

```
In [14]: import numpy as np
import pandas as pd
from pathlib import Path
import os.path
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
import tensorflow as tf

from sklearn.metrics import confusion_matrix ,classification_report
```

```
In [15]: image_dir=Path('/kaggle/input/indian-food-classification')
```

```
In [16]: filepaths=list(image_dir.glob(r'**/*.jpg'))
labels=list(map(lambda x:os.path.split(os.path.split(x)[0])[1],filepaths))
filepaths=pd.Series(filepaths,name='Filepaths').astype(str)
labels=pd.Series(labels,name='Label')
images=pd.concat([filepaths,labels],axis=1)
category_samples=[]

for category in images['Label'].unique():
    category_slices=images.query('Label==@category')
    category_samples.append(category_slices.sample(frac=1,random_state=1))
image_df=pd.concat(category_samples,axis=0).sample(frac=1.0,random_state=1).
```

```
In [17]: image_df
```

```
Out[17]:
```

	Filepaths	Label
0	/kaggle/input/indian-food-classification/Food ...	chole_bhature
1	/kaggle/input/indian-food-classification/Food ...	chapati
2	/kaggle/input/indian-food-classification/Food ...	kadai_paneer
3	/kaggle/input/indian-food-classification/Food ...	kulfi
4	/kaggle/input/indian-food-classification/Food ...	samosa
...	...	...
6248	/kaggle/input/indian-food-classification/Food ...	chole_bhature
6249	/kaggle/input/indian-food-classification/Food ...	butter_naam
6250	/kaggle/input/indian-food-classification/Food ...	idli
6251	/kaggle/input/indian-food-classification/Food ...	momos
6252	/kaggle/input/indian-food-classification/Food ...	butter_naam

6253 rows × 2 columns

```
In [18]: image_df['Label'].value_counts()
```

```
Out[18]: Label
chapati      413
kadai_paneer 412
chole_bhature 410
chai         381
fried_rice   353
pav_bhaji    353
butter_naan  328
dal_makhani  319
momos        319
masala_dosa  311
burger       309
idli         306
kaathi_rolls 293
jalebi       293
dhokla       289
pakode       278
pizza        261
samosa       261
kulfi        234
paani_puri   130
Name: count, dtype: int64
```

```
In [19]: # This operation would be performed on the actual 'image_df' DataFrame
label_counts = image_df['Label'].value_counts().reset_index()
label_counts.columns = ['Label', 'Counts']
label_counts
```

Out[19]:

	Label	Counts
0	chapati	413
1	kadai_paneer	412
2	chole_bhature	410
3	chai	381
4	fried_rice	353
5	pav_bhaji	353
6	butter_naam	328
7	dal_makhani	319
8	momos	319
9	masala_dosa	311
10	burger	309
11	idli	306
12	kaathi_rolls	293
13	jalebi	293
14	dhokla	289
15	pakode	278
16	pizza	261
17	samosa	261
18	kulfi	234
19	paani_puri	130

```
In [20]: IMG_SIZE = (256, 256)
         BATCH_SIZE = 32
         SEED = 42
         PATH = "../input/indian-food-classification/Food Classification"
```

```
In [21]: data_ds = tf.keras.preprocessing.image_dataset_from_directory(
         PATH,
         seed=SEED,
         image_size=IMG_SIZE,
         batch_size=BATCH_SIZE,
         label_mode="categorical"
         )
```

Found 6269 files belonging to 20 classes.

```
In [22]: # Get class names
         classes = data_ds.class_names
```

```
In [23]: classes
```

```
Out[23]: ['burger',
          'butter_naan',
          'chai',
          'chapati',
          'chole_bhature',
          'dal_makhani',
          'dhokla',
          'fried_rice',
          'idli',
          'jalebi',
          'kaathi_rolls',
          'kadai_paneer',
          'kulfi',
          'masala_dosa',
          'momos',
          'paani_puri',
          'pakode',
          'pav_bhaji',
          'pizza',
          'samosa']
```

```
In [24]: # Get class names
         classes = data_ds.class_names
```

```
In [25]: classes
```

```
Out[25]: ['burger',
          'butter_naan',
          'chai',
          'chapati',
          'chole_bhature',
          'dal_makhani',
          'dhokla',
          'fried_rice',
          'idli',
          'jalebi',
          'kaathi_rolls',
          'kadai_paneer',
          'kulfi',
          'masala_dosa',
          'momos',
          'paani_puri',
          'pakode',
          'pav_bhaji',
          'pizza',
          'samosa']
```

## Detecting outliers

### Unusual Image size

This code identifies images in a dataset that have unusual sizes compared to the rest of the dataset. It begins by defining a path to the dataset and listing all the subdirectories, which correspond to different classes of images. For each image in these folders, the code reads the image using OpenCV, calculates its size in pixels by multiplying the width and height, and stores these sizes in a list. After converting the list of sizes into a NumPy array, the code calculates the z-scores—a statistical measure that indicates how many standard deviations a data point is from the mean of the dataset. It then identifies outliers as those images whose sizes have z-scores that exceed an absolute value of 3, which is a common threshold for outlier detection. These outliers are likely to be either much larger or smaller than the typical image size in the dataset. The indices of these outliers in the dataset are printed out, which can be used to further examine or process these specific images.

```
In [26]: import os
import numpy as np
import cv2
from scipy.stats import zscore
import matplotlib.pyplot as plt

# Define the path to the dataset
PATH = "../input/indian-food-classification/Food Classification"

# List all the folders in the dataset
folders = [folder for folder in os.listdir(PATH) if os.path.isdir(os.path.join(PATH, folder))]

# Initialize a list to store image sizes and file paths
image_sizes = []
file_paths = []

# Loop through each folder (which corresponds to a class)
for folder in folders:
    # Get a list of all the file names of images in this folder
    image_files = os.listdir(os.path.join(PATH, folder))
    # Loop through each image file
    for image_file in image_files:
        # Define the path to the image
        image_path = os.path.join(PATH, folder, image_file)
        # Load the image
        image = cv2.imread(image_path)
        # Check if the image was loaded correctly
        if image is not None:
            # Get the size (in pixels) of the image
            size = image.shape[0] * image.shape[1]
            # Append the size and file path to our lists
            image_sizes.append(size)
            file_paths.append(image_path)

# Convert the list to a NumPy array
image_sizes = np.array(image_sizes)
```

```

# Calculate z-scores for the image sizes
z_scores = zscore(image_sizes)

# Define a threshold for what we consider to be an outlier
threshold = 3

# Find indices of outliers
outlier_indices = np.where((z_scores > threshold) | (z_scores < -threshold))

# Exclude outliers
non_outlier_indices = np.setdiff1d(np.arange(len(file_paths)), outlier_indices)
non_outlier_file_paths_1 = [file_paths[index] for index in non_outlier_indices]

# Display the number of outliers
print(f"Found {len(outlier_indices)} outliers based on image size.")

# print the paths of the outliers
for index in outlier_indices[:5]:
    print(file_paths[index])

# Display individual outliers
for index in outlier_indices[:5]:
    image = cv2.cvtColor(cv2.imread(file_paths[index]), cv2.COLOR_BGR2RGB)
    plt.figure()
    plt.imshow(image)
    plt.axis('off')
    plt.title(f"Outlier {index}")
    plt.show()

# Now can use non_outlier_file_paths for further processing

```

Found 168 outliers based on image size.

```

../input/indian-food-classification/Food Classification/momos/002.jpg
../input/indian-food-classification/Food Classification/momos/064.jpg
../input/indian-food-classification/Food Classification/momos/086.jpg
../input/indian-food-classification/Food Classification/momos/017.jpg
../input/indian-food-classification/Food Classification/momos/216.jpg

```

Outlier 62



Outlier 114



Outlier 126



Outlier 129





Outlier 176



## Image Quality Outliers (using blur detection with Laplacian variance)

This code calculates the quality of images based on the variance of the Laplacian filter, which is a measure of sharpness or focus. Images with low variance in the Laplacian are typically blurry, and high variance indicates a sharp image. The code computes this focus measure for each image in the dataset and then uses a z-score to detect outliers, which are images that are significantly blurrier or sharper than the average. It displays the first few outlier images based on image quality.

```
In [27]: import os
import cv2
import numpy as np
from scipy.stats import zscore
import matplotlib.pyplot as plt

# Define the path to the dataset
PATH = "../input/indian-food-classification/Food Classification"

# Function to calculate image quality using Laplacian variance
def calculate_image_quality(image_path):
    image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
    if image is None:
        return None
    focus_measure = cv2.Laplacian(image, cv2.CV_64F).var()
```

```

    return focus_measure

# Load images and compute quality
file_paths = []
quality_scores = []

for folder in os.listdir(PATH):
    folder_path = os.path.join(PATH, folder)
    if os.path.isdir(folder_path):
        for image_file in os.listdir(folder_path):
            image_path = os.path.join(folder_path, image_file)
            quality = calculate_image_quality(image_path)
            if quality is not None: # Only add if the image was successful
                file_paths.append(image_path)
                quality_scores.append(quality)

# Convert to NumPy array
quality_scores = np.array(quality_scores)

# Detect outliers
quality_outliers = np.where(zscore(quality_scores) > 3)[0]

# Exclude outliers from the dataset
non_outlier_indices = np.setdiff1d(np.arange(len(file_paths)), quality_outliers)
non_outlier_file_paths_2 = [file_paths[index] for index in non_outlier_indices]
non_outlier_quality_scores = quality_scores[non_outlier_indices]

# Display outliers
def display_outliers(outlier_indices, file_paths, title):
    if len(outlier_indices) == 0:
        print(f"No outliers detected for {title}.")
        return

    plt.figure(figsize=(15, 5))
    for i, idx in enumerate(outlier_indices[:5]): # Displaying only first 5
        image = cv2.cvtColor(cv2.imread(file_paths[idx]), cv2.COLOR_BGR2RGB)
        plt.subplot(1, 5, i + 1)
        plt.imshow(image)
        plt.title(f"{title} {idx}")
        plt.axis('off')
    plt.tight_layout()
    plt.show()

# Call the display function
print("Image Quality Outliers:")
display_outliers(quality_outliers, file_paths, "Quality")

# You can now use non_outlier_file_paths and non_outlier_quality_scores for

```

Image Quality Outliers:



## Color histogram

This code calculates the color histograms of the images. A color histogram represents the distribution of colors in an image; it counts how many times each color appears. This is done across the RGB color channels. The code then uses a z-score to detect images whose color distribution is significantly different from that of the majority of images. Outliers could be images that are overly bright, dark, or have an unusual color cast. It displays the first few outlier images based on color distribution.

```
In [28]: import os
import cv2
import numpy as np
from scipy.stats import zscore
import matplotlib.pyplot as plt

# Define the path to the dataset
PATH = "../input/indian-food-classification/Food Classification" # Replace

# Function to calculate color histogram
def calculate_color_histogram(image_path):
    image = cv2.imread(image_path)
    if image is None:
        return None
    hist = cv2.calcHist([image], [0, 1, 2], None, [8, 8, 8], [0, 256, 0, 256])
    return hist.flatten()

# Function to display outliers
def display_outliers(outlier_indices, file_paths, title):
    plt.figure(figsize=(15, 5))
    for i, idx in enumerate(outlier_indices[:5]): # Show up to the first 5
        image = cv2.imread(file_paths[idx])
        if image is None:
            continue
        image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
        plt.subplot(1, 5, i+1)
        plt.imshow(image)
        plt.axis('off')
        plt.title(f"{title} Outlier {idx}")
    plt.show()
```

```

# Load images and compute color histograms
file_paths_3 = []
color_histograms = []

for folder in os.listdir(PATH):
    folder_path = os.path.join(PATH, folder)
    if os.path.isdir(folder_path):
        for image_file in os.listdir(folder_path):
            image_path = os.path.join(folder_path, image_file)
            hist = calculate_color_histogram(image_path)
            if hist is not None: # Skip if the image could not be read
                file_paths_3.append(image_path)
                color_histograms.append(hist)

# Convert to NumPy array
color_histograms = np.array(color_histograms)

# Detect outliers
color_outliers = np.where(zscore(color_histograms) > 3)[0]

# Filter out the outlier file paths
non_outlier_file_paths_3 = [file_paths_3[i] for i in range(len(file_paths_3))

# Call the display function
print("Color Histogram Outliers:")
display_outliers(color_outliers, file_paths_3, "Color")

# The non_outlier_file_paths_3 now contains the file paths without the outliers
print("Stored non-outlier file paths.")

```

Color Histogram Outliers:



Stored non-outlier file paths.

```

In [29]: print(non_outlier_file_paths_1[:5])
         print(non_outlier_file_paths_2[:5])
         print(non_outlier_file_paths_3[:5])

```

```
['../input/indian-food-classification/Food Classification/momos/208.jpg',
 '../input/indian-food-classification/Food Classification/momos/029.jpg',
 '../input/indian-food-classification/Food Classification/momos/014.jpg',
 '../input/indian-food-classification/Food Classification/momos/275.jpg',
 '../input/indian-food-classification/Food Classification/momos/212.jpg']
['../input/indian-food-classification/Food Classification/momos/208.jpg',
 '../input/indian-food-classification/Food Classification/momos/029.jpg',
 '../input/indian-food-classification/Food Classification/momos/014.jpg',
 '../input/indian-food-classification/Food Classification/momos/275.jpg',
 '../input/indian-food-classification/Food Classification/momos/212.jpg']
['../input/indian-food-classification/Food Classification/momos/208.jpg',
 '../input/indian-food-classification/Food Classification/momos/029.jpg',
 '../input/indian-food-classification/Food Classification/momos/014.jpg',
 '../input/indian-food-classification/Food Classification/momos/275.jpg',
 '../input/indian-food-classification/Food Classification/momos/239.jpg']
```

```
In [30]: # Convert lists to sets
set_fp1 = set(non_outlier_file_paths_1)
set_fp2 = set(non_outlier_file_paths_2)
set_fp3 = set(non_outlier_file_paths_3)

# Find the intersection of the three sets
common_file_paths = list(set_fp1.intersection(set_fp2, set_fp3))
```

```
In [31]: #print(len(common_file_paths))
cf = common_file_paths
```

```
In [39]: type(cf)
```

```
Out[39]: list
```

```
In [40]: k=cf[0].split('/')
k=k[-2]
```

```
In [46]:
```

```
Out[46]: 4292
```

```
In [47]: i=0
img_lable_after_removing_outlier=[]
for i in range(len(common_file_paths)):
    k=cf[i].split('/')
    img_lable_after_removing_outlier.append(k[-2])
```

```
In [50]: import pandas as pd
import pandas as pd

series = pd.Series(img_lable_after_removing_outlier)

img_counts_after_removing_outlier = series.value_counts()
```

```
In [51]: img_counts_after_removing_outlier
```

```
Out[51]: chapati      303
         chole_bhature 290
         chai        273
         kadai_paneer 270
         pav_bhaji    254
         momos        243
         butter_naan  242
         fried_rice   241
         dal_makhani  231
         idli         215
         kaathi_rolls 204
         masala_dosa   199
         pakode       192
         samosa       191
         burger       187
         jalebi       184
         dhokla       177
         kulfi        158
         pizza        154
         paani_puri    84
         Name: count, dtype: int64
```

## Preview of some images

```
In [23]: IMG_SIZE = (224, 224) # VGG16 default image size

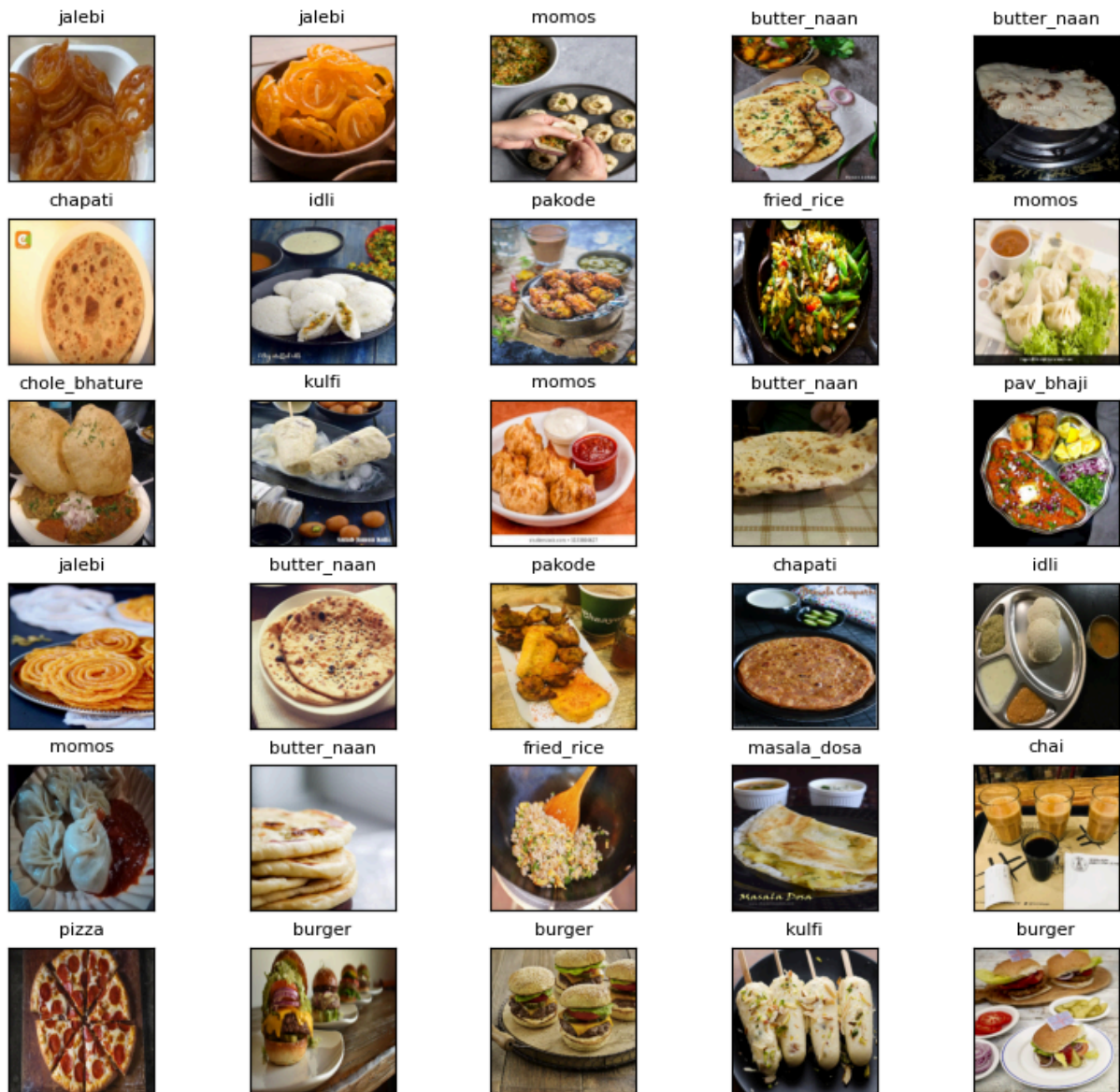
# Load images and labels
img = []
lbl = []
for file_path in common_file_paths: # Replace with your actual file paths
    image = cv2.imread(file_path)
    if image is None:
        continue # Skip files that aren't valid images
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    image = cv2.resize(image, IMG_SIZE)
    img.append(image)

    label = file_path.split(os.path.sep)[-2]
    lbl.append(label)
```

```
In [24]: # Display some pictures of the dataset
fig, axes = plt.subplots(nrows=6, ncols=5, figsize=(8, 7),
                        subplot_kw={'xticks': [], 'yticks': []})

for i, ax in enumerate(axes.flat):
    # ax.imshow(plt.imread(df_unique.Filepath[i]))
    ax.imshow(img[i])
    ax.set_title(lbl[i], fontsize = 8)
    # ax.set_title(df_unique.Labels[i], fontsize = 8)
plt.tight_layout(pad=0.5)
plt.show()
```





In [ ]:

## Transfer Learning with the EfficientNetV2L

```
In [25]: import numpy as np
import cv2
import os
from sklearn.model_selection import train_test_split
from tensorflow.keras.applications import EfficientNetV2L
from tensorflow.keras.applications.efficientnet_v2 import preprocess_input
from tensorflow.keras.layers import Flatten, Dense, Dropout, BatchNormalizat
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import Adam
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt
```

```

IMG_SIZE = (224, 224)

images = []
labels = []
for file_path in common_file_paths:
    image = cv2.imread(file_path)
    if image is None:
        continue
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    image = cv2.resize(image, IMG_SIZE)
    images.append(image)
    label = file_path.split(os.path.sep)[-2]
    labels.append(label)

images = np.array(images, dtype=np.float32)
labels = np.array(labels)
images = preprocess_input(images)

label_encoder = LabelEncoder()
labels_encoded = label_encoder.fit_transform(labels)
labels_one_hot = to_categorical(labels_encoded)

X_train, X_val, y_train, y_val = train_test_split(images, labels_one_hot, test_size=0.2, random_state=42)

data_gen = ImageDataGenerator(rotation_range=20, zoom_range=0.15, width_shift_range=0.1, height_shift_range=0.1,
                              shear_range=0.2, fill_mode='nearest')

base_model = EfficientNetV2L(weights='imagenet', include_top=False, input_shape=(224, 224, 3))

# Freeze base model layers and unfreeze the last 4 layers
for layer in base_model.layers[:-4]:
    layer.trainable = False

x = base_model.output
x = Flatten()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
x = BatchNormalization()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
predictions = Dense(len(label_encoder.classes_), activation='softmax')(x)
model = Model(inputs=base_model.input, outputs=predictions)

model.compile(optimizer=Adam(learning_rate=1e-5), loss='categorical_crossentropy', metrics=['accuracy'])

history = model.fit(data_gen.flow(X_train, y_train, batch_size=32), validation_data=(X_val, y_val), epochs=100)

```



```

val_loss, val_accuracy = model.evaluate(X_val, y_val)
print(f'Validation accuracy: {val_accuracy * 100:.2f}%')

plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()

plt.show()

```

Downloading data from [https://storage.googleapis.com/tensorflow/keras-applications/efficientnet\\_v2/efficientnetv2-l\\_notop.h5](https://storage.googleapis.com/tensorflow/keras-applications/efficientnet_v2/efficientnetv2-l_notop.h5)

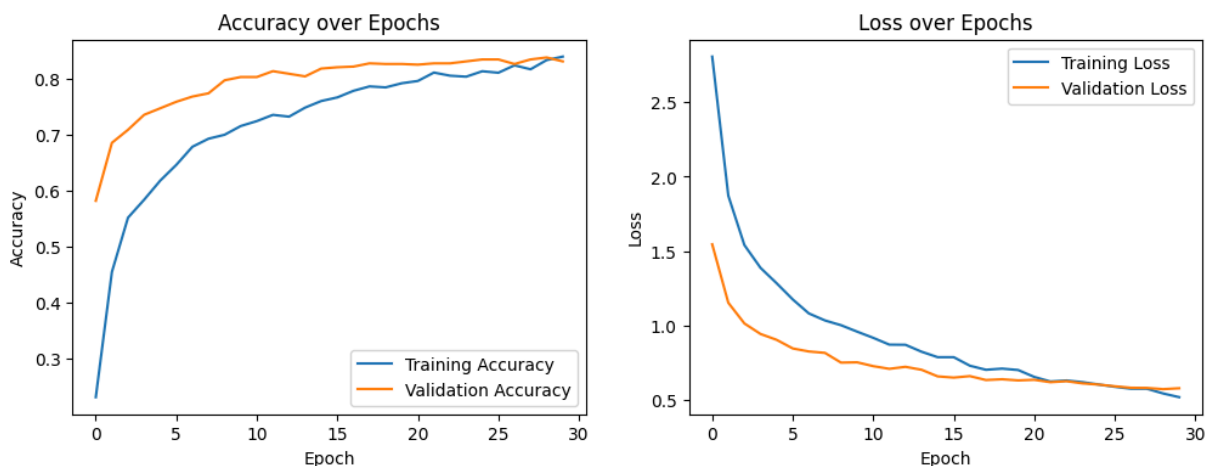
473176280/473176280 [=====] - 2s 0us/step

Epoch 1/30

2024-05-06 17:56:09.158838: E tensorflow/core/grappler/optimizers/meta\_optimizer.cc:954] layout failed: INVALID\_ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape in model/block1b\_drop/dropout/SelectV2-2-TransposeNHWCToNCHW-LayoutOptimizer

107/107 [=====] - 81s 467ms/step - loss: 2.8053 - accuracy: 0.2311 - val\_loss: 1.5458 - val\_accuracy: 0.5821  
Epoch 2/30  
107/107 [=====] - 42s 388ms/step - loss: 1.8720 - accuracy: 0.4549 - val\_loss: 1.1536 - val\_accuracy: 0.6857  
Epoch 3/30  
107/107 [=====] - 41s 382ms/step - loss: 1.5405 - accuracy: 0.5519 - val\_loss: 1.0148 - val\_accuracy: 0.7090  
Epoch 4/30  
107/107 [=====] - 41s 383ms/step - loss: 1.3880 - accuracy: 0.5839 - val\_loss: 0.9442 - val\_accuracy: 0.7357  
Epoch 5/30  
107/107 [=====] - 41s 387ms/step - loss: 1.2845 - accuracy: 0.6181 - val\_loss: 0.9050 - val\_accuracy: 0.7474  
Epoch 6/30  
107/107 [=====] - 41s 382ms/step - loss: 1.1761 - accuracy: 0.6463 - val\_loss: 0.8471 - val\_accuracy: 0.7590  
Epoch 7/30  
107/107 [=====] - 41s 386ms/step - loss: 1.0822 - accuracy: 0.6786 - val\_loss: 0.8268 - val\_accuracy: 0.7683  
Epoch 8/30  
107/107 [=====] - 42s 387ms/step - loss: 1.0350 - accuracy: 0.6930 - val\_loss: 0.8174 - val\_accuracy: 0.7742  
Epoch 9/30  
107/107 [=====] - 41s 383ms/step - loss: 1.0030 - accuracy: 0.7001 - val\_loss: 0.7520 - val\_accuracy: 0.7974  
Epoch 10/30  
107/107 [=====] - 41s 381ms/step - loss: 0.9598 - accuracy: 0.7157 - val\_loss: 0.7540 - val\_accuracy: 0.8033  
Epoch 11/30  
107/107 [=====] - 41s 378ms/step - loss: 0.9183 - accuracy: 0.7245 - val\_loss: 0.7282 - val\_accuracy: 0.8033  
Epoch 12/30  
107/107 [=====] - 41s 383ms/step - loss: 0.8727 - accuracy: 0.7357 - val\_loss: 0.7103 - val\_accuracy: 0.8137  
Epoch 13/30  
107/107 [=====] - 41s 383ms/step - loss: 0.8717 - accuracy: 0.7324 - val\_loss: 0.7237 - val\_accuracy: 0.8091  
Epoch 14/30  
107/107 [=====] - 41s 383ms/step - loss: 0.8252 - accuracy: 0.7486 - val\_loss: 0.7041 - val\_accuracy: 0.8044  
Epoch 15/30  
107/107 [=====] - 41s 382ms/step - loss: 0.7885 - accuracy: 0.7604 - val\_loss: 0.6596 - val\_accuracy: 0.8184  
Epoch 16/30  
107/107 [=====] - 41s 382ms/step - loss: 0.7885 - accuracy: 0.7668 - val\_loss: 0.6515 - val\_accuracy: 0.8207  
Epoch 17/30  
107/107 [=====] - 41s 386ms/step - loss: 0.7312 - accuracy: 0.7786 - val\_loss: 0.6614 - val\_accuracy: 0.8219  
Epoch 18/30  
107/107 [=====] - 41s 383ms/step - loss: 0.7040 - accuracy: 0.7865 - val\_loss: 0.6352 - val\_accuracy: 0.8277  
Epoch 19/30  
107/107 [=====] - 41s 382ms/step - loss: 0.7117 - accuracy: 0.7848 - val\_loss: 0.6404 - val\_accuracy: 0.8265

Epoch 20/30  
 107/107 [=====] - 41s 380ms/step - loss: 0.7024 - accuracy: 0.7921 - val\_loss: 0.6327 - val\_accuracy: 0.8265  
 Epoch 21/30  
 107/107 [=====] - 41s 380ms/step - loss: 0.6559 - accuracy: 0.7962 - val\_loss: 0.6367 - val\_accuracy: 0.8254  
 Epoch 22/30  
 107/107 [=====] - 41s 380ms/step - loss: 0.6252 - accuracy: 0.8112 - val\_loss: 0.6206 - val\_accuracy: 0.8277  
 Epoch 23/30  
 107/107 [=====] - 41s 385ms/step - loss: 0.6319 - accuracy: 0.8056 - val\_loss: 0.6272 - val\_accuracy: 0.8277  
 Epoch 24/30  
 107/107 [=====] - 41s 381ms/step - loss: 0.6210 - accuracy: 0.8039 - val\_loss: 0.6126 - val\_accuracy: 0.8312  
 Epoch 25/30  
 107/107 [=====] - 41s 380ms/step - loss: 0.6051 - accuracy: 0.8136 - val\_loss: 0.6045 - val\_accuracy: 0.8347  
 Epoch 26/30  
 107/107 [=====] - 41s 385ms/step - loss: 0.5898 - accuracy: 0.8109 - val\_loss: 0.5931 - val\_accuracy: 0.8347  
 Epoch 27/30  
 107/107 [=====] - 41s 384ms/step - loss: 0.5767 - accuracy: 0.8242 - val\_loss: 0.5831 - val\_accuracy: 0.8265  
 Epoch 28/30  
 107/107 [=====] - 41s 381ms/step - loss: 0.5767 - accuracy: 0.8171 - val\_loss: 0.5821 - val\_accuracy: 0.8347  
 Epoch 29/30  
 107/107 [=====] - 41s 383ms/step - loss: 0.5451 - accuracy: 0.8336 - val\_loss: 0.5739 - val\_accuracy: 0.8382  
 Epoch 30/30  
 107/107 [=====] - 41s 383ms/step - loss: 0.5202 - accuracy: 0.8398 - val\_loss: 0.5799 - val\_accuracy: 0.8312  
 27/27 [=====] - 6s 216ms/step - loss: 0.5799 - accuracy: 0.8312  
 Validation accuracy: 83.12%



```
In [26]: y_pred = model.predict(X_val)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_val, axis=1)
conf_matrix = confusion_matrix(y_true, y_pred_classes)
plt.figure(figsize=(10, 8))
```

```

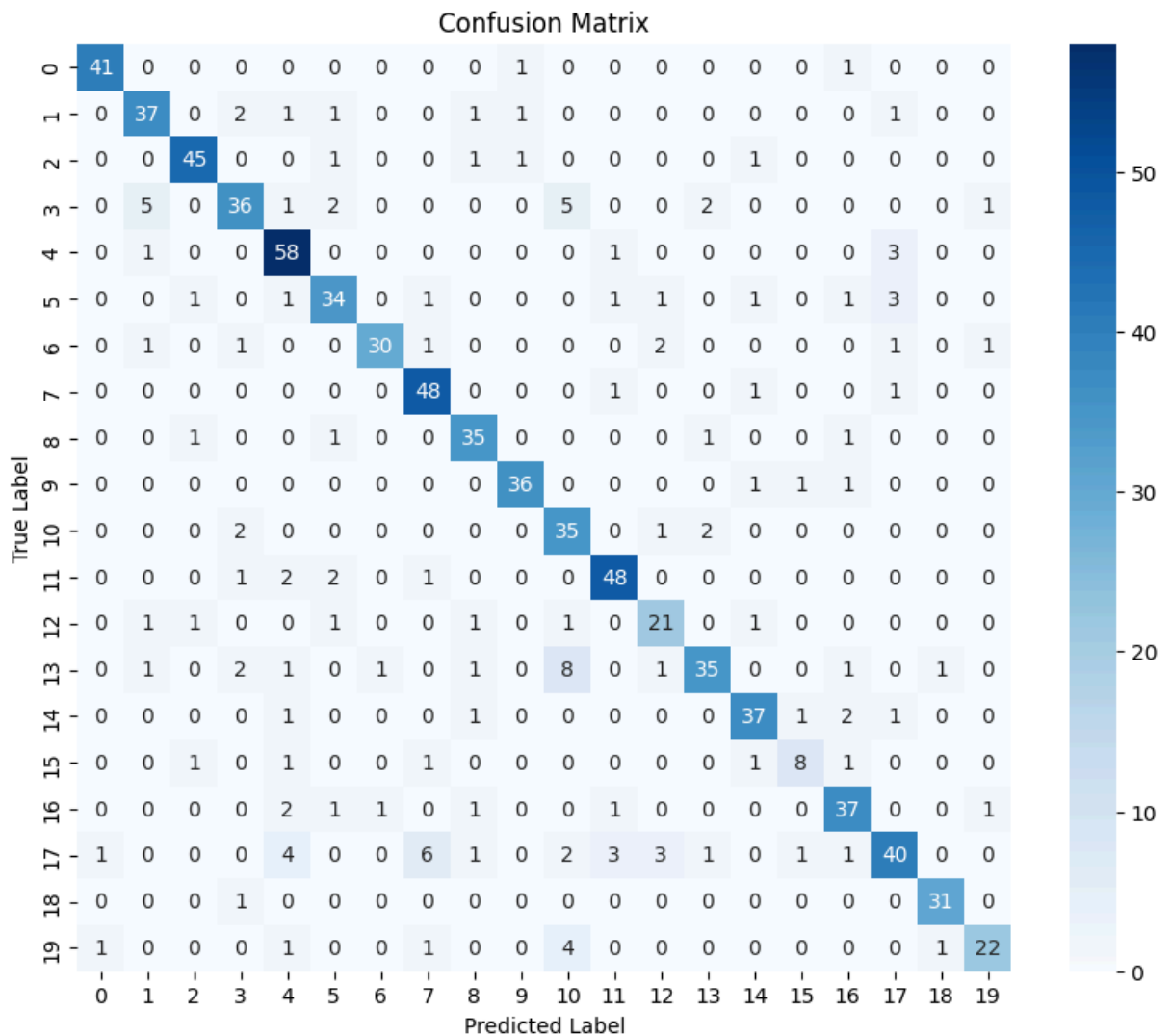
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.show()

# Calculate precision, recall, F1 score, and accuracy
report = classification_report(y_true, y_pred_classes, output_dict=True)
accuracy = report['accuracy']
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']

# Display the calculated metrics
print(f'Validation Accuracy: {accuracy:.4f}')
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1_score:.4f}')

```

27/27 [=====] - 12s 210ms/step



Validation Accuracy: 0.8312  
 Precision: 0.8334  
 Recall: 0.8270  
 F1 Score: 0.8266

```
In [27]: model.summary()
```

Model: "model"

Layer (type) d to	Output Shape	Param #	Connecte
=====			
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
rescaling (Rescaling) 1[0][0]']	(None, 224, 224, 3)	0	['input_
stem_conv (Conv2D) ing[0][0]']	(None, 112, 112, 32)	864	['rescal
stem_bn (BatchNormalizatio onv[0][0]') n)	(None, 112, 112, 32)	128	['stem_c
stem_activation (Activatio n[0][0]') n)	(None, 112, 112, 32)	0	['stem_b
block1a_project_conv (Conv ctivation[0][0]') 2D)	(None, 112, 112, 32)	9216	['stem_a
block1a_project_bn (BatchN a_project_conv[0][0]') ormalization)	(None, 112, 112, 32)	128	['block1
block1a_project_activation a_project_bn[0][0]') (Activation)	(None, 112, 112, 32)	0	['block1
block1a_add (Add) a_project_activation[0	(None, 112, 112, 32)	0	['block1
ctivation[0][0]']			][0]', 'stem_a
block1b_project_conv (Conv a_add[0][0]') 2D)	(None, 112, 112, 32)	9216	['block1
block1b_project_bn (BatchN b_project_conv[0][0]') ormalization)	(None, 112, 112, 32)	128	['block1
block1b_project_activation b_project_bn[0][0]') (Activation)	(None, 112, 112, 32)	0	['block1
block1b_drop (Dropout) b_project_activation[0	(None, 112, 112, 32)	0	['block1
			][0]']

block1b_add (Add) b_drop[0][0]', a_add[0][0]']	(None, 112, 112, 32)	0	['block1  'block1
block1c_project_conv (Conv b_add[0][0]'] 2D)	(None, 112, 112, 32)	9216	['block1
block1c_project_bn (BatchN c_project_conv[0][0]'] ormalization)	(None, 112, 112, 32)	128	['block1
block1c_project_activation c_project_bn[0][0]'] (Activation)	(None, 112, 112, 32)	0	['block1
block1c_drop (Dropout) c_project_activation[0	(None, 112, 112, 32)	0	['block1  ][0]']
block1c_add (Add) c_drop[0][0]', b_add[0][0]']	(None, 112, 112, 32)	0	['block1  'block1
block1d_project_conv (Conv c_add[0][0]'] 2D)	(None, 112, 112, 32)	9216	['block1
block1d_project_bn (BatchN d_project_conv[0][0]'] ormalization)	(None, 112, 112, 32)	128	['block1
block1d_project_activation d_project_bn[0][0]'] (Activation)	(None, 112, 112, 32)	0	['block1
block1d_drop (Dropout) d_project_activation[0	(None, 112, 112, 32)	0	['block1  ][0]']
block1d_add (Add) d_drop[0][0]', c_add[0][0]']	(None, 112, 112, 32)	0	['block1  'block1
block2a_expand_conv (Conv2 d_add[0][0]'] D)	(None, 56, 56, 128)	36864	['block1
block2a_expand_bn (BatchNo a_expand_conv[0][0]'] rmalization)	(None, 56, 56, 128)	512	['block2

block2a_expand_activation a_expand_bn[0][0]'] (Activation)	(None, 56, 56, 128)	0	['block2
block2a_project_conv a_expand_activation[0] 2D)	(Conv (None, 56, 56, 64)	8192	['block2 [0]']
block2a_project_bn a_project_conv[0][0]'] ormalization)	(BatchN (None, 56, 56, 64)	256	['block2
block2b_expand_conv a_project_bn[0][0]'] D)	(Conv2 (None, 56, 56, 256)	147456	['block2
block2b_expand_bn b_expand_conv[0][0]'] rmalization)	(BatchNo (None, 56, 56, 256)	1024	['block2
block2b_expand_activation b_expand_bn[0][0]'] (Activation)	(None, 56, 56, 256)	0	['block2
block2b_project_conv b_expand_activation[0] 2D)	(Conv (None, 56, 56, 64)	16384	['block2 [0]']
block2b_project_bn b_project_conv[0][0]'] ormalization)	(BatchN (None, 56, 56, 64)	256	['block2
block2b_drop b_project_bn[0][0]']	(Dropout (None, 56, 56, 64)	0	['block2
block2b_add b_drop[0][0]', a_project_bn[0][0]']	(Add (None, 56, 56, 64)	0	['block2 'block2
block2c_expand_conv b_add[0][0]'] D)	(Conv2 (None, 56, 56, 256)	147456	['block2
block2c_expand_bn c_expand_conv[0][0]'] rmalization)	(BatchNo (None, 56, 56, 256)	1024	['block2
block2c_expand_activation c_expand_bn[0][0]'] (Activation)	(None, 56, 56, 256)	0	['block2
block2c_project_conv c_expand_activation[0] 2D)	(Conv (None, 56, 56, 64)	16384	['block2 [0]']



block2c_project_bn (BatchNorm2d)	(None, 56, 56, 64)	256	['block2c_project_conv[0][0]']
block2c_drop (Dropout)	(None, 56, 56, 64)	0	['block2c_project_bn[0][0]']
block2c_add (Add)	(None, 56, 56, 64)	0	['block2c_drop[0][0]', 'block2b_add[0][0]']
block2d_expand_conv (Conv2d)	(None, 56, 56, 256)	147456	['block2c_add[0][0]']
block2d_expand_bn (BatchNorm2d)	(None, 56, 56, 256)	1024	['block2d_expand_conv[0][0]']
block2d_expand_activation (Activation)	(None, 56, 56, 256)	0	['block2d_expand_bn[0][0]']
block2d_project_conv (Conv2d)	(None, 56, 56, 64)	16384	['block2d_expand_activation[0][0]']
block2d_project_bn (BatchNorm2d)	(None, 56, 56, 64)	256	['block2d_project_conv[0][0]']
block2d_drop (Dropout)	(None, 56, 56, 64)	0	['block2d_project_bn[0][0]']
block2d_add (Add)	(None, 56, 56, 64)	0	['block2d_drop[0][0]', 'block2c_add[0][0]']
block2e_expand_conv (Conv2d)	(None, 56, 56, 256)	147456	['block2d_add[0][0]']
block2e_expand_bn (BatchNorm2d)	(None, 56, 56, 256)	1024	['block2e_expand_conv[0][0]']
block2e_expand_activation (Activation)	(None, 56, 56, 256)	0	['block2e_expand_bn[0][0]']
block2e_project_conv (Conv2d)	(None, 56, 56, 64)	16384	['block2e_expand_activation[0][0]']

block2e_project_bn (BatchNormal ization)	(None, 56, 56, 64)	256	['block2 e_project_conv[0][0]']
block2e_drop (Dropout)	(None, 56, 56, 64)	0	['block2 e_project_bn[0][0]']
block2e_add (Add)	(None, 56, 56, 64)	0	['block2 e_drop[0][0]',  'block2 d_add[0][0]']
block2f_expand_conv (Conv2 D)	(None, 56, 56, 256)	147456	['block2 e_add[0][0]']
block2f_expand_bn (BatchNormal ization)	(None, 56, 56, 256)	1024	['block2 f_expand_conv[0][0]']
block2f_expand_activation (Activation)	(None, 56, 56, 256)	0	['block2 f_expand_bn[0][0]']
block2f_project_conv (Conv 2D)	(None, 56, 56, 64)	16384	['block2 f_expand_activation[0] [0]']
block2f_project_bn (BatchNormal ization)	(None, 56, 56, 64)	256	['block2 f_project_conv[0][0]']
block2f_drop (Dropout)	(None, 56, 56, 64)	0	['block2 f_project_bn[0][0]']
block2f_add (Add)	(None, 56, 56, 64)	0	['block2 f_drop[0][0]',  'block2 e_add[0][0]']
block2g_expand_conv (Conv2 D)	(None, 56, 56, 256)	147456	['block2 f_add[0][0]']
block2g_expand_bn (BatchNormal ization)	(None, 56, 56, 256)	1024	['block2 g_expand_conv[0][0]']
block2g_expand_activation (Activation)	(None, 56, 56, 256)	0	['block2 g_expand_bn[0][0]']
block2g_project_conv (Conv 2D)	(None, 56, 56, 64)	16384	['block2 g_expand_activation[0] [0]']

block2g_project_bn (Batch Normalization)	(None, 56, 56, 64)	256	['block2g_project_conv[0][0]']
block2g_drop (Dropout)	(None, 56, 56, 64)	0	['block2g_project_bn[0][0]']
block2g_add (Add)	(None, 56, 56, 64)	0	['block2g_drop[0][0]', 'block2f_add[0][0]']
block3a_expand_conv (Conv2D)	(None, 28, 28, 256)	147456	['block2g_add[0][0]']
block3a_expand_bn (Batch Normalization)	(None, 28, 28, 256)	1024	['block3a_expand_conv[0][0]']
block3a_expand_activation (Activation)	(None, 28, 28, 256)	0	['block3a_expand_bn[0][0]']
block3a_project_conv (Conv2D)	(None, 28, 28, 96)	24576	['block3a_expand_activation[0][0]']
block3a_project_bn (Batch Normalization)	(None, 28, 28, 96)	384	['block3a_project_conv[0][0]']
block3b_expand_conv (Conv2D)	(None, 28, 28, 384)	331776	['block3a_project_bn[0][0]']
block3b_expand_bn (Batch Normalization)	(None, 28, 28, 384)	1536	['block3b_expand_conv[0][0]']
block3b_expand_activation (Activation)	(None, 28, 28, 384)	0	['block3b_expand_bn[0][0]']
block3b_project_conv (Conv2D)	(None, 28, 28, 96)	36864	['block3b_expand_activation[0][0]']
block3b_project_bn (Batch Normalization)	(None, 28, 28, 96)	384	['block3b_project_conv[0][0]']
block3b_drop (Dropout)	(None, 28, 28, 96)	0	['block3b_project_bn[0][0]']
block3b_add (Add)	(None, 28, 28, 96)	0	['block3b_drop[0][0]', 'block3b_project_bn[0][0]']

b_drop[0][0]',			'block3
a_project_bn[0][0]']			
block3c_expand_conv (Conv2D)	(None, 28, 28, 384)	331776	['block3
b_add[0][0]']			
block3c_expand_bn (BatchNormalization)	(None, 28, 28, 384)	1536	['block3
c_expand_conv[0][0]']			
block3c_expand_activation (Activation)	(None, 28, 28, 384)	0	['block3
c_expand_bn[0][0]']			
block3c_project_conv (Conv2D)	(None, 28, 28, 96)	36864	['block3
c_expand_activation[0]			[0]']
block3c_project_bn (BatchNormalization)	(None, 28, 28, 96)	384	['block3
c_project_conv[0][0]']			
block3c_drop (Dropout)	(None, 28, 28, 96)	0	['block3
c_project_bn[0][0]']			
block3c_add (Add)	(None, 28, 28, 96)	0	['block3
c_drop[0][0]',			'block3
b_add[0][0]']			
block3d_expand_conv (Conv2D)	(None, 28, 28, 384)	331776	['block3
c_add[0][0]']			
block3d_expand_bn (BatchNormalization)	(None, 28, 28, 384)	1536	['block3
d_expand_conv[0][0]']			
block3d_expand_activation (Activation)	(None, 28, 28, 384)	0	['block3
d_expand_bn[0][0]']			
block3d_project_conv (Conv2D)	(None, 28, 28, 96)	36864	['block3
d_expand_activation[0]			[0]']
block3d_project_bn (BatchNormalization)	(None, 28, 28, 96)	384	['block3
d_project_conv[0][0]']			
block3d_drop (Dropout)	(None, 28, 28, 96)	0	['block3
d_project_bn[0][0]']			
block3d_add (Add)	(None, 28, 28, 96)	0	['block3

d_drop[0][0]',			'block3
c_add[0][0]']			
block3e_expand_conv (Conv2D)	(None, 28, 28, 384)	331776	['block3
d_add[0][0]']			
block3e_expand_bn (BatchNormalization)	(None, 28, 28, 384)	1536	['block3
e_expand_conv[0][0]']			
block3e_expand_activation (Activation)	(None, 28, 28, 384)	0	['block3
e_expand_bn[0][0]']			
block3e_project_conv (Conv2D)	(None, 28, 28, 96)	36864	['block3
e_expand_activation[0]			[0]']
block3e_project_bn (BatchNormalization)	(None, 28, 28, 96)	384	['block3
e_project_conv[0][0]']			
block3e_drop (Dropout)	(None, 28, 28, 96)	0	['block3
e_project_bn[0][0]']			
block3e_add (Add)	(None, 28, 28, 96)	0	['block3
e_drop[0][0]',			'block3
d_add[0][0]']			
block3f_expand_conv (Conv2D)	(None, 28, 28, 384)	331776	['block3
e_add[0][0]']			
block3f_expand_bn (BatchNormalization)	(None, 28, 28, 384)	1536	['block3
f_expand_conv[0][0]']			
block3f_expand_activation (Activation)	(None, 28, 28, 384)	0	['block3
f_expand_bn[0][0]']			
block3f_project_conv (Conv2D)	(None, 28, 28, 96)	36864	['block3
f_expand_activation[0]			[0]']
block3f_project_bn (BatchNormalization)	(None, 28, 28, 96)	384	['block3
f_project_conv[0][0]']			
block3f_drop (Dropout)	(None, 28, 28, 96)	0	['block3
f_project_bn[0][0]']			
block3f_add (Add)	(None, 28, 28, 96)	0	['block3

f_drop[0][0]',			'block3
e_add[0][0]']			
block3g_expand_conv (Conv2D)	(None, 28, 28, 384)	331776	['block3
f_add[0][0]']			
block3g_expand_bn (BatchNormalization)	(None, 28, 28, 384)	1536	['block3
g_expand_conv[0][0]']			
block3g_expand_activation (Activation)	(None, 28, 28, 384)	0	['block3
g_expand_bn[0][0]']			
block3g_project_conv (Conv2D)	(None, 28, 28, 96)	36864	['block3
g_expand_activation[0]			[0]']
block3g_project_bn (BatchNormalization)	(None, 28, 28, 96)	384	['block3
g_project_conv[0][0]']			
block3g_drop (Dropout)	(None, 28, 28, 96)	0	['block3
g_project_bn[0][0]']			
block3g_add (Add)	(None, 28, 28, 96)	0	['block3
g_drop[0][0]',			'block3
f_add[0][0]']			
block4a_expand_conv (Conv2D)	(None, 28, 28, 384)	36864	['block3
g_add[0][0]']			
block4a_expand_bn (BatchNormalization)	(None, 28, 28, 384)	1536	['block4
a_expand_conv[0][0]']			
block4a_expand_activation (Activation)	(None, 28, 28, 384)	0	['block4
a_expand_bn[0][0]']			
block4a_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 384)	3456	['block4
a_expand_activation[0]			[0]']
block4a_bn (BatchNormalization)	(None, 14, 14, 384)	1536	['block4
a_dwconv2[0][0]']			
block4a_activation (Activation)	(None, 14, 14, 384)	0	['block4
a_bn[0][0]']			

block4a_se_squeeze (Global a_activation[0][0]') AveragePooling2D)	(None, 384)	0	['block4
block4a_se_reshape (Reshap a_se_squeeze[0][0]') e)	(None, 1, 1, 384)	0	['block4
block4a_se_reduce (Conv2D) a_se_reshape[0][0]')	(None, 1, 1, 24)	9240	['block4
block4a_se_expand (Conv2D) a_se_reduce[0][0]')	(None, 1, 1, 384)	9600	['block4
block4a_se_excite (Multipl a_activation[0][0]'), y) a_se_expand[0][0]')	(None, 14, 14, 384)	0	['block4  'block4
block4a_project_conv (Conv a_se_excite[0][0]') 2D)	(None, 14, 14, 192)	73728	['block4
block4a_project_bn (BatchN a_project_conv[0][0]') ormalization)	(None, 14, 14, 192)	768	['block4
block4b_expand_conv (Conv2 a_project_bn[0][0]') D)	(None, 14, 14, 768)	147456	['block4
block4b_expand_bn (BatchNo b_expand_conv[0][0]') rmalization)	(None, 14, 14, 768)	3072	['block4
block4b_expand_activation b_expand_bn[0][0]') (Activation)	(None, 14, 14, 768)	0	['block4
block4b_dwconv2 (Depthwise b_expand_activation[0] Conv2D)	(None, 14, 14, 768)	6912	['block4  [0]']
block4b_bn (BatchNormaliza b_dwconv2[0][0]') tion)	(None, 14, 14, 768)	3072	['block4
block4b_activation (Activa b_bn[0][0]') tion)	(None, 14, 14, 768)	0	['block4
block4b_se_squeeze (Global b_activation[0][0]') AveragePooling2D)	(None, 768)	0	['block4
block4b_se_reshape (Reshap (None, 1, 1, 768)	(None, 1, 1, 768)	0	['block4

b_se_squeeze[0][0]'] e)			
block4b_se_reduce (Conv2D) b_se_reshape[0][0]']	(None, 1, 1, 48)	36912	['block4
block4b_se_expand (Conv2D) b_se_reduce[0][0]']	(None, 1, 1, 768)	37632	['block4
block4b_se_excite (Multipl b_activation[0][0]'], y) b_se_expand[0][0]']	(None, 14, 14, 768)	0	['block4  'block4
block4b_project_conv (Conv b_se_excite[0][0]'] 2D)	(None, 14, 14, 192)	147456	['block4
block4b_project_bn (BatchN b_project_conv[0][0]'] ormalization)	(None, 14, 14, 192)	768	['block4
block4b_drop (Dropout) b_project_bn[0][0]']	(None, 14, 14, 192)	0	['block4
block4b_add (Add) b_drop[0][0]'], a_project_bn[0][0]']	(None, 14, 14, 192)	0	['block4  'block4
block4c_expand_conv (Conv2 b_add[0][0]'] D)	(None, 14, 14, 768)	147456	['block4
block4c_expand_bn (BatchNo c_expand_conv[0][0]'] rmalization)	(None, 14, 14, 768)	3072	['block4
block4c_expand_activation c_expand_bn[0][0]'] (Activation)	(None, 14, 14, 768)	0	['block4
block4c_dwconv2 (Depthwise c_expand_activation[0] Conv2D)	(None, 14, 14, 768)	6912	['block4  [0]']
block4c_bn (BatchNormaliza c_dwconv2[0][0]'] tion)	(None, 14, 14, 768)	3072	['block4
block4c_activation (Activa c_bn[0][0]'] tion)	(None, 14, 14, 768)	0	['block4
block4c_se_squeeze (Global c_activation[0][0]']	(None, 768)	0	['block4



AveragePooling2D)				
block4c_se_reshape (Reshape)	(None, 1, 1, 768)	0	['block4c_se_squeeze[0][0]']	
c_se_squeeze[0][0]') e)				
block4c_se_reduce (Conv2D)	(None, 1, 1, 48)	36912	['block4c_se_reshape[0][0]']	
c_se_reshape[0][0]']				
block4c_se_expand (Conv2D)	(None, 1, 1, 768)	37632	['block4c_se_reduce[0][0]']	
c_se_reduce[0][0]']				
block4c_se_excite (Multipl	(None, 14, 14, 768)	0	['block4c_activation[0][0]']	
c_activation[0][0]']			y)	
y)			['block4c_se_expand[0][0]']	
c_se_expand[0][0]']				
block4c_project_conv (Conv	(None, 14, 14, 192)	147456	['block4c_se_excite[0][0]']	
c_se_excite[0][0]']			2D)	
2D)				
block4c_project_bn (BatchN	(None, 14, 14, 192)	768	['block4c_project_conv[0][0]']	
c_project_conv[0][0]']			ormalization)	
ormalization)				
block4c_drop (Dropout)	(None, 14, 14, 192)	0	['block4c_project_bn[0][0]']	
c_project_bn[0][0]']				
block4c_add (Add)	(None, 14, 14, 192)	0	['block4c_drop[0][0]']	
c_drop[0][0]']			b_add[0][0]']	
b_add[0][0]']				
block4d_expand_conv (Conv2	(None, 14, 14, 768)	147456	['block4c_add[0][0]']	
c_add[0][0]']			D)	
D)				
block4d_expand_bn (BatchNo	(None, 14, 14, 768)	3072	['block4d_expand_conv[0][0]']	
d_expand_conv[0][0]']			ormalization)	
ormalization)				
block4d_expand_activation	(None, 14, 14, 768)	0	['block4d_expand_bn[0][0]']	
d_expand_bn[0][0]']			(Activation)	
(Activation)				
block4d_dwconv2 (Depthwise	(None, 14, 14, 768)	6912	['block4d_expand_activation[0]	
d_expand_activation[0]			Conv2D)	
Conv2D)			[0]']	
[0]']				
block4d_bn (BatchNormaliza	(None, 14, 14, 768)	3072	['block4d_dwconv2[0][0]']	
d_dwconv2[0][0]']			tion)	
tion)				
block4d_activation (Activa	(None, 14, 14, 768)	0	['block4d_bn[0][0]']	
d_bn[0][0]']			tion)	
tion)				

block4d_se_squeeze (Global d_activation[0][0]') AveragePooling2D)	(None, 768)	0	['block4
block4d_se_reshape (Reshap d_se_squeeze[0][0]') e)	(None, 1, 1, 768)	0	['block4
block4d_se_reduce (Conv2D) d_se_reshape[0][0]')	(None, 1, 1, 48)	36912	['block4
block4d_se_expand (Conv2D) d_se_reduce[0][0]')	(None, 1, 1, 768)	37632	['block4
block4d_se_excite (Multipl d_activation[0][0]'), y) d_se_expand[0][0]')	(None, 14, 14, 768)	0	['block4 'block4
block4d_project_conv (Conv d_se_excite[0][0]') 2D)	(None, 14, 14, 192)	147456	['block4
block4d_project_bn (BatchN d_project_conv[0][0]') ormalization)	(None, 14, 14, 192)	768	['block4
block4d_drop (Dropout) d_project_bn[0][0]')	(None, 14, 14, 192)	0	['block4
block4d_add (Add) d_drop[0][0]'), c_add[0][0]')	(None, 14, 14, 192)	0	['block4 'block4
block4e_expand_conv (Conv2 d_add[0][0]') D)	(None, 14, 14, 768)	147456	['block4
block4e_expand_bn (BatchNo e_expand_conv[0][0]') rmalization)	(None, 14, 14, 768)	3072	['block4
block4e_expand_activation e_expand_bn[0][0]') (Activation)	(None, 14, 14, 768)	0	['block4
block4e_dwconv2 (Depthwise e_expand_activation[0] Conv2D)	(None, 14, 14, 768)	6912	['block4 [0]']
block4e_bn (BatchNormaliza e_dwconv2[0][0]') tion)	(None, 14, 14, 768)	3072	['block4

block4e_activation (Activation) e_bn[0][0]']	(None, 14, 14, 768)	0	['block4
block4e_se_squeeze (Global e_activation[0][0]'] AveragePooling2D)	(None, 768)	0	['block4
block4e_se_reshape (Reshape) e_se_squeeze[0][0]'] e)	(None, 1, 1, 768)	0	['block4
block4e_se_reduce (Conv2D) e_se_reshape[0][0]']	(None, 1, 1, 48)	36912	['block4
block4e_se_expand (Conv2D) e_se_reduce[0][0]']	(None, 1, 1, 768)	37632	['block4
block4e_se_excite (Multiply) e_activation[0][0]'], y) e_se_expand[0][0]']	(None, 14, 14, 768)	0	['block4  'block4
block4e_project_conv (Conv2D) e_se_excite[0][0]'] 2D)	(None, 14, 14, 192)	147456	['block4
block4e_project_bn (BatchNormalization) e_project_conv[0][0]']	(None, 14, 14, 192)	768	['block4
block4e_drop (Dropout) e_project_bn[0][0]']	(None, 14, 14, 192)	0	['block4
block4e_add (Add) e_drop[0][0]'], d_add[0][0]']	(None, 14, 14, 192)	0	['block4  'block4
block4f_expand_conv (Conv2D) e_add[0][0]'] D)	(None, 14, 14, 768)	147456	['block4
block4f_expand_bn (BatchNormalization) f_expand_conv[0][0]']	(None, 14, 14, 768)	3072	['block4
block4f_expand_activation f_expand_bn[0][0]'] (Activation)	(None, 14, 14, 768)	0	['block4
block4f_dwconv2 (Depthwise f_expand_activation[0] Conv2D)	(None, 14, 14, 768)	6912	['block4  [0]']
block4f_bn (BatchNormalization)	(None, 14, 14, 768)	3072	['block4

f_dwconv2[0][0]'] tion)			
block4f_activation (Active (None, 14, 14, 768) f_bn[0][0]'] tion)	0		['block4
block4f_se_squeeze (Global (None, 768) f_activation[0][0]'] AveragePooling2D)	0		['block4
block4f_se_reshape (Reshap (None, 1, 1, 768) f_se_squeeze[0][0]'] e)	0		['block4
block4f_se_reduce (Conv2D) (None, 1, 1, 48) f_se_reshape[0][0]']	36912		['block4
block4f_se_expand (Conv2D) (None, 1, 1, 768) f_se_reduce[0][0]']	37632		['block4
block4f_se_excite (Multipl (None, 14, 14, 768) f_activation[0][0]',' y) f_se_expand[0][0]']	0		['block4 'block4
block4f_project_conv (Conv (None, 14, 14, 192) f_se_excite[0][0]'] 2D)	147456		['block4
block4f_project_bn (BatchN (None, 14, 14, 192) f_project_conv[0][0]'] ormalization)	768		['block4
block4f_drop (Dropout) (None, 14, 14, 192) f_project_bn[0][0]']	0		['block4
block4f_add (Add) (None, 14, 14, 192) f_drop[0][0]',' e_add[0][0]']	0		['block4 'block4
block4g_expand_conv (Conv2 (None, 14, 14, 768) f_add[0][0]'] D)	147456		['block4
block4g_expand_bn (BatchNo (None, 14, 14, 768) g_expand_conv[0][0]'] rmalization)	3072		['block4
block4g_expand_activation (None, 14, 14, 768) g_expand_bn[0][0]'] (Activation)	0		['block4
block4g_dwconv2 (Depthwise (None, 14, 14, 768) g_expand_activation[0]	6912		['block4

Conv2D)			[0]']
block4g_bn (BatchNormaliza g_dwconv2[0][0]') tion)	(None, 14, 14, 768)	3072	['block4
block4g_activation (Activa g_bn[0][0]') tion)	(None, 14, 14, 768)	0	['block4
block4g_se_squeeze (Global g_activation[0][0]') AveragePooling2D)	(None, 768)	0	['block4
block4g_se_reshape (Reshap g_se_squeeze[0][0]') e)	(None, 1, 1, 768)	0	['block4
block4g_se_reduce (Conv2D) g_se_reshape[0][0]')	(None, 1, 1, 48)	36912	['block4
block4g_se_expand (Conv2D) g_se_reduce[0][0]')	(None, 1, 1, 768)	37632	['block4
block4g_se_excite (Multipl g_activation[0][0]', y) g_se_expand[0][0]')	(None, 14, 14, 768)	0	['block4 'block4
block4g_project_conv (Conv g_se_excite[0][0]') 2D)	(None, 14, 14, 192)	147456	['block4
block4g_project_bn (BatchN g_project_conv[0][0]') ormalization)	(None, 14, 14, 192)	768	['block4
block4g_drop (Dropout) g_project_bn[0][0]')	(None, 14, 14, 192)	0	['block4
block4g_add (Add) g_drop[0][0]', f_add[0][0]')	(None, 14, 14, 192)	0	['block4 'block4
block4h_expand_conv (Conv2 g_add[0][0]') D)	(None, 14, 14, 768)	147456	['block4
block4h_expand_bn (BatchNo h_expand_conv[0][0]') rmalization)	(None, 14, 14, 768)	3072	['block4
block4h_expand_activation h_expand_bn[0][0]') (Activation)	(None, 14, 14, 768)	0	['block4

block4h_dwconv2 (Depthwise h_expand_activation[0] Conv2D)	(None, 14, 14, 768)	6912	['block4 [0]']
block4h_bn (BatchNormaliza h_dwconv2[0][0]') tion)	(None, 14, 14, 768)	3072	['block4
block4h_activation (Activa h_bn[0][0]') tion)	(None, 14, 14, 768)	0	['block4
block4h_se_squeeze (Global h_activation[0][0]') AveragePooling2D)	(None, 768)	0	['block4
block4h_se_reshape (Reshap h_se_squeeze[0][0]') e)	(None, 1, 1, 768)	0	['block4
block4h_se_reduce (Conv2D) h_se_reshape[0][0]')	(None, 1, 1, 48)	36912	['block4
block4h_se_expand (Conv2D) h_se_reduce[0][0]')	(None, 1, 1, 768)	37632	['block4
block4h_se_excite (Multipl h_activation[0][0]', y) h_se_expand[0][0]')	(None, 14, 14, 768)	0	['block4  'block4
block4h_project_conv (Conv h_se_excite[0][0]') 2D)	(None, 14, 14, 192)	147456	['block4
block4h_project_bn (BatchN h_project_conv[0][0]') ormalization)	(None, 14, 14, 192)	768	['block4
block4h_drop (Dropout) h_project_bn[0][0]')	(None, 14, 14, 192)	0	['block4
block4h_add (Add) h_drop[0][0]', g_add[0][0]')	(None, 14, 14, 192)	0	['block4  'block4
block4i_expand_conv (Conv2 h_add[0][0]') D)	(None, 14, 14, 768)	147456	['block4
block4i_expand_bn (BatchNo i_expand_conv[0][0]') rmalization)	(None, 14, 14, 768)	3072	['block4

block4i_expand_activation i_expand_bn[0][0]'] (Activation)	(None, 14, 14, 768)	0	['block4
block4i_dwconv2 (Depthwise i_expand_activation[0] Conv2D)	(None, 14, 14, 768)	6912	['block4 [0]']
block4i_bn (BatchNormaliza i_dwconv2[0][0]'] tion)	(None, 14, 14, 768)	3072	['block4
block4i_activation (Activa i_bn[0][0]'] tion)	(None, 14, 14, 768)	0	['block4
block4i_se_squeeze (Global i_activation[0][0]'] AveragePooling2D)	(None, 768)	0	['block4
block4i_se_reshape (Reshap i_se_squeeze[0][0]'] e)	(None, 1, 1, 768)	0	['block4
block4i_se_reduce (Conv2D) i_se_reshape[0][0]']	(None, 1, 1, 48)	36912	['block4
block4i_se_expand (Conv2D) i_se_reduce[0][0]']	(None, 1, 1, 768)	37632	['block4
block4i_se_excite (Multipl i_activation[0][0]'], y) i_se_expand[0][0]']	(None, 14, 14, 768)	0	['block4 'block4
block4i_project_conv (Conv i_se_excite[0][0]'] 2D)	(None, 14, 14, 192)	147456	['block4
block4i_project_bn (BatchN i_project_conv[0][0]'] ormalization)	(None, 14, 14, 192)	768	['block4
block4i_drop (Dropout) i_project_bn[0][0]']	(None, 14, 14, 192)	0	['block4
block4i_add (Add) i_drop[0][0]'], h_add[0][0]']	(None, 14, 14, 192)	0	['block4 'block4
block4j_expand_conv (Conv2 i_add[0][0]'] D)	(None, 14, 14, 768)	147456	['block4
block4j_expand_bn (BatchNo i_expand_conv[0][0]']	(None, 14, 14, 768)	3072	['block4

j_expand_conv[0][0]'] rmalization)			
block4j_expand_activation j_expand_bn[0][0]'] (Activation)	(None, 14, 14, 768)	0	['block4
block4j_dwconv2 (Depthwise j_expand_activation[0] Conv2D)	(None, 14, 14, 768)	6912	['block4 [0]']
block4j_bn (BatchNormaliza j_dwconv2[0][0]'] tion)	(None, 14, 14, 768)	3072	['block4
block4j_activation (Activa j_bn[0][0]'] tion)	(None, 14, 14, 768)	0	['block4
block4j_se_squeeze (Global j_activation[0][0]'] AveragePooling2D)	(None, 768)	0	['block4
block4j_se_reshape (Reshap j_se_squeeze[0][0]'] e)	(None, 1, 1, 768)	0	['block4
block4j_se_reduce (Conv2D) j_se_reshape[0][0]']	(None, 1, 1, 48)	36912	['block4
block4j_se_expand (Conv2D) j_se_reduce[0][0]']	(None, 1, 1, 768)	37632	['block4
block4j_se_excite (Multipl j_activation[0][0]'], y) j_se_expand[0][0]']	(None, 14, 14, 768)	0	['block4 'block4
block4j_project_conv (Conv j_se_excite[0][0]'] 2D)	(None, 14, 14, 192)	147456	['block4
block4j_project_bn (BatchN j_project_conv[0][0]'] ormalization)	(None, 14, 14, 192)	768	['block4
block4j_drop (Dropout) j_project_bn[0][0]']	(None, 14, 14, 192)	0	['block4
block4j_add (Add) j_drop[0][0]'], i_add[0][0]']	(None, 14, 14, 192)	0	['block4 'block4
block5a_expand_conv (Conv2 j_add[0][0]']	(None, 14, 14, 1152)	221184	['block4



D)

block5a_expand_bn (BatchNormaliza a_expand_conv[0][0]') rmalization)	(None, 14, 14, 1152)	4608	['block5
block5a_expand_activation a_expand_bn[0][0]') (Activation)	(None, 14, 14, 1152)	0	['block5
block5a_dwconv2 (Depthwise a_expand_activation[0] Conv2D)	(None, 14, 14, 1152)	10368	['block5 [0]']
block5a_bn (BatchNormaliza a_dwconv2[0][0]') tion)	(None, 14, 14, 1152)	4608	['block5
block5a_activation (Activa a_bn[0][0]') tion)	(None, 14, 14, 1152)	0	['block5
block5a_se_squeeze (Global a_activation[0][0]') AveragePooling2D)	(None, 1152)	0	['block5
block5a_se_reshape (Reshap a_se_squeeze[0][0]') e)	(None, 1, 1, 1152)	0	['block5
block5a_se_reduce (Conv2D) a_se_reshape[0][0]')	(None, 1, 1, 48)	55344	['block5
block5a_se_expand (Conv2D) a_se_reduce[0][0]')	(None, 1, 1, 1152)	56448	['block5
block5a_se_excite (Multipl a_activation[0][0]'), y) a_se_expand[0][0]')	(None, 14, 14, 1152)	0	['block5 'block5
block5a_project_conv (Conv a_se_excite[0][0]') 2D)	(None, 14, 14, 224)	258048	['block5
block5a_project_bn (BatchN a_project_conv[0][0]') ormalization)	(None, 14, 14, 224)	896	['block5
block5b_expand_conv (Conv2 a_project_bn[0][0]') D)	(None, 14, 14, 1344)	301056	['block5
block5b_expand_bn (BatchNo b_expand_conv[0][0]') rmalization)	(None, 14, 14, 1344)	5376	['block5

block5b_expand_activation b_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5b_dwconv2 (Depthwise b_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5b_bn (BatchNormaliza b_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5b_activation (Activa b_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5
block5b_se_squeeze (Global b_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5b_se_reshape (Reshap b_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5b_se_reduce (Conv2D) b_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5b_se_expand (Conv2D) b_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5
block5b_se_excite (Multipl b_activation[0][0]'], y) b_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5 'block5
block5b_project_conv (Conv b_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5b_project_bn (BatchN b_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5
block5b_drop (Dropout) b_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5b_add (Add) b_drop[0][0]'], a_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5 'block5
block5c_expand_conv (Conv2 b_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5

block5c_expand_bn (BatchNormalization)	(None, 14, 14, 1344)	5376	['block5c_expand_conv[0][0]']
block5c_expand_activation (Activation)	(None, 14, 14, 1344)	0	['block5c_expand_bn[0][0]']
block5c_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 1344)	12096	['block5c_expand_activation[0][0]']
block5c_bn (BatchNormalization)	(None, 14, 14, 1344)	5376	['block5c_dwconv2[0][0]']
block5c_activation (Activation)	(None, 14, 14, 1344)	0	['block5c_bn[0][0]']
block5c_se_squeeze (Global AveragePooling2D)	(None, 1344)	0	['block5c_activation[0][0]']
block5c_se_reshape (Reshape)	(None, 1, 1, 1344)	0	['block5c_se_squeeze[0][0]']
block5c_se_reduce (Conv2D)	(None, 1, 1, 56)	75320	['block5c_se_reshape[0][0]']
block5c_se_expand (Conv2D)	(None, 1, 1, 1344)	76608	['block5c_se_reduce[0][0]']
block5c_se_excite (Multiply)	(None, 14, 14, 1344)	0	['block5c_se_expand[0][0]']
block5c_project_conv (Conv2D)	(None, 14, 14, 224)	301056	['block5c_se_excite[0][0]']
block5c_project_bn (BatchNormalization)	(None, 14, 14, 224)	896	['block5c_project_conv[0][0]']
block5c_drop (Dropout)	(None, 14, 14, 224)	0	['block5c_project_bn[0][0]']
block5c_add (Add)	(None, 14, 14, 224)	0	['block5c_drop[0][0]']
block5d_expand_conv (Conv2D)	(None, 14, 14, 1344)	301056	['block5c_add[0][0]']

c_add[0][0]'] D)			
block5d_expand_bn (BatchNo d_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5d_expand_activation d_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5d_dwconv2 (Depthwise d_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5d_bn (BatchNormaliza d_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5d_activation (Activa d_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5
block5d_se_squeeze (Global d_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5d_se_reshape (Reshap d_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5d_se_reduce (Conv2D) d_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5d_se_expand (Conv2D) d_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5
block5d_se_excite (Multipl d_activation[0][0]'], y) d_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5 'block5
block5d_project_conv (Conv d_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5d_project_bn (BatchN d_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5
block5d_drop (Dropout) d_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5d_add (Add) d_drop[0][0]'],	(None, 14, 14, 224)	0	['block5 'block5

c_add[0][0]']			
block5e_expand_conv (Conv2D)	(None, 14, 14, 1344)	301056	['block5d_add[0][0]']
block5e_expand_bn (BatchNormalization)	(None, 14, 14, 1344)	5376	['block5e_expand_conv[0][0]']
block5e_expand_activation (Activation)	(None, 14, 14, 1344)	0	['block5e_expand_bn[0][0]']
block5e_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 1344)	12096	['block5e_expand_activation[0][0]']
block5e_bn (BatchNormalization)	(None, 14, 14, 1344)	5376	['block5e_dwconv2[0][0]']
block5e_activation (Activation)	(None, 14, 14, 1344)	0	['block5e_bn[0][0]']
block5e_se_squeeze (Global AveragePooling2D)	(None, 1344)	0	['block5e_activation[0][0]']
block5e_se_reshape (Reshape)	(None, 1, 1, 1344)	0	['block5e_se_squeeze[0][0]']
block5e_se_reduce (Conv2D)	(None, 1, 1, 56)	75320	['block5e_se_reshape[0][0]']
block5e_se_expand (Conv2D)	(None, 1, 1, 1344)	76608	['block5e_se_reduce[0][0]']
block5e_se_excite (Multiply)	(None, 14, 14, 1344)	0	['block5e_se_expand[0][0]']
block5e_project_conv (Conv2D)	(None, 14, 14, 224)	301056	['block5e_se_excite[0][0]']
block5e_project_bn (BatchNormalization)	(None, 14, 14, 224)	896	['block5e_project_conv[0][0]']
block5e_drop (Dropout)	(None, 14, 14, 224)	0	['block5e_project_bn[0][0]']

block5e_add (Add) e_drop[0][0]'], d_add[0][0]']	(None, 14, 14, 224)	0	['block5  'block5
block5f_expand_conv (Conv2 e_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5
block5f_expand_bn (BatchNo f_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5f_expand_activation f_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5f_dwconv2 (Depthwise f_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5  [0]']
block5f_bn (BatchNormaliza f_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5f_activation (Activa f_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5
block5f_se_squeeze (Global f_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5f_se_reshape (Reshap f_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5f_se_reduce (Conv2D) f_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5f_se_expand (Conv2D) f_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5
block5f_se_excite (Multipl f_activation[0][0]'], y) f_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5  'block5
block5f_project_conv (Conv f_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5f_project_bn (BatchN f_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5

block5f_drop (Dropout) f_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5f_add (Add) f_drop[0][0]', e_add[0][0]']	(None, 14, 14, 224)	0	['block5 'block5
block5g_expand_conv (Conv2 f_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5
block5g_expand_bn (BatchNo g_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5g_expand_activation g_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5g_dwconv2 (Depthwise g_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5g_bn (BatchNormaliza g_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5g_activation (Activa g_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5
block5g_se_squeeze (Global g_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5g_se_reshape (Reshap g_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5g_se_reduce (Conv2D) g_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5g_se_expand (Conv2D) g_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5
block5g_se_excite (Multipl g_activation[0][0]'], y) g_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5 'block5
block5g_project_conv (Conv g_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5g_project_bn (BatchN	(None, 14, 14, 224)	896	['block5

g_project_conv[0][0]'] ormalization)			
block5g_drop (Dropout) g_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5g_add (Add) g_drop[0][0]',	(None, 14, 14, 224)	0	['block5
f_add[0][0]']			'block5
block5h_expand_conv (Conv2 g_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5
block5h_expand_bn (BatchNo h_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5h_expand_activation h_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5h_dwconv2 (Depthwise h_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5h_bn (BatchNormaliza h_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5h_activation (Activa h_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5
block5h_se_squeeze (Global h_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5h_se_reshape (Reshap h_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5h_se_reduce (Conv2D) h_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5h_se_expand (Conv2D) h_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5
block5h_se_excite (Multipl h_activation[0][0]', y) h_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5 'block5
block5h_project_conv (Conv h_se_excite[0][0]']	(None, 14, 14, 224)	301056	['block5



2D)			
block5h_project_bn (Batch Normalization)	(None, 14, 14, 224)	896	['block5h_project_conv[0][0]']
block5h_drop (Dropout)	(None, 14, 14, 224)	0	['block5h_project_bn[0][0]']
block5h_add (Add)	(None, 14, 14, 224)	0	['block5h_drop[0][0]', 'block5g_add[0][0]']
block5i_expand_conv (Conv2D)	(None, 14, 14, 1344)	301056	['block5h_add[0][0]']
block5i_expand_bn (Batch Normalization)	(None, 14, 14, 1344)	5376	['block5i_expand_conv[0][0]']
block5i_expand_activation (Activation)	(None, 14, 14, 1344)	0	['block5i_expand_bn[0][0]']
block5i_dwconv2 (Depthwise Conv2D)	(None, 14, 14, 1344)	12096	['block5i_expand_activation[0][0]']
block5i_bn (Batch Normalization)	(None, 14, 14, 1344)	5376	['block5i_dwconv2[0][0]']
block5i_activation (Activation)	(None, 14, 14, 1344)	0	['block5i_bn[0][0]']
block5i_se_squeeze (Global AveragePooling2D)	(None, 1344)	0	['block5i_activation[0][0]']
block5i_se_reshape (Reshape)	(None, 1, 1, 1344)	0	['block5i_se_squeeze[0][0]']
block5i_se_reduce (Conv2D)	(None, 1, 1, 56)	75320	['block5i_se_reshape[0][0]']
block5i_se_expand (Conv2D)	(None, 1, 1, 1344)	76608	['block5i_se_reduce[0][0]']
block5i_se_excite (Multiply)	(None, 14, 14, 1344)	0	['block5i_se_expand[0][0]', 'block5i_activation[0][0]']

block5i_project_conv (Conv2D) i_se_excite[0][0]'	(None, 14, 14, 224)	301056	['block5
block5i_project_bn (BatchNormalization) i_project_conv[0][0]'	(None, 14, 14, 224)	896	['block5
block5i_drop (Dropout) i_project_bn[0][0]'	(None, 14, 14, 224)	0	['block5
block5i_add (Add) i_drop[0][0]', h_add[0][0]'	(None, 14, 14, 224)	0	['block5  'block5
block5j_expand_conv (Conv2D) i_add[0][0]'	(None, 14, 14, 1344)	301056	['block5
block5j_expand_bn (BatchNormalization) j_expand_conv[0][0]'	(None, 14, 14, 1344)	5376	['block5
block5j_expand_activation j_expand_bn[0][0]' (Activation)	(None, 14, 14, 1344)	0	['block5
block5j_dwconv2 (Depthwise Conv2D) j_expand_activation[0]	(None, 14, 14, 1344)	12096	['block5  [0]'
block5j_bn (BatchNormalization) j_dwconv2[0][0]'	(None, 14, 14, 1344)	5376	['block5
block5j_activation (Activation) j_bn[0][0]'	(None, 14, 14, 1344)	0	['block5
block5j_se_squeeze (Global AveragePooling2D) j_activation[0][0]'	(None, 1344)	0	['block5
block5j_se_reshape (Reshape) j_se_squeeze[0][0]'	(None, 1, 1, 1344)	0	['block5
block5j_se_reduce (Conv2D) j_se_reshape[0][0]'	(None, 1, 1, 56)	75320	['block5
block5j_se_expand (Conv2D) j_se_reduce[0][0]'	(None, 1, 1, 1344)	76608	['block5
block5j_se_excite (Multipl	(None, 14, 14, 1344)	0	['block5

j_activation[0][0]', y)			'block5
j_se_expand[0][0]']			
block5j_project_conv (Conv j_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5j_project_bn (BatchN j_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5
block5j_drop (Dropout) j_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5j_add (Add) j_drop[0][0]'],	(None, 14, 14, 224)	0	['block5
i_add[0][0]']			'block5
block5k_expand_conv (Conv2 j_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5
block5k_expand_bn (BatchNo k_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5k_expand_activation k_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5k_dwconv2 (Depthwise k_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5k_bn (BatchNormaliza k_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5k_activation (Activa k_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5
block5k_se_squeeze (Global k_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5k_se_reshape (Reshap k_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5k_se_reduce (Conv2D) k_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5k_se_expand (Conv2D)	(None, 1, 1, 1344)	76608	['block5

k_se_reduce[0][0]']			
block5k_se_excite (Multipl k_activation[0][0]', y) k_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5 'block5
block5k_project_conv (Conv k_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5k_project_bn (BatchN k_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5
block5k_drop (Dropout) k_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5k_add (Add) k_drop[0][0]', j_add[0][0]']	(None, 14, 14, 224)	0	['block5 'block5
block5l_expand_conv (Conv2 k_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5
block5l_expand_bn (BatchNo l_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5l_expand_activation l_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5l_dwconv2 (Depthwise l_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5l_bn (BatchNormaliza l_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5l_activation (Activa l_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5
block5l_se_squeeze (Global l_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5l_se_reshape (Reshap l_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5l_se_reduce (Conv2D)	(None, 1, 1, 56)	75320	['block5

l_se_reshape[0][0]']				
block5l_se_expand (Conv2D) l_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5	
block5l_se_excite (Multipl l_activation[0][0]',' y) l_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5 'block5	
block5l_project_conv (Conv l_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5	
block5l_project_bn (BatchN l_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5	
block5l_drop (Dropout) l_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5	
block5l_add (Add) l_drop[0][0]',' k_add[0][0]']	(None, 14, 14, 224)	0	['block5 'block5	
block5m_expand_conv (Conv2 l_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5	
block5m_expand_bn (BatchNo m_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5	
block5m_expand_activation m_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5	
block5m_dwconv2 (Depthwise m_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']	
block5m_bn (BatchNormaliza m_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5	
block5m_activation (Activa m_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5	
block5m_se_squeeze (Global m_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5	
block5m_se_reshape (Reshap m_se_squeeze[0][0]']	(None, 1, 1, 1344)	0	['block5	

e)

block5m_se_reduce (Conv2D)	(None, 1, 1, 56)	75320	['block5 m_se_reshape[0][0]']
block5m_se_expand (Conv2D)	(None, 1, 1, 1344)	76608	['block5 m_se_reduce[0][0]']
block5m_se_excite (Multipl m_activation[0][0]'	(None, 14, 14, 1344)	0	['block5 y) m_se_expand[0][0]']
block5m_project_conv (Conv m_se_excite[0][0]'	(None, 14, 14, 224)	301056	['block5 2D)
block5m_project_bn (BatchN m_project_conv[0][0]'	(None, 14, 14, 224)	896	['block5 ormalization)
block5m_drop (Dropout)	(None, 14, 14, 224)	0	['block5 m_project_bn[0][0]']
block5m_add (Add)	(None, 14, 14, 224)	0	['block5 m_drop[0][0]'
l_add[0][0]']			['block5 l_add[0][0]']
block5n_expand_conv (Conv2 m_add[0][0]']	(None, 14, 14, 1344)	301056	['block5 D)
block5n_expand_bn (BatchNo n_expand_conv[0][0]'	(None, 14, 14, 1344)	5376	['block5 rmalization)
block5n_expand_activation n_expand_bn[0][0]']	(None, 14, 14, 1344)	0	['block5 (Activation)
block5n_dwconv2 (Depthwise n_expand_activation[0]	(None, 14, 14, 1344)	12096	['block5 Conv2D)
block5n_bn (BatchNormaliza n_dwconv2[0][0]']	(None, 14, 14, 1344)	5376	['block5 tion)
block5n_activation (Activa n_bn[0][0]']	(None, 14, 14, 1344)	0	['block5 tion)
block5n_se_squeeze (Global n_activation[0][0]']	(None, 1344)	0	['block5 AveragePooling2D)

block5n_se_reshape (Reshape) n_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5n_se_reduce (Conv2D) n_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5n_se_expand (Conv2D) n_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5
block5n_se_excite (Multipl n_activation[0][0]'], y) n_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5  'block5
block5n_project_conv (Conv n_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5n_project_bn (BatchN n_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5
block5n_drop (Dropout) n_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5n_add (Add) n_drop[0][0]'], m_add[0][0]']	(None, 14, 14, 224)	0	['block5  'block5
block5o_expand_conv (Conv2 n_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5
block5o_expand_bn (BatchNo o_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5o_expand_activation o_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5o_dwconv2 (Depthwise o_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5  [0]']
block5o_bn (BatchNormaliza o_dwconv2[0][0]'] tion)	(None, 14, 14, 1344)	5376	['block5
block5o_activation (Activa o_bn[0][0]'] tion)	(None, 14, 14, 1344)	0	['block5

block5o_se_squeeze (Global o_activation[0][0]') AveragePooling2D)	(None, 1344)	0	['block5
block5o_se_reshape (Reshap o_se_squeeze[0][0]') e)	(None, 1, 1, 1344)	0	['block5
block5o_se_reduce (Conv2D) o_se_reshape[0][0]')	(None, 1, 1, 56)	75320	['block5
block5o_se_expand (Conv2D) o_se_reduce[0][0]')	(None, 1, 1, 1344)	76608	['block5
block5o_se_excite (Multipl o_activation[0][0]', y) o_se_expand[0][0]')	(None, 14, 14, 1344)	0	['block5  'block5
block5o_project_conv (Conv o_se_excite[0][0]') 2D)	(None, 14, 14, 224)	301056	['block5
block5o_project_bn (BatchN o_project_conv[0][0]') ormalization)	(None, 14, 14, 224)	896	['block5
block5o_drop (Dropout) o_project_bn[0][0]')	(None, 14, 14, 224)	0	['block5
block5o_add (Add) o_drop[0][0]', n_add[0][0]')	(None, 14, 14, 224)	0	['block5  'block5
block5p_expand_conv (Conv2 o_add[0][0]') D)	(None, 14, 14, 1344)	301056	['block5
block5p_expand_bn (BatchNo p_expand_conv[0][0]') rmalization)	(None, 14, 14, 1344)	5376	['block5
block5p_expand_activation p_expand_bn[0][0]') (Activation)	(None, 14, 14, 1344)	0	['block5
block5p_dwconv2 (Depthwise p_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5  [0]')
block5p_bn (BatchNormaliza p_dwconv2[0][0]') tion)	(None, 14, 14, 1344)	5376	['block5
block5p_activation (Activa	(None, 14, 14, 1344)	0	['block5



p_bn[0][0]'] tion)			
block5p_se_squeeze (Global p_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block5
block5p_se_reshape (Reshap p_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block5
block5p_se_reduce (Conv2D) p_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block5
block5p_se_expand (Conv2D) p_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block5
block5p_se_excite (Multipl p_activation[0][0]'], y) p_se_expand[0][0]']	(None, 14, 14, 1344)	0	['block5 'block5
block5p_project_conv (Conv p_se_excite[0][0]'] 2D)	(None, 14, 14, 224)	301056	['block5
block5p_project_bn (BatchN p_project_conv[0][0]'] ormalization)	(None, 14, 14, 224)	896	['block5
block5p_drop (Dropout) p_project_bn[0][0]']	(None, 14, 14, 224)	0	['block5
block5p_add (Add) p_drop[0][0]'], o_add[0][0]']	(None, 14, 14, 224)	0	['block5 'block5
block5q_expand_conv (Conv2 p_add[0][0]'] D)	(None, 14, 14, 1344)	301056	['block5
block5q_expand_bn (BatchNo q_expand_conv[0][0]'] rmalization)	(None, 14, 14, 1344)	5376	['block5
block5q_expand_activation q_expand_bn[0][0]'] (Activation)	(None, 14, 14, 1344)	0	['block5
block5q_dwconv2 (Depthwise q_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5q_bn (BatchNormaliza q_dwconv2[0][0]']	(None, 14, 14, 1344)	5376	['block5

tion)			
block5q_activation (Activation_bn[0][0]')	(None, 14, 14, 1344)	0	['block5
tion)			
block5q_se_squeeze (Global_q_activation[0][0]')	(None, 1344)	0	['block5
AveragePooling2D)			
block5q_se_reshape (Reshape_q_se_squeeze[0][0]')	(None, 1, 1, 1344)	0	['block5
e)			
block5q_se_reduce (Conv2D_q_se_reshape[0][0]')	(None, 1, 1, 56)	75320	['block5
block5q_se_expand (Conv2D_q_se_reduce[0][0]')	(None, 1, 1, 1344)	76608	['block5
block5q_se_excite (Multiply_q_activation[0][0]')	(None, 14, 14, 1344)	0	['block5
y)			'block5
q_se_expand[0][0]')			
block5q_project_conv (Conv2D_q_se_excite[0][0]')	(None, 14, 14, 224)	301056	['block5
2D)			
block5q_project_bn (BatchNormalization_q_project_conv[0][0]')	(None, 14, 14, 224)	896	['block5
block5q_drop (Dropout_q_project_bn[0][0]')	(None, 14, 14, 224)	0	['block5
block5q_add (Add_q_drop[0][0]')	(None, 14, 14, 224)	0	['block5
			'block5
p_add[0][0]')			
block5r_expand_conv (Conv2D_q_add[0][0]')	(None, 14, 14, 1344)	301056	['block5
D)			
block5r_expand_bn (BatchNormalization_r_expand_conv[0][0]')	(None, 14, 14, 1344)	5376	['block5
rmalization)			
block5r_expand_activation (Activation_r_expand_bn[0][0]')	(None, 14, 14, 1344)	0	['block5
block5r_dwconv2 (Depthwise_r_expand_activation[0]	(None, 14, 14, 1344)	12096	['block5
Conv2D)			[0]']

block5r_bn (BatchNormalization)	(None, 14, 14, 1344)	5376	['block5r_dwconv2[0][0]']
block5r_activation (Activation)	(None, 14, 14, 1344)	0	['block5r_bn[0][0]']
block5r_se_squeeze (GlobalAveragePooling2D)	(None, 1344)	0	['block5r_activation[0][0]']
block5r_se_reshape (Reshape)	(None, 1, 1, 1344)	0	['block5r_se_squeeze[0][0]']
block5r_se_reduce (Conv2D)	(None, 1, 1, 56)	75320	['block5r_se_reshape[0][0]']
block5r_se_expand (Conv2D)	(None, 1, 1, 1344)	76608	['block5r_se_reduce[0][0]']
block5r_se_excite (Multiply)	(None, 14, 14, 1344)	0	['block5r_activation[0][0]', 'block5r_se_expand[0][0]']
block5r_project_conv (Conv2D)	(None, 14, 14, 224)	301056	['block5r_se_excite[0][0]']
block5r_project_bn (BatchNormalization)	(None, 14, 14, 224)	896	['block5r_project_conv[0][0]']
block5r_drop (Dropout)	(None, 14, 14, 224)	0	['block5r_project_bn[0][0]']
block5r_add (Add)	(None, 14, 14, 224)	0	['block5r_drop[0][0]', 'block5q_add[0][0]']
block5s_expand_conv (Conv2D)	(None, 14, 14, 1344)	301056	['block5r_add[0][0]']
block5s_expand_bn (BatchNormalization)	(None, 14, 14, 1344)	5376	['block5s_expand_conv[0][0]']
block5s_expand_activation (Activation)	(None, 14, 14, 1344)	0	['block5s_expand_bn[0][0]']

block5s_dwconv2 (Depthwise s_expand_activation[0] Conv2D)	(None, 14, 14, 1344)	12096	['block5 [0]']
block5s_bn (BatchNormaliza s_dwconv2[0][0]') tion)	(None, 14, 14, 1344)	5376	['block5
block5s_activation (Activa s_bn[0][0]') tion)	(None, 14, 14, 1344)	0	['block5
block5s_se_squeeze (Global s_activation[0][0]') AveragePooling2D)	(None, 1344)	0	['block5
block5s_se_reshape (Reshap s_se_squeeze[0][0]') e)	(None, 1, 1, 1344)	0	['block5
block5s_se_reduce (Conv2D) s_se_reshape[0][0]')	(None, 1, 1, 56)	75320	['block5
block5s_se_expand (Conv2D) s_se_reduce[0][0]')	(None, 1, 1, 1344)	76608	['block5
block5s_se_excite (Multipl s_activation[0][0]', y) s_se_expand[0][0]')	(None, 14, 14, 1344)	0	['block5 'block5
block5s_project_conv (Conv s_se_excite[0][0]') 2D)	(None, 14, 14, 224)	301056	['block5
block5s_project_bn (BatchN s_project_conv[0][0]') ormalization)	(None, 14, 14, 224)	896	['block5
block5s_drop (Dropout) s_project_bn[0][0]')	(None, 14, 14, 224)	0	['block5
block5s_add (Add) s_drop[0][0]', r_add[0][0]')	(None, 14, 14, 224)	0	['block5 'block5
block6a_expand_conv (Conv2 s_add[0][0]') D)	(None, 14, 14, 1344)	301056	['block5
block6a_expand_bn (BatchNo a_expand_conv[0][0]') rmalization)	(None, 14, 14, 1344)	5376	['block6
block6a_expand_activation	(None, 14, 14, 1344)	0	['block6

a_expand_bn[0][0]'] (Activation)				
block6a_dwconv2 (Depthwise a_expand_activation[0] Conv2D)	(None, 7, 7, 1344)	12096	['block6 [0]']	
block6a_bn (BatchNormaliza a_dwconv2[0][0]'] tion)	(None, 7, 7, 1344)	5376	['block6	
block6a_activation (Activa a_bn[0][0]'] tion)	(None, 7, 7, 1344)	0	['block6	
block6a_se_squeeze (Global a_activation[0][0]'] AveragePooling2D)	(None, 1344)	0	['block6	
block6a_se_reshape (Reshap a_se_squeeze[0][0]'] e)	(None, 1, 1, 1344)	0	['block6	
block6a_se_reduce (Conv2D) a_se_reshape[0][0]']	(None, 1, 1, 56)	75320	['block6	
block6a_se_expand (Conv2D) a_se_reduce[0][0]']	(None, 1, 1, 1344)	76608	['block6	
block6a_se_excite (Multipl a_activation[0][0]'], y) a_se_expand[0][0]']	(None, 7, 7, 1344)	0	['block6  'block6	
block6a_project_conv (Conv a_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	516096	['block6	
block6a_project_bn (BatchN a_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6	
block6b_expand_conv (Conv2 a_project_bn[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6	
block6b_expand_bn (BatchNo b_expand_conv[0][0]'] rmalization)	(None, 7, 7, 2304)	9216	['block6	
block6b_expand_activation b_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6	
block6b_dwconv2 (Depthwise b_expand_activation[0]	(None, 7, 7, 2304)	20736	['block6	

Conv2D)			[0]']
block6b_bn (BatchNormaliza b_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6b_activation (Activa b_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6b_se_squeeze (Global b_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6
block6b_se_reshape (Reshap b_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6b_se_reduce (Conv2D) b_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6b_se_expand (Conv2D) b_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6b_se_excite (Multipl b_activation[0][0]'], y) b_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6b_project_conv (Conv b_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6b_project_bn (BatchN b_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6b_drop (Dropout) b_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6b_add (Add) b_drop[0][0]'], a_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6c_expand_conv (Conv2 b_add[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6
block6c_expand_bn (BatchNo c_expand_conv[0][0]'] rmalization)	(None, 7, 7, 2304)	9216	['block6
block6c_expand_activation c_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6

block6c_dwconv2 (Depthwise c_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6c_bn (BatchNormaliza c_dwconv2[0][0]') tion)	(None, 7, 7, 2304)	9216	['block6
block6c_activation (Activa c_bn[0][0]') tion)	(None, 7, 7, 2304)	0	['block6
block6c_se_squeeze (Global c_activation[0][0]') AveragePooling2D)	(None, 2304)	0	['block6
block6c_se_reshape (Reshap c_se_squeeze[0][0]') e)	(None, 1, 1, 2304)	0	['block6
block6c_se_reduce (Conv2D) c_se_reshape[0][0]')	(None, 1, 1, 96)	221280	['block6
block6c_se_expand (Conv2D) c_se_reduce[0][0]')	(None, 1, 1, 2304)	223488	['block6
block6c_se_excite (Multipl c_activation[0][0]', y) c_se_expand[0][0]')	(None, 7, 7, 2304)	0	['block6  'block6
block6c_project_conv (Conv c_se_excite[0][0]') 2D)	(None, 7, 7, 384)	884736	['block6
block6c_project_bn (BatchN c_project_conv[0][0]') ormalization)	(None, 7, 7, 384)	1536	['block6
block6c_drop (Dropout) c_project_bn[0][0]')	(None, 7, 7, 384)	0	['block6
block6c_add (Add) c_drop[0][0]', b_add[0][0]')	(None, 7, 7, 384)	0	['block6  'block6
block6d_expand_conv (Conv2 c_add[0][0]') D)	(None, 7, 7, 2304)	884736	['block6
block6d_expand_bn (BatchNo d_expand_conv[0][0]') rmalization)	(None, 7, 7, 2304)	9216	['block6

block6d_expand_activation d_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6d_dwconv2 (Depthwise d_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6d_bn (BatchNormaliza d_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6d_activation (Activa d_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6d_se_squeeze (Global d_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6
block6d_se_reshape (Reshap d_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6d_se_reduce (Conv2D) d_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6d_se_expand (Conv2D) d_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6d_se_excite (Multipl d_activation[0][0]'], y) d_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6d_project_conv (Conv d_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6d_project_bn (BatchN d_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6d_drop (Dropout) d_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6d_add (Add) d_drop[0][0]'], c_add[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6e_expand_conv (Conv2 d_add[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6
block6e_expand_bn (BatchNo d_expand_conv[0][0]']	(None, 7, 7, 2304)	9216	['block6



e_expand_conv[0][0]'] rmalization)			
block6e_expand_activation e_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6e_dwconv2 (Depthwise e_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6e_bn (BatchNormaliza e_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6e_activation (Activa e_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6e_se_squeeze (Global e_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6
block6e_se_reshape (Reshap e_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6e_se_reduce (Conv2D) e_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6e_se_expand (Conv2D) e_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6e_se_excite (Multipl e_activation[0][0]'], y) e_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6e_project_conv (Conv e_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6e_project_bn (BatchN e_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6e_drop (Dropout) e_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6e_add (Add) e_drop[0][0]'], d_add[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6f_expand_conv (Conv2 e_add[0][0]']	(None, 7, 7, 2304)	884736	['block6

D)			
block6f_expand_bn (BatchNormalization)	(None, 7, 7, 2304)	9216	['block6f_expand_conv[0][0]']
block6f_expand_activation (Activation)	(None, 7, 7, 2304)	0	['block6f_expand_bn[0][0]']
block6f_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 2304)	20736	['block6f_expand_activation[0][0]']
block6f_bn (BatchNormalization)	(None, 7, 7, 2304)	9216	['block6f_dwconv2[0][0]']
block6f_activation (Activation)	(None, 7, 7, 2304)	0	['block6f_bn[0][0]']
block6f_se_squeeze (Global AveragePooling2D)	(None, 2304)	0	['block6f_activation[0][0]']
block6f_se_reshape (Reshape)	(None, 1, 1, 2304)	0	['block6f_se_squeeze[0][0]']
block6f_se_reduce (Conv2D)	(None, 1, 1, 96)	221280	['block6f_se_reshape[0][0]']
block6f_se_expand (Conv2D)	(None, 1, 1, 2304)	223488	['block6f_se_reduce[0][0]']
block6f_se_excite (Multiply)	(None, 7, 7, 2304)	0	['block6f_se_expand[0][0]']
block6f_project_conv (Conv2D)	(None, 7, 7, 384)	884736	['block6f_se_excite[0][0]']
block6f_project_bn (BatchNormalization)	(None, 7, 7, 384)	1536	['block6f_project_conv[0][0]']
block6f_drop (Dropout)	(None, 7, 7, 384)	0	['block6f_project_bn[0][0]']
block6f_add (Add)	(None, 7, 7, 384)	0	['block6f_drop[0][0]']
e_add[0][0]']			['block6f_add[0][0]']

block6g_expand_conv (Conv2 f_add[0][0]') D)	(None, 7, 7, 2304)	884736	['block6
block6g_expand_bn (BatchNo g_expand_conv[0][0]') rmalization)	(None, 7, 7, 2304)	9216	['block6
block6g_expand_activation g_expand_bn[0][0]') (Activation)	(None, 7, 7, 2304)	0	['block6
block6g_dwconv2 (Depthwise g_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6g_bn (BatchNormaliza g_dwconv2[0][0]') tion)	(None, 7, 7, 2304)	9216	['block6
block6g_activation (Activa g_bn[0][0]') tion)	(None, 7, 7, 2304)	0	['block6
block6g_se_squeeze (Global g_activation[0][0]') AveragePooling2D)	(None, 2304)	0	['block6
block6g_se_reshape (Reshap g_se_squeeze[0][0]') e)	(None, 1, 1, 2304)	0	['block6
block6g_se_reduce (Conv2D) g_se_reshape[0][0]')	(None, 1, 1, 96)	221280	['block6
block6g_se_expand (Conv2D) g_se_reduce[0][0]')	(None, 1, 1, 2304)	223488	['block6
block6g_se_excite (Multipl g_activation[0][0]'), y) g_se_expand[0][0]')	(None, 7, 7, 2304)	0	['block6 'block6
block6g_project_conv (Conv g_se_excite[0][0]') 2D)	(None, 7, 7, 384)	884736	['block6
block6g_project_bn (BatchN g_project_conv[0][0]') ormalization)	(None, 7, 7, 384)	1536	['block6
block6g_drop (Dropout) g_project_bn[0][0]')	(None, 7, 7, 384)	0	['block6
block6g_add (Add)	(None, 7, 7, 384)	0	['block6

g_drop[0][0]',			'block6
f_add[0][0]']			
block6h_expand_conv (Conv2D)	(None, 7, 7, 2304)	884736	['block6
g_add[0][0]']			
block6h_expand_bn (BatchNormalization)	(None, 7, 7, 2304)	9216	['block6
h_expand_conv[0][0]']			
block6h_expand_activation (Activation)	(None, 7, 7, 2304)	0	['block6
h_expand_bn[0][0]']			
block6h_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 2304)	20736	['block6
h_expand_activation[0]			[0]']
block6h_bn (BatchNormalization)	(None, 7, 7, 2304)	9216	['block6
h_dwconv2[0][0]']			
block6h_activation (Activation)	(None, 7, 7, 2304)	0	['block6
h_bn[0][0]']			
block6h_se_squeeze (Global AveragePooling2D)	(None, 2304)	0	['block6
h_activation[0][0]']			
block6h_se_reshape (Reshape)	(None, 1, 1, 2304)	0	['block6
h_se_squeeze[0][0]']			
block6h_se_reduce (Conv2D)	(None, 1, 1, 96)	221280	['block6
h_se_reshape[0][0]']			
block6h_se_expand (Conv2D)	(None, 1, 1, 2304)	223488	['block6
h_se_reduce[0][0]']			
block6h_se_excite (Multiply)	(None, 7, 7, 2304)	0	['block6
h_activation[0][0]',			'block6
h_se_expand[0][0]']			
block6h_project_conv (Conv2D)	(None, 7, 7, 384)	884736	['block6
h_se_excite[0][0]']			
block6h_project_bn (BatchNormalization)	(None, 7, 7, 384)	1536	['block6
h_project_conv[0][0]']			
block6h_drop (Dropout)	(None, 7, 7, 384)	0	['block6

h_project_bn[0][0]']			
block6h_add (Add)	(None, 7, 7, 384)	0	['block6
h_drop[0][0]',			'block6
g_add[0][0]']			
block6i_expand_conv (Conv2	(None, 7, 7, 2304)	884736	['block6
h_add[0][0]']			
D)			
block6i_expand_bn (BatchNo	(None, 7, 7, 2304)	9216	['block6
i_expand_conv[0][0]']			
rmalization)			
block6i_expand_activation	(None, 7, 7, 2304)	0	['block6
i_expand_bn[0][0]']			
(Activation)			
block6i_dwconv2 (Depthwise	(None, 7, 7, 2304)	20736	['block6
i_expand_activation[0]			
Conv2D)			[0]']
block6i_bn (BatchNormaliza	(None, 7, 7, 2304)	9216	['block6
i_dwconv2[0][0]']			
tion)			
block6i_activation (Activa	(None, 7, 7, 2304)	0	['block6
i_bn[0][0]']			
tion)			
block6i_se_squeeze (Global	(None, 2304)	0	['block6
i_activation[0][0]']			
AveragePooling2D)			
block6i_se_reshape (Reshap	(None, 1, 1, 2304)	0	['block6
i_se_squeeze[0][0]']			
e)			
block6i_se_reduce (Conv2D)	(None, 1, 1, 96)	221280	['block6
i_se_reshape[0][0]']			
block6i_se_expand (Conv2D)	(None, 1, 1, 2304)	223488	['block6
i_se_reduce[0][0]']			
block6i_se_excite (Multipl	(None, 7, 7, 2304)	0	['block6
i_activation[0][0]',			
y)			'block6
i_se_expand[0][0]']			
block6i_project_conv (Conv	(None, 7, 7, 384)	884736	['block6
i_se_excite[0][0]']			
2D)			
block6i_project_bn (BatchN	(None, 7, 7, 384)	1536	['block6
i_project_conv[0][0]']			

ormalization)			
block6i_drop (Dropout) i_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6i_add (Add) i_drop[0][0]', h_add[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6j_expand_conv (Conv2 i_add[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6
block6j_expand_bn (BatchNo j_expand_conv[0][0]'] rmalization)	(None, 7, 7, 2304)	9216	['block6
block6j_expand_activation j_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6j_dwconv2 (Depthwise j_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6j_bn (BatchNormaliza j_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6j_activation (Activa j_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6j_se_squeeze (Global j_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6
block6j_se_reshape (Reshap j_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6j_se_reduce (Conv2D) j_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6j_se_expand (Conv2D) j_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6j_se_excite (Multipl j_activation[0][0]'], y) j_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6j_project_conv (Conv j_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6

block6j_project_bn (Batch Normalization)	(None, 7, 7, 384)	1536	['block6j_project_conv[0][0]']
block6j_drop (Dropout)	(None, 7, 7, 384)	0	['block6j_project_bn[0][0]']
block6j_add (Add)	(None, 7, 7, 384)	0	['block6j_drop[0][0]', 'block6i_add[0][0]']
block6k_expand_conv (Conv2D)	(None, 7, 7, 2304)	884736	['block6j_add[0][0]']
block6k_expand_bn (Batch Normalization)	(None, 7, 7, 2304)	9216	['block6k_expand_conv[0][0]']
block6k_expand_activation (Activation)	(None, 7, 7, 2304)	0	['block6k_expand_bn[0][0]']
block6k_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 2304)	20736	['block6k_expand_activation[0][0]']
block6k_bn (Batch Normalization)	(None, 7, 7, 2304)	9216	['block6k_dwconv2[0][0]']
block6k_activation (Activation)	(None, 7, 7, 2304)	0	['block6k_bn[0][0]']
block6k_se_squeeze (Global AveragePooling2D)	(None, 2304)	0	['block6k_activation[0][0]']
block6k_se_reshape (Reshape)	(None, 1, 1, 2304)	0	['block6k_se_squeeze[0][0]']
block6k_se_reduce (Conv2D)	(None, 1, 1, 96)	221280	['block6k_se_reshape[0][0]']
block6k_se_expand (Conv2D)	(None, 1, 1, 2304)	223488	['block6k_se_reduce[0][0]']
block6k_se_excite (Multiply)	(None, 7, 7, 2304)	0	['block6k_se_expand[0][0]', 'block6k_activation[0][0]']

block6k_project_conv (Conv2D) k_se_excite[0][0]']	(None, 7, 7, 384)	884736	['block6
block6k_project_bn (BatchNormalization) k_project_conv[0][0]']	(None, 7, 7, 384)	1536	['block6
block6k_drop (Dropout) k_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6k_add (Add) k_drop[0][0]', j_add[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6l_expand_conv (Conv2D) k_add[0][0]']	(None, 7, 7, 2304)	884736	['block6
block6l_expand_bn (BatchNormalization) l_expand_conv[0][0]']	(None, 7, 7, 2304)	9216	['block6
block6l_expand_activation (Activation) l_expand_bn[0][0]']	(None, 7, 7, 2304)	0	['block6
block6l_dwconv2 (Depthwise Conv2D) l_expand_activation[0][0]']	(None, 7, 7, 2304)	20736	['block6 [0]']
block6l_bn (BatchNormalization) l_dwconv2[0][0]']	(None, 7, 7, 2304)	9216	['block6
block6l_activation (Activation) l_bn[0][0]']	(None, 7, 7, 2304)	0	['block6
block6l_se_squeeze (Global AveragePooling2D) l_activation[0][0]']	(None, 2304)	0	['block6
block6l_se_reshape (Reshape) l_se_squeeze[0][0]']	(None, 1, 1, 2304)	0	['block6
block6l_se_reduce (Conv2D) l_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6l_se_expand (Conv2D) l_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6l_se_excite (Multiply) l_activation[0][0]',	(None, 7, 7, 2304)	0	['block6



y)			'block6
l_se_expand[0][0]']			
block6l_project_conv (Conv	(None, 7, 7, 384)	884736	['block6
l_se_excite[0][0]']			
2D)			
block6l_project_bn (BatchN	(None, 7, 7, 384)	1536	['block6
l_project_conv[0][0]']			
ormalization)			
block6l_drop (Dropout)	(None, 7, 7, 384)	0	['block6
l_project_bn[0][0]']			
block6l_add (Add)	(None, 7, 7, 384)	0	['block6
l_drop[0][0]',			
			'block6
k_add[0][0]']			
block6m_expand_conv (Conv2	(None, 7, 7, 2304)	884736	['block6
l_add[0][0]']			
D)			
block6m_expand_bn (BatchNo	(None, 7, 7, 2304)	9216	['block6
m_expand_conv[0][0]']			
rmalization)			
block6m_expand_activation	(None, 7, 7, 2304)	0	['block6
m_expand_bn[0][0]']			
(Activation)			
block6m_dwconv2 (Depthwise	(None, 7, 7, 2304)	20736	['block6
m_expand_activation[0]			
Conv2D)			[0]']
block6m_bn (BatchNormaliza	(None, 7, 7, 2304)	9216	['block6
m_dwconv2[0][0]']			
tion)			
block6m_activation (Activa	(None, 7, 7, 2304)	0	['block6
m_bn[0][0]']			
tion)			
block6m_se_squeeze (Global	(None, 2304)	0	['block6
m_activation[0][0]']			
AveragePooling2D)			
block6m_se_reshape (Reshap	(None, 1, 1, 2304)	0	['block6
m_se_squeeze[0][0]']			
e)			
block6m_se_reduce (Conv2D)	(None, 1, 1, 96)	221280	['block6
m_se_reshape[0][0]']			
block6m_se_expand (Conv2D)	(None, 1, 1, 2304)	223488	['block6
m_se_reduce[0][0]']			

block6m_se_excite (Multipl m_activation[0][0]',' y) m_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6m_project_conv (Conv m_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6m_project_bn (BatchN m_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6m_drop (Dropout) m_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6m_add (Add) m_drop[0][0]',' l_add[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6n_expand_conv (Conv2 m_add[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6
block6n_expand_bn (BatchNo n_expand_conv[0][0]'] rmalization)	(None, 7, 7, 2304)	9216	['block6
block6n_expand_activation n_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6n_dwconv2 (Depthwise n_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6n_bn (BatchNormaliza n_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6n_activation (Activa n_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6n_se_squeeze (Global n_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6
block6n_se_reshape (Reshap n_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6n_se_reduce (Conv2D) n_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6

block6n_se_expand (Conv2D)	(None, 1, 1, 2304)	223488	['block6
n_se_reduce[0][0]']			
block6n_se_excite (Multipl	(None, 7, 7, 2304)	0	['block6
n_activation[0][0]',			
y)			'block6
n_se_expand[0][0]']			
block6n_project_conv (Conv	(None, 7, 7, 384)	884736	['block6
n_se_excite[0][0]']			
2D)			
block6n_project_bn (BatchN	(None, 7, 7, 384)	1536	['block6
n_project_conv[0][0]']			
ormalization)			
block6n_drop (Dropout)	(None, 7, 7, 384)	0	['block6
n_project_bn[0][0]']			
block6n_add (Add)	(None, 7, 7, 384)	0	['block6
n_drop[0][0]',			
			'block6
m_add[0][0]']			
block6o_expand_conv (Conv2	(None, 7, 7, 2304)	884736	['block6
n_add[0][0]']			
D)			
block6o_expand_bn (BatchNo	(None, 7, 7, 2304)	9216	['block6
o_expand_conv[0][0]']			
rmalization)			
block6o_expand_activation	(None, 7, 7, 2304)	0	['block6
o_expand_bn[0][0]']			
(Activation)			
block6o_dwconv2 (Depthwise	(None, 7, 7, 2304)	20736	['block6
o_expand_activation[0]			
Conv2D)			[0]']
block6o_bn (BatchNormaliza	(None, 7, 7, 2304)	9216	['block6
o_dwconv2[0][0]']			
tion)			
block6o_activation (Activa	(None, 7, 7, 2304)	0	['block6
o_bn[0][0]']			
tion)			
block6o_se_squeeze (Global	(None, 2304)	0	['block6
o_activation[0][0]']			
AveragePooling2D)			
block6o_se_reshape (Reshap	(None, 1, 1, 2304)	0	['block6
o_se_squeeze[0][0]']			
e)			

block6o_se_reduce (Conv2D) o_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6o_se_expand (Conv2D) o_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6o_se_excite (Multipl o_activation[0][0]',' y) o_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6  'block6
block6o_project_conv (Conv o_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6o_project_bn (BatchN o_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6o_drop (Dropout) o_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6o_add (Add) o_drop[0][0]',' n_add[0][0]']	(None, 7, 7, 384)	0	['block6  'block6
block6p_expand_conv (Conv2 o_add[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6
block6p_expand_bn (BatchNo p_expand_conv[0][0]'] rmalization)	(None, 7, 7, 2304)	9216	['block6
block6p_expand_activation p_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6p_dwconv2 (Depthwise p_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6  [0]']
block6p_bn (BatchNormaliza p_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6p_activation (Activa p_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6p_se_squeeze (Global p_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6

block6p_se_reshape (Reshape) p_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6p_se_reduce (Conv2D) p_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6p_se_expand (Conv2D) p_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6p_se_excite (Multipl p_activation[0][0]'], y) p_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6  'block6
block6p_project_conv (Conv p_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6p_project_bn (BatchN p_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6p_drop (Dropout) p_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6p_add (Add) p_drop[0][0]'], o_add[0][0]']	(None, 7, 7, 384)	0	['block6  'block6
block6q_expand_conv (Conv2 p_add[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6
block6q_expand_bn (BatchNo q_expand_conv[0][0]'] rmalization)	(None, 7, 7, 2304)	9216	['block6
block6q_expand_activation q_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6q_dwconv2 (Depthwise q_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6  [0]']
block6q_bn (BatchNormaliza q_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6q_activation (Activa q_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6q_se_squeeze (Global	(None, 2304)	0	['block6

q_activation[0][0]'] AveragePooling2D)			
block6q_se_reshape (Reshape q_se_squeeze[0][0]') e)	(None, 1, 1, 2304)	0	['block6
block6q_se_reduce (Conv2D) q_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6q_se_expand (Conv2D) q_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6q_se_excite (Multipl q_activation[0][0]', y) q_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6  'block6
block6q_project_conv (Conv q_se_excite[0][0]') 2D)	(None, 7, 7, 384)	884736	['block6
block6q_project_bn (BatchN q_project_conv[0][0]') ormalization)	(None, 7, 7, 384)	1536	['block6
block6q_drop (Dropout) q_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6q_add (Add) q_drop[0][0]', p_add[0][0]']	(None, 7, 7, 384)	0	['block6  'block6
block6r_expand_conv (Conv2 q_add[0][0]') D)	(None, 7, 7, 2304)	884736	['block6
block6r_expand_bn (BatchNo r_expand_conv[0][0]') rmalization)	(None, 7, 7, 2304)	9216	['block6
block6r_expand_activation r_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6r_dwconv2 (Depthwise r_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6  [0]']
block6r_bn (BatchNormaliza r_dwconv2[0][0]') tion)	(None, 7, 7, 2304)	9216	['block6
block6r_activation (Activa r_bn[0][0]']	(None, 7, 7, 2304)	0	['block6

tion)			
block6r_se_squeeze (Global r_activation[0][0]') AveragePooling2D)	(None, 2304)	0	['block6
block6r_se_reshape (Reshap r_se_squeeze[0][0]') e)	(None, 1, 1, 2304)	0	['block6
block6r_se_reduce (Conv2D) r_se_reshape[0][0]')	(None, 1, 1, 96)	221280	['block6
block6r_se_expand (Conv2D) r_se_reduce[0][0]')	(None, 1, 1, 2304)	223488	['block6
block6r_se_excite (Multipl r_activation[0][0]', y) r_se_expand[0][0]')	(None, 7, 7, 2304)	0	['block6 'block6
block6r_project_conv (Conv r_se_excite[0][0]') 2D)	(None, 7, 7, 384)	884736	['block6
block6r_project_bn (BatchN r_project_conv[0][0]') ormalization)	(None, 7, 7, 384)	1536	['block6
block6r_drop (Dropout) r_project_bn[0][0]')	(None, 7, 7, 384)	0	['block6
block6r_add (Add) r_drop[0][0]', q_add[0][0]')	(None, 7, 7, 384)	0	['block6 'block6
block6s_expand_conv (Conv2 r_add[0][0]') D)	(None, 7, 7, 2304)	884736	['block6
block6s_expand_bn (BatchNo s_expand_conv[0][0]') rmalization)	(None, 7, 7, 2304)	9216	['block6
block6s_expand_activation s_expand_bn[0][0]') (Activation)	(None, 7, 7, 2304)	0	['block6
block6s_dwconv2 (Depthwise s_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]')
block6s_bn (BatchNormaliza s_dwconv2[0][0]') tion)	(None, 7, 7, 2304)	9216	['block6

block6s_activation (Activation) s_bn[0][0]')	(None, 7, 7, 2304)	0	['block6
block6s_se_squeeze (Global AveragePooling2D) s_activation[0][0]')	(None, 2304)	0	['block6
block6s_se_reshape (Reshape) s_se_squeeze[0][0]')	(None, 1, 1, 2304)	0	['block6
block6s_se_reduce (Conv2D) s_se_reshape[0][0]')	(None, 1, 1, 96)	221280	['block6
block6s_se_expand (Conv2D) s_se_reduce[0][0]')	(None, 1, 1, 2304)	223488	['block6
block6s_se_excite (Multiply) s_activation[0][0]', s_se_expand[0][0]')	(None, 7, 7, 2304)	0	['block6 'block6
block6s_project_conv (Conv2D) s_se_excite[0][0]')	(None, 7, 7, 384)	884736	['block6
block6s_project_bn (Batch Normalization) s_project_conv[0][0]')	(None, 7, 7, 384)	1536	['block6
block6s_drop (Dropout) s_project_bn[0][0]')	(None, 7, 7, 384)	0	['block6
block6s_add (Add) s_drop[0][0]', r_add[0][0]')	(None, 7, 7, 384)	0	['block6 'block6
block6t_expand_conv (Conv2D) s_add[0][0]')	(None, 7, 7, 2304)	884736	['block6
block6t_expand_bn (Batch Normalization) t_expand_conv[0][0]')	(None, 7, 7, 2304)	9216	['block6
block6t_expand_activation (Activation) t_expand_bn[0][0]')	(None, 7, 7, 2304)	0	['block6
block6t_dwconv2 (Depthwise Conv2D) t_expand_activation[0]	(None, 7, 7, 2304)	20736	['block6 [0]']



block6t_bn (BatchNormalization) t_dwconv2[0][0]']	(None, 7, 7, 2304)	9216	['block6
block6t_activation (Activation) t_bn[0][0]']	(None, 7, 7, 2304)	0	['block6
block6t_se_squeeze (GlobalAveragePooling2D) t_activation[0][0]']	(None, 2304)	0	['block6
block6t_se_reshape (Reshape) t_se_squeeze[0][0]']	(None, 1, 1, 2304)	0	['block6
block6t_se_reduce (Conv2D) t_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6t_se_expand (Conv2D) t_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6t_se_excite (Multiply) t_activation[0][0]'], t_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6t_project_conv (Conv2D) t_se_excite[0][0]']	(None, 7, 7, 384)	884736	['block6
block6t_project_bn (BatchNormalization) t_project_conv[0][0]']	(None, 7, 7, 384)	1536	['block6
block6t_drop (Dropout) t_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6t_add (Add) t_drop[0][0]'], s_add[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6u_expand_conv (Conv2D) t_add[0][0]']	(None, 7, 7, 2304)	884736	['block6
block6u_expand_bn (BatchNormalization) u_expand_conv[0][0]']	(None, 7, 7, 2304)	9216	['block6
block6u_expand_activation (Activation) u_expand_bn[0][0]']	(None, 7, 7, 2304)	0	['block6
block6u_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 2304)	20736	['block6

u_expand_activation[0] Conv2D)			[0]']
block6u_bn (BatchNormaliza u_dwconv2[0][0]') tion)	(None, 7, 7, 2304)	9216	['block6
block6u_activation (Activa u_bn[0][0]') tion)	(None, 7, 7, 2304)	0	['block6
block6u_se_squeeze (Global u_activation[0][0]') AveragePooling2D)	(None, 2304)	0	['block6
block6u_se_reshape (Reshap u_se_squeeze[0][0]') e)	(None, 1, 1, 2304)	0	['block6
block6u_se_reduce (Conv2D) u_se_reshape[0][0]')	(None, 1, 1, 96)	221280	['block6
block6u_se_expand (Conv2D) u_se_reduce[0][0]')	(None, 1, 1, 2304)	223488	['block6
block6u_se_excite (Multipl u_activation[0][0]'), y) u_se_expand[0][0]')	(None, 7, 7, 2304)	0	['block6 'block6
block6u_project_conv (Conv u_se_excite[0][0]') 2D)	(None, 7, 7, 384)	884736	['block6
block6u_project_bn (BatchN u_project_conv[0][0]') ormalization)	(None, 7, 7, 384)	1536	['block6
block6u_drop (Dropout) u_project_bn[0][0]')	(None, 7, 7, 384)	0	['block6
block6u_add (Add) u_drop[0][0]'), t_add[0][0]')	(None, 7, 7, 384)	0	['block6 'block6
block6v_expand_conv (Conv2 u_add[0][0]') D)	(None, 7, 7, 2304)	884736	['block6
block6v_expand_bn (BatchNo v_expand_conv[0][0]') rmalization)	(None, 7, 7, 2304)	9216	['block6
block6v_expand_activation v_expand_bn[0][0]')	(None, 7, 7, 2304)	0	['block6

(Activation)			
block6v_dwconv2 (Depthwise v_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6v_bn (BatchNormaliza v_dwconv2[0][0]') tion)	(None, 7, 7, 2304)	9216	['block6
block6v_activation (Activa v_bn[0][0]') tion)	(None, 7, 7, 2304)	0	['block6
block6v_se_squeeze (Global v_activation[0][0]') AveragePooling2D)	(None, 2304)	0	['block6
block6v_se_reshape (Reshap v_se_squeeze[0][0]') e)	(None, 1, 1, 2304)	0	['block6
block6v_se_reduce (Conv2D) v_se_reshape[0][0]')	(None, 1, 1, 96)	221280	['block6
block6v_se_expand (Conv2D) v_se_reduce[0][0]')	(None, 1, 1, 2304)	223488	['block6
block6v_se_excite (Multipl v_activation[0][0]', y) v_se_expand[0][0]')	(None, 7, 7, 2304)	0	['block6 'block6
block6v_project_conv (Conv v_se_excite[0][0]') 2D)	(None, 7, 7, 384)	884736	['block6
block6v_project_bn (BatchN v_project_conv[0][0]') ormalization)	(None, 7, 7, 384)	1536	['block6
block6v_drop (Dropout) v_project_bn[0][0]')	(None, 7, 7, 384)	0	['block6
block6v_add (Add) v_drop[0][0]', u_add[0][0]')	(None, 7, 7, 384)	0	['block6 'block6
block6w_expand_conv (Conv2 v_add[0][0]') D)	(None, 7, 7, 2304)	884736	['block6
block6w_expand_bn (BatchNo w_expand_conv[0][0]') rmalization)	(None, 7, 7, 2304)	9216	['block6

block6w_expand_activation w_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6w_dwconv2 (Depthwise w_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6w_bn (BatchNormaliza w_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6w_activation (Activa w_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6w_se_squeeze (Global w_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6
block6w_se_reshape (Reshap w_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6w_se_reduce (Conv2D) w_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6w_se_expand (Conv2D) w_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6w_se_excite (Multipl w_activation[0][0]'], y) w_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6w_project_conv (Conv w_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6w_project_bn (BatchN w_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6w_drop (Dropout) w_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6w_add (Add) w_drop[0][0]'], v_add[0][0]']	(None, 7, 7, 384)	0	['block6 'block6
block6x_expand_conv (Conv2 w_add[0][0]'] D)	(None, 7, 7, 2304)	884736	['block6

block6x_expand_bn (BatchNormalization)	(None, 7, 7, 2304)	9216	['block6x_expand_conv[0][0]']
block6x_expand_activation (Activation)	(None, 7, 7, 2304)	0	['block6x_expand_bn[0][0]']
block6x_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 2304)	20736	['block6x_expand_activation[0][0]']
block6x_bn (BatchNormalization)	(None, 7, 7, 2304)	9216	['block6x_dwconv2[0][0]']
block6x_activation (Activation)	(None, 7, 7, 2304)	0	['block6x_bn[0][0]']
block6x_se_squeeze (Global AveragePooling2D)	(None, 2304)	0	['block6x_activation[0][0]']
block6x_se_reshape (Reshape)	(None, 1, 1, 2304)	0	['block6x_se_squeeze[0][0]']
block6x_se_reduce (Conv2D)	(None, 1, 1, 96)	221280	['block6x_se_reshape[0][0]']
block6x_se_expand (Conv2D)	(None, 1, 1, 2304)	223488	['block6x_se_reduce[0][0]']
block6x_se_excite (Multiply)	(None, 7, 7, 2304)	0	['block6x_se_expand[0][0]']
block6x_project_conv (Conv2D)	(None, 7, 7, 384)	884736	['block6x_se_excite[0][0]']
block6x_project_bn (BatchNormalization)	(None, 7, 7, 384)	1536	['block6x_project_conv[0][0]']
block6x_drop (Dropout)	(None, 7, 7, 384)	0	['block6x_project_bn[0][0]']
block6x_add (Add)	(None, 7, 7, 384)	0	['block6x_drop[0][0]']
block6y_expand_conv (Conv2D)	(None, 7, 7, 2304)	884736	['block6x_add[0][0]']

x_add[0][0]'] D)			
block6y_expand_bn (BatchNo y_expand_conv[0][0]'] rmalization)	(None, 7, 7, 2304)	9216	['block6
block6y_expand_activation y_expand_bn[0][0]'] (Activation)	(None, 7, 7, 2304)	0	['block6
block6y_dwconv2 (Depthwise y_expand_activation[0] Conv2D)	(None, 7, 7, 2304)	20736	['block6 [0]']
block6y_bn (BatchNormaliza y_dwconv2[0][0]'] tion)	(None, 7, 7, 2304)	9216	['block6
block6y_activation (Activa y_bn[0][0]'] tion)	(None, 7, 7, 2304)	0	['block6
block6y_se_squeeze (Global y_activation[0][0]'] AveragePooling2D)	(None, 2304)	0	['block6
block6y_se_reshape (Reshap y_se_squeeze[0][0]'] e)	(None, 1, 1, 2304)	0	['block6
block6y_se_reduce (Conv2D) y_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block6
block6y_se_expand (Conv2D) y_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block6
block6y_se_excite (Multipl y_activation[0][0]'], y) y_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block6 'block6
block6y_project_conv (Conv y_se_excite[0][0]'] 2D)	(None, 7, 7, 384)	884736	['block6
block6y_project_bn (BatchN y_project_conv[0][0]'] ormalization)	(None, 7, 7, 384)	1536	['block6
block6y_drop (Dropout) y_project_bn[0][0]']	(None, 7, 7, 384)	0	['block6
block6y_add (Add) y_drop[0][0]'],	(None, 7, 7, 384)	0	['block6 'block6

x_add[0][0]']			
block7a_expand_conv (Conv2D) y_add[0][0]']	(None, 7, 7, 2304)	884736	['block6
block7a_expand_bn (BatchNormalization) a_expand_conv[0][0]']	(None, 7, 7, 2304)	9216	['block7
block7a_expand_activation a_expand_bn[0][0]']	(None, 7, 7, 2304)	0	['block7
block7a_dwconv2 (Depthwise Conv2D) a_expand_activation[0]	(None, 7, 7, 2304)	20736	['block7
block7a_bn (BatchNormalization) a_dwconv2[0][0]']	(None, 7, 7, 2304)	9216	['block7
block7a_activation (Activation) a_bn[0][0]']	(None, 7, 7, 2304)	0	['block7
block7a_se_squeeze (Global AveragePooling2D) a_activation[0][0]']	(None, 2304)	0	['block7
block7a_se_reshape (Reshape) a_se_squeeze[0][0]']	(None, 1, 1, 2304)	0	['block7
block7a_se_reduce (Conv2D) a_se_reshape[0][0]']	(None, 1, 1, 96)	221280	['block7
block7a_se_expand (Conv2D) a_se_reduce[0][0]']	(None, 1, 1, 2304)	223488	['block7
block7a_se_excite (Multiply) a_activation[0][0]', y) a_se_expand[0][0]']	(None, 7, 7, 2304)	0	['block7 'block7
block7a_project_conv (Conv2D) a_se_excite[0][0]']	(None, 7, 7, 640)	1474560	['block7
block7a_project_bn (BatchNormalization) a_project_conv[0][0]']	(None, 7, 7, 640)	2560	['block7
block7b_expand_conv (Conv2D) a_project_bn[0][0]']	(None, 7, 7, 3840)	2457600	['block7

block7b_expand_bn (BatchNormaliza b_expand_conv[0][0]') rmalization)	(None, 7, 7, 3840)	15360	['block7
block7b_expand_activation b_expand_bn[0][0]') (Activation)	(None, 7, 7, 3840)	0	['block7
block7b_dwconv2 (Depthwise b_expand_activation[0] Conv2D)	(None, 7, 7, 3840)	34560	['block7 [0]']
block7b_bn (BatchNormaliza b_dwconv2[0][0]') tion)	(None, 7, 7, 3840)	15360	['block7
block7b_activation (Activa b_bn[0][0]') tion)	(None, 7, 7, 3840)	0	['block7
block7b_se_squeeze (Global b_activation[0][0]') AveragePooling2D)	(None, 3840)	0	['block7
block7b_se_reshape (Reshap b_se_squeeze[0][0]') e)	(None, 1, 1, 3840)	0	['block7
block7b_se_reduce (Conv2D) b_se_reshape[0][0]')	(None, 1, 1, 160)	614560	['block7
block7b_se_expand (Conv2D) b_se_reduce[0][0]')	(None, 1, 1, 3840)	618240	['block7
block7b_se_excite (Multipl b_activation[0][0]'), y) b_se_expand[0][0]')	(None, 7, 7, 3840)	0	['block7 'block7
block7b_project_conv (Conv b_se_excite[0][0]') 2D)	(None, 7, 7, 640)	2457600	['block7
block7b_project_bn (BatchN b_project_conv[0][0]') ormalization)	(None, 7, 7, 640)	2560	['block7
block7b_drop (Dropout) b_project_bn[0][0]')	(None, 7, 7, 640)	0	['block7
block7b_add (Add) b_drop[0][0]'), a_project_bn[0][0]')	(None, 7, 7, 640)	0	['block7 'block7



block7c_expand_conv (Conv2D) b_add[0][0]']	(None, 7, 7, 3840)	2457600	['block7
block7c_expand_bn (BatchNormalization) c_expand_conv[0][0]']	(None, 7, 7, 3840)	15360	['block7
block7c_expand_activation c_expand_bn[0][0]'] (Activation)	(None, 7, 7, 3840)	0	['block7
block7c_dwconv2 (Depthwise Conv2D) c_expand_activation[0]	(None, 7, 7, 3840)	34560	['block7 [0]']
block7c_bn (BatchNormalization) c_dwconv2[0][0]']	(None, 7, 7, 3840)	15360	['block7
block7c_activation (Activation) c_bn[0][0]']	(None, 7, 7, 3840)	0	['block7
block7c_se_squeeze (Global AveragePooling2D) c_activation[0][0]']	(None, 3840)	0	['block7
block7c_se_reshape (Reshape) c_se_squeeze[0][0]']	(None, 1, 1, 3840)	0	['block7
block7c_se_reduce (Conv2D) c_se_reshape[0][0]']	(None, 1, 1, 160)	614560	['block7
block7c_se_expand (Conv2D) c_se_reduce[0][0]']	(None, 1, 1, 3840)	618240	['block7
block7c_se_excite (Multiply) c_activation[0][0]', y) c_se_expand[0][0]']	(None, 7, 7, 3840)	0	['block7 'block7
block7c_project_conv (Conv2D) c_se_excite[0][0]']	(None, 7, 7, 640)	2457600	['block7
block7c_project_bn (BatchNormalization) c_project_conv[0][0]']	(None, 7, 7, 640)	2560	['block7
block7c_drop (Dropout) c_project_bn[0][0]']	(None, 7, 7, 640)	0	['block7
block7c_add (Add) c_drop[0][0]',	(None, 7, 7, 640)	0	['block7

b_add[0][0]']			'block7
block7d_expand_conv (Conv2D)	(None, 7, 7, 3840)	2457600	['block7c_add[0][0]']
block7d_expand_bn (BatchNormalization)	(None, 7, 7, 3840)	15360	['block7d_expand_conv[0][0]']
block7d_expand_activation (Activation)	(None, 7, 7, 3840)	0	['block7d_expand_bn[0][0]']
block7d_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 3840)	34560	['block7d_expand_activation[0][0]']
block7d_bn (BatchNormalization)	(None, 7, 7, 3840)	15360	['block7d_dwconv2[0][0]']
block7d_activation (Activation)	(None, 7, 7, 3840)	0	['block7d_bn[0][0]']
block7d_se_squeeze (Global AveragePooling2D)	(None, 3840)	0	['block7d_activation[0][0]']
block7d_se_reshape (Reshape)	(None, 1, 1, 3840)	0	['block7d_se_squeeze[0][0]']
block7d_se_reduce (Conv2D)	(None, 1, 1, 160)	614560	['block7d_se_reshape[0][0]']
block7d_se_expand (Conv2D)	(None, 1, 1, 3840)	618240	['block7d_se_reduce[0][0]']
block7d_se_excite (Multiply)	(None, 7, 7, 3840)	0	['block7d_activation[0][0]', 'block7d_se_expand[0][0]']
block7d_project_conv (Conv2D)	(None, 7, 7, 640)	2457600	['block7d_se_excite[0][0]']
block7d_project_bn (BatchNormalization)	(None, 7, 7, 640)	2560	['block7d_project_conv[0][0]']
block7d_drop (Dropout)	(None, 7, 7, 640)	0	['block7d_project_bn[0][0]']

block7d_add (Add)	(None, 7, 7, 640)	0	['block7
d_drop[0][0]',			'block7
c_add[0][0]']			
block7e_expand_conv (Conv2D)	(None, 7, 7, 3840)	2457600	['block7
d_add[0][0]']			
block7e_expand_bn (BatchNormalization)	(None, 7, 7, 3840)	15360	['block7
e_expand_conv[0][0]']			
block7e_expand_activation (Activation)	(None, 7, 7, 3840)	0	['block7
e_expand_bn[0][0]']			
block7e_dwconv2 (Depthwise Conv2D)	(None, 7, 7, 3840)	34560	['block7
e_expand_activation[0]			[0]']
block7e_bn (BatchNormalization)	(None, 7, 7, 3840)	15360	['block7
e_dwconv2[0][0]']			
block7e_activation (Activation)	(None, 7, 7, 3840)	0	['block7
e_bn[0][0]']			
block7e_se_squeeze (Global AveragePooling2D)	(None, 3840)	0	['block7
e_activation[0][0]']			
block7e_se_reshape (Reshape)	(None, 1, 1, 3840)	0	['block7
e_se_squeeze[0][0]']			
block7e_se_reduce (Conv2D)	(None, 1, 1, 160)	614560	['block7
e_se_reshape[0][0]']			
block7e_se_expand (Conv2D)	(None, 1, 1, 3840)	618240	['block7
e_se_reduce[0][0]']			
block7e_se_excite (Multiply)	(None, 7, 7, 3840)	0	['block7
e_activation[0][0]',			'block7
e_se_expand[0][0]']			
block7e_project_conv (Conv2D)	(None, 7, 7, 640)	2457600	['block7
e_se_excite[0][0]']			
block7e_project_bn (BatchNormalization)	(None, 7, 7, 640)	2560	['block7
e_project_conv[0][0]']			

block7e_drop (Dropout)	(None, 7, 7, 640)	0	['block7
e_project_bn[0][0]']			
block7e_add (Add)	(None, 7, 7, 640)	0	['block7
e_drop[0][0]',			
d_add[0][0]']			'block7
block7f_expand_conv (Conv2	(None, 7, 7, 3840)	2457600	['block7
e_add[0][0]']			
D)			
block7f_expand_bn (BatchNo	(None, 7, 7, 3840)	15360	['block7
f_expand_conv[0][0]']			
rmalization)			
block7f_expand_activation	(None, 7, 7, 3840)	0	['block7
f_expand_bn[0][0]']			
(Activation)			
block7f_dwconv2 (Depthwise	(None, 7, 7, 3840)	34560	['block7
f_expand_activation[0]			
Conv2D)			[0]']
block7f_bn (BatchNormaliza	(None, 7, 7, 3840)	15360	['block7
f_dwconv2[0][0]']			
tion)			
block7f_activation (Activa	(None, 7, 7, 3840)	0	['block7
f_bn[0][0]']			
tion)			
block7f_se_squeeze (Global	(None, 3840)	0	['block7
f_activation[0][0]']			
AveragePooling2D)			
block7f_se_reshape (Reshap	(None, 1, 1, 3840)	0	['block7
f_se_squeeze[0][0]']			
e)			
block7f_se_reduce (Conv2D)	(None, 1, 1, 160)	614560	['block7
f_se_reshape[0][0]']			
block7f_se_expand (Conv2D)	(None, 1, 1, 3840)	618240	['block7
f_se_reduce[0][0]']			
block7f_se_excite (Multipl	(None, 7, 7, 3840)	0	['block7
f_activation[0][0]',			
y)			'block7
f_se_expand[0][0]']			
block7f_project_conv (Conv	(None, 7, 7, 640)	2457600	['block7
f_se_excite[0][0]']			
2D)			

block7f_project_bn (BatchNormaliza f_project_conv[0][0]') ormalization)	(None, 7, 7, 640)	2560	['block7
block7f_drop (Dropout) f_project_bn[0][0]')	(None, 7, 7, 640)	0	['block7
block7f_add (Add) f_drop[0][0]', e_add[0][0]')	(None, 7, 7, 640)	0	['block7 'block7
block7g_expand_conv (Conv2D) f_add[0][0]')	(None, 7, 7, 3840)	2457600	['block7
block7g_expand_bn (BatchNormaliza g_expand_conv[0][0]') ormalization)	(None, 7, 7, 3840)	15360	['block7
block7g_expand_activation g_expand_bn[0][0]') (Activation)	(None, 7, 7, 3840)	0	['block7
block7g_dwconv2 (Depthwise g_expand_activation[0] Conv2D)	(None, 7, 7, 3840)	34560	['block7 [0]')
block7g_bn (BatchNormalization) g_dwconv2[0][0]')	(None, 7, 7, 3840)	15360	['block7
block7g_activation (Activation) g_bn[0][0]') tion)	(None, 7, 7, 3840)	0	['block7
block7g_se_squeeze (Global g_activation[0][0]') AveragePooling2D)	(None, 3840)	0	['block7
block7g_se_reshape (Reshape) g_se_squeeze[0][0]') e)	(None, 1, 1, 3840)	0	['block7
block7g_se_reduce (Conv2D) g_se_reshape[0][0]')	(None, 1, 1, 160)	614560	['block7
block7g_se_expand (Conv2D) g_se_reduce[0][0]')	(None, 1, 1, 3840)	618240	['block7
block7g_se_excite (Multiply) g_activation[0][0]', y) g_se_expand[0][0]')	(None, 7, 7, 3840)	0	['block7 'block7
block7g_project_conv (Conv2D)	(None, 7, 7, 640)	2457600	['block7

g_se_excite[0][0]'] 2D)			
block7g_project_bn (BatchN g_project_conv[0][0]'] ormalization)	(None, 7, 7, 640)	2560	['block7
block7g_drop (Dropout) g_project_bn[0][0]']	(None, 7, 7, 640)	0	['block7
block7g_add (Add) g_drop[0][0]'], f_add[0][0]']	(None, 7, 7, 640)	0	['block7 'block7
top_conv (Conv2D) g_add[0][0]']	(None, 7, 7, 1280)	819200	['block7
top_bn (BatchNormalization nv[0][0]'] )	(None, 7, 7, 1280)	5120	['top_co
top_activation (Activation [0][0]'] )	(None, 7, 7, 1280)	0	['top_bn
flatten (Flatten) tivation[0][0]']	(None, 62720)	0	['top_ac
dense (Dense) n[0][0]']	(None, 4096)	2569052	['flatte
		16	
dropout (Dropout) [0][0]']	(None, 4096)	0	['dense
batch_normalization (Batch t[0][0]'] Normalization)	(None, 4096)	16384	['dropou
dense_1 (Dense) normalization[0][0]']	(None, 4096)	1678131	['batch_
		2	
dropout_1 (Dropout) 1[0][0]']	(None, 4096)	0	['dense_
dense_2 (Dense) t_1[0][0]']	(None, 20)	81940	['dropou

=====

Total params: 391531700 (1.46 GB)  
Trainable params: 274598420 (1.02 GB)  
Non-trainable params: 116933280 (446.07 MB)

```
In [32]: print(base_model.input)
print(base_model.outputs)
```

```
KerasTensor(type_spec=TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name='input_1'), name='input_1', description="created by layer 'input_1'")
[<KerasTensor: shape=(None, 7, 7, 1280) dtype=float32 (created by layer 'top_activation')>]
```

```
In [28]: # import pickle
# with open('EfficientNetV2L_Transferlearning_20_cusine.pkl', 'wb') as f:
#     pickle.dump(model, f)
```

```
In [30]: # from IPython.display import FileLink
# FileLink('EfficientNetV2L_Transferlearning_20_cusine.pkl')
```

```
Out[30]: EfficientNetV2L_Transferlearning_20_cusine.pkl
```

```
In [ ]:
```

```
In [ ]:
```

=====

## Transfer Learning

```
In [ ]: import numpy as np
import cv2
from sklearn.model_selection import train_test_split
from tensorflow.keras.applications import VGG16
from tensorflow.keras.applications.vgg16 import preprocess_input
from tensorflow.keras.layers import Flatten, Dense, Input
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import Adam
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.utils import to_categorical

# Define constants
IMG_SIZE = (224, 224) # VGG16 default image size

# Load images and labels
images = []
labels = []
for file_path in common_file_paths: # Replace with your actual file paths
    image = cv2.imread(file_path)
    if image is None:
        continue # Skip files that aren't valid images
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
```

```

    image = cv2.resize(image, IMG_SIZE)
    images.append(image)

    label = file_path.split(os.path.sep)[-2]
    labels.append(label)

# Convert lists to NumPy arrays
images = np.array(images, dtype=np.float32)
labels = np.array(labels)

# Preprocess images for VGG16
images = preprocess_input(images)

# Encode labels to integers
label_encoder = LabelEncoder()
labels_encoded = label_encoder.fit_transform(labels)
labels_one_hot = to_categorical(labels_encoded)

# Split the data into training and validation sets
X_train, X_val, y_train, y_val = train_test_split(images, labels_one_hot, te

# Load VGG16 pre-trained model without the top classification layer
base_model = VGG16(weights='imagenet', include_top=False, input_tensor=Input

# Freeze the layers of the base model
for layer in base_model.layers:
    layer.trainable = False

# Create the custom top layers for our dataset
x = base_model.output
x = Flatten()(x)
x = Dense(4096, activation='relu')(x)
x = Dense(4096, activation='relu')(x)
predictions = Dense(len(label_encoder.classes_), activation='softmax')(x)

# This is the model we will train
model = Model(inputs=base_model.input, outputs=predictions)

# Compile the model
model.compile(optimizer=Adam(learning_rate=1e-4), loss='categorical_crossentropy')

# Train the model
history = model.fit(X_train, y_train, validation_data=(X_val, y_val), epochs

# Evaluate the model on the validation set
val_loss, val_accuracy = model.evaluate(X_val, y_val)
print(f'Validation accuracy: {val_accuracy * 100:.2f}%')

# Optionally plot the training history
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()
plt.title('Accuracy over epochs')
plt.xlabel('Epoch')

```



```

plt.ylabel('Accuracy')

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.legend()
plt.title('Loss over epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')

plt.show()

```

```

In [ ]: from sklearn.metrics import confusion_matrix, classification_report
import seaborn as sns
import matplotlib.pyplot as plt

# Predict labels for validation set
y_pred = model.predict(X_val)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true_classes = np.argmax(y_val, axis=1)

# Calculate the confusion matrix
conf_matrix = confusion_matrix(y_true_classes, y_pred_classes)

# Plot the confusion matrix
plt.figure(figsize=(10, 10))
sns.heatmap(conf_matrix, annot=True, fmt='g', cmap='RdPu')
plt.xlabel('Predicted labels')
plt.ylabel('True labels')
plt.title('Confusion Matrix')
plt.show()

# Calculate precision, recall, F1 score, and accuracy
report = classification_report(y_true_classes, y_pred_classes, output_dict=True)
accuracy = report['accuracy']
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']

# Display the calculated metrics
print(f'Validation Accuracy: {accuracy:.4f}')
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1_score:.4f}')

```

```

In [ ]: import numpy as np
import cv2
import os
from sklearn.model_selection import train_test_split
from tensorflow.keras.applications import VGG16
from tensorflow.keras.applications.vgg16 import preprocess_input
from tensorflow.keras.layers import Flatten, Dense, Dropout, BatchNormalization
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import Adam
from sklearn.preprocessing import LabelEncoder

```

```

from tensorflow.keras.utils import to_categorical
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt

# Constants
IMG_SIZE = (224, 224) # VGG16 default image size

# Load and preprocess images
images = []
labels = []
for file_path in common_file_paths: # Replace with your actual file paths
    image = cv2.imread(file_path)
    if image is None:
        continue
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    image = cv2.resize(image, IMG_SIZE)
    images.append(image)
    label = file_path.split(os.path.sep)[-2]
    labels.append(label)

# Convert to NumPy arrays and preprocess
images = np.array(images, dtype=np.float32)
labels = np.array(labels)
images = preprocess_input(images)

# Encode labels
label_encoder = LabelEncoder()
labels_encoded = label_encoder.fit_transform(labels)
labels_one_hot = to_categorical(labels_encoded)

# Split data
X_train, X_val, y_train, y_val = train_test_split(images, labels_one_hot, te

# Data augmentation
data_gen = ImageDataGenerator(rotation_range=20, zoom_range=0.15, width_shif

# Load VGG16 model
base_model = VGG16(weights='imagenet', include_top=False, input_tensor=Input

# Freeze base model layers and unfreeze the last 4 layers
for layer in base_model.layers[:-4]:
    layer.trainable = False

# Model architecture
x = base_model.output
x = Flatten()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
x = BatchNormalization()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
predictions = Dense(len(label_encoder.classes_), activation='softmax')(x)
model = Model(inputs=base_model.input, outputs=predictions)

# Compile model
model.compile(optimizer=Adam(learning_rate=1e-5), loss='categorical_crossentropy

```

```

# Train model
history = model.fit(data_gen.flow(X_train, y_train, batch_size=32), validation_data=(X_val, y_val))

# Evaluate model
val_loss, val_accuracy = model.evaluate(X_val, y_val)
print(f'Validation accuracy: {val_accuracy * 100:.2f}%')

# Plotting training history
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()

plt.show()

```

```

In [ ]: from sklearn.metrics import confusion_matrix, classification_report
import seaborn as sns
import matplotlib.pyplot as plt

# Predict labels for validation set
y_pred = model.predict(X_val)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true_classes = np.argmax(y_val, axis=1)

# Calculate the confusion matrix
conf_matrix = confusion_matrix(y_true_classes, y_pred_classes)

# Plot the confusion matrix
plt.figure(figsize=(10, 10))
sns.heatmap(conf_matrix, annot=True, fmt='g', cmap='RdPu')
plt.xlabel('Predicted labels')
plt.ylabel('True labels')
plt.title('Confusion Matrix')
plt.show()

# Calculate precision, recall, F1 score, and accuracy
report = classification_report(y_true_classes, y_pred_classes, output_dict=True)
accuracy = report['accuracy']
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']

# Display the calculated metrics

```

```
print(f'Validation Accuracy: {accuracy:.4f}')
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1_score:.4f}')
```

## CNN with transfer learning

```
In [ ]: import os
import cv2
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from tensorflow.keras.applications.vgg16 import VGG16
from tensorflow.keras.layers import Flatten, Dense, Dropout, BatchNormalization
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.callbacks import ReduceLROnPlateau, EarlyStopping
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.models import Sequential
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from sklearn.metrics import confusion_matrix
import seaborn as sns

# Define image size and batch size
IMG_SIZE = (128, 128)
BATCH_SIZE = 32

# Load images and labels
images = []
labels = []
for file_path in common_file_paths:
    image = cv2.imread(file_path)
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    image = cv2.resize(image, IMG_SIZE)
    images.append(image)
    label = file_path.split(os.path.sep)[-2]
    labels.append(label)

# Convert lists to NumPy arrays and normalize
images = np.array(images) / 255.0
labels = np.array(labels)

# One-hot encoding of labels
label_encoder = LabelEncoder()
labels_encoded = label_encoder.fit_transform(labels)
labels_one_hot = to_categorical(labels_encoded)

# Split the data
X_train, X_val, y_train, y_val = train_test_split(images, labels_one_hot, te

# Data augmentation
data_augmentation = ImageDataGenerator(
    rotation_range=20,
    zoom_range=0.15,
```

```

        width_shift_range=0.2,
        height_shift_range=0.2,
        shear_range=0.15,
        horizontal_flip=True,
        fill_mode="nearest"
    )

    # Load pre-trained VGG16 model
    base_model = VGG16(weights='imagenet', include_top=False, input_shape=(IMG_S

    # Freeze layers in the base model
    for layer in base_model.layers:
        layer.trainable = False

    # Building the model
    model = Sequential([
        base_model,
        GlobalAveragePooling2D(),
        Dense(512, activation='relu'),
        BatchNormalization(),
        Dropout(0.5),
        Dense(256, activation='relu'),
        BatchNormalization(),
        Dropout(0.5),
        Dense(len(label_encoder.classes_), activation='softmax')
    ])

    # Compile the model
    optimizer = Adam(lr=1e-3)
    model.compile(optimizer=optimizer, loss='categorical_crossentropy', metrics=

    # Callbacks
    reduce_lr = ReduceLRonPlateau(monitor='val_loss', factor=0.2, patience=5, mi
    early_stopping = EarlyStopping(monitor='val_loss', patience=10, verbose=1, r

    # Train the model
    history = model.fit(
        data_augmentation.flow(X_train, y_train, batch_size=BATCH_SIZE),
        validation_data=(X_val, y_val),
        epochs=50,
        callbacks=[reduce_lr, early_stopping]
    )

    # Evaluate the model
    val_loss, val_accuracy = model.evaluate(X_val, y_val)
    print(f'Validation accuracy: {val_accuracy * 100:.2f}%')

    # Plot training history
    plt.figure(figsize=(12, 5))
    plt.subplot(1, 2, 1)
    plt.plot(history.history['accuracy'], label='Training Accuracy')
    plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
    plt.title('Accuracy over Epochs')
    plt.xlabel('Epoch')
    plt.ylabel('Accuracy')
    plt.legend()

```

```

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()

# Confusion Matrix
y_pred = model.predict(X_val)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_val, axis=1)
conf_matrix = confusion_matrix(y_true, y_pred_classes)
plt.figure(figsize=(10, 8))
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.show()

# Calculate precision, recall, F1 score, and accuracy
report = classification_report(y_true, y_pred_classes, output_dict=True)
accuracy = report['accuracy']
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']

# Display the calculated metrics
print(f'Validation Accuracy: {accuracy:.4f}')
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1_score:.4f}')

```

```

In [ ]: import os
import cv2
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from tensorflow.keras.applications.vgg16 import VGG16
from tensorflow.keras.layers import Flatten, Dense, Dropout, BatchNormalizat
from tensorflow.keras.optimizers import Adam, RMSprop
from tensorflow.keras.callbacks import ReduceLROnPlateau, EarlyStopping
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.models import Sequential
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from sklearn.metrics import confusion_matrix
import seaborn as sns

# Define image size and batch size
IMG_SIZE = (128, 128)
BATCH_SIZE = 32

# Load images and labels

```

```

images = []
labels = []
for file_path in common_file_paths:
    image = cv2.imread(file_path)
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    image = cv2.resize(image, IMG_SIZE)
    images.append(image)
    label = file_path.split(os.path.sep)[-2]
    labels.append(label)

# Convert lists to NumPy arrays and normalize
images = np.array(images) / 255.0
labels = np.array(labels)

# One-hot encoding of labels
label_encoder = LabelEncoder()
labels_encoded = label_encoder.fit_transform(labels)
labels_one_hot = to_categorical(labels_encoded)

# Split the data
X_train, X_val, y_train, y_val = train_test_split(images, labels_one_hot, te

# Data augmentation
data_augmentation = ImageDataGenerator(
    rotation_range=20,
    zoom_range=0.15,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.15,
    horizontal_flip=True,
    fill_mode="nearest"
)

# Load pre-trained VGG16 model
base_model = VGG16(weights='imagenet', include_top=False, input_shape=(IMG_S

# Freeze layers in the base model
for layer in base_model.layers:
    layer.trainable = False

# Building the model
model = Sequential([
    base_model,
    GlobalAveragePooling2D(),
    Dense(512, activation='relu'),
    BatchNormalization(),
    Dropout(0.5),
    Dense(256, activation='relu'),
    BatchNormalization(),
    Dropout(0.5),
    Dense(len(label_encoder.classes_), activation='softmax')
])

# Compile the model
optimizer = RMSprop(lr=1e-5)
model.compile(optimizer=optimizer, loss='categorical_crossentropy', metrics=

```

```

# Callbacks
reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.2, patience=5, mi
early_stopping = EarlyStopping(monitor='val_loss', patience=10, verbose=1, r

# Train the model
history = model.fit(
    data_augmentation.flow(X_train, y_train, batch_size=BATCH_SIZE),
    validation_data=(X_val, y_val),
    epochs=50,
    callbacks=[reduce_lr, early_stopping]
)

# Evaluate the model
val_loss, val_accuracy = model.evaluate(X_val, y_val)
print(f'Validation accuracy: {val_accuracy * 100:.2f}%')

# Plot training history
plt.figure(figsize=(12, 5))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()

# Confusion Matrix
y_pred = model.predict(X_val)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_val, axis=1)
conf_matrix = confusion_matrix(y_true, y_pred_classes)
plt.figure(figsize=(10, 8))
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.show()

# Calculate precision, recall, F1 score, and accuracy
report = classification_report(y_true_classes, y_pred_classes, output_dict=True)
accuracy = report['accuracy']
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']

# Display the calculated metrics
print(f'Validation Accuracy: {accuracy:.4f}')

```



```
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1_score:.4f}')
```

## transfer learning with new imagenet models

```
In [ ]: import numpy as np
import cv2
import os
from sklearn.model_selection import train_test_split
from tensorflow.keras.applications import EfficientNetV2M
from tensorflow.keras.applications.efficientnet_v2 import preprocess_input
from tensorflow.keras.layers import Flatten, Dense, Dropout, BatchNormalization
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import Adam
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt

# Constants
IMG_SIZE = (224, 224) # VGG16 default image size

# Load and preprocess images
images = []
labels = []
for file_path in common_file_paths: # Replace with your actual file paths
    image = cv2.imread(file_path)
    if image is None:
        continue
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    image = cv2.resize(image, IMG_SIZE)
    images.append(image)
    label = file_path.split(os.path.sep)[-2]
    labels.append(label)

# Convert to NumPy arrays and preprocess
images = np.array(images, dtype=np.float32)
labels = np.array(labels)
images = preprocess_input(images)

# Encode labels
label_encoder = LabelEncoder()
labels_encoded = label_encoder.fit_transform(labels)
labels_one_hot = to_categorical(labels_encoded)

# Split data
X_train, X_val, y_train, y_val = train_test_split(images, labels_one_hot, te

# Data augmentation
data_gen = ImageDataGenerator(rotation_range=20, zoom_range=0.15, width_shif
```

```

# Load EfficientNetV2M model
base_model = EfficientNetV2M(weights='imagenet', include_top=False, input_te

# Freeze base model layers and unfreeze the last 4 layers
for layer in base_model.layers[:-4]:
    layer.trainable = False

# Model architecture
x = base_model.output
x = Flatten()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
x = BatchNormalization()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
predictions = Dense(len(label_encoder.classes_), activation='softmax')(x)
model = Model(inputs=base_model.input, outputs=predictions)

# Compile model
model.compile(optimizer=Adam(learning_rate=1e-5), loss='categorical_crossentropy')

# Train model
history = model.fit(data_gen.flow(X_train, y_train, batch_size=32), validation_data=(X_val, y_val))

# Evaluate model
val_loss, val_accuracy = model.evaluate(X_val, y_val)
print(f'Validation accuracy: {val_accuracy * 100:.2f}%')

# Plotting training history
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()

plt.show()

```

```

In [ ]: # Confusion Matrix
y_pred = model.predict(X_val)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_val, axis=1)
conf_matrix = confusion_matrix(y_true, y_pred_classes)
plt.figure(figsize=(10, 8))
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')

```

```

plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.show()

# Calculate precision, recall, F1 score, and accuracy
report = classification_report(y_true, y_pred_classes, output_dict=True)
accuracy = report['accuracy']
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']

# Display the calculated metrics
print(f'Validation Accuracy: {accuracy:.4f}')
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1_score:.4f}')

```

```

In [ ]: import numpy as np
import cv2
import os
from sklearn.model_selection import train_test_split
from tensorflow.keras.applications import EfficientNetV2L
from tensorflow.keras.applications.efficientnet_v2 import preprocess_input
from tensorflow.keras.layers import Flatten, Dense, Dropout, BatchNormalization
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import Adam
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt

# Constants
IMG_SIZE = (224, 224) # VGG16 default image size

# Load and preprocess images
images = []
labels = []
for file_path in common_file_paths: # Replace with your actual file paths
    image = cv2.imread(file_path)
    if image is None:
        continue
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    image = cv2.resize(image, IMG_SIZE)
    images.append(image)
    label = file_path.split(os.path.sep)[-2]
    labels.append(label)

# Convert to NumPy arrays and preprocess
images = np.array(images, dtype=np.float32)
labels = np.array(labels)
images = preprocess_input(images)

# Encode labels
label_encoder = LabelEncoder()
labels_encoded = label_encoder.fit_transform(labels)
labels_one_hot = to_categorical(labels_encoded)

```

```

# Split data
X_train, X_val, y_train, y_val = train_test_split(images, labels_one_hot, te

# Data augmentation
data_gen = ImageDataGenerator(rotation_range=20, zoom_range=0.15, width_shif

# Load EfficientNetV2M model
base_model = EfficientNetV2L(weights='imagenet', include_top=False, input_te

# Freeze base model layers and unfreeze the last 4 layers
for layer in base_model.layers[:-4]:
    layer.trainable = False

# Model architecture
x = base_model.output
x = Flatten()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
x = BatchNormalization()(x)
x = Dense(4096, activation='relu')(x)
x = Dropout(0.5)(x)
predictions = Dense(len(label_encoder.classes_), activation='softmax')(x)
model = Model(inputs=base_model.input, outputs=predictions)

# Compile model
model.compile(optimizer=Adam(learning_rate=1e-5), loss='categorical_crossentropy')

# Train model
history = model.fit(data_gen.flow(X_train, y_train, batch_size=32), validati

# Evaluate model
val_loss, val_accuracy = model.evaluate(X_val, y_val)
print(f'Validation accuracy: {val_accuracy * 100:.2f}%')

# Plotting training history
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()

plt.show()

```

```
In [ ]: # Confusion Matrix
y_pred = model.predict(X_val)
y_pred_classes = np.argmax(y_pred, axis=1)
y_true = np.argmax(y_val, axis=1)
conf_matrix = confusion_matrix(y_true, y_pred_classes)
plt.figure(figsize=(10, 8))
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.show()

# Calculate precision, recall, F1 score, and accuracy
report = classification_report(y_true, y_pred_classes, output_dict=True)
accuracy = report['accuracy']
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']

# Display the calculated metrics
print(f'Validation Accuracy: {accuracy:.4f}')
print(f'Precision: {precision:.4f}')
print(f'Recall: {recall:.4f}')
print(f'F1 Score: {f1_score:.4f}')
```

In [ ]:

In [ ]: