: 0.2332 - acc: 0.9218
Epoch 23: val loss did not improve from 0.51804
32/32 [=======] - 8s 261ms/step - loss: 0.2332 - acc: 0.9218 - val loss: 0.5214 - val ac
c: 0.8190
Epoch 24/50
Epoch 24: val loss improved from 0.51804 to 0.47747, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
INFO:tensorflow:Assets written to: models/DenseNet121/assets
32/32 [=========] - 35s 1s/step - loss: 0.2259 - acc: 0.9209 - val loss: 0.4775 - val acc:
0.8326
Epoch 25/50
Epoch 25: val loss did not improve from 0.47747
32/32 [====
      c: 0.7971
Epoch 26/50
Epoch 26: val_loss did not improve from 0.47747
32/32 [=========] - 8s 264ms/step - loss: 0.2990 - acc: 0.8916 - val loss: 0.4852 - val ac
c: 0.8421
Epoch 27/50
Epoch 27: val_loss did not improve from 0.47747
c: 0.8232
Epoch 28/50
Epoch 28: val loss did not improve from 0.47747
32/32 [=======] - 8s 265ms/step - loss: 0.2582 - acc: 0.9131 - val loss: 0.4904 - val ac
c: 0.8410
Epoch 29/50
Epoch 29: val loss did not improve from 0.47747
c: 0.8232
Epoch 30/50
32/32 [=============] - ETA: 0s - loss: 0.2714 - acc: 0.9089
Epoch 30: val_loss did not improve from 0.47747
c: 0.8389
Epoch 31/50
Epoch 31: val loss did not improve from 0.47747
32/32 [===
            :========] - 8s 268ms/step - loss: 0.2255 - acc: 0.9268 - val_loss: 0.5929 - val_ac
c: 0.8117
Epoch 32/50
32/32 [==:
           Epoch 32: val loss improved from 0.47747 to 0.47249, saving model to models/DenseNet121
INFO:tensorflow:Assets written to: models/DenseNet121/assets
```

```
INFO:tensorflow:Assets written to: models/DenseNet121/assets
32/32 [========] - 37s 1s/step - loss: 0.2335 - acc: 0.9189 - val loss: 0.4725 - val acc:
0.8441
Epoch 33/50
32/32 [==
         ========= ] - ETA: Os - loss: 0.2505 - acc: 0.9131
Epoch 33: val loss did not improve from 0.47249
32/32 [=======] - 8s 239ms/step - loss: 0.2505 - acc: 0.9131 - val loss: 0.4756 - val ac
Epoch 34/50
      32/32 [==
Epoch 34: val loss did not improve from 0.47249
32/32 [==
       c: 0.8305
Epoch 35/50
Epoch 35: val loss did not improve from 0.47249
c: 0.8295
Epoch 36/50
Epoch 36: val loss did not improve from 0.47249
c: 0.8295
Epoch 37/50
Epoch 37: val loss did not improve from 0.47249
32/32 [========] - 9s 272ms/step - loss: 0.1994 - acc: 0.9336 - val loss: 0.5376 - val ac
c: 0.8305
Epoch 38/50
32/32 [===
           ========] - ETA: 0s - loss: 0.1911 - acc: 0.9307
Epoch 38: val_loss did not improve from 0.47249
c: 0.8358
Epoch 39/50
Epoch 39: val loss did not improve from 0.47249
c: 0.8054
Epoch 40/50
32/32 [============] - ETA: 0s - loss: 0.1662 - acc: 0.9434
Epoch 40: val loss did not improve from 0.47249
c: 0.8274
Epoch 41/50
Epoch 41: val_loss did not improve from 0.47249
32/32 [==========] - 9s 279ms/step - loss: 0.1921 - acc: 0.9337 - val loss: 0.6453 - val ac
c: 0.7992
Epoch 42/50
32/32 [=====
      Epoch 42: val loss did not improve from 0.47249
c: 0.8274
Epoch 43/50
Epoch 43: val loss did not improve from 0.47249
      32/32 [=====
c: 0.8180
Epoch 44/50
Epoch 44: val loss did not improve from 0.47249
c: 0.8295
Epoch 45/50
Epoch 45: val_loss did not improve from 0.47249
32/32 [===
      c: 0.7939
Epoch 46/50
31/32 [===
      =====================>.] - ETA: Os - loss: 0.2573 - acc: 0.9083
Epoch 46: val_loss did not improve from 0.47249
32/32 [=======] - 9s 272ms/step - loss: 0.2535 - acc: 0.9092 - val loss: 0.6066 - val ac
c: 0.7887
Epoch 47/50
Epoch 47: val loss did not improve from 0.47249
c: 0.8169
Epoch 48/50
       32/32 [===
Epoch 48: val loss did not improve from 0.47249
c: 0.8285
Epoch 49/50
Epoch 49: val loss did not improve from 0.47249
```

32/32 [=== c: 0.8169

```
Epoch 50/50
32/32 [=========] - ETA: 0s - loss: 0.1996 - acc: 0.9336
Epoch 50: val_loss did not improve from 0.47249
32/32 [==========] - 8s 263ms/step - loss: 0.1996 - acc: 0.9336 - val_loss: 0.4909 - val_acc: 0.8536
```

Obtenemos métricas para el dataset de validación

```
In [70]: # Tamaño entrada
         input\_shape = (224, 224, 3)
         # Callbacks y guardado de grádicas
         callbacks = get_callbacks("DenseNet121")
         # Cargar el modelo preentrenado de DenseNet121 sin incluir las capas densas (fully connected)
         base_model = DenseNet121(weights='imagenet', include_top=False, input_shape=input_shape)
         # Congelar las capas preentrenadas
         for layer in base model.layers:
             layer.trainable = False
         # Crear un nuevo modelo Sequential
         model = Sequential()
         # Agregar la base preentrenada de DenseNet121
         model.add(base_model)
         # Añadir capas adicionales para la clasificación
         model.add(Flatten())
         model.add(Dense(128, activation='relu'))
         model.add(Dense(12, activation='softmax'))
         # Compilar el modelo
         model.compile(optimizer='adam', loss='categorical crossentropy', metrics=['acc'])
         # Mostrar la arquitectura del modelo
         model.summary()
         # Sacar prediccion
         model.load weights("models/DenseNet121")
         prediction = model.predict(valid tfdataset)
         Model: "sequential 20"
         Layer (type)
                                     Output Shape
                                                              Param #
         _____
          densenet121 (Functional)
                                   (None, 7, 7, 1024)
                                                              7037504
          flatten 8 (Flatten)
                                    (None, 50176)
          dense 26 (Dense)
                                     (None, 128)
                                                              6422656
         dense 27 (Dense)
                                     (None, 12)
                                                              1548
         Total params: 13461708 (51.35 MB)
         Trainable params: 6424204 (24.51 MB)
         Non-trainable params: 7037504 (26.85 MB)
         2023-11-26 20:36:44.491259: W tensorflow/core/util/tensor slice reader.cc:98] Could not open models/DenseNet121
         : FAILED PRECONDITION: models/DenseNet121; Is a directory: perhaps your file is in a different file format and
         you need to use a different restore operator?
         30/30 [===
                                      ======] - 6s 159ms/step
In [71]: # Mostrar classification report
         print("Classification DenseNet121")
```

print classification report(df dataset valid["label"].to numpy(), prediction)

```
Classification DenseNet121
Classification Report:
               precision
                            recall f1-score support
                   0.79
           0
                              0.87
                                        0.83
                                                    131
                   0.86
           1
                              0.88
                                        0.87
                                                    123
                   0.82
                              0.92
                                        0.87
                                                    104
           3
                   0.90
                              0.94
                                        0.92
                                                    100
                              0.87
           4
                   0.92
                                        0.90
                                                    95
           5
                   0.91
                              0.94
                                        0.92
                                                     78
           6
                   0.91
                              0.88
                                        0.89
                                                     77
                   0.96
                              0.78
                                        0.86
           7
                                                     58
           8
                   0.55
                              0.53
                                        0.54
                                                     53
           9
                   0.81
                              0.62
                                        0.70
                                                     47
                              0.58
          10
                   0.87
                                        0.69
                                                     45
                   0.77
                              0.96
                                        0.85
                                                    45
          11
                                        0.84
                                                    956
    accuracy
                   0.84
                              0.81
                                        0.82
                                                    956
   macro avg
weighted avg
                   0.85
                              0.84
                                        0.84
                                                    956
```

VGG16

```
In [35]: from tensorflow.keras.applications import VGG16
         # Tamaño entrada
         input shape = (224, 224, 3)
         # Callbacks y guardado de grádicas
         callbacks = get_callbacks("VGG16")
         # Cargar el modelo preentrenado de DenseNet121 sin incluir las capas densas (fully connected)
         base model = VGG16(weights='imagenet', include top=False, input shape=input shape)
         # Congelar las capas preentrenadas
         for layer in base model.layers:
             layer.trainable = False
         # Crear un nuevo modelo Sequential
         model = Sequential()
         # Agregar la base preentrenada de VGG16
         model.add(base model)
         # Añadir capas adicionales para la clasificación
         model.add(Flatten())
         model.add(Dense(128, activation='relu'))
         model.add(Dense(12, activation='softmax'))
         # Compilar el modelo
         model.compile(optimizer='adam', loss='categorical crossentropy', metrics=['acc'])
         # Mostrar la arquitectura del modelo
         model.summary()
         history = model.fit(
             train_tfdataset.repeat(),
             steps_per_epoch = 32,
             epochs = 50,
             validation_data = valid_tfdataset,
             callbacks = callbacks
```

Model: "sequential_13"

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14714688
flatten_1 (Flatten)	(None, 25088)	0
dense_2 (Dense)	(None, 128)	3211392
dense_3 (Dense)	(None, 12)	1548
Total params: 17927628 (68.3 Trainable params: 3212940 (1 Non-trainable params: 147146	2.26 MB)	

Epoch 1/50
32/32 [========] - ETA: 0s - loss: 2.9209 -

INFO:tensorflow:Assets written to: models/VGG16/assets

```
INFO:tensorflow:Assets written to: models/VGG16/assets
cc: 0.2939
Epoch 2/50
Epoch 2: val loss improved from 2.10990 to 1.86994, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [=========] - 12s 370ms/step - loss: 2.0168 - acc: 0.3096 - val_loss: 1.8699 - val a
cc: 0.4153
Epoch 3/50
                       =====] - ETA: Os - loss: 1.8000 - acc: 0.4043
32/32 [==
Epoch 3: val loss improved from 1.86994 to 1.81284, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [==========] - 11s 336ms/step - loss: 1.8000 - acc: 0.4043 - val loss: 1.8128 - val a
cc: 0.3672
Epoch 4/50
                      ======] - ETA: Os - loss: 1.6699 - acc: 0.4416
32/32 [==:
Epoch 4: val_loss improved from 1.81284 to 1.62056, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [=========] - 10s 311ms/step - loss: 1.6699 - acc: 0.4416 - val loss: 1.6206 - val a
cc: 0.4404
Epoch 5/50
32/32 [====
                ========] - ETA: 0s - loss: 1.4949 - acc: 0.5039
Epoch 5: val loss improved from 1.62056 to 1.45303, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [===========] - 11s 354ms/step - loss: 1.4949 - acc: 0.5039 - val_loss: 1.4530 - val_a
cc: 0.5481
Epoch 6/50
32/32 [=============] - ETA: 0s - loss: 1.5012 - acc: 0.5107
Epoch 6: val loss improved from 1.45303 to 1.38130, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
cc: 0.5429
Epoch 7/50
32/32 [=============] - ETA: 0s - loss: 1.3280 - acc: 0.5781
Epoch 7: val_loss improved from 1.38130 to 1.31300, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
cc: 0.5565
Epoch 8/50
32/32 [============] - ETA: 0s - loss: 1.2789 - acc: 0.5703
Epoch 8: val_loss improved from 1.31300 to 1.25748, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
cc: 0.5868
Epoch 9/50
32/32 [=========] - ETA: 0s - loss: 1.2034 - acc: 0.5957
Epoch 9: val loss improved from 1.25748 to 1.20577, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
cc: 0.6109
Epoch 10/50
Epoch 10: val loss improved from 1.20577 to 1.13618, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
c: 0.6119
Epoch 11/50
Epoch 11: val loss improved from 1.13618 to 1.03131, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
```

```
32/32 [==========] - 11s 338ms/step - loss: 1.0796 - acc: 0.6377 - val loss: 1.0313 - val a
cc: 0.6642
Epoch 12/50
Epoch 12: val loss did not improve from 1.03131
c: 0.6339
Epoch 13/50
Epoch 13: val loss did not improve from 1.03131
c: 0.6119
Epoch 14/50
Epoch 14: val loss improved from 1.03131 to 1.02398, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [============] - 10s 319ms/step - loss: 0.9892 - acc: 0.6768 - val_loss: 1.0240 - val a
cc: 0.6517
Epoch 15/50
Epoch 15: val loss improved from 1.02398 to 0.98896, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [=========] - 11s 347ms/step - loss: 0.9496 - acc: 0.6921 - val loss: 0.9890 - val a
cc: 0.6496
Epoch 16/50
Epoch 16: val loss did not improve from 0.98896
        32/32 [===
cc: 0.6684
Epoch 17/50
```

```
2023-11-26 19:59:05.281453: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
1073741824 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:05.281524: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 1073741824
2023-11-26 19:59:05.440712: E tensorflow/compiler/xla/stream executor/cuda/cuda_driver.cc:809] failed to alloc
966367744 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:05.440769: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 966367744
2023-11-26 19:59:05.599051: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
869731072 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:05.599099: W ./tensorflow/compiler/xla/stream executor/device host allocator.h:52] could not a
llocate pinned host memory of size: 869731072
2023-11-26 19:59:05.742493: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
782758144 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:05.742555: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 782758144
2023-11-26 19:59:05.884883: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
704482304 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:05.884935: W ./tensorflow/compiler/xla/stream executor/device host allocator.h:52] could not a
llocate pinned host memory of size: 704482304
2023-11-26 19:59:06.037201: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
634034176 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:06.037257: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 634034176
2023-11-26 19:59:06.171800: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
570630912 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:06.171856: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 570630912
2023-11-26 19:59:06.300902: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
513568000 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:06.300946: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 513568000
2023-11-26 19:59:06.429364: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
462211328 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:06.429422: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 462211328
2023-11-26 19:59:06.598022: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
415990272 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:06.598066: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 415990272 2023-11-26 19:59:06.737634: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
374391296 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:06.737689: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 374391296
2023-11-26 19:59:06.901586: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
336952320 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:06.901637: W ./tensorflow/compiler/xla/stream executor/device host allocator.h:52] could not a
llocate pinned host memory of size: 336952320
2023-11-26 19:59:07.052610: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
303257088 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:07.052658: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 303257088
2023-11-26 19:59:07.172372: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
272931584 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:07.172420: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 272931584
2023-11-26 19:59:07.311385: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
245638656 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:07.311448: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 245638656
2023-11-26 19:59:07.455055: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
221074944 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:07.455109: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 221074944
2023-11-26 19:59:07.586139: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
198967552 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:07.586190: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 198967552
2023-11-26 19:59:07.722924: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
179070976 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:07.722981: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 179070976
2023-11-26 19:59:07.868829: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
161164032 bytes on host: CUDA ERROR OUT OF MEMORY: out of memory
2023-11-26 19:59:07.869001: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a llocate pinned host memory of size: 161164032
2023-11-26 19:59:08.002276: E tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:809] failed to alloc
145047808 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:08.002329: W ./tensorflow/compiler/xla/stream_executor/device_host_allocator.h:52] could not a
llocate pinned host memory of size: 145047808
2023-11-26 19:59:08.135124: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:809] failed to alloc
130543104 bytes on host: CUDA_ERROR_OUT_OF_MEMORY: out of memory
2023-11-26 19:59:08.135169: W ./tensorflow/compiler/xla/stream executor/device host allocator.h:52] could not a
llocate pinned host memory of size: 130543104
Epoch 17: val loss improved from 0.98896 to 0.94046, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
```

INFO:tensorflow:Assets written to: models/VGG16/assets

```
32/32 [==========] - 13s 433ms/step - loss: 0.8813 - acc: 0.7100 - val loss: 0.9405 - val a
cc: 0.6872
Epoch 18/50
              ========= ] - ETA: Os - loss: 0.8699 - acc: 0.7256
32/32 [==
Epoch 18: val loss improved from 0.94046 to 0.86848, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [=========] - 9s 297ms/step - loss: 0.8699 - acc: 0.7256 - val_loss: 0.8685 - val_ac
c: 0.6935
Epoch 19/50
32/32 [============] - ETA: 0s - loss: 0.7864 - acc: 0.7634
Epoch 19: val_loss did not improve from 0.86848
c: 0.6851
Epoch 20/50
Epoch 20: val loss improved from 0.86848 to 0.84392, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
          cc: 0.7301
Epoch 21/50
Epoch 21: val loss did not improve from 0.84392
32/32 [===
            =========] - 8s 254ms/step - loss: 0.7733 - acc: 0.7578 - val loss: 0.8445 - val ac
c: 0.7123
Epoch 22/50
32/32 [==
                  =======] - ETA: 0s - loss: 0.7629 - acc: 0.7559
Epoch 22: val loss improved from 0.84392 to 0.82143, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
cc: 0.7312
Epoch 23/50
32/32 [==
                  ======] - ETA: Os - loss: 0.6726 - acc: 0.7762
Epoch 23: val loss improved from 0.82143 to 0.79773, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [==========] - 11s 352ms/step - loss: 0.6726 - acc: 0.7762 - val_loss: 0.7977 - val_a
cc: 0.7207
Epoch 24/50
                      ===] - ETA: 0s - loss: 0.7315 - acc: 0.7646
32/32 [==
Epoch 24: val loss did not improve from 0.79773
32/32 [============= ] - 9s 275ms/step - loss: 0.7315 - acc: 0.7646 - val loss: 0.8084 - val ac
c: 0.7176
Epoch 25/50
Epoch 25: val loss improved from 0.79773 to 0.76293, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
c: 0.7385
Epoch 26/50
Epoch 26: val_loss did not improve from 0.76293
32/32 [===
                    c: 0.7374
Epoch 27/50
Epoch 27: val loss did not improve from 0.76293
c: 0.6925
Epoch 28/50
Epoch 28: val loss did not improve from 0.76293
c: 0.6914
Epoch 29/50
Epoch 29: val loss improved from 0.76293 to 0.74871, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
```

```
32/32 [===========] - 10s 327ms/step - loss: 0.6259 - acc: 0.7998 - val loss: 0.7487 - val a
cc: 0.7427
Epoch 30/50
Epoch 30: val loss did not improve from 0.74871
c: 0.7113
Epoch 31/50
Epoch 31: val loss did not improve from 0.74871
32/32 [=============] - 10s 304ms/step - loss: 0.5783 - acc: 0.8164 - val loss: 0.8804 - val a
cc: 0.6946
Epoch 32/50
Epoch 32: val loss did not improve from 0.74871
c: 0.6726
Epoch 33/50
Epoch 33: val loss did not improve from 0.74871
32/32 [=======] - 8s 256ms/step - loss: 0.6408 - acc: 0.7930 - val loss: 0.7654 - val ac
c: 0.7312
Epoch 34/50
32/32 [===
            ======== ] - ETA: 0s - loss: 0.6262 - acc: 0.7980
Epoch 34: val loss did not improve from 0.74871
c: 0.7165
Epoch 35/50
Epoch 35: val loss improved from 0.74871 to 0.71677, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [=========] - 10s 328ms/step - loss: 0.6261 - acc: 0.7900 - val_loss: 0.7168 - val_a
cc: 0.7437
Epoch 36/50
Epoch 36: val loss did not improve from 0.71677
        32/32 [==
c: 0.7207
Epoch 37/50
Epoch 37: val loss did not improve from 0.71677
c: 0.7458
Epoch 38/50
Epoch 38: val loss did not improve from 0.71677
c: 0.7270
Epoch 39/50
       32/32 [===
Epoch 39: val loss did not improve from 0.71677
32/32 [=======] - 9s 275ms/step - loss: 0.5294 - acc: 0.8301 - val loss: 0.7299 - val ac
c: 0.7490
Epoch 40/50
Epoch 40: val_loss did not improve from 0.71677
c: 0.7197
Epoch 41/50
Epoch 41: val loss did not improve from 0.71677
32/32 [============== ] - 9s 279ms/step - loss: 0.5901 - acc: 0.8168 - val loss: 0.7856 - val ac
c: 0.7333
Epoch 42/50
32/32 [=========] - ETA: 0s - loss: 0.5469 - acc: 0.8271
Epoch 42: val loss did not improve from 0.71677
c: 0.7448
Fnoch 43/50
Epoch 43: val_loss did not improve from 0.71677
32/32 [=======] - 12s 400ms/step - loss: 0.5503 - acc: 0.8145 - val loss: 0.8011 - val a
cc: 0.7270
Epoch 44/50
           ========] - ETA: 0s - loss: 0.4901 - acc: 0.8496
32/32 [===
Epoch 44: val loss improved from 0.71677 to 0.68165, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
```

INFO:tensorflow:Assets written to: models/VGG16/assets

```
32/32 [==========] - 10s 308ms/step - loss: 0.4901 - acc: 0.8496 - val loss: 0.6817 - val a
cc: 0.7720
Epoch 45/50
Epoch 45: val loss did not improve from 0.68165
c: 0.7531
Epoch 46/50
                 ======] - ETA: 0s - loss: 0.4587 - acc: 0.8486
32/32 [==
Epoch 46: val loss improved from 0.68165 to 0.65698, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [===
               ========] - 15s 470ms/step - loss: 0.4587 - acc: 0.8486 - val loss: 0.6570 - val a
cc: 0.7626
Epoch 47/50
Epoch 47: val_loss improved from 0.65698 to 0.64047, saving model to models/VGG16
INFO:tensorflow:Assets written to: models/VGG16/assets
INFO:tensorflow:Assets written to: models/VGG16/assets
32/32 [===========] - 10s 330ms/step - loss: 0.4966 - acc: 0.8291 - val loss: 0.6405 - val a
cc: 0.7688
Fnoch 48/50
Epoch 48: val_loss did not improve from 0.64047
c: 0.7249
Epoch 49/50
        32/32 [====
Epoch 49: val_loss did not improve from 0.64047
c: 0.7144
Epoch 50/50
Epoch 50: val loss did not improve from 0.64047
c: 0.7646
```

Obtenemos métricas para el dataset de validación

```
In [65]: # Cargar modelos en memoria
         # Tamaño entrada
         input shape = (224, 224, 3)
         # Callbacks y guardado de grádicas
         callbacks = get_callbacks("VGG16")
         # Cargar el modelo preentrenado de DenseNet121 sin incluir las capas densas (fully connected)
         base model = VGG16(weights='imagenet', include top=False, input shape=input shape)
         # Congelar las capas preentrenadas
         for layer in base model.layers:
             layer.trainable = False
         # Crear un nuevo modelo Sequential
         model = Sequential()
         # Agregar la base preentrenada de VGG16
         model.add(base model)
         # Añadir capas adicionales para la clasificación
         model.add(Flatten())
         model.add(Dense(128, activation='relu'))
         model.add(Dense(12, activation='softmax'))
         # Compilar el modelo
         model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['acc'])
         # Sacar prediccion
         model.load weights("models/VGG16")
         model.summary()
         prediction = model.predict(valid tfdataset)
```

Model: "sequential_17"		
Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14714688
flatten_5 (Flatten)	(None, 25088)	0
dense_20 (Dense)	(None, 128)	3211392
dense_21 (Dense)	(None, 12)	1548
Total params: 17927628 (6 Trainable params: 3212940 Non-trainable params: 147	(12.26 MB)	
2023-11-26 20:33:55.15580	7: W tensorflow/core/util	

r.cc:98] Could not open models/VGG16: FAIL ED_PRECONDITION: models/VGG16; Is a directory: perhaps your file is in a different file format and you need to use a different restore operator?

30/30 [=========] - 5s 155ms/step

```
In [66]: # Mostrar classification report
         print("Classification VGG16")
         print classification report(df dataset valid["label"].to numpy(), prediction)
```

Classification VGG16 Classification Report:

	precision	recall	f1-score	support
0	0.66	0.84	0.74	131
1	0.88	0.78	0.83	123
2	0.74	0.72	0.73	104
3	0.92	0.97	0.94	100
4	0.72	0.72	0.72	95
5	0.92	0.85	0.88	78
6	0.80	0.88	0.84	77
7	0.80	0.74	0.77	58
8	0.61	0.47	0.53	53
9	0.54	0.60	0.57	47
10	0.69	0.56	0.62	45
11	0.92	0.76	0.83	45
accuracy			0.77	956
macro avg	0.77	0.74	0.75	956
weighted avg	0.77	0.77	0.77	956

EfficientNetV2 B0 (ImageNet Weights)

```
In [42]:
         def load efficienNetV2(
             num_classes = int,
             imagenet weights = bool,
         ):
             Loads EfficientNetV2 model with customized top layers for a specific number of classes.
                num classes (int): Number of classes for the final classification layer.
             Returns:
             tf.keras.models.Model: An instance of the EfficientNetV2 model with customized top layers.
             if imagenet_weights:
                 print(' > Loading iamgenet weights')
                 weights_name = 'imagenet'
             else:
                 weights_name =None
             redEffi = effnV2.EfficientNetV2B0(
                 include top=False,
                 weights=weights_name,
                 input_shape=(224, 224, 3),
             x = redEffi.outputs[0]
             x_{mean} = GlobalAveragePooling2D()(x)
             x max = GlobalMaxPooling2D()(x)
             x = Concatenate()([x_mean, x_max])
             x = Dropout(0.3)(x)
             x = Dense(num classes, activation = 'softmax')(x)
             model = tf.keras.models.Model(
                     inputs=redEffi.inputs, outputs=[x])
             return model
```

```
In []: # Cargamos el modelo de prueba
model = load_efficienNetV2(
    num_classes=12,
    imagenet_weights=True)

model.summary()

model.compile(
    loss = tf.keras.losses.categorical_crossentropy,
    optimizer = tf.keras.optimizers.Adam(0.0001),
    metrics = ['acc'])

callbacks = get_callbacks("efficintB0_cw_imagenet")

# Entrenamos el modelo
history = model.fit(train_tfdataset.repeat(),
    validation_data=valid_tfdataset,
    class_weight = class_weight_dict,
    epochs = 50,
    steps_per_epoch=50,
    callbacks = callbacks,
    verbose=1)
```

> Loading iamgenet weights
Model: "model_2"

Layer (type)	Output Shape	Param #	Connected to
input_3 (InputLayer)	[(None, 224, 224, 3)]	0	[]
rescaling_2 (Rescaling)	(None, 224, 224, 3)	0	['input_3[0][0]']
normalization_2 (Normalization)	(None, 224, 224, 3)	0	['rescaling_2[0][0]']
stem_conv (Conv2D)	(None, 112, 112, 32)	864	['normalization_2[0][0]']
<pre>stem_bn (BatchNormalizatio n)</pre>	(None, 112, 112, 32)	128	['stem_conv[0][0]']
<pre>stem_activation (Activatio n)</pre>	(None, 112, 112, 32)	0	['stem_bn[0][0]']
block1a_project_conv (Conv 2D)	(None, 112, 112, 16)	4608	['stem_activation[0][0]']
<pre>block1a_project_bn (BatchN ormalization)</pre>	(None, 112, 112, 16)	64	['block1a_project_conv[0][0]']
<pre>block1a_project_activation (Activation)</pre>	(None, 112, 112, 16)	0	['block1a_project_bn[0][0]']
block2a_expand_conv (Conv2 D)	(None, 56, 56, 64)	9216	['block1a_project_activation[0][0]']
<pre>block2a_expand_bn (BatchNo rmalization)</pre>	(None, 56, 56, 64)	256	['block2a_expand_conv[0][0]']
<pre>block2a_expand_activation (Activation)</pre>	(None, 56, 56, 64)	0	['block2a_expand_bn[0][0]']
block2a_project_conv (Conv 2D)	(None, 56, 56, 32)	2048	<pre>['block2a_expand_activation[0] [0]']</pre>
<pre>block2a_project_bn (BatchN ormalization)</pre>	(None, 56, 56, 32)	128	['block2a_project_conv[0][0]']
block2b_expand_conv (Conv2 D)	(None, 56, 56, 128)	36864	['block2a_project_bn[0][0]']
<pre>block2b_expand_bn (BatchNo rmalization)</pre>	(None, 56, 56, 128)	512	['block2b_expand_conv[0][0]']
<pre>block2b_expand_activation (Activation)</pre>	(None, 56, 56, 128)	0	['block2b_expand_bn[0][0]']
block2b_project_conv (Conv 2D)	(None, 56, 56, 32)	4096	<pre>['block2b_expand_activation[0] [0]']</pre>
<pre>block2b_project_bn (BatchN ormalization)</pre>	(None, 56, 56, 32)	128	['block2b_project_conv[0][0]']
block2b_drop (Dropout)	(None, 56, 56, 32)	0	['block2b_project_bn[0][0]']
block2b_add (Add)	(None, 56, 56, 32)	0	['block2b_drop[0][0]', 'block2a_project_bn[0][0]']
block3a_expand_conv (Conv2	(None, 28, 28, 128)	36864	['block2b_add[0][0]']

D)

<pre>block3a_expand_bn (BatchNo rmalization)</pre>	(None, 28, 28, 128)	512	['block3a_expand_conv[0][0]']
<pre>block3a_expand_activation (Activation)</pre>	(None, 28, 28, 128)	0	['block3a_expand_bn[0][0]']
block3a_project_conv (Conv 2D)	(None, 28, 28, 48)	6144	<pre>['block3a_expand_activation[0] [0]']</pre>
<pre>block3a_project_bn (BatchN ormalization)</pre>	(None, 28, 28, 48)	192	['block3a_project_conv[0][0]']
block3b_expand_conv (Conv2 D)	(None, 28, 28, 192)	82944	['block3a_project_bn[0][0]']
<pre>block3b_expand_bn (BatchNo rmalization)</pre>	(None, 28, 28, 192)	768	['block3b_expand_conv[0][0]']
<pre>block3b_expand_activation (Activation)</pre>	(None, 28, 28, 192)	0	['block3b_expand_bn[0][0]']
<pre>block3b_project_conv (Conv 2D)</pre>	(None, 28, 28, 48)	9216	<pre>['block3b_expand_activation[0] [0]']</pre>
<pre>block3b_project_bn (BatchN ormalization)</pre>	(None, 28, 28, 48)	192	['block3b_project_conv[0][0]']
block3b_drop (Dropout)	(None, 28, 28, 48)	0	['block3b_project_bn[0][0]']
block3b_add (Add)	(None, 28, 28, 48)	0	['block3b_drop[0][0]', 'block3a_project_bn[0][0]']
block4a_expand_conv (Conv2 D)	(None, 28, 28, 192)	9216	['block3b_add[0][0]']
<pre>block4a_expand_bn (BatchNo rmalization)</pre>	(None, 28, 28, 192)	768	['block4a_expand_conv[0][0]']
<pre>block4a_expand_activation (Activation)</pre>	(None, 28, 28, 192)	0	['block4a_expand_bn[0][0]']
block4a_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 192)	1728	<pre>['block4a_expand_activation[0] [0]']</pre>
<pre>block4a_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 192)	768	['block4a_dwconv2[0][0]']
<pre>block4a_activation (Activa tion)</pre>	(None, 14, 14, 192)	0	['block4a_bn[0][0]']
<pre>block4a_se_squeeze (Global AveragePooling2D)</pre>	(None, 192)	0	['block4a_activation[0][0]']
<pre>block4a_se_reshape (Reshap e)</pre>	(None, 1, 1, 192)	0	['block4a_se_squeeze[0][0]']
block4a_se_reduce (Conv2D)	(None, 1, 1, 12)	2316	['block4a_se_reshape[0][0]']
block4a_se_expand (Conv2D)	(None, 1, 1, 192)	2496	['block4a_se_reduce[0][0]']
<pre>block4a_se_excite (Multipl y)</pre>	(None, 14, 14, 192)	0	<pre>['block4a_activation[0][0]', 'block4a_se_expand[0][0]']</pre>
<pre>block4a_project_conv (Conv 2D)</pre>	(None, 14, 14, 96)	18432	['block4a_se_excite[0][0]']
<pre>block4a_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 96)	384	['block4a_project_conv[0][0]']
block4b_expand_conv (Conv2 D)	(None, 14, 14, 384)	36864	['block4a_project_bn[0][0]']
12 141 15 15	(None, 14, 14, 384)	1536	['block4b_expand_conv[0][0]']
<pre>block4b_expand_bn (BatchNo rmalization)</pre>			
	(None, 14, 14, 384)	0	['block4b_expand_bn[0][0]']
<pre>rmalization) block4b_expand_activation</pre>	(None, 14, 14, 384) (None, 14, 14, 384)	0 3456	<pre>['block4b_expand_bn[0][0]'] ['block4b_expand_activation[0] [0]']</pre>
<pre>rmalization) block4b_expand_activation (Activation) block4b_dwconv2 (Depthwise)</pre>			['block4b_expand_activation[0]

block4b_se_squeeze (Global AveragePooling2D)	(None, 384)	0	['block4b_activation[0][0]']
<pre>block4b_se_reshape (Reshap e)</pre>	(None, 1, 1, 384)	0	['block4b_se_squeeze[0][0]']
block4b_se_reduce (Conv2D)	(None, 1, 1, 24)	9240	['block4b_se_reshape[0][0]']
block4b_se_expand (Conv2D)	(None, 1, 1, 384)	9600	['block4b_se_reduce[0][0]']
<pre>block4b_se_excite (Multipl y)</pre>	(None, 14, 14, 384)	0	['block4b_activation[0][0]', 'block4b_se_expand[0][0]']
<pre>block4b_project_conv (Conv 2D)</pre>	(None, 14, 14, 96)	36864	['block4b_se_excite[0][0]']
<pre>block4b_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 96)	384	['block4b_project_conv[0][0]']
block4b_drop (Dropout)	(None, 14, 14, 96)	0	['block4b_project_bn[0][0]']
block4b_add (Add)	(None, 14, 14, 96)	0	['block4b_drop[0][0]', 'block4a_project_bn[0][0]']
block4c_expand_conv (Conv2 D)	(None, 14, 14, 384)	36864	['block4b_add[0][0]']
<pre>block4c_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 384)	1536	['block4c_expand_conv[0][0]']
<pre>block4c_expand_activation (Activation)</pre>	(None, 14, 14, 384)	0	['block4c_expand_bn[0][0]']
<pre>block4c_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 384)	3456	['block4c_expand_activation[0] [0]']
<pre>block4c_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 384)	1536	['block4c_dwconv2[0][0]']
<pre>block4c_activation (Activa tion)</pre>	(None, 14, 14, 384)	0	['block4c_bn[0][0]']
<pre>block4c_se_squeeze (Global AveragePooling2D)</pre>	(None, 384)	0	['block4c_activation[0][0]']
<pre>block4c_se_reshape (Reshap e)</pre>	(None, 1, 1, 384)	0	['block4c_se_squeeze[0][0]']
block4c_se_reduce (Conv2D)	(None, 1, 1, 24)	9240	['block4c_se_reshape[0][0]']
block4c_se_expand (Conv2D)	(None, 1, 1, 384)	9600	['block4c_se_reduce[0][0]']
<pre>block4c_se_excite (Multipl y)</pre>	(None, 14, 14, 384)	0	<pre>['block4c_activation[0][0]', 'block4c_se_expand[0][0]']</pre>
block4c_project_conv (Conv 2D)	(None, 14, 14, 96)	36864	['block4c_se_excite[0][0]']
<pre>block4c_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 96)	384	['block4c_project_conv[0][0]']
block4c_drop (Dropout)	(None, 14, 14, 96)	0	['block4c_project_bn[0][0]']
block4c_add (Add)	(None, 14, 14, 96)	0	['block4c_drop[0][0]', 'block4b_add[0][0]']
block5a_expand_conv (Conv2 D)	(None, 14, 14, 576)	55296	['block4c_add[0][0]']
<pre>block5a_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 576)	2304	['block5a_expand_conv[0][0]']
<pre>block5a_expand_activation (Activation)</pre>	(None, 14, 14, 576)	0	['block5a_expand_bn[0][0]']
<pre>block5a_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 576)	5184	<pre>['block5a_expand_activation[0] [0]']</pre>
block5a_bn (BatchNormaliza tion)	(None, 14, 14, 576)	2304	['block5a_dwconv2[0][0]']
block5a_activation (Activa tion)	(None, 14, 14, 576)	0	['block5a_bn[0][0]']
block5a_se_squeeze (Global AveragePooling2D)	(None, 576)	0	['block5a_activation[0][0]']
<pre>block5a_se_reshape (Reshap e)</pre>	(None, 1, 1, 576)	0	['block5a_se_squeeze[0][0]']

block5a_se_reduce (Conv2D)	(None, 1, 1, 24)	13848	['block5a_se_reshape[0][0]']
block5a_se_expand (Conv2D)	(None, 1, 1, 576)	14400	['block5a_se_reduce[0][0]']
<pre>block5a_se_excite (Multipl y)</pre>	(None, 14, 14, 576)	0	<pre>['block5a_activation[0][0]', 'block5a_se_expand[0][0]']</pre>
block5a_project_conv (Conv 2D)	(None, 14, 14, 112)	64512	['block5a_se_excite[0][0]']
<pre>block5a_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 112)	448	['block5a_project_conv[0][0]']
block5b_expand_conv (Conv2 D)	(None, 14, 14, 672)	75264	['block5a_project_bn[0][0]']
<pre>block5b_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block5b_expand_conv[0][0]']
<pre>block5b_expand_activation (Activation)</pre>	(None, 14, 14, 672)	0	['block5b_expand_bn[0][0]']
<pre>block5b_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 672)	6048	<pre>['block5b_expand_activation[0] [0]']</pre>
<pre>block5b_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 672)	2688	['block5b_dwconv2[0][0]']
<pre>block5b_activation (Activa tion)</pre>	(None, 14, 14, 672)	0	['block5b_bn[0][0]']
<pre>block5b_se_squeeze (Global AveragePooling2D)</pre>	(None, 672)	0	['block5b_activation[0][0]']
<pre>block5b_se_reshape (Reshap e)</pre>	(None, 1, 1, 672)	0	['block5b_se_squeeze[0][0]']
block5b_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5b_se_reshape[0][0]']
block5b_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5b_se_reduce[0][0]']
<pre>block5b_se_excite (Multipl y)</pre>	(None, 14, 14, 672)	0	<pre>['block5b_activation[0][0]', 'block5b_se_expand[0][0]']</pre>
block5b_project_conv (Conv 2D)	(None, 14, 14, 112)	75264	['block5b_se_excite[0][0]']
<pre>block5b_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 112)	448	['block5b_project_conv[0][0]']
block5b_drop (Dropout)	(None, 14, 14, 112)	0	['block5b_project_bn[0][0]']
block5b_add (Add)	(None, 14, 14, 112)	0	['block5b_drop[0][0]', 'block5a_project_bn[0][0]']
<pre>block5c_expand_conv (Conv2 D)</pre>	(None, 14, 14, 672)	75264	['block5b_add[0][0]']
<pre>block5c_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block5c_expand_conv[0][0]']
<pre>block5c_expand_activation (Activation)</pre>	(None, 14, 14, 672)	0	['block5c_expand_bn[0][0]']
<pre>block5c_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 672)	6048	<pre>['block5c_expand_activation[0] [0]']</pre>
<pre>block5c_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 672)	2688	['block5c_dwconv2[0][0]']
<pre>block5c_activation (Activa tion)</pre>	(None, 14, 14, 672)	0	['block5c_bn[0][0]']
<pre>block5c_se_squeeze (Global AveragePooling2D)</pre>	(None, 672)	0	['block5c_activation[0][0]']
<pre>block5c_se_reshape (Reshap e)</pre>	(None, 1, 1, 672)	0	['block5c_se_squeeze[0][0]']
block5c_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5c_se_reshape[0][0]']
block5c_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5c_se_reduce[0][0]']
<pre>block5c_se_excite (Multipl y)</pre>	(None, 14, 14, 672)	0	<pre>['block5c_activation[0][0]', 'block5c_se_expand[0][0]']</pre>
<pre>block5c_project_conv (Conv 2D)</pre>	(None, 14, 14, 112)	75264	['block5c_se_excite[0][0]']

blockSc add (Add) (None, 14, 14, 112) 8 ['blockSc add[0][0]'] 1 1 1 1 1 1 1 1 1		(None, 14, 14, 112)	448	['block5c_project_conv[0][0]']
	block5c_drop (Dropout)	(None, 14, 14, 112)	0	['block5c_project_bn[0][0]']
DiockSd expand bn (BatchNo (None, 14, 14, 672) 2688 ['blockSd expand conv[0][0]'] ramalization (None, 14, 14, 672) 0 ['blockSd expand bn[0][0]'] (Activation) DiockSd dynamid activation (None, 14, 14, 672) 0 ['blockSd expand activation[0] Conv20) DiockSd bn (BatchNormaliza (None, 14, 14, 672) 2688 ['blockSd_buconv2[0][0]'] DiockSd sh (BatchNormaliza (None, 14, 14, 672) 0 ['blockSd_buconv2[0][0]'] DiockSd se squeeze (Global (None, 672) 0 ['blockSd_buf][0]'] DiockSd_se_greenape (Reshap (None, 1, 1, 672) 0 ['blockSd_se_squeeze[0][0]'] DiockSd_se_greenape (Reshap (None, 1, 1, 672) 0 ['blockSd_se_greenape (Reshap (None, 14, 14, 672) 0 ['blockSd_se_greenape (Reshap (None, 14, 14, 112) 0 ['blockSd_project_conv(0][0]'] DiockSd_project_bn (BatchNo (None, 14, 14, 112) 0 ['blockSd_project_bn (0][0]'] DiockSd_project_bn (Reshap (None, 14, 14, 672) 75264 ['blockSd_add[0][0]'] DiockSd_project_bn (Reshap (None, 14, 14, 672) 2688 ['blockSe_grand_conv[0][0]'] DiockSe_grand_activation (Reshap (None, 14, 14, 672) 2688 ['blockSe_grand_activation (Reshap (None, 14, 14, 672) 0 ['blockSe_grand_activation (0][0]'] DiockSe_grand_activation (None, 14, 14, 672) 0 ['blockSe_grand_activation (0][0]'] DiockSe_grand_activation (None, 14, 14, 672) 0 ['blockSe_grand_activation (0][0]'] DiockSe_grand_conv2 (Dione, 11, 1, 672) 0 ['blockSe_grand_activation (0][0]'] DiockSe_grand_conv2 (None, 14, 14, 672) 0 ['blockSe_grand_conv2 (0][0]'] DiockSe_grand_conv2 (None, 14, 14, 672) 0 ['blockSe_grand_conv2	block5c_add (Add)	(None, 14, 14, 112)	0	
		(None, 14, 14, 672)	75264	['block5c_add[0][0]']
DickSd_dwconv2 (Depthwise (None, 14, 14, 672) 6048 ['blockSd_expand_activation[0] Clorv2D Clorv2D	_ · _ ·	(None, 14, 14, 672)	2688	['block5d_expand_conv[0][0]']
		(None, 14, 14, 672)	0	['block5d_expand_bn[0][0]']
block5d activation (Activa (None, 14, 14, 672) 0 ['block5d bn[0][0]']		(None, 14, 14, 672)	6048	
	- '	(None, 14, 14, 672)	2688	['block5d_dwconv2[0][0]']
BlockSd_se_reshape (Reshap (None, 1, 1, 672) 0	-	(None, 14, 14, 672)	0	['block5d_bn[0][0]']
	'	(None, 672)	0	['block5d_activation[0][0]']
block5d_se_expand (Conv2D) (None, 1, 1, 672) 19488 ['block5d_se_educe[0][0]'] block5d_se_excite (Multipl (None, 14, 14, 672) 0 ['block5d_se_expand[0][0]'] block5d_se_excite (Multipl (None, 14, 14, 112) 75264 ['block5d_se_expand[0][0]'] block5d_project_conv (Conv (None, 14, 14, 112) 75264 ['block5d_se_excite[0][0]'] block5d_project_bn (BatchN (None, 14, 14, 112) 0 ['block5d_project_conv[0][0]'] block5d_drop (Dropout) (None, 14, 14, 112) 0 ['block5d_drop[0][0]', 'block5d_add (Add) (None, 14, 14, 112) 0 ['block5d_drop[0][0]', 'block5d_add[0][0]'] block5e_expand_conv (Conv2 (None, 14, 14, 672) 75264 ['block5d_add[0][0]'] block5e_expand_bn (BatchNo (None, 14, 14, 672) 2688 ['block5e_expand_conv[0][0]'] chlock5e_expand_activation (None, 14, 14, 672) 6048 ['block5e_expand_activation[0][0]'] chlock5e_expand_activation (None, 14, 14, 672) 2688 ['block5e_expand_activation[0][0]'] chlock5e_expand_activation (Activa (None, 14, 14, 672) 2688 ['block5e_expand_activation[0][0]'] chlock5e_expand_activation (Activa (None, 14, 14, 672) 0 ['block5e_expand_activation[0][0]'] chlock5e_expend_expand_activation (Activa (None, 14, 14, 672) 0 ['block5e_expand_expand_activation[0][0]'] chlock5e_expend_expand		(None, 1, 1, 672)	0	['block5d_se_squeeze[0][0]']
BlockSd_se_excite (Multiple (None, 14, 14, 672) BlockSd_se_expand[0][0]',	block5d_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5d_se_reshape[0][0]']
block5d_project_conv (Conv (None, 14, 14, 112) 75264 ['block5d_se_excite[0][0]']	block5d_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5d_se_reduce[0][0]']
DiockSd project_bn (BatchN (None, 14, 14, 112)		(None, 14, 14, 672)	0	
block5d_drop (Dropout) (None, 14, 14, 112) 0 ['block5d_project_bn[0][0]'] block5d_add (Add) (None, 14, 14, 112) 0 ['block5d_drop[0][0]', 'block5c_add[0][0]'] block5e_expand_conv (Conv2 (None, 14, 14, 672) 75264 ['block5d_add[0][0]'] block5e_expand_bn (BatchNo (None, 14, 14, 672) 2688 ['block5e_expand_conv[0][0]'] malization) block5e_expand_activation (None, 14, 14, 672) 0 ['block5e_expand_bn[0][0]'] block5e_expand_activation (None, 14, 14, 672) 6048 ['block5e_expand_bn[0][0]'] block5e_bn (BatchNormaliza (None, 14, 14, 672) 2688 ['block5e_expand_activation[0][0]'] block5e_bn (BatchNormaliza (None, 14, 14, 672) 2688 ['block5e_dwconv2[0][0]'] block5e_activation (Activa (None, 14, 14, 672) 0 ['block5e_bn[0][0]'] block5e_se_squeeze (Global (None, 672) 0 ['block5e_se_reshape[0][0]'] block5e_se_reshape (Reshap (None, 1, 1, 672) 0 ['block5e_se_reshape[0][0]'] block5e_se_expand (Conv2D) (None, 1, 1, 672) 19488 ['block5e_se_reshape[0][0]'] block5e_se_excite (Multipl (None, 14, 14, 672) 19488 ['block5e_se_reduce[0][0]'] block5e_se_excite (Multipl (None, 14, 14, 672) 19488 ['block5e_se_excite[0][0]'] block5e_se_project_conv (Conv (None, 14, 14, 112) 75264 ['block5e_se_excite[0][0]'] block5e_project_conv (Conv (None, 14, 14, 112) 75264 ['block5e_se_excite[0][0]']		(None, 14, 14, 112)	75264	['block5d_se_excite[0][0]']
block5d_add (Add) (None, 14, 14, 112) 0 ['block5d_drop[0][0]', 'block5c_add[0][0]'] block5e_expand_conv (Conv2 (None, 14, 14, 672) 75264 ['block5d_add[0][0]'] block5e_expand_bn (BatchNo (None, 14, 14, 672) 2688 ['block5e_expand_conv[0][0]'] rmalization) (None, 14, 14, 672) 0 ['block5e_expand_bn[0][0]'] block5e_expand_activation (Activation) (None, 14, 14, 672) 6048 ['block5e_expand_activation[0][0]'] block5e_bn (BatchNormaliza (None, 14, 14, 672) 2688 ['block5e_dwconv2[0][0]'] block5e_activation (Activa (None, 14, 14, 672) 0 ['block5e_dwconv2[0][0]'] block5e_se_squeeze (Global AveragePooling2D) (None, 672) 0 ['block5e_activation[0][0]'] block5e_se_reshape (Reshap (None, 1, 1, 672) 0 ['block5e_se_reshape[0][0]'] block5e_se_expand (Conv2D) (None, 1, 1, 672) 19488 ['block5e_se_reduce[0][0]'] block5e_se_excite (Multipl (None, 14, 14, 672) 0 ['block5e_se_reduce[0][0]'] block5e_project_conv (Conv (None, 14, 14, 112) 75264 ['block5e_se_excite[0][0]'] block5e_project_bn (BatchN (None, 14, 14, 112) 75264 ['block5e_project_conv[0][0]']		(None, 14, 14, 112)	448	['block5d_project_conv[0][0]']
block5e_expand_conv (Conv2 (None, 14, 14, 672) 75264 ['block5d_add[0][0]']	block5d_drop (Dropout)	(None, 14, 14, 112)	0	['block5d_project_bn[0][0]']
D	block5d_add (Add)	(None, 14, 14, 112)	0	
block5e_expand_activation (Activation) (None, 14, 14, 672) 0 ['block5e_expand_bn[0][0]'] block5e_dwconv2 (Depthwise (None, 14, 14, 672) 6048 ['block5e_expand_activation[0] [0]'] block5e_bn (BatchNormaliza (None, 14, 14, 672) 2688 ['block5e_dwconv2[0][0]'] block5e_activation (Activa (None, 14, 14, 672) 0 ['block5e_bn[0][0]'] block5e_se_squeeze (Global (None, 672) 0 ['block5e_activation[0][0]'] AveragePooling2D) 0 ['block5e_se_squeeze[0][0]'] block5e_se_reshape (Reshap (None, 1, 1, 672) 0 ['block5e_se_squeeze[0][0]'] block5e_se_excide (Conv2D) (None, 1, 1, 672) 19488 ['block5e_se_reshape[0][0]'] block5e_se_excite (Multipl (None, 14, 14, 672) 0 ['block5e_se_excite[0][0]'] block5e_project_conv (Conv (None, 14, 14, 112) 75264 ['block5e_se_excite[0][0]'] block5e_project_bn (BatchN (None, 14, 14, 112) 448 ['block5e_project_conv[0][0]']		(None, 14, 14, 672)	75264	['block5d_add[0][0]']
Block5e_dwconv2 (Depthwise (None, 14, 14, 672) 6048		(None, 14, 14, 672)	2688	['block5e_expand_conv[0][0]']
Conv2D) [0]'] block5e_bn (BatchNormaliza (None, 14, 14, 672) tion) 2688 ['block5e_dwconv2[0][0]'] block5e_activation (Activa (None, 14, 14, 672) tion) 0 ['block5e_bn[0][0]'] block5e_se_squeeze (Global AveragePooling2D) (None, 672)		(None, 14, 14, 672)	0	['block5e_expand_bn[0][0]']
tion) block5e_activation (Activa (None, 14, 14, 672)		(None, 14, 14, 672)	6048	_ · _ · _
block5e_se_squeeze (Global (None, 672) AveragePooling2D) block5e_se_reshape (Reshap (None, 1, 1, 672) e) block5e_se_reduce (Conv2D) (None, 1, 1, 28) block5e_se_expand (Conv2D) (None, 1, 1, 672) block5e_se_expand (Conv2D) (None, 1, 1, 672) block5e_se_expand (Conv2D) (None, 1, 1, 672) block5e_se_excite (Multipl (None, 14, 14, 672) y) block5e_se_excite (Multipl (None, 14, 14, 112) block5e_project_conv (Conv (None, 14, 14, 112) block5e_project_bn (BatchN (None, 14, 14, 112) block5e_project_bn (BatchN (None, 14, 14, 112) block5e_project_conv[0][0]']		(None, 14, 14, 672)	2688	['block5e_dwconv2[0][0]']
AveragePooling2D) block5e_se_reshape (Reshap (None, 1, 1, 672)	-	(None, 14, 14, 672)	0	['block5e_bn[0][0]']
block5e_se_reduce (Conv2D) (None, 1, 1, 28)		(None, 672)	0	['block5e_activation[0][0]']
block5e_se_expand (Conv2D) (None, 1, 1, 672) 19488 ['block5e_se_reduce[0][0]'] block5e_se_excite (Multipl (None, 14, 14, 672) y) 0 ['block5e_activation[0][0]', 'block5e_se_expand[0][0]'] block5e_project_conv (Conv (None, 14, 14, 112) 75264 ['block5e_se_excite[0][0]'] block5e_project_bn (BatchN (None, 14, 14, 112) ormalization) 448 ['block5e_project_conv[0][0]']		(None, 1, 1, 672)	0	['block5e_se_squeeze[0][0]']
block5e_se_excite (Multipl (None, 14, 14, 672) 0 ['block5e_activation[0][0]', 'block5e_se_expand[0][0]'] block5e_project_conv (Conv (None, 14, 14, 112) 75264 ['block5e_se_excite[0][0]'] block5e_project_bn (BatchN (None, 14, 14, 112) 448 ['block5e_project_conv[0][0]'] ormalization)	block5e_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5e_se_reshape[0][0]']
y) 'block5e_se_expand[0][0]'] block5e_project_conv (Conv (None, 14, 14, 112) 75264 ['block5e_se_excite[0][0]'] 2D)	block5e_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5e_se_reduce[0][0]']
2D) block5e_project_bn (BatchN (None, 14, 14, 112)		(None, 14, 14, 672)	0	
ormalization)		(None, 14, 14, 112)	75264	['block5e_se_excite[0][0]']
block5e_drop (Dropout) (None, 14, 14, 112) 0 ['block5e_project_bn[0][0]']		(None, 14, 14, 112)	448	['block5e_project_conv[0][0]']
	block5e_drop (Dropout)	(None, 14, 14, 112)	0	['block5e_project_bn[0][0]']

block5e_add (Add)	(None, 14, 14, 112)	0	['block5e_drop[0][0]', 'block5d_add[0][0]']
block6a_expand_conv (Conv2 D)	(None, 14, 14, 672)	75264	_ ['block5e_add[0][0]']
<pre>block6a_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block6a_expand_conv[0][0]']
<pre>block6a_expand_activation (Activation)</pre>	(None, 14, 14, 672)	0	['block6a_expand_bn[0][0]']
block6a_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 672)	6048	<pre>['block6a_expand_activation[0] [0]']</pre>
<pre>block6a_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 672)	2688	['block6a_dwconv2[0][0]']
<pre>block6a_activation (Activa tion)</pre>	(None, 7, 7, 672)	Θ	['block6a_bn[0][0]']
<pre>block6a_se_squeeze (Global AveragePooling2D)</pre>	(None, 672)	0	['block6a_activation[0][0]']
<pre>block6a_se_reshape (Reshap e)</pre>	(None, 1, 1, 672)	0	['block6a_se_squeeze[0][0]']
block6a_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block6a_se_reshape[0][0]']
block6a_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block6a_se_reduce[0][0]']
<pre>block6a_se_excite (Multipl y)</pre>	(None, 7, 7, 672)	0	<pre>['block6a_activation[0][0]', 'block6a_se_expand[0][0]']</pre>
<pre>block6a_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	129024	['block6a_se_excite[0][0]']
<pre>block6a_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6a_project_conv[0][0]']
<pre>block6b_expand_conv (Conv2 D)</pre>	(None, 7, 7, 1152)	221184	['block6a_project_bn[0][0]']
<pre>block6b_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6b_expand_conv[0][0]']
<pre>block6b_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	0	['block6b_expand_bn[0][0]']
<pre>block6b_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6b_expand_activation[0] [0]']</pre>
<pre>block6b_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6b_dwconv2[0][0]']
<pre>block6b_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6b_bn[0][0]']
<pre>block6b_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	0	['block6b_activation[0][0]']
<pre>block6b_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6b_se_squeeze[0][0]']
<pre>block6b_se_reduce (Conv2D)</pre>	(None, 1, 1, 48)	55344	['block6b_se_reshape[0][0]']
block6b_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6b_se_reduce[0][0]']
<pre>block6b_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	<pre>['block6b_activation[0][0]', 'block6b_se_expand[0][0]']</pre>
<pre>block6b_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6b_se_excite[0][0]']
<pre>block6b_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6b_project_conv[0][0]']
block6b_drop (Dropout)	(None, 7, 7, 192)	0	['block6b_project_bn[0][0]']
block6b_add (Add)	(None, 7, 7, 192)	0	['block6b_drop[0][0]', 'block6a_project_bn[0][0]']
block6c_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6b_add[0][0]']
<pre>block6c_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6c_expand_conv[0][0]']

block6c_expand_activation (Activation)	(None, 7, 7, 1152)	0	['block6c_expand_bn[0][0]']
<pre>block6c_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6c_expand_activation[0] [0]']</pre>
<pre>block6c_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6c_dwconv2[0][0]']
<pre>block6c_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6c_bn[0][0]']
<pre>block6c_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	0	['block6c_activation[0][0]']
<pre>block6c_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6c_se_squeeze[0][0]']
block6c_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6c_se_reshape[0][0]']
block6c_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6c_se_reduce[0][0]']
<pre>block6c_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	<pre>['block6c_activation[0][0]', 'block6c_se_expand[0][0]']</pre>
<pre>block6c_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6c_se_excite[0][0]']
<pre>block6c_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6c_project_conv[0][0]']
block6c_drop (Dropout)	(None, 7, 7, 192)	0	['block6c_project_bn[0][0]']
block6c_add (Add)	(None, 7, 7, 192)	0	['block6c_drop[0][0]', 'block6b_add[0][0]']
<pre>block6d_expand_conv (Conv2 D)</pre>	(None, 7, 7, 1152)	221184	['block6c_add[0][0]']
<pre>block6d_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6d_expand_conv[0][0]']
<pre>block6d_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	0	['block6d_expand_bn[0][0]']
<pre>block6d_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6d_expand_activation[0] [0]']</pre>
<pre>block6d_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6d_dwconv2[0][0]']
<pre>block6d_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6d_bn[0][0]']
<pre>block6d_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	0	['block6d_activation[0][0]']
<pre>block6d_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6d_se_squeeze[0][0]']
block6d_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6d_se_reshape[0][0]']
block6d_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6d_se_reduce[0][0]']
<pre>block6d_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	<pre>['block6d_activation[0][0]', 'block6d_se_expand[0][0]']</pre>
<pre>block6d_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6d_se_excite[0][0]']
<pre>block6d_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6d_project_conv[0][0]']
block6d_drop (Dropout)	(None, 7, 7, 192)	0	['block6d_project_bn[0][0]']
block6d_add (Add)	(None, 7, 7, 192)	0	['block6d_drop[0][0]', 'block6c_add[0][0]']
block6e_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6d_add[0][0]']
<pre>block6e_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6e_expand_conv[0][0]']
<pre>block6e_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	0	['block6e_expand_bn[0][0]']
<pre>block6e_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6e_expand_activation[0] [0]']</pre>

<pre>block6e_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6e_dwconv2[0][0]']
<pre>block6e_activation (Activa tion)</pre>	(None, 7, 7, 1152)	Θ	['block6e_bn[0][0]']
block6e_se_squeeze (Global AveragePooling2D)	(None, 1152)	0	['block6e_activation[0][0]']
block6e_se_reshape (Reshap	(None, 1, 1, 1152)	Θ	['block6e_se_squeeze[0][0]']
block6e_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6e_se_reshape[0][0]']
block6e_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6e_se_reduce[0][0]']
<pre>block6e_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	['block6e_activation[0][0]', 'block6e_se_expand[0][0]']
<pre>block6e_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6e_se_excite[0][0]']
<pre>block6e_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6e_project_conv[0][0]']
<pre>block6e_drop (Dropout)</pre>	(None, 7, 7, 192)	0	['block6e_project_bn[0][0]']
block6e_add (Add)	(None, 7, 7, 192)	0	['block6e_drop[0][0]', 'block6d_add[0][0]']
<pre>block6f_expand_conv (Conv2 D)</pre>	(None, 7, 7, 1152)	221184	['block6e_add[0][0]']
<pre>block6f_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6f_expand_conv[0][0]']
<pre>block6f_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	Θ	['block6f_expand_bn[0][0]']
<pre>block6f_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6f_expand_activation[0] [0]']</pre>
<pre>block6f_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6f_dwconv2[0][0]']
<pre>block6f_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6f_bn[0][0]']
<pre>block6f_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	0	['block6f_activation[0][0]']
<pre>block6f_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6f_se_squeeze[0][0]']
block6f_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6f_se_reshape[0][0]']
block6f_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6f_se_reduce[0][0]']
<pre>block6f_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	<pre>['block6f_activation[0][0]', 'block6f_se_expand[0][0]']</pre>
<pre>block6f_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6f_se_excite[0][0]']
<pre>block6f_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6f_project_conv[0][0]']
block6f_drop (Dropout)	(None, 7, 7, 192)	0	['block6f_project_bn[0][0]']
block6f_add (Add)	(None, 7, 7, 192)	0	['block6f_drop[0][0]', 'block6e_add[0][0]']
block6g_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6f_add[0][0]']
<pre>block6g_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6g_expand_conv[0][0]']
<pre>block6g_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	Θ	['block6g_expand_bn[0][0]']
block6g_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 1152)	10368	<pre>['block6g_expand_activation[0] [0]']</pre>
<pre>block6g_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6g_dwconv2[0][0]']
block6g_activation (Activa	(None, 7, 7, 1152)	0	['block6g_bn[0][0]']

tion)			
<pre>block6g_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	0	['block6g_activation[0][0]']
<pre>block6g_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6g_se_squeeze[0][0]']
block6g_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6g_se_reshape[0][0]']
block6g_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6g_se_reduce[0][0]']
<pre>block6g_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	<pre>['block6g_activation[0][0]', 'block6g_se_expand[0][0]']</pre>
<pre>block6g_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6g_se_excite[0][0]']
<pre>block6g_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6g_project_conv[0][0]']
<pre>block6g_drop (Dropout)</pre>	(None, 7, 7, 192)	0	['block6g_project_bn[0][0]']
block6g_add (Add)	(None, 7, 7, 192)	0	['block6g_drop[0][0]', 'block6f_add[0][0]']
<pre>block6h_expand_conv (Conv2 D)</pre>	(None, 7, 7, 1152)	221184	['block6g_add[0][0]']
<pre>block6h_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6h_expand_conv[0][0]']
<pre>block6h_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	0	['block6h_expand_bn[0][0]']
<pre>block6h_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6h_expand_activation[0] [0]']</pre>
<pre>block6h_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6h_dwconv2[0][0]']
<pre>block6h_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6h_bn[0][0]']
<pre>block6h_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	0	['block6h_activation[0][0]']
<pre>block6h_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6h_se_squeeze[0][0]']
block6h_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6h_se_reshape[0][0]']
block6h_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6h_se_reduce[0][0]']
<pre>block6h_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	<pre>['block6h_activation[0][0]', 'block6h_se_expand[0][0]']</pre>
<pre>block6h_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6h_se_excite[0][0]']
<pre>block6h_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6h_project_conv[0][0]']
block6h_drop (Dropout)	(None, 7, 7, 192)	0	['block6h_project_bn[0][0]']
block6h_add (Add)	(None, 7, 7, 192)	0	['block6h_drop[0][0]', 'block6g_add[0][0]']
top_conv (Conv2D)	(None, 7, 7, 1280)	245760	['block6h_add[0][0]']
top_bn (BatchNormalization)	(None, 7, 7, 1280)	5120	['top_conv[0][0]']
$\begin{array}{ll} top_activation & (Activation \\) \end{array}$	(None, 7, 7, 1280)	0	['top_bn[0][0]']
<pre>global_average_pooling2d_2 (GlobalAveragePooling2D)</pre>	(None, 1280)	0	['top_activation[0][0]']
<pre>global_max_pooling2d_2 (Gl obalMaxPooling2D)</pre>	(None, 1280)	0	['top_activation[0][0]']

['global_average_pooling2d_2[0
][0]',
 'global_max_pooling2d_2[0][0]
']

['concatenate_2[0][0]']

0

concatenate_2 (Concatenate (None, 2560)
)

(None, 2560)

dropout_2 (Dropout)

```
dense 2 (Dense)
                      (None, 12)
                                            30732
                                                    ['dropout 2[0][0]']
Total params: 5950044 (22.70 MB)
Trainable params: 5889436 (22.47 MB)
Non-trainable params: 60608 (236.75 KB)
Epoch 1/50
                     =======] - ETA: Os - loss: 4.4812 - acc: 0.1256
50/50 [===
Epoch 1: val loss improved from inf to 3.59153, saving model to models\efficintB0 cw imagenet
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
            50/50 [===
: 0.1088
Epoch 2/50
50/50 [=====
         Epoch 2: val_loss improved from 3.59153 to 2.28200, saving model to models\efficintB0_cw_imagenet
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
50/50 [==
                  =======] - 130s 3s/step - loss: 2.9424 - acc: 0.2544 - val loss: 2.2820 - val acc
: 0.2981
Epoch 3/50
Epoch 3: val_loss improved from 2.28200 to 1.90023, saving model to models\efficintB0_cw_imagenet
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
50/50 [==
                  =========] - 129s 3s/step - loss: 2.0486 - acc: 0.3897 - val loss: 1.9002 - val acc
: 0.3630
Epoch 4/50
Epoch 4: val_loss improved from 1.90023 to 0.72259, saving model to models\efficintB0_cw_imagenet
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
50/50 [==============] - 130s 3s/step - loss: 1.5630 - acc: 0.5250 - val loss: 0.7226 - val acc
: 0.7458
Fnoch 5/50
Epoch 5: val loss did not improve from 0.72259
50/50 [=============] - 114s 2s/step - loss: 1.2692 - acc: 0.5984 - val loss: 1.0895 - val acc
: 0.6318
Epoch 6/50
50/50 [====
          Epoch 6: val loss did not improve from 0.72259
50/50 [===========] - 115s 2s/step - loss: 0.9674 - acc: 0.6819 - val_loss: 0.7446 - val acc
: 0.7395
Epoch 7/50
```

Epoch 9: val_loss improved from 0.72259 to 0.64226, saving model to models\efficintB0_cw_imagenet

 ${\tt Epoch 10: val_loss improved from 0.64226 to 0.34639, saving model to models \verb|\efficintB0_cw_imagenet| }$

Epoch 7: val loss did not improve from 0.72259

Epoch 8: val loss did not improve from 0.72259

INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets

INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets

50/50 [===: : 0.5628 Epoch 8/50

: 0.6977 Epoch 9/50

: 0.7699 Epoch 10/50

```
50/50 [============= ] - 131s 3s/step - loss: 0.5614 - acc: 0.7982 - val loss: 0.3464 - val acc
: 0.8734
Epoch 11/50
Epoch 11: val loss did not improve from 0.34639
: 0.7699
Epoch 12/50
        50/50 [=====
Epoch 12: val loss did not improve from 0.34639
: 0.7835
Epoch 13/50
Epoch 13: val loss did not improve from 0.34639
                     ====] - 118s 2s/step - loss: 0.4365 - acc: 0.8419 - val loss: 0.3903 - val acc
50/50 [==
: 0.8609
Epoch 14/50
50/50 [============] - ETA: 0s - loss: 0.3857 - acc: 0.8506
Epoch 14: val_loss did not improve from 0.34639
50/50 [===
       : 0.7029
Epoch 15/50
Epoch 15: val loss did not improve from 0.34639
: 0.8577
Epoch 16/50
50/50 [=====
         Epoch 16: val loss improved from 0.34639 to 0.33762, saving model to models\efficintB0 cw imagenet
INFO:tensorflow:Assets written to: models\efficintBO cw imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
50/50 [==
              :=========] - 134s 3s/step - loss: 0.3286 - acc: 0.8850 - val loss: 0.3376 - val acc
: 0.8808
Epoch 17/50
Epoch 17: val loss improved from 0.33762 to 0.20352, saving model to models\efficintB0 cw imagenet
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
50/50 [==
              :========] - 133s 3s/step - loss: 0.3153 - acc: 0.8758 - val loss: 0.2035 - val acc
: 0.9268
Epoch 18/50
Epoch 18: val loss did not improve from 0.20352
50/50 [===========] - 115s 2s/step - loss: 0.2895 - acc: 0.8944 - val_loss: 0.2415 - val acc
: 0.9142
Epoch 19/50
Epoch 19: val_loss did not improve from 0.20352
50/50 [============ ] - 115s 2s/step - loss: 0.3004 - acc: 0.8888 - val loss: 0.4898 - val acc
: 0.8264
Epoch 20/50
50/50 [===
         Epoch 20: val_loss improved from 0.20352 to 0.19889, saving model to models\efficintB0_cw_imagenet INFO:tensorflow:Assets written to: models\efficintB0_cw_imagenet\assets
INFO:tensorflow:Assets written to: models\efficintB0 cw imagenet\assets
```

```
50/50 [============ ] - 130s 3s/step - loss: 0.2390 - acc: 0.9061 - val loss: 0.1989 - val acc
: 0.9236
Epoch 21/50
Epoch 21: val loss did not improve from 0.19889
: 0.8243
Epoch 22/50
      50/50 [==
Epoch 22: val loss did not improve from 0.19889
: 0.8787
Epoch 23/50
Epoch 23: val loss did not improve from 0.19889
                ===] - 118s 2s/step - loss: 0.2463 - acc: 0.9056 - val loss: 0.3327 - val acc
50/50 [==
: 0.8849
Epoch 24/50
50/50 [=====
      Epoch 24: val loss did not improve from 0.19889
: 0.8860
Fnoch 25/50
Epoch 25: val_loss did not improve from 0.19889
50/50 [===
     : 0.8996
Epoch 26/50
          =========] - ETA: Os - loss: 0.2094 - acc: 0.9200
50/50 [===
Epoch 26: val loss did not improve from 0.19889
: 0.8870
Epoch 27/50
50/50 [==
             =======] - ETA: 0s - loss: 0.1950 - acc: 0.9206
Epoch 27: val loss did not improve from 0.19889
50/50 [=============] - 117s 2s/step - loss: 0.1950 - acc: 0.9206 - val loss: 0.5372 - val acc
: 0.8065
Epoch 28/50
Epoch 28: val loss did not improve from 0.19889
      : 0.8755
Epoch 29/50
50/50 [=====
      Epoch 29: val_loss did not improve from 0.19889
50/50 [==========] - 117s 2s/step - loss: 0.1788 - acc: 0.9325 - val loss: 0.2020 - val acc
: 0.9320
Epoch 30/50
Epoch 30: val loss did not improve from 0.19889
: 0.9132
Epoch 30: early stopping
```

Obtenemos métricas para el dataset de validación

> Loading iamgenet weights

Model: "model 9"

```
In [63]: # Cargar modelos en memoria
# Cargamos el modelo de prueba
model = load_efficienNetV2(
    num_classes=12,
    imagenet_weights=True)

model.summary()

model.compile(
    loss = tf.keras.losses.categorical_crossentropy,
    optimizer = tf.keras.optimizers.Adam(0.0001),
    metrics = ['acc'])

# Sacar prediccion
model.load_weights("models/efficintB0_cw_imagenet")
prediction = model.predict(valid_tfdataset)
```

```
Layer (type)
                            Output Shape
                                                         Param # Connected to
input 18 (InputLayer)
                           [(None, 224, 224, 3)]
rescaling_12 (Rescaling)
                            (None, 224, 224, 3)
                                                         0
                                                                   ['input_18[0][0]']
normalization 12 (Normaliz (None, 224, 224, 3)
                                                         0
                                                                   ['rescaling 12[0][0]']
ation)
stem conv (Conv2D)
                                                        864
                           (None, 112, 112, 32)
                                                                   ['normalization 12[0][0]']
```

stem_bn (BatchNormalization)	(None, 112, 112, 32)	128	['stem_conv[0][0]']
<pre>stem_activation (Activatio n)</pre>	(None, 112, 112, 32)	0	['stem_bn[0][0]']
block1a_project_conv (Conv 2D)	(None, 112, 112, 16)	4608	['stem_activation[0][0]']
<pre>blockla_project_bn (BatchN ormalization)</pre>	(None, 112, 112, 16)	64	['blockla_project_conv[0][0]']
<pre>blockla_project_activation (Activation)</pre>	(None, 112, 112, 16)	0	['blockla_project_bn[0][0]']
block2a_expand_conv (Conv2 D)	(None, 56, 56, 64)	9216	['blockla_project_activation[0][0]']
<pre>block2a_expand_bn (BatchNo rmalization)</pre>	(None, 56, 56, 64)	256	['block2a_expand_conv[0][0]']
<pre>block2a_expand_activation (Activation)</pre>	(None, 56, 56, 64)	0	['block2a_expand_bn[0][0]']
<pre>block2a_project_conv (Conv 2D)</pre>	(None, 56, 56, 32)	2048	<pre>['block2a_expand_activation[0] [0]']</pre>
<pre>block2a_project_bn (BatchN ormalization)</pre>	(None, 56, 56, 32)	128	['block2a_project_conv[0][0]']
<pre>block2b_expand_conv (Conv2 D)</pre>	(None, 56, 56, 128)	36864	['block2a_project_bn[0][0]']
<pre>block2b_expand_bn (BatchNo rmalization)</pre>	(None, 56, 56, 128)	512	['block2b_expand_conv[0][0]']
<pre>block2b_expand_activation (Activation)</pre>	(None, 56, 56, 128)	0	['block2b_expand_bn[0][0]']
<pre>block2b_project_conv (Conv 2D)</pre>	(None, 56, 56, 32)	4096	<pre>['block2b_expand_activation[0] [0]']</pre>
<pre>block2b_project_bn (BatchN ormalization)</pre>	(None, 56, 56, 32)	128	['block2b_project_conv[0][0]']
block2b_drop (Dropout)	(None, 56, 56, 32)	0	['block2b_project_bn[0][0]']
block2b_add (Add)	(None, 56, 56, 32)	0	['block2b_drop[0][0]', 'block2a_project_bn[0][0]']
block3a_expand_conv (Conv2 D)	(None, 28, 28, 128)	36864	['block2b_add[0][0]']
<pre>block3a_expand_bn (BatchNo rmalization)</pre>	(None, 28, 28, 128)	512	['block3a_expand_conv[0][0]']
<pre>block3a_expand_activation (Activation)</pre>	(None, 28, 28, 128)	0	['block3a_expand_bn[0][0]']
<pre>block3a_project_conv (Conv 2D)</pre>	(None, 28, 28, 48)	6144	<pre>['block3a_expand_activation[0] [0]']</pre>
<pre>block3a_project_bn (BatchN ormalization)</pre>	(None, 28, 28, 48)	192	['block3a_project_conv[0][0]']
<pre>block3b_expand_conv (Conv2 D)</pre>	(None, 28, 28, 192)	82944	['block3a_project_bn[0][0]']
<pre>block3b_expand_bn (BatchNo rmalization)</pre>	(None, 28, 28, 192)	768	['block3b_expand_conv[0][0]']
<pre>block3b_expand_activation (Activation)</pre>	(None, 28, 28, 192)	0	['block3b_expand_bn[0][0]']
<pre>block3b_project_conv (Conv 2D)</pre>	(None, 28, 28, 48)	9216	<pre>['block3b_expand_activation[0] [0]']</pre>
<pre>block3b_project_bn (BatchN ormalization)</pre>	(None, 28, 28, 48)	192	['block3b_project_conv[0][0]']
block3b_drop (Dropout)	(None, 28, 28, 48)	0	['block3b_project_bn[0][0]']
block3b_add (Add)	(None, 28, 28, 48)	0	['block3b_drop[0][0]', 'block3a_project_bn[0][0]']
			prock2a_brolecr_pu[6][6]]
block4a_expand_conv (Conv2 D)	(None, 28, 28, 192)	9216	['block3b_add[0][0]']

rmalization)			
block4a_expand_activation (Activation)	(None, 28, 28, 192)	0	['block4a_expand_bn[0][0]']
block4a_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 192)	1728	<pre>['block4a_expand_activation[0] [0]']</pre>
block4a_bn (BatchNormaliza tion)	(None, 14, 14, 192)	768	['block4a_dwconv2[0][0]']
<pre>block4a_activation (Activa tion)</pre>	(None, 14, 14, 192)	Θ	['block4a_bn[0][0]']
block4a_se_squeeze (Global AveragePooling2D)	(None, 192)	0	['block4a_activation[0][0]']
block4a_se_reshape (Reshap	(None, 1, 1, 192)	0	['block4a_se_squeeze[0][0]']
block4a_se_reduce (Conv2D)	(None, 1, 1, 12)	2316	['block4a_se_reshape[0][0]']
block4a_se_expand (Conv2D)	(None, 1, 1, 192)	2496	['block4a_se_reduce[0][0]']
<pre>block4a_se_excite (Multipl y)</pre>	(None, 14, 14, 192)	0	['block4a_activation[0][0]', 'block4a_se_expand[0][0]']
block4a_project_conv (Conv 2D)	(None, 14, 14, 96)	18432	['block4a_se_excite[0][0]']
<pre>block4a_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 96)	384	['block4a_project_conv[0][0]']
block4b_expand_conv (Conv2 D)	(None, 14, 14, 384)	36864	['block4a_project_bn[0][0]']
<pre>block4b_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 384)	1536	['block4b_expand_conv[0][0]']
block4b_expand_activation (Activation)	(None, 14, 14, 384)	0	['block4b_expand_bn[0][0]']
block4b_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 384)	3456	<pre>['block4b_expand_activation[0] [0]']</pre>
<pre>block4b_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 384)	1536	['block4b_dwconv2[0][0]']
<pre>block4b_activation (Activa tion)</pre>	(None, 14, 14, 384)	0	['block4b_bn[0][0]']
<pre>block4b_se_squeeze (Global AveragePooling2D)</pre>	(None, 384)	0	['block4b_activation[0][0]']
<pre>block4b_se_reshape (Reshap e)</pre>	(None, 1, 1, 384)	0	['block4b_se_squeeze[0][0]']
block4b_se_reduce (Conv2D)	(None, 1, 1, 24)	9240	['block4b_se_reshape[0][0]']
block4b_se_expand (Conv2D)	(None, 1, 1, 384)	9600	['block4b_se_reduce[0][0]']
<pre>block4b_se_excite (Multipl y)</pre>	(None, 14, 14, 384)	0	<pre>['block4b_activation[0][0]', 'block4b_se_expand[0][0]']</pre>
<pre>block4b_project_conv (Conv 2D)</pre>	(None, 14, 14, 96)	36864	['block4b_se_excite[0][0]']
<pre>block4b_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 96)	384	['block4b_project_conv[0][0]']
block4b_drop (Dropout)	(None, 14, 14, 96)	0	['block4b_project_bn[0][0]']
block4b_add (Add)	(None, 14, 14, 96)	0	['block4b_drop[0][0]', 'block4a_project_bn[0][0]']
block4c_expand_conv (Conv2 D)	(None, 14, 14, 384)	36864	['block4b_add[0][0]']
<pre>block4c_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 384)	1536	['block4c_expand_conv[0][0]']
<pre>block4c_expand_activation (Activation)</pre>	(None, 14, 14, 384)	0	['block4c_expand_bn[0][0]']
<pre>block4c_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 384)	3456	<pre>['block4c_expand_activation[0] [0]']</pre>
<pre>block4c_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 384)	1536	['block4c_dwconv2[0][0]']

<pre>block4c_activation (Activa tion)</pre>	(None, 14, 14, 384)	0	['block4c_bn[0][0]']
<pre>block4c_se_squeeze (Global AveragePooling2D)</pre>	(None, 384)	0	['block4c_activation[0][0]']
<pre>block4c_se_reshape (Reshap e)</pre>	(None, 1, 1, 384)	0	['block4c_se_squeeze[0][0]']
block4c_se_reduce (Conv2D)	(None, 1, 1, 24)	9240	['block4c_se_reshape[0][0]']
block4c_se_expand (Conv2D)	(None, 1, 1, 384)	9600	['block4c_se_reduce[0][0]']
<pre>block4c_se_excite (Multipl y)</pre>	(None, 14, 14, 384)	0	<pre>['block4c_activation[0][0]', 'block4c_se_expand[0][0]']</pre>
<pre>block4c_project_conv (Conv 2D)</pre>	(None, 14, 14, 96)	36864	['block4c_se_excite[0][0]']
<pre>block4c_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 96)	384	['block4c_project_conv[0][0]']
block4c_drop (Dropout)	(None, 14, 14, 96)	0	['block4c_project_bn[0][0]']
block4c_add (Add)	(None, 14, 14, 96)	0	['block4c_drop[0][0]', 'block4b_add[0][0]']
<pre>block5a_expand_conv (Conv2 D)</pre>	(None, 14, 14, 576)	55296	['block4c_add[0][0]']
<pre>block5a_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 576)	2304	['block5a_expand_conv[0][0]']
<pre>block5a_expand_activation (Activation)</pre>	(None, 14, 14, 576)	0	['block5a_expand_bn[0][0]']
<pre>block5a_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 576)	5184	<pre>['block5a_expand_activation[0] [0]']</pre>
<pre>block5a_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 576)	2304	['block5a_dwconv2[0][0]']
<pre>block5a_activation (Activa tion)</pre>	(None, 14, 14, 576)	0	['block5a_bn[0][0]']
<pre>block5a_se_squeeze (Global AveragePooling2D)</pre>	(None, 576)	0	['block5a_activation[0][0]']
<pre>block5a_se_reshape (Reshap e)</pre>	(None, 1, 1, 576)	0	['block5a_se_squeeze[0][0]']
block5a_se_reduce (Conv2D)	(None, 1, 1, 24)	13848	['block5a_se_reshape[0][0]']
block5a_se_expand (Conv2D)	(None, 1, 1, 576)	14400	['block5a_se_reduce[0][0]']
<pre>block5a_se_excite (Multipl y)</pre>	(None, 14, 14, 576)	0	['block5a_activation[0][0]', 'block5a_se_expand[0][0]']
<pre>block5a_project_conv (Conv 2D)</pre>	(None, 14, 14, 112)	64512	['block5a_se_excite[0][0]']
<pre>block5a_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 112)	448	['block5a_project_conv[0][0]']
<pre>block5b_expand_conv (Conv2 D)</pre>	(None, 14, 14, 672)	75264	['block5a_project_bn[0][0]']
<pre>block5b_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block5b_expand_conv[0][0]']
<pre>block5b_expand_activation (Activation)</pre>	(None, 14, 14, 672)	0	['block5b_expand_bn[0][0]']
<pre>block5b_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 672)	6048	<pre>['block5b_expand_activation[0] [0]']</pre>
<pre>block5b_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 672)	2688	['block5b_dwconv2[0][0]']
<pre>block5b_activation (Activa tion)</pre>	(None, 14, 14, 672)	0	['block5b_bn[0][0]']
block5b_se_squeeze (Global AveragePooling2D)	(None, 672)	0	['block5b_activation[0][0]']
block5b_se_reshape (Reshap e)	(None, 1, 1, 672)	0	['block5b_se_squeeze[0][0]']

block5b_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5b_se_reshape[0][0]']
block5b_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5b_se_reduce[0][0]']
<pre>block5b_se_excite (Multipl y)</pre>	(None, 14, 14, 672)	0	<pre>['block5b_activation[0][0]', 'block5b_se_expand[0][0]']</pre>
<pre>block5b_project_conv (Conv 2D)</pre>	(None, 14, 14, 112)	75264	['block5b_se_excite[0][0]']
<pre>block5b_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 112)	448	['block5b_project_conv[0][0]']
block5b_drop (Dropout)	(None, 14, 14, 112)	0	['block5b_project_bn[0][0]']
block5b_add (Add)	(None, 14, 14, 112)	0	['block5b_drop[0][0]', 'block5a_project_bn[0][0]']
block5c_expand_conv (Conv2 D)	(None, 14, 14, 672)	75264	['block5b_add[0][0]']
<pre>block5c_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block5c_expand_conv[0][0]']
<pre>block5c_expand_activation (Activation)</pre>	(None, 14, 14, 672)	Θ	['block5c_expand_bn[0][0]']
block5c_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 672)	6048	<pre>['block5c_expand_activation[0] [0]']</pre>
<pre>block5c_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 672)	2688	['block5c_dwconv2[0][0]']
<pre>block5c_activation (Activa tion)</pre>	(None, 14, 14, 672)	0	['block5c_bn[0][0]']
<pre>block5c_se_squeeze (Global AveragePooling2D)</pre>	(None, 672)	0	['block5c_activation[0][0]']
<pre>block5c_se_reshape (Reshap e)</pre>	(None, 1, 1, 672)	0	['block5c_se_squeeze[0][0]']
block5c_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5c_se_reshape[0][0]']
block5c_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5c_se_reduce[0][0]']
<pre>block5c_se_excite (Multipl y)</pre>	(None, 14, 14, 672)	0	<pre>['block5c_activation[0][0]', 'block5c_se_expand[0][0]']</pre>
<pre>block5c_project_conv (Conv 2D)</pre>	(None, 14, 14, 112)	75264	['block5c_se_excite[0][0]']
<pre>block5c_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 112)	448	['block5c_project_conv[0][0]']
block5c_drop (Dropout)	(None, 14, 14, 112)	0	['block5c_project_bn[0][0]']
block5c_add (Add)	(None, 14, 14, 112)	0	['block5c_drop[0][0]', 'block5b_add[0][0]']
block5d_expand_conv (Conv2 D)	(None, 14, 14, 672)	75264	['block5c_add[0][0]']
<pre>block5d_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block5d_expand_conv[0][0]']
<pre>block5d_expand_activation (Activation)</pre>	(None, 14, 14, 672)	0	['block5d_expand_bn[0][0]']
<pre>block5d_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 672)	6048	<pre>['block5d_expand_activation[0] [0]']</pre>
<pre>block5d_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 672)	2688	['block5d_dwconv2[0][0]']
<pre>block5d_activation (Activa tion)</pre>	(None, 14, 14, 672)	0	['block5d_bn[0][0]']
block5d_se_squeeze (Global AveragePooling2D)	(None, 672)	0	['block5d_activation[0][0]']
block5d_se_reshape (Reshap e)	(None, 1, 1, 672)	0	['block5d_se_squeeze[0][0]']
block5d_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5d_se_reshape[0][0]']
block5d_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5d_se_reduce[0][0]']
block5d se excite (Multipl	(None, 14, 14, 672)	0	<pre>['block5d activation[0][0]',</pre>
, , , , , ,	. ,		

y)			'block5d_se_expand[0][0]']
block5d_project_conv (Conv 2D)	(None, 14, 14, 112)	75264	['block5d_se_excite[0][0]']
<pre>block5d_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 112)	448	['block5d_project_conv[0][0]']
block5d_drop (Dropout)	(None, 14, 14, 112)	0	['block5d_project_bn[0][0]']
block5d_add (Add)	(None, 14, 14, 112)	0	['block5d_drop[0][0]', 'block5c_add[0][0]']
block5e_expand_conv (Conv2 D)	(None, 14, 14, 672)	75264	['block5d_add[0][0]']
<pre>block5e_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block5e_expand_conv[0][0]']
<pre>block5e_expand_activation (Activation)</pre>	(None, 14, 14, 672)	0	['block5e_expand_bn[0][0]']
<pre>block5e_dwconv2 (Depthwise Conv2D)</pre>	(None, 14, 14, 672)	6048	<pre>['block5e_expand_activation[0] [0]']</pre>
<pre>block5e_bn (BatchNormaliza tion)</pre>	(None, 14, 14, 672)	2688	['block5e_dwconv2[0][0]']
<pre>block5e_activation (Activa tion)</pre>	(None, 14, 14, 672)	0	['block5e_bn[0][0]']
<pre>block5e_se_squeeze (Global AveragePooling2D)</pre>	(None, 672)	0	['block5e_activation[0][0]']
<pre>block5e_se_reshape (Reshap e)</pre>	(None, 1, 1, 672)	0	['block5e_se_squeeze[0][0]']
block5e_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5e_se_reshape[0][0]']
block5e_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5e_se_reduce[0][0]']
<pre>block5e_se_excite (Multipl y)</pre>	(None, 14, 14, 672)	0	<pre>['block5e_activation[0][0]', 'block5e_se_expand[0][0]']</pre>
<pre>block5e_project_conv (Conv 2D)</pre>	(None, 14, 14, 112)	75264	['block5e_se_excite[0][0]']
<pre>block5e_project_bn (BatchN ormalization)</pre>	(None, 14, 14, 112)	448	['block5e_project_conv[0][0]']
block5e_drop (Dropout)	(None, 14, 14, 112)	0	['block5e_project_bn[0][0]']
block5e_add (Add)	(None, 14, 14, 112)	0	['block5e_drop[0][0]', 'block5d_add[0][0]']
block6a_expand_conv (Conv2 D)	(None, 14, 14, 672)	75264	['block5e_add[0][0]']
<pre>block6a_expand_bn (BatchNo rmalization)</pre>	(None, 14, 14, 672)	2688	['block6a_expand_conv[0][0]']
<pre>block6a_expand_activation (Activation)</pre>	(None, 14, 14, 672)	0	['block6a_expand_bn[0][0]']
<pre>block6a_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 672)	6048	<pre>['block6a_expand_activation[0] [0]']</pre>
<pre>block6a_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 672)	2688	['block6a_dwconv2[0][0]']
<pre>block6a_activation (Activa tion)</pre>	(None, 7, 7, 672)	0	['block6a_bn[0][0]']
block6a_se_squeeze (Global AveragePooling2D)	(None, 672)	0	['block6a_activation[0][0]']
<pre>block6a_se_reshape (Reshap e)</pre>	(None, 1, 1, 672)	0	['block6a_se_squeeze[0][0]']
block6a_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block6a_se_reshape[0][0]']
block6a_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block6a_se_reduce[0][0]']
<pre>block6a_se_excite (Multipl y)</pre>	(None, 7, 7, 672)	0	<pre>['block6a_activation[0][0]', 'block6a_se_expand[0][0]']</pre>
<pre>block6a_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	129024	['block6a_se_excite[0][0]']

<pre>block6a_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6a_project_conv[0][0]']
block6b_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6a_project_bn[0][0]']
<pre>block6b_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6b_expand_conv[0][0]']
<pre>block6b_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	0	['block6b_expand_bn[0][0]']
block6b_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 1152)	10368	<pre>['block6b_expand_activation[0] [0]']</pre>
<pre>block6b_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6b_dwconv2[0][0]']
$\begin{array}{c} {\tt block6b_activation} \ \ ({\tt Activation}) \end{array}$	(None, 7, 7, 1152)	0	['block6b_bn[0][0]']
block6b_se_squeeze (Global AveragePooling2D)	(None, 1152)	0	['block6b_activation[0][0]']
<pre>block6b_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6b_se_squeeze[0][0]']
block6b_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6b_se_reshape[0][0]']
block6b_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6b_se_reduce[0][0]']
<pre>block6b_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	Θ	['block6b_activation[0][0]', 'block6b_se_expand[0][0]']
block6b_project_conv (Conv 2D)	(None, 7, 7, 192)	221184	['block6b_se_excite[0][0]']
<pre>block6b_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6b_project_conv[0][0]']
block6b_drop (Dropout)	(None, 7, 7, 192)	0	['block6b_project_bn[0][0]']
block6b_add (Add)	(None, 7, 7, 192)	0	['block6b_drop[0][0]', 'block6a_project_bn[0][0]']
block6c_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6b_add[0][0]']
<pre>block6c_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6c_expand_conv[0][0]']
<pre>block6c_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	0	['block6c_expand_bn[0][0]']
<pre>block6c_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6c_expand_activation[0] [0]']</pre>
<pre>block6c_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6c_dwconv2[0][0]']
<pre>block6c_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6c_bn[0][0]']
<pre>block6c_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	0	['block6c_activation[0][0]']
<pre>block6c_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6c_se_squeeze[0][0]']
block6c_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6c_se_reshape[0][0]']
block6c_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6c_se_reduce[0][0]']
<pre>block6c_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	['block6c_activation[0][0]', 'block6c_se_expand[0][0]']
<pre>block6c_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6c_se_excite[0][0]']
<pre>block6c_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6c_project_conv[0][0]']
block6c_drop (Dropout)	(None, 7, 7, 192)	Θ	['block6c_project_bn[0][0]']
block6c_add (Add)	(None, 7, 7, 192)	Θ	['block6c_drop[0][0]', 'block6b_add[0][0]']
block6d_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6c_add[0][0]']

<pre>block6d_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7,	1152)	4608	['block6d_expand_conv[0][0]']
<pre>block6d_expand_activation (Activation)</pre>	(None, 7, 7,	1152)	0	['block6d_expand_bn[0][0]']
block6d_dwconv2 (Depthwise Conv2D)	(None, 7, 7,	1152)	10368	['block6d_expand_activation[0] [0]']
<pre>block6d_bn (BatchNormaliza tion)</pre>	(None, 7, 7,	1152)	4608	['block6d_dwconv2[0][0]']
<pre>block6d_activation (Activa tion)</pre>	(None, 7, 7,	1152)	0	['block6d_bn[0][0]']
<pre>block6d_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)		0	['block6d_activation[0][0]']
<pre>block6d_se_reshape (Reshap e)</pre>	(None, 1, 1,	1152)	0	['block6d_se_squeeze[0][0]']
block6d_se_reduce (Conv2D)	(None, 1, 1,	48)	55344	['block6d_se_reshape[0][0]']
block6d se expand (Conv2D)	(None, 1, 1,	1152)	56448	['block6d se reduce[0][0]']
block6d_se_excite (Multiply)	(None, 7, 7,		0	['block6d_activation[0][0]', 'block6d_se_expand[0][0]']
block6d_project_conv (Conv 2D)	(None, 7, 7,	192)	221184	['block6d_se_excite[0][0]']
block6d_project_bn (BatchN ormalization)	(None, 7, 7,	192)	768	['block6d_project_conv[0][0]']
block6d_drop (Dropout)	(None, 7, 7,	192)	Θ	['block6d_project_bn[0][0]']
block6d_add (Add)	(None, 7, 7,	192)	0	['block6d_drop[0][0]', 'block6c_add[0][0]']
block6e_expand_conv (Conv2 D)	(None, 7, 7,	1152)	221184	['block6d_add[0][0]']
<pre>block6e_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7,	1152)	4608	['block6e_expand_conv[0][0]']
block6e_expand_activation (Activation)	(None, 7, 7,	1152)	0	['block6e_expand_bn[0][0]']
<pre>block6e_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7,	1152)	10368	<pre>['block6e_expand_activation[0] [0]']</pre>
<pre>block6e_bn (BatchNormaliza tion)</pre>	(None, 7, 7,	1152)	4608	['block6e_dwconv2[0][0]']
<pre>block6e_activation (Activa tion)</pre>	(None, 7, 7,	1152)	0	['block6e_bn[0][0]']
<pre>block6e_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)		0	['block6e_activation[0][0]']
<pre>block6e_se_reshape (Reshap e)</pre>	(None, 1, 1,	1152)	0	['block6e_se_squeeze[0][0]']
block6e_se_reduce (Conv2D)	(None, 1, 1,	48)	55344	['block6e_se_reshape[0][0]']
block6e se expand (Conv2D)	(None, 1, 1,	1152)	56448	['block6e se reduce[0][0]']
block6e_se_excite (Multipl y)	(None, 7, 7,	1152)	0	['block6e_activation[0][0]', 'block6e_se_expand[0][0]']
block6e_project_conv (Conv 2D)	(None, 7, 7,	192)	221184	['block6e_se_excite[0][0]']
block6e_project_bn (BatchN ormalization)	(None, 7, 7,	192)	768	['block6e_project_conv[0][0]']
block6e drop (Dropout)	(None, 7, 7,	192)	0	['block6e project bn[0][0]']
block6e_add (Add)	(None, 7, 7,	192)	0	 ['block6e_drop[0][0]', 'block6d add[0][0]']
block6f_expand_conv (Conv2 D)	(None, 7, 7,	1152)	221184	['block6e_add[0][0]']
block6f_expand_bn (BatchNo rmalization)	(None, 7, 7,	1152)	4608	['block6f_expand_conv[0][0]']
block6f_expand_activation	(None, 7, 7,	1152)	0	['block6f_expand_bn[0][0]']

(Activation)			
<pre>block6f_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6f_expand_activation[0] [0]']</pre>
<pre>block6f_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6f_dwconv2[0][0]']
<pre>block6f_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6f_bn[0][0]']
<pre>block6f_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	Θ	['block6f_activation[0][0]']
<pre>block6f_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6f_se_squeeze[0][0]']
block6f_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6f_se_reshape[0][0]']
block6f_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6f_se_reduce[0][0]']
<pre>block6f_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	['block6f_activation[0][0]', 'block6f_se_expand[0][0]']
<pre>block6f_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6f_se_excite[0][0]']
<pre>block6f_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6f_project_conv[0][0]']
block6f_drop (Dropout)	(None, 7, 7, 192)	0	['block6f_project_bn[0][0]']
block6f_add (Add)	(None, 7, 7, 192)	0	['block6f_drop[0][0]', 'block6e_add[0][0]']
block6g_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6f_add[0][0]']
<pre>block6g_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6g_expand_conv[0][0]']
<pre>block6g_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	Θ	['block6g_expand_bn[0][0]']
<pre>block6g_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6g_expand_activation[0] [0]']</pre>
<pre>block6g_bn (BatchNormaliza tion)</pre>	(None, 7, 7, 1152)	4608	['block6g_dwconv2[0][0]']
<pre>block6g_activation (Activa tion)</pre>	(None, 7, 7, 1152)	0	['block6g_bn[0][0]']
<pre>block6g_se_squeeze (Global AveragePooling2D)</pre>	(None, 1152)	Θ	['block6g_activation[0][0]']
<pre>block6g_se_reshape (Reshap e)</pre>	(None, 1, 1, 1152)	0	['block6g_se_squeeze[0][0]']
block6g_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6g_se_reshape[0][0]']
block6g_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6g_se_reduce[0][0]']
<pre>block6g_se_excite (Multipl y)</pre>	(None, 7, 7, 1152)	0	['block6g_activation[0][0]', 'block6g_se_expand[0][0]']
<pre>block6g_project_conv (Conv 2D)</pre>	(None, 7, 7, 192)	221184	['block6g_se_excite[0][0]']
<pre>block6g_project_bn (BatchN ormalization)</pre>	(None, 7, 7, 192)	768	['block6g_project_conv[0][0]']
block6g_drop (Dropout)	(None, 7, 7, 192)	0	['block6g_project_bn[0][0]']
block6g_add (Add)	(None, 7, 7, 192)	0	['block6g_drop[0][0]', 'block6f_add[0][0]']
block6h_expand_conv (Conv2 D)	(None, 7, 7, 1152)	221184	['block6g_add[0][0]']
<pre>block6h_expand_bn (BatchNo rmalization)</pre>	(None, 7, 7, 1152)	4608	['block6h_expand_conv[0][0]']
<pre>block6h_expand_activation (Activation)</pre>	(None, 7, 7, 1152)	0	['block6h_expand_bn[0][0]']
<pre>block6h_dwconv2 (Depthwise Conv2D)</pre>	(None, 7, 7, 1152)	10368	<pre>['block6h_expand_activation[0] [0]']</pre>

```
block6h_bn (BatchNormaliza (None, 7, 7, 1152)
                                                           4608
                                                                     ['block6h_dwconv2[0][0]']
tion)
block6h activation (Activa (None, 7, 7, 1152)
                                                           0
                                                                     ['block6h bn[0][0]']
tion)
block6h_se_squeeze (Global
                            (None, 1152)
                                                           0
                                                                     ['block6h_activation[0][0]']
AveragePooling2D)
block6h_se_reshape (Reshap
                            (None, 1, 1, 1152)
                                                           Θ
                                                                     ['block6h se squeeze[0][0]']
block6h se reduce (Conv2D)
                            (None, 1, 1, 48)
                                                           55344
                                                                     ['block6h se reshape[0][0]']
block6h se expand (Conv2D)
                             (None, 1, 1, 1152)
                                                           56448
                                                                     ['block6h se reduce[0][0]']
                            (None, 7, 7, 1152)
block6h se excite (Multipl
                                                                     ['block6h_activation[0][0]',
                                                           0
                                                                      'block6h se expand[0][0]']
y)
block6h project conv (Conv
                            (None, 7, 7, 192)
                                                           221184
                                                                     ['block6h se excite[0][0]']
2D)
block6h project bn (BatchN
                            (None, 7, 7, 192)
                                                           768
                                                                     ['block6h project conv[0][0]']
ormalization)
block6h drop (Dropout)
                             (None, 7, 7, 192)
                                                                     ['block6h project bn[0][0]']
block6h add (Add)
                             (None, 7, 7, 192)
                                                           0
                                                                     ['block6h_drop[0][0]',
                                                                      'block6g add[0][0]']
                             (None, 7, 7, 1280)
                                                           245760
top_conv (Conv2D)
                                                                     ['block6h_add[0][0]']
top bn (BatchNormalization
                            (None, 7, 7, 1280)
                                                           5120
                                                                     ['top conv[0][0]']
top activation (Activation
                            (None, 7, 7, 1280)
                                                           0
                                                                     ['top bn[0][0]']
global_average_pooling2d_1
                            (None, 1280)
                                                           0
                                                                     ['top_activation[0][0]']
1 (GlobalAveragePooling2D)
global_max_pooling2d_10 (G (None, 1280)
                                                           0
                                                                     ['top activation[0][0]']
lobalMaxPooling2D)
concatenate_9 (Concatenate (None, 2560)
                                                           0
                                                                     ['global_average_pooling2d_11[
                                                                     0][0]',
                                                                       'global max pooling2d 10[0][0
dropout 45 (Dropout)
                             (None, 2560)
                                                           0
                                                                     ['concatenate 9[0][0]']
dense_19 (Dense)
                             (None, 12)
                                                           30732
                                                                     ['dropout_45[0][0]']
```

Total params: 5950044 (22.70 MB) Trainable params: 5889436 (22.47 MB) Non-trainable params: 60608 (236.75 KB)

2023-11-26 20:33:15.205783: W tensorflow/core/util/tensor_slice_reader.cc:98] Could not open models/efficintB0_cw_imagenet: FAILED_PRECONDITION: models/efficintB0_cw_imagenet; Is a directory: perhaps your file is in a different file format and you need to use a different restore operator?
30/30 [==========] - 6s 155ms/step

```
In [64]: # Mostrar classification report
print("Classification Report EfficientNetV2 B0 (ImageNet Weights)")
print_classification_report(df_dataset_valid["label"].to_numpy(), prediction)
```

Classification Report EfficientNetV2 B0 (ImageNet Weights) Classification Report:

	precision	recall	f1-score	support
0	0.88	0.79	0.83	131
1	1.00	0.98	0.99	123
2	0.95	0.98	0.97	104
3	1.00	0.99	0.99	100
4	0.95	0.95	0.95	95
5	0.96	0.97	0.97	78
6	0.95	0.95	0.95	77
7	1.00	0.93	0.96	58
8	0.52	0.60	0.56	53
9	0.96	0.98	0.97	47
10	0.81	0.96	0.88	45
11	0.98	1.00	0.99	45
accuracy macro avg weighted avg	0.91 0.93	0.92 0.92	0.92 0.92 0.92	956 956 956

6.1 Evaluación del modelo predictivo y planteamiento de la siguiente prueba experimental (from scratch)

Estrategia 1: Entrenar desde cero o from scratch

Para realizar esta parte de la práctica se han utilizado un total de 6 arquitecturas diferentes. Para hacer experimentos comparativos y poder ver cómo afectan las técnicas de regularización a las arquitecturas, todas tienen el mismo top model para hacer la clasificación y comparten algunos métodos de regularización que se mantienen a lo largo de todas las arquitecturas.

Se definen ciertos parámetros comunes en todos los experimentos:

Hiperparámetros de entrenamiento:

• Paso por época: 32

• Épocas: 50

Top model:

- Capa densa de 128 unidades y activación ReLU
- Capa de salida con activación softmax para la clasificación multiclase

Class Weights:

El dataset tiene un desequilibrio elevado en las clases y por lo tanto, se puede producir un sesgo en el entrenamiento. Para poder evitarlo, se ha aplicado a cada entrenamiento el parámetro "class_weights" de SkLearn. Al hacer esto, se le aplican diferentes pesos a las clases en la función de pérdidas durante el entrenamiento.

A continuación, se describen las arquitecturas utilizadas junto con su nombre:

Model1:

Capas convolucionales:

- Block 1: 64 filtros, kernel 3x3, Maxpooling 2x2, Batch Normalization, Dropout (0.1)
- Block 2: 128 filtros, kernel 3x3, Maxpooling 2x2, Batch Normalization, Dropout (0.1)
- Block 3: 256 filtros, kernel 3x3, Maxpooling 2x2, Batch Normalization, Dropout (0.1)

Técnicas de regularización:

• Uso de Dropout y Batch Normalization en todas las capas convolucionales

Model1 nobatchnorm:

Capas convolucionales:

- Block 1: 64 filtros, kernel 3x3, Maxpooling 2x2, Dropout (0.1)
- Block 2: 128 filtros, kernel 3x3, Maxpooling 2x2, Dropout (0.1)
- Block 3: 256 filtros, kernel 3x3, Maxpooling 2x2, Dropout (0.1)

Técnicas de regularización:

• Uso de Dropout en todas las capas convolucionales

Model2:

Capas convolucionales:

- Block 1: 64 filtros, kernel 5x5, Maxpooling 2x2, Batch Normalization, Dropout (0.1)
- Block 2: 128 filtros, kernel 5x5, Maxpooling 2x2, Batch Normalization, Dropout (0.1)

Técnicas de regularización:

• Uso de Dropout y Batch Normalization en todas las capas convolucionales

Model2 nobatchnorm:

Capas convolucionales:

• Block 1: 64 filtros, kernel 5x5, Maxpooling 2x2, Dropout (0.1)

• Block 2: 128 filtros, kernel 5x5, Maxpooling 2x2, Dropout (0.1)

Técnicas de regularización:

• Uso de Dropout en todas las capas convolucionales

Model3:

Capas convolucionales:

- Block 1: 64 filtros, kernel 3x3, Maxpooling 2x2, Batch Normalization, Dropout (0.1)
- Block 2: 128 filtros, kernel 3x3, Maxpooling 2x2, Batch Normalization, Dropout (0.1)
- Block 3: 256 filtros, kernel 3x3, Maxpooling 2x2, Batch Normalization, Dropout (0.1)
- Block 4: 512 filtros, kernel 3x3, Maxpooling 2x2, Batch Normalization, Dropout (0.1)

Técnicas de regularización:

• Uso de Dropout y Batch Normalization en todas las capas convolucionales

Model3_nobatchnorm:

Capas convolucionales:

- Block 1: 64 filtros, kernel 3x3, Maxpooling 2x2, Dropout (0.1)
- Block 2: 128 filtros, kernel 3x3, Maxpooling 2x2, Dropout (0.1)
- Block 3: 256 filtros, kernel 3x3, Maxpooling 2x2, Dropout (0.1)
- Block 4: 512 filtros, kernel 3x3, Maxpooling 2x2, Dropout (0.1)

Técnicas de regularización:

• Uso de Dropout en todas las capas convolucionales

Análisis de resultados:

Se puede observar que el Model3 con Batch Normalization tiene unas métricas mejores que el resto de modelos. Esto se puede deber tanto a la propia regularización de la red como al hecho de que tiene más número de parámetros. Al no tener un dataset de test con etiquetas, no podemos comprobar la fiabilidad de las métricas fuera del entrenamiento.

A continuación, se muestran las gráficas de entrenamiento para cada modelo:

Model1:



Model1 No Batch Normalization:



Model2:



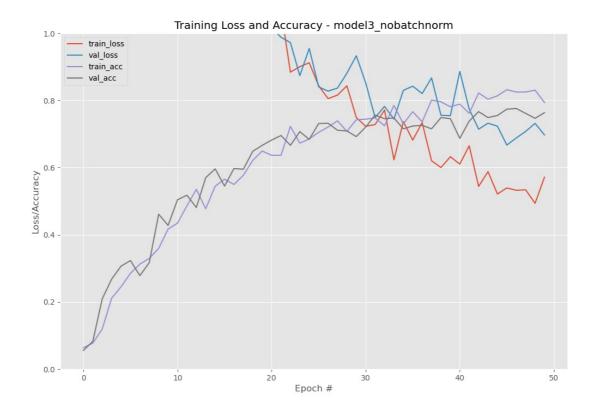
Model2 No Batch Normalization:



Model3:



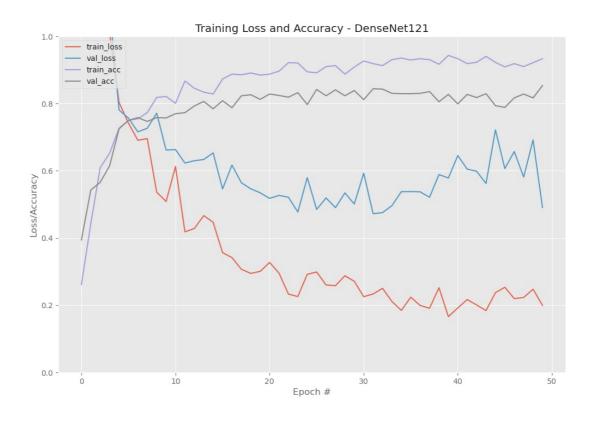
Model3 No Batch Normalization:



6.2 Evaluación del modelo predictivo y planteamiento de la siguiente prueba experimental (pre-entrenada)

Para realizar una comparación de los modelos vamos a observar las curvas de la función de pérdida y la precisión.

DenseNet121:





EfficientNetV2 B0:



La razón por la cual EfficientNetV2 ha dado los mejores resultados puede estar relacionada con varias razones:

- Mayor capacidad representativa: EfficientNetV2 puede capturar patrones más complejos y representaciones semánticas más
 profundas en comparación con arquitecturas más simples como VGG16. La arquitectura presenta capas de skip-connections
 mientras que la VGG16 no. Esto es especialmente beneficioso para tareas de clasificación en conjuntos de datos complejos.
- Arquitectura eficiente: EfficientNetV2 se diseñó específicamente para ser eficiente en términos de recursos computacionales y
 parámetros, manteniendo un buen rendimiento. Puede lograr una mayor eficiencia con respecto al uso de recursos como memoria y
 potencia de cálculo en comparación con arquitecturas más antiguas o menos eficientes. El resto de arquitecturas están
 sobredimensionadas para el problema propuesto

A pesar de que la EfficientNet ha dado los mejores resultados para resolver este problema, cabe destacar que los entrenamientos son muy erráticos y las curvas de entrenamiento son muy oscilatorias. No es el enfoque óptimo para resolver el problema, pero pensamos que una buena optimización de las redes pre-entrenadas podría darnos mejores resultados.