E-waste classification Using Machine Learning

Model: EfficientNetV2B2, ResNet50

```
import tensorflow as tf
from tensorflow.keras import layers, models, optimizers, callbacks
from tensorflow.keras.models import Sequential, load_model
from tensorflow.keras.applications import EfficientNetV2B0
from tensorflow.keras.applications.efficientnet import
preprocess_input
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import confusion_matrix, classification_report
from tensorflow.keras.applications import EfficientNetV2B2,
MobileNetV2, ResNet50
import os
from PIL import Image
```

Dataset Analysis

```
dataset path = r"C:\Users\Kanishk Pareek\Downloads\archive\modified-
dataset"
train path = r"C:\Users\Kanishk Pareek\Downloads\archive\modified-
dataset\train"
test path = r"C:\Users\Kanishk Pareek\Downloads\archive\modified-
dataset\test"
val path = r"C:\Users\Kanishk Pareek\Downloads\archive\modified-
dataset\val"
# Retrieve the names of all folders in the Train dataset
garbage types = os.listdir(train path)
for garbage type in garbage types:
    folder path = os.path.join(train path, garbage type)
    if os.path.isdir(folder_path):
        image files = [f for f in os.listdir(folder path) if
f.endswith(('jpg', 'jpeg'))]
        # For displaying the count of images in the current folder
        num images = len(image files)
        print(f"{garbage type} folder contains {num images} images.")
print("-" * 40)
# Retrieve the names of all folders in the Test dataset
garbage types = os.listdir(test path)
for garbage type in garbage types:
    folder path = os.path.join(test path, garbage type)
    if os.path.isdir(folder path):
```

```
image files = [f for f in os.listdir(folder path) if
f.endswith(('jpg', 'jpeg'))]
        # For displaying the count of images in the current folder
        num images = len(image files)
        print(f"{garbage_type} folder contains {num images} images.")
print("-" * 40)
# Retrieve the names of all folders in the val dataset
garbage types = os.listdir(val path)
for garbage type in garbage types:
    folder path = os.path.join(val path, garbage type)
    if os.path.isdir(folder path):
        image files = [f for f in os.listdir(folder path) if
f.endswith(('jpg', 'jpeg'))]
        # For displaying the count of images in the current folder
        num images = len(image_files)
        print(f"{garbage type} folder contains {num images} images.")
Battery folder contains 240 images.
Keyboard folder contains 240 images.
Microwave folder contains 240 images.
Mobile folder contains 240 images.
Mouse folder contains 240 images.
PCB folder contains 240 images.
Player folder contains 240 images.
Printer folder contains 240 images.
Television folder contains 240 images.
Washing Machine folder contains 240 images.
Battery folder contains 30 images.
Keyboard folder contains 30 images.
Microwave folder contains 30 images.
Mobile folder contains 30 images.
Mouse folder contains 30 images.
PCB folder contains 30 images.
Player folder contains 30 images.
Printer folder contains 30 images.
Television folder contains 30 images.
Washing Machine folder contains 30 images.
Battery folder contains 30 images.
Keyboard folder contains 30 images.
Microwave folder contains 30 images.
Mobile folder contains 30 images.
Mouse folder contains 30 images.
PCB folder contains 30 images.
Player folder contains 30 images.
Printer folder contains 30 images.
```

```
Television folder contains 30 images.
Washing Machine folder contains 30 images.
datatrain=
tf.keras.utils.image dataset from directory(train path,shuffle = True,
image size = (128, 128), batch size = 32, validation split= False)
Found 2400 files belonging to 10 classes.
datatest=tf.keras.utils.image dataset from directory(test path,shuffle
= False, image size = (128,128), batch size = 32, validation split=
False)
Found 300 files belonging to 10 classes.
datavalid =
tf.keras.utils.image dataset from directory(val path, shuffle = True,
image size = (128, 128), batch size = 32, validation split= False)
Found 300 files belonging to 10 classes.
# Retrieve all folders (categories) in the train dataset
garbage types = os.listdir(train path)
# Displaying 3 sample images from each class
for garbage type in garbage types:
    folder path = os.path.join(train path, garbage type)
    if os.path.isdir(folder path):
        image files = [f for f in os.listdir(folder path) if
f.endswith(('jpg', 'jpeg', 'png'))]
        num images = len(image files)
        print(f"{garbage type} folder contains {num images} images.")
        # Display up to 3 images from each class
        sample images = image files[:3]
        plt.figure(figsize=(12, 4))
        for i, image file in enumerate(sample images):
            img path = os.path.join(folder path, image file)
            img = Image.open(img_path)
            plt.subplot(1, 3, i + 1)
            plt.imshow(img)
            plt.title(garbage type)
            plt.axis('off')
        plt.suptitle(f"Sample Images from '{garbage type}' Class",
fontsize=14)
        plt.show()
Battery folder contains 240 images.
```

Sample Images from 'Battery' Class

Battery







Keyboard folder contains 240 images.



Microwave folder contains 240 images.



Mobile folder contains 240 images.

Sample Images from 'Mobile' Class







Mouse folder contains 240 images.

Sample Images from 'Mouse' Class







PCB folder contains 240 images.

Sample Images from 'PCB' Class







Player folder contains 240 images.

Sample Images from 'Player' Class







Printer folder contains 240 images.

Sample Images from 'Printer' Class

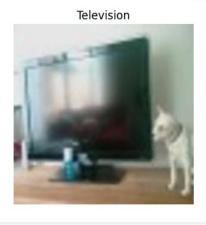






Television folder contains 240 images.

Sample Images from 'Television' Class







Washing Machine folder contains 240 images.

Sample Images from 'Washing Machine' Class

Washing Machine



Washing Machine



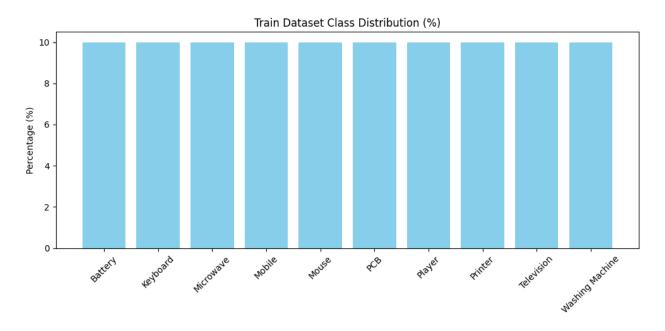
Washing Machine

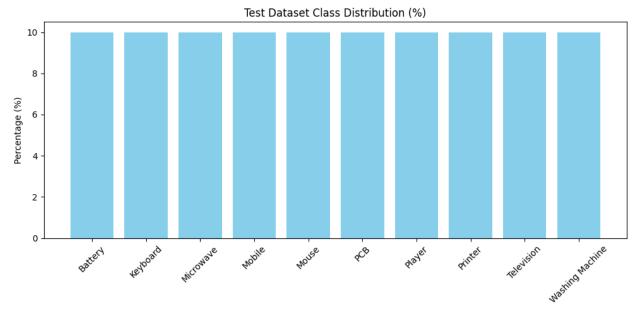


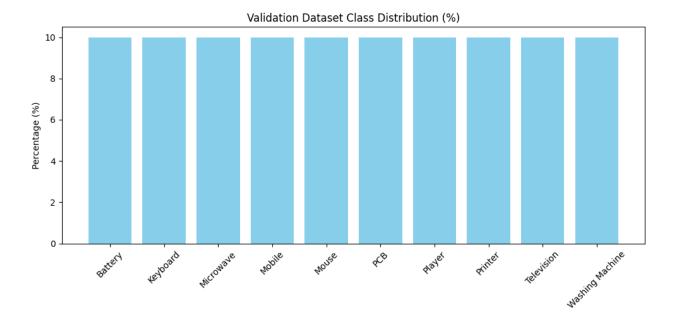
```
import os
def get class distribution percent(data path):
    Calculates the percentage distribution of image classes in a given
dataset split.
    Args:
        data path (str): Path to dataset split (e.g., train path,
test path, val path)
    Returns:
        dict: {class name: percentage of total images (rounded to 2
decimals)}
    class counts = {}
    total images = 0
    # Loop through each class folder
    for class name in os.listdir(data path):
        folder path = os.path.join(data_path, class_name)
        if os.path.isdir(folder path):
            image files = [f for f in os.listdir(folder path) if
f.lower().endswith(('jpg', 'jpeg'))]
            count = len(image files)
            class counts[class name] = count
            total images += count
    # Convert counts to percentages
    class percentages = {
        class name: round((count / total images) * 100, 2)
        for class name, count in class counts.items()
    }
    return class percentages
```

```
train percent = get class distribution percent(train path)
test percent = get class distribution percent(test path)
val percent = get class distribution percent(val path)
print("Train Class Distribution (%):")
print(train percent)
print("-" * 40)
print("Test Class Distribution (%):")
print(test percent)
print("-" \overline{*} 40)
print("Validation Class Distribution (%):")
print(val percent)
Train Class Distribution (%):
{'Battery': 10.0, 'Keyboard': 10.0, 'Microwave': 10.0, 'Mobile': 10.0,
'Mouse': 10.0, 'PCB': 10.0, 'Player': 10.0, 'Printer': 10.0,
'Television': 10.0, 'Washing Machine': 10.0}
Test Class Distribution (%):
{'Battery': 10.0, 'Keyboard': 10.0, 'Microwave': 10.0, 'Mobile': 10.0,
'Mouse': 10.0, 'PCB': 10.0, 'Player': 10.0, 'Printer': 10.0,
'Television': 10.0, 'Washing Machine': 10.0}
Validation Class Distribution (%):
{'Battery': 10.0, 'Keyboard': 10.0, 'Microwave': 10.0, 'Mobile': 10.0,
'Mouse': 10.0, 'PCB': 10.0, 'Player': 10.0, 'Printer': 10.0,
'Television': 10.0, 'Washing Machine': 10.0}
def plot class distribution(percent dict, title="Class Distribution")
(%)"):
    Plots a bar chart of class distribution percentages.
        percent dict (dict): class name -> percentage
        title (str): Title of the plot
    labels = list(percent dict.keys())
    percentages = list(percent dict.values())
    plt.figure(figsize=(10, 5))
    plt.bar(labels, percentages, color='skyblue')
    plt.vlabel("Percentage (%)")
    plt.title(title)
    plt.xticks(rotation=45)
    plt.tight layout()
    plt.show()
```

plot_class_distribution(train_percent, "Train Dataset Class
Distribution (%)")
plot_class_distribution(test_percent, "Test Dataset Class Distribution
(%)")
plot_class_distribution(val_percent, "Validation Dataset Class
Distribution (%)")







Data Augmetation

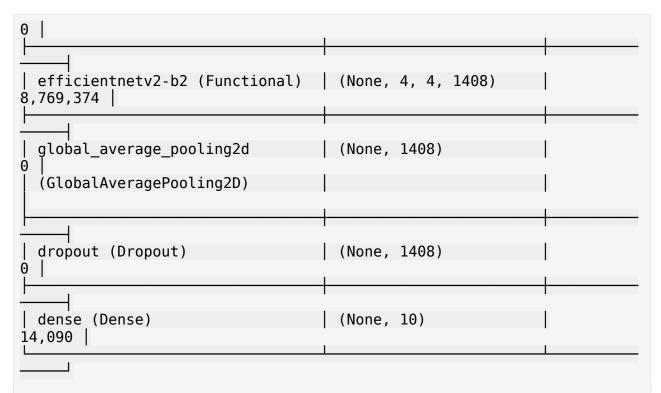
```
data_augmentation = tf.keras.Sequential([
    tf.keras.layers.RandomFlip("horizontal"),
    tf.keras.layers.RandomRotation(0.1),
    tf.keras.layers.RandomZoom(0.1),
])
```

Model Training(EfficientNetV2B2)

```
base model = tf.keras.applications.EfficientNetV2B2(
    input shape=(128, 128, 3),
    include top=False,
    weights='imagenet'
base model.trainable = True
for layer in base model.layers[:100]:
    layer.trainable = False
model = tf.keras.Sequential([
    tf.keras.layers.Input(shape=(128, 128, 3)),
    data augmentation,
    base model,
    tf.keras.layers.GlobalAveragePooling2D(),
    tf.keras.layers.Dropout(0.2),
    tf.keras.layers.Dense(10, activation='softmax')
])
model.compile(optimizer=tf.keras.optimizers.Adam(learning rate=0.0001)
,loss = tf.keras.losses.SparseCategoricalCrossentropy(),
metrics=['Accuracy'])
```

```
# Define an EarlyStopping callback to stop training when validation
loss stops improving
early = tf.keras.callbacks.EarlyStopping(
   monitor='val_loss', # Metric to monitor (validation
loss here)
                         # Stop training if there's no
   patience=3,
improvement after this 3 epochs
   restore_best_weights=True  # After stopping, restore the model
weights from the epoch with the best val loss
# Set the number of epochs
epochs = 20
# Train the model on the training dataset 'datatrain'
history = model.fit(
                            # Training data generator or
   datatrain,
dataset
   validation_data=datavalid, # Validation data generator or
dataset
   epochs=epochs, # Number of training epochs
batch_size=30, # Number of samples per gradient
update
   callbacks=[early],
   verbose = 1
                           # List of callbacks to apply during
training (e.g., early stopping)
Epoch 1/20
2.1132 - val Accuracy: 0.8333 - val loss: 1.0403
Epoch 2/20 75/75 41s 542ms/step - Accuracy: 0.8082 - loss:
0.9550 - val Accuracy: 0.9167 - val loss: 0.4449
Epoch 3/20
               39s 514ms/step - Accuracy: 0.8839 - loss:
0.4577 - val Accuracy: 0.9367 - val loss: 0.2626
Epoch 4/20
               39s 526ms/step - Accuracy: 0.9048 - loss:
75/75 —
0.3233 - val Accuracy: 0.9467 - val loss: 0.1993
0.2333 - val Accuracy: 0.9533 - val loss: 0.1673
0.1583 - val Accuracy: 0.9533 - val loss: 0.1545
0.1378 - val Accuracy: 0.9633 - val_loss: 0.1542
Epoch 8/20
```

```
45s 594ms/step - Accuracy: 0.9786 - loss:
0.0879 - val Accuracy: 0.9600 - val loss: 0.1424
Epoch 9/20
                _____ 58s 770ms/step - Accuracy: 0.9726 - loss:
75/75 —
0.0922 - val Accuracy: 0.9633 - val loss: 0.1392
Epoch 10/20 75/75 70s 928ms/step - Accuracy: 0.9826 - loss:
0.0716 - val Accuracy: 0.9600 - val loss: 0.1382
Epoch 11/20
75/75 — 71s 945ms/step - Accuracy: 0.9851 - loss:
0.0589 - val Accuracy: 0.9600 - val loss: 0.1520
Epoch 12/20
              68s 912ms/step - Accuracy: 0.9893 - loss:
75/75 ———
0.0477 - val Accuracy: 0.9700 - val loss: 0.1339
Epoch 13/20
               75/75 ———
0.0466 - val_Accuracy: 0.9667 - val_loss: 0.1409
Epoch 14/20
                  —— 65s 868ms/step - Accuracy: 0.9883 - loss:
0.0431 - val Accuracy: 0.9700 - val_loss: 0.1365
Epoch 15/20
               65s 872ms/step - Accuracy: 0.9897 - loss:
75/75 —
0.0399 - val Accuracy: 0.9633 - val_loss: 0.1293
Epoch 16/20 63s 844ms/step - Accuracy: 0.9850 - loss:
0.0476 - val Accuracy: 0.9700 - val loss: 0.1303
0.0478 - val Accuracy: 0.9633 - val loss: 0.1275
Epoch 18/20 66s 875ms/step - Accuracy: 0.9939 - loss:
0.0231 - val_Accuracy: 0.9600 - val_loss: 0.1342
Epoch 19/20
              65s 868ms/step - Accuracy: 0.9945 - loss:
75/75 ———
0.0261 - val Accuracy: 0.9633 - val loss: 0.1326
Epoch 20/20
               75/75 ———
0.0283 - val Accuracy: 0.9700 - val_loss: 0.1302
model.summary()
Model: "sequential 1"
Layer (type)
                            Output Shape
Param #
| sequential (Sequential)
                            (None, 128, 128, 3)
```



Total params: 24,744,022 (94.39 MB)

Trainable params: 7,980,278 (30.44 MB)

Non-trainable params: 803,186 (3.06 MB)

Optimizer params: 15,960,558 (60.88 MB)

base_model.summary()

Model: "efficientnetv2-b2"

| Layer (type) | Output Shape | Param # | Connected to |
|---|--------------------------|---------|--------------|
| input_layer (InputLayer) | (None, 128, 128, 3) | 0 | - |
| rescaling input_layer[0][0] (Rescaling) | (None, 128, 128, 3) | 0 | |

| normalization [0] (Normalization) | (None, 128, 128, 3) | 0 | rescaling[0] |
|--|-----------------------|---------------|---------------|
| stem_conv (Conv2D) normalization[0] | (None, 64, 64, 32) | 864 | |
| stem_bn [0] (BatchNormalizatio | (None, 64, 64, 32) | 128 | stem_conv[0] |
| stem_activation (Activation) | (None, 64, 64, 32) | 0 | stem_bn[0][0] |
| block1a_project_co stem_activation[(Conv2D) | (None, 64, 64, 16) | 4,608 | |
| blockla_project_bnblockla_project (BatchNormalizatio | (None, 64, 64, 16) | 64 | |
| blockla_project_ac blockla_project (Activation) | (None, 64, 64, 16) | 0 | |
| block1b_project_co block1a_project (Conv2D) | (None, 64, 64, 16) | 2,304 | |
| | | | |

| block1b_project_bn block1b_project (BatchNormalizatio | (None, 64, 64, 16) | 64 | |
|---|-----------------------|--------|--|
| block1b_project_ac block1b_project (Activation) | (None, 64, 64, 16) | 0 | |
| block1b_drop block1b_project (Dropout) | (None, 64, 64, 16) | 0 | |
| block1b_add (Add) block1b_drop[0][block1a_project | (None, 64, 64, 16) | 0 | |
| block2a_expand_conv block1b_add[0][0] (Conv2D) | (None, 32, 32, 64) | 9,216 | |
| block2a_expand_bn block2a_expand_c (BatchNormalizatio | (None, 32, 32, 64) | 256 | |
| block2a_expand_act block2a_expand_b (Activation) | (None, 32, 32, 64) | 0 | |
| block2a_project_co block2a_expand_a (Conv2D) | (None, 32, 32, 32) | 2,048 | |
| block2a_project_bn | (None, 32, 32, | 128 | |

| block2a_project (BatchNormalizatio | 32) | | |
|---|------------------------|--------|--|
| block2b_expand_convblock2a_project (Conv2D) | (None, 32, 32, | 36,864 | |
| block2b_expand_bn block2b_expand_c (BatchNormalizatio | (None, 32, 32, 128) | 512 | |
| block2b_expand_act block2b_expand_b (Activation) | (None, 32, 32, 128) | 0 | |
| block2b_project_co block2b_expand_a (Conv2D) | (None, 32, 32, 32) | 4,096 | |
| <pre></pre> | (None, 32, 32, 32) | 128 | |
| block2b_drop block2b_project (Dropout) | (None, 32, 32, 32) | 0 | |
| block2b_add (Add) block2b_drop[0][block2a_project | (None, 32, 32, 32) | 0 | |
| block2c_expand_conv block2b_add[0][0] | (None, 32, 32, | 36,864 | |

| (Conv2D) | 128) | | |
|---|--------------------|------------|--|
| block2c_expand_bn block2c_expand_c (BatchNormalizatio | (None, 32, 32, | 512 | |
| block2c_expand_act block2c_expand_b (Activation) | (None, 32, 32, | 0 | |
| block2c_project_co block2c_expand_a (Conv2D) | (None, 32, 32, 32) | 4,096 | |
| block2c_project_bn block2c_project (BatchNormalizatio | (None, 32, 32, 32) | 128 | |
| block2c_drop block2c_project (Dropout) | (None, 32, 32, 32) | 0 | |
| block2c_add (Add) block2c_drop[0][| (None, 32, 32, 32) | 0 | |
| block3a_expand_conv block2c_add[0][0] (Conv2D) | (None, 16, 16, | 36,864 | |
| block3a_expand_bnblock3a_expand_c (BatchNormalizatio | (None, 16, 16, | 512 | |

| block3a_expand_act block3a_expand_b (Activation) | (None, 16, 16, 128) | 0 | |
|---|------------------------|---------|--|
| block3a_project_co block3a_expand_a (Conv2D) | (None, 16, 16, 56) | 7,168 | |
| block3a_project_bnblock3a_project (BatchNormalizatio | (None, 16, 16, 56) | 224 | |
| block3b_expand_conv block3a_project (Conv2D) | (None, 16, 16, 224) | 112,896 | |
| block3b_expand_c | (None, 16, 16, 224) | 896 | |
| block3b_expand_act block3b_expand_b (Activation) | (None, 16, 16, 224) | 0 | |
| block3b_project_co block3b_expand_a (Conv2D) | (None, 16, 16, 56) | 12,544 | |
| block3b_project_bn block3b_project (BatchNormalizatio | (None, 16, 16, 56) | 224 | |

| block3b_drop block3b_project (Dropout) | (None, 16, 16, 56) | 0 | |
|---|------------------------|---------------|--|
| block3b_add (Add) block3b_drop[0][block3a_project | (None, 16, 16, 56) | 0 | |
| block3c_expand_conv block3b_add[0][0] (Conv2D) | (None, 16, 16, 224) | 112,896 | |
| block3c_expand_bn block3c_expand_c (BatchNormalizatio | (None, 16, 16, 224) | 896 | |
| block3c_expand_act block3c_expand_b (Activation) | (None, 16, 16, 224) | 0 | |
| block3c_project_co block3c_expand_a (Conv2D) | (None, 16, 16, 56) | 12,544 | |
| block3c_project_bn block3c_project (BatchNormalizatio | (None, 16, 16, 56) | 224 | |
| block3c_drop block3c_project (Dropout) | (None, 16, 16, 56) | 0 | |

| block3c_drop[0][| (None, 16, 16, 56) | 0 | |
|--|-----------------------|--------|---------------|
| block4a_expand_conv block3c_add[0][0] (Conv2D) | (None, 16, 16, 224) | 12,544 | |
| block4a_expand_bn block4a_expand_c (BatchNormalizatio | (None, 16, 16, 224) | 896 | |
| block4a_expand_act block4a_expand_b (Activation) | (None, 16, 16, 224) | 0 | |
| block4a_dwconv2 block4a_expand_a (DepthwiseConv2D) | (None, 8, 8, 224) | 2,016 | |
| block4a_bn block4a_dwconv2[(BatchNormalizatio | (None, 8, 8, 224) | 896 | |
| block4a_activation [0] (Activation) | (None, 8, 8, 224) | 0 | block4a_bn[0] |
| block4a_se_squeeze block4a_activati (GlobalAveragePool | (None, 224) | 0 | |
| | | | |

| block4a_se_reshape block4a_se_squee (Reshape) | (None, 1, 1, 224) | 0 | |
|---|-------------------|--------------|--|
| block4a_se_reduce block4a_se_resha (Conv2D) | (None, 1, 1, 14) | 3,150 | |
| block4a_se_expand block4a_se_reduc (Conv2D) | (None, 1, 1, 224) | 3,360 | |
| block4a_se_excite block4a_activati (Multiply) block4a_se_expan | (None, 8, 8, 224) | 0 | |
| block4a_project_co block4a_se_excit (Conv2D) | (None, 8, 8, 104) | 23,296 | |
| block4a_project_bn block4a_project (BatchNormalizatio | (None, 8, 8, 104) | 416 | |
| block4b_expand_conv block4a_project (Conv2D) | (None, 8, 8, 416) | 43,264 | |
| block4b_expand_bn block4b_expand_c (BatchNormalizatio | (None, 8, 8, 416) | 1,664 | |
| block4b_expand_act | (None, 8, 8, 416) | 0 | |

| block4b_expand_b (Activation) | | | |
|--|-----------------------|-------------|---------------|
| block4b_dwconv2 block4b_expand_a (DepthwiseConv2D) | (None, 8, 8, 416) | 3,744 | |
| block4b_bn block4b_dwconv2[(BatchNormalizatio | (None, 8, 8, 416) | 1,664 | |
| block4b_activation [0] (Activation) | (None, 8, 8, 416) | 0 | block4b_bn[0] |
| block4b_se_squeeze block4b_activati (GlobalAveragePool | (None, 416) | 0 | |
| block4b_se_reshape block4b_se_squee (Reshape) | (None, 1, 1, 416) | 0 | |
| block4b_se_reduce block4b_se_resha (Conv2D) | (None, 1, 1, 26) | 10,842 | |
| block4b_se_expand block4b_se_reduc (Conv2D) | (None, 1, 1, 416) | 11,232 | |
| block4b_se_excite block4b_activati | (None, 8, 8, 416) | 0 | |

| (Multiply) block4b_se_expan… | | | |
|--|-------------------|------------------|--|
| block4b_project_co block4b_se_excit (Conv2D) | (None, 8, 8, 104) | 43,264 | |
| block4b_project_bn block4b_project (BatchNormalizatio | (None, 8, 8, 104) | 416 | |
| block4b_drop block4b_project (Dropout) | (None, 8, 8, 104) | 0 | |
| block4b_add (Add) block4b_drop[0][block4a_project | (None, 8, 8, 104) | 0 | |
| block4c_expand_conv block4b_add[0][0] (Conv2D) | (None, 8, 8, 416) | 43,264 | |
| | (None, 8, 8, 416) | 1,664 | |
| block4c_expand_act block4c_expand_b (Activation) | (None, 8, 8, 416) | 0 | |
| block4c_dwconv2 block4c_expand_a (DepthwiseConv2D) | (None, 8, 8, 416) | 3,744 | |

| block4c_bn block4c_dwconv2[(BatchNormalizatio | (None, 8, 8, 416) | 1,664 | |
|--|-------------------|-----------|---------------|
| block4c_activation [0] (Activation) | (None, 8, 8, 416) | 0 | block4c_bn[0] |
| block4c_se_squeeze block4c_activati (GlobalAveragePool | (None, 416) | 0 | |
| block4c_se_reshape block4c_se_squee (Reshape) | (None, 1, 1, 416) | 0 | |
| block4c_se_reduce block4c_se_resha (Conv2D) | (None, 1, 1, 26) | 10,842 | |
| block4c_se_expand block4c_se_reduc (Conv2D) | (None, 1, 1, 416) | 11,232 | |
| block4c_se_excite block4c_activati (Multiply) block4c_se_expan | (None, 8, 8, 416) | 0 | |
| block4c_project_co block4c_se_excit (Conv2D) | (None, 8, 8, 104) | 43,264 | |

| (None, 8, 8, 104) | 416 | |
|-------------------|---|--|
| (None, 8, 8, 104) | 0 | |
| (None, 8, 8, 104) | 0 | |
| | | |
| (None, 8, 8, 416) | 43,264 | |
| (None, 8, 8, 416) | 1,664 | |
| (None, 8, 8, 416) | 0 | |
| (None, 8, 8, 416) | 3,744 | |
| (None, 8, 8, 416) | 1,664 | |
| | | |
| | (None, 8, 8, 104) (None, 8, 8, 416) (None, 8, 8, 416) (None, 8, 8, 416) (None, 8, 8, 416) | (None, 8, 8, 104) 0 (None, 8, 8, 416) 43,264 (None, 8, 8, 416) 1,664 (None, 8, 8, 416) 0 (None, 8, 8, 416) 3,744 |

| block4d_activation [0] (Activation) | (None, 8, 8, 416) | 0 | block4d_bn[0] |
|---|-------------------|------------------|---------------|
| block4d_se_squeeze block4d_activati (GlobalAveragePool | (None, 416) | 0 | |
| block4d_se_reshape block4d_se_squee (Reshape) | (None, 1, 1, 416) | 0 | |
| block4d_se_reduce block4d_se_resha (Conv2D) | (None, 1, 1, 26) | 10,842 | |
| block4d_se_expand block4d_se_reduc (Conv2D) | (None, 1, 1, 416) | 11,232 | |
| block4d_se_excite block4d_activati (Multiply) block4d_se_expan | (None, 8, 8, 416) | 0 | |
| block4d_project_co block4d_se_excit (Conv2D) | (None, 8, 8, 104) | 43,264 | |
| block4d_project_bn block4d_project (BatchNormalizatio | (None, 8, 8, 104) | 416 | |
| block4d_drop | (None, 8, 8, 104) | 0 | |

| block4d_project (Dropout) | | | |
|--|-------------------|---------------------|---------------|
| block4d_add (Add) block4d_drop[0][| (None, 8, 8, 104) | 0 | |
| block4c_add[0][0] | | | |
| block5a_expand_convblock4d_add[0][0] (Conv2D) | (None, 8, 8, 624) | 64,896 | |
| block5a_expand_bnblock5a_expand_c (BatchNormalizatio | (None, 8, 8, 624) | 2,496 | |
| block5a_expand_actblock5a_expand_b (Activation) | (None, 8, 8, 624) | 0 | |
| block5a_dwconv2 block5a_expand_a (DepthwiseConv2D) | (None, 8, 8, 624) | 5,616 | |
| block5a_bn block5a_dwconv2[(BatchNormalizatio | (None, 8, 8, 624) | 2,496 | |
| block5a_activation [0] (Activation) | (None, 8, 8, 624) | 0 | block5a_bn[0] |
| block5a_se_squeeze block5a_activati | (None, 624) | 0 | |

| (GlobalAveragePool | | | |
|--|-----------------------|------------|--|
| block5a_se_reshape block5a_se_squee (Reshape) | (None, 1, 1, 624) | 0 | |
| block5a_se_reduce block5a_se_resha (Conv2D) | (None, 1, 1, 26) | 16,250 | |
| block5a_se_expand block5a_se_reduc (Conv2D) | (None, 1, 1, 624) | 16,848 | |
| block5a_se_excite block5a_activati (Multiply) block5a_se_expan | (None, 8, 8, 624) | 0 | |
| block5a_project_co block5a_se_excit (Conv2D) | (None, 8, 8, 120) | 74,880 | |
| block5a_project_bnblock5a_project (BatchNormalizatio | (None, 8, 8, 120) | 480 | |
| block5b_expand_conv block5a_project (Conv2D) | (None, 8, 8, 720) | 86,400 | |
| block5b_expand_bn block5b_expand_c (BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |

| block5b_expand_act block5b_expand_b (Activation) | (None, 8, 8, 720) | 0 | |
|--|-------------------|--------|---------------|
| block5b_dwconv2 block5b_expand_a (DepthwiseConv2D) | (None, 8, 8, 720) | 6,480 | |
| block5b_bn block5b_dwconv2[(BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| block5b_activation [0] (Activation) | (None, 8, 8, 720) | 0 | block5b_bn[0] |
| block5b_se_squeeze block5b_activati (GlobalAveragePool | (None, 720) | 0 | |
| block5b_se_reshape block5b_se_squee (Reshape) | (None, 1, 1, 720) | 0 | |
| block5b_se_reduce | (None, 1, 1, 30) | 21,630 | |
| block5b_se_resha (Conv2D) | | | |

| | | I | |
|--|-------------------|------------------|--|
| block5b_se_excite block5b_activati (Multiply) block5b_se_expan | (None, 8, 8, 720) | 0 | |
| block5b_project_co block5b_se_excit (Conv2D) | (None, 8, 8, 120) | 86,400 | |
| block5b_project_bn block5b_project (BatchNormalizatio | (None, 8, 8, 120) | 480 | |
| block5b_drop block5b_project (Dropout) | (None, 8, 8, 120) | 0 | |
| block5b_add (Add) block5b_drop[0][| (None, 8, 8, 120) | 0 | |
| block5a_project | (None, 8, 8, 720) | 86,400 | |
| block5c_expand_bn block5c_expand_c (BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| block5c_expand_act block5c_expand_b (Activation) | (None, 8, 8, 720) | 0 | |

| block5c_expand_a | | | | |
|--|------------------------------------|-------------------|--------|---------------|
| block5c_dwconv2[(BatchNormalizatio block5c_activation (None, 8, 8, 720) 0 block5c_bn[0] [0] (Activation) 0 block5c_se_squeeze (None, 720) 0 block5c_se_squeeze (None, 720) 0 block5c_se_reshape (None, 1, 1, 720) 0 block5c_se_reshape (None, 1, 1, 720) 0 block5c_se_reshae (Reshape) (Reshape) (None, 1, 1, 30) 21,630 block5c_se_resha (Conv2D) block5c_se_reduc (Conv2D) block5c_se_reduc (Conv2D) | block5c_expand_a | (None, 8, 8, 720) | 6,480 | |
| Conv2D C | block5c_dwconv2[| (None, 8, 8, 720) | 2,880 | |
| block5c_activati (GlobalAveragePool block5c_se_reshape (None, 1, 1, 720) 0 block5c_se_squee (Reshape) block5c_se_reduce (None, 1, 1, 30) 21,630 block5c_se_resha (Conv2D) block5c_se_expand (None, 1, 1, 720) 22,320 block5c_se_reduc (Conv2D) block5c_se_excite (None, 8, 8, 720) 0 block5c_activati (Multiply) | [0] | (None, 8, 8, 720) | 0 | block5c_bn[0] |
| block5c_se_squee (Reshape) | block5c_activati | (None, 720) | 0 | |
| block5c_se_resha (Conv2D) | block5c_se_squee | (None, 1, 1, 720) | 0 | |
| block5c_se_reduc (Conv2D) | block5c_se_resha | (None, 1, 1, 30) | 21,630 | |
| block5c_activati (Multiply) | block5c_se_reduc | (None, 1, 1, 720) | 22,320 | |
| | block5c_activati (Multiply) | (None, 8, 8, 720) | 0 | |

| block5c_project_co block5c_se_excit (Conv2D) | (None, 8, 8, 120) | 86,400 | |
|--|-------------------|--------------|--|
| block5c_project_bn block5c_project (BatchNormalizatio | (None, 8, 8, 120) | 480 | |
| block5c_drop block5c_project (Dropout) | (None, 8, 8, 120) | 0 | |
| block5c_add (Add) block5c_drop[0][block5b_add[0][0] | (None, 8, 8, 120) | 0 | |
| block5d_expand_conv block5c_add[0][0] (Conv2D) | (None, 8, 8, 720) | 86,400 | |
| block5d_expand_bnblock5d_expand_c (BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| block5d_expand_act block5d_expand_b (Activation) | (None, 8, 8, 720) | 0 | |
| block5d_dwconv2 block5d_expand_a (DepthwiseConv2D) | (None, 8, 8, 720) | 6,480 | |
| block5d_bn | (None, 8, 8, 720) | 2,880 | |

| block5d_dwconv2[(BatchNormalizatio | | | |
|--|-------------------|------------|---------------|
| block5d_activation [0] (Activation) | (None, 8, 8, 720) | 0 | block5d_bn[0] |
| block5d_se_squeeze block5d_activati (GlobalAveragePool | (None, 720) | 0 | |
| block5d_se_reshape block5d_se_squee (Reshape) | (None, 1, 1, 720) | 0 | |
| block5d_se_reduce block5d_se_resha (Conv2D) | (None, 1, 1, 30) | 21,630 | |
| block5d_se_expand block5d_se_reduc (Conv2D) | (None, 1, 1, 720) | 22,320 | |
| block5d_se_excite block5d_activati (Multiply) block5d_se_expan | (None, 8, 8, 720) | 0 | |
| block5d_project_co block5d_se_excit (Conv2D) | (None, 8, 8, 120) | 86,400 | |
| block5d_project_bn block5d_project | (None, 8, 8, 120) | 480 | |

| (BatchNormalizatio | | | |
|--|-------------------|------------------|---------------|
| block5d_drop block5d_project (Dropout) | (None, 8, 8, 120) | 0 | |
| | (None, 8, 8, 120) | 0 | |
| block5e_expand_convblock5d_add[0][0] (Conv2D) | (None, 8, 8, 720) | 86,400 | |
| block5e_expand_bnblock5e_expand_c (BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| block5e_expand_act block5e_expand_b (Activation) | (None, 8, 8, 720) | 0 | |
| block5e_dwconv2 block5e_expand_a (DepthwiseConv2D) | (None, 8, 8, 720) | 6,480 | |
| block5e_bn block5e_dwconv2[(BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| block5e_activation [0] (Activation) | (None, 8, 8, 720) | 0 | block5e_bn[0] |

| block5e_se_squeeze block5e_activati (GlobalAveragePool | (None, 720) | | |
|--|-------------------|--------|--|
| block5e_se_reshape block5e_se_squee (Reshape) | (None, 1, 1, 720) | 0 | |
| block5e_se_reduce block5e_se_resha (Conv2D) | (None, 1, 1, 30) | 21,630 | |
| block5e_se_expand block5e_se_reduc (Conv2D) | (None, 1, 1, 720) | 22,320 | |
| block5e_se_excite block5e_activati (Multiply) block5e_se_expan | (None, 8, 8, 720) | 0 | |
| block5e_project_co block5e_se_excit (Conv2D) | (None, 8, 8, 120) | 86,400 | |
| block5e_project_bn block5e_project (BatchNormalizatio | (None, 8, 8, 120) | 480 | |
| block5e_drop block5e_project (Dropout) | (None, 8, 8, 120) | 0 | |

| block5e_add (Add) block5e_drop[0][| (None, 8, 8, 120) | 0 | |
|--|-------------------|------------------|---------------|
| block5d_add[0][0] | | | |
| block5f_expand_conv block5e_add[0][0] (Conv2D) | (None, 8, 8, 720) | 86,400 | |
| block5f_expand_bn block5f_expand_c (BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| block5f_expand_act block5f_expand_b (Activation) | (None, 8, 8, 720) | 0 | |
| block5f_dwconv2 block5f_expand_a (DepthwiseConv2D) | (None, 8, 8, 720) | 6,480 | |
| block5f_bn block5f_dwconv2[(BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| | (None, 8, 8, 720) | 0 | block5f_bn[0] |
| block5f_se_squeeze block5f_activati (GlobalAveragePool | (None, 720) | 0 | |
| | | | |

| block5f_se_reduce (None, 1, 1, 30) 21,630 block5f_se_resha (Conv2D) | block5f_se_reshape block5f_se_squee (Reshape) | (None, 1, 1, 720) | | |
|---|---|-------------------|------------------|--|
| block5f_se_reduc (Conv2D) | block5f_se_resha | (None, 1, 1, 30) | 21,630 | |
| block5f_activati (Multiply) block5f_se_expan | block5f_se_reduc | (None, 1, 1, 720) | 22,320 | |
| | block5f_activati (Multiply) | (None, 8, 8, 720) | 0 | |
| block5f_project_co (None, 8, 8, 120) 86,400 block5f_se_excit (Conv2D) | block5f_se_excit | (None, 8, 8, 120) | 86,400 | |
| block5f_project_bn (None, 8, 8, 120) 480 block5f_project (BatchNormalizatio | block5f_project | (None, 8, 8, 120) | 480 | |
| block5f_drop (None, 8, 8, 120) 0 block5f_project (Dropout) | block5f_project | (None, 8, 8, 120) | 0 | |
| block5f_add (Add) | | (None, 8, 8, 120) | 0 | |
| block5e_add[0][0] | | (None 8 8 720) | 86 400 | |

| block5f_add[0][0] (Conv2D) | | | |
|--|-------------------|--------|---------------|
| block6a_expand_bn block6a_expand_c (BatchNormalizatio | (None, 8, 8, 720) | 2,880 | |
| block6a_expand_actblock6a_expand_b (Activation) | (None, 8, 8, 720) | 0 | |
| block6a_dwconv2 block6a_expand_a (DepthwiseConv2D) | (None, 4, 4, 720) | 6,480 | |
| block6a_bn block6a_dwconv2[(BatchNormalizatio | (None, 4, 4, 720) | 2,880 | |
| block6a_activation [0] (Activation) | (None, 4, 4, 720) | 0 | block6a_bn[0] |
| block6a_se_squeeze block6a_activati (GlobalAveragePool | (None, 720) | 0 | |
| block6a_se_reshape block6a_se_squee (Reshape) | (None, 1, 1, 720) | 0 | |
| block6a_se_reduce block6a_se_resha | (None, 1, 1, 30) | 21,630 | |

| (Conv2D) | | | |
|---|-----------------------|-------------------|--|
| block6a_se_expand block6a_se_reduc (Conv2D) | (None, 1, 1, 720) | 22,320 | |
| block6a_se_excite block6a_activati (Multiply) block6a_se_expan | (None, 4, 4, 720) | 0 | |
| block6a_project_co block6a_se_excit (Conv2D) | (None, 4, 4, 208) | 149,760 | |
| block6a_project_bn block6a_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
| block6b_expand_conv block6a_project (Conv2D) | (None, 4, 4, 1248) | 259,584 | |
| block6b_expand_bn block6b_expand_c (BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6b_expand_act block6b_expand_b (Activation) | (None, 4, 4, 1248) | 0 | |
| block6b_dwconv2 block6b_expand_a (DepthwiseConv2D) | (None, 4, 4, 1248) | 11,232 | |

| block6b_bn block6b_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
|--|-----------------------|-------------|---------------|
| block6b_activation [0] (Activation) | (None, 4, 4, 1248) | 0 | block6b_bn[0] |
| block6b_se_squeeze block6b_activati (GlobalAveragePool | (None, 1248) | 0 | |
| block6b_se_reshape block6b_se_squee (Reshape) | (None, 1, 1, 1248) | 0 | |
| block6b_se_reduce block6b_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
| block6b_se_expand block6b_se_reduc (Conv2D) | (None, 1, 1, | 66,144 | |
| block6b_se_excite block6b_activati (Multiply) block6b_se_expan | (None, 4, 4, 1248) | 0 | |
| block6b_project_co block6b_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |

| block6b_project_bn block6b_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
|---|-----------------------|---------|--|
| block6b_drop block6b_project (Dropout) | (None, 4, 4, 208) | 0 | |
| block6b_add (Add) block6b_drop[0][| (None, 4, 4, 208) | 0 | |
| block6a_project | | | |
| block6c_expand_conv block6b_add[0][0] (Conv2D) | (None, 4, 4, 1248) | 259,584 | |
| block6c_expand_bn block6c_expand_c (BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| | (None, 4, 4, 1248) | 0 | |
| block6c_dwconv2 block6c_expand_a (DepthwiseConv2D) | (None, 4, 4, 1248) | 11,232 | |
| block6c_bn block6c_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |

| block6c_activation [0] | (None, 4, 4, | 0 | block6c_bn[0] |
|--|-------------------|---------|---------------|
| (Activation) | 1248) | | |
| | | | |
| block6c_se_squeeze block6c_activati | (None, 1248) | 0 | |
| (GlobalAveragePool | | | |
| | | | |
| block6c_se_reshape | (None, 1, 1, | 0 | |
| block6c_se_squee (Reshape) | 1248) | | |
| <u> </u> | | | |
| block6c_se_reduce | (None, 1, 1, 52) | 64,948 | |
| block6c_se_resha… (Conv2D) | | | |
| | | | |
| block6c_se_expand | (None, 1, 1, | 66,144 | |
| block6c_se_reduc (Conv2D) | 1248) | , | ' [|
| (comv2b) | 12 10 / | | |
| | (None, 4, 4, | 0 | |
| block6c_activati | | | l I |
| (Multiply) block6c_se_expan | 1248) | | |
| | | | |
| block6c_project_co… block6c_se_excit… | (None, 4, 4, 208) | 259,584 | |
| (Conv2D) | | | |
| | | | |
| block6c_project_bn block6c_project | (None, 4, 4, 208) | 832 | |
| (BatchNormalizatio | | | |
| | | | |
| | | | |

| block6c_drop block6c_project (Dropout) | (None, 4, 4, 208) | 0 | |
|--|-----------------------|-------------------|---------------|
| block6c_add (Add) block6c_drop[0][| (None, 4, 4, 208) | 0 | |
| block6b_add[0][0] | | | |
| block6d_expand_conv block6c_add[0][0] (Conv2D) | (None, 4, 4, 1248) | 259,584 | |
| | (None, 4, 4, 1248) | 4,992 | |
| block6d_expand_act block6d_expand_b (Activation) | (None, 4, 4, | 0 | |
| | (None, 4, 4, 1248) | 11,232 | |
| block6d_bn block6d_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6d_activation [0] (Activation) | (None, 4, 4, 1248) | 0 | block6d_bn[0] |
| block6d_se_squeeze | (None, 1248) | 0 | |

| block6d_activati (GlobalAveragePool | | | |
|---|-----------------------|-------------|--|
| block6d_se_squee | (None, 1, 1, 1, 1248) | 0 | |
| block6d_se_reduce block6d_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
| block6d_se_expand block6d_se_reduc (Conv2D) | (None, 1, 1, 1248) | 66,144 | |
| block6d_se_excite block6d_activati (Multiply) block6d_se_expan | (None, 4, 4, 1248) | 0 | |
| block6d_project_co block6d_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |
| block6d_project_bn block6d_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
| block6d_drop block6d_project (Dropout) | (None, 4, 4, 208) | 0 | |
| block6d_add (Add) | (None, 4, 4, 208) | 0 | |

| block6c_add[0][0] | | | |
|--|-----------------------|-----------------|---------------|
| block6e_expand_conv block6d_add[0][0] (Conv2D) | (None, 4, 4, | 259,584 | |
| | (None, 4, 4, | 4,992 | |
| block6e_expand_act block6e_expand_b (Activation) | (None, 4, 4, 1248) | 0 | |
| block6e_dwconv2 block6e_expand_a (DepthwiseConv2D) | (None, 4, 4, 1248) | 11,232 | |
| block6e_bn block6e_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6e_activation [0] (Activation) | (None, 4, 4, 1248) | 0 | block6e_bn[0] |
| block6e_se_squeeze block6e_activati (GlobalAveragePool | (None, 1248) | 0 | |
| block6e_se_reshape block6e_se_squee (Reshape) | (None, 1, 1, | 0 | |

| block6e_se_reduce block6e_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
|--|-----------------------|---------|--|
| block6e_se_expand block6e_se_reduc (Conv2D) | (None, 1, 1, 1248) | 66,144 | |
| block6e_se_excite block6e_activati (Multiply) block6e_se_expan | (None, 4, 4, 1248) | 0 | |
| block6e_project_co block6e_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |
| block6e_project_bn block6e_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
| block6e_drop block6e_project (Dropout) | (None, 4, 4, 208) | 0 | |
| | (None, 4, 4, 208) | 0 | |
| | (None, 4, 4, 1248) | 259,584 | |

| <pre> block6f_expand_bn block6f_expand_c (BatchNormalizatio </pre> | (None, 4, 4, 1248) | 4,992 | |
|--|-----------------------|-----------|---------------|
| block6f_expand_act block6f_expand_b (Activation) | (None, 4, 4, | 0 | |
| block6f_dwconv2 block6f_expand_a (DepthwiseConv2D) | (None, 4, 4, 1248) | 11,232 | |
| block6f_bn block6f_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6f_activation [0] (Activation) | (None, 4, 4, 1248) | 0 | block6f_bn[0] |
| block6f_se_squeeze block6f_activati (GlobalAveragePool | (None, 1248) | 0 | |
| block6f_se_reshape block6f_se_squee (Reshape) | (None, 1, 1, 1, 1248) | 0 | |
| block6f_se_reduce block6f_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
| | | | |

| block6f_se_expand block6f_se_reduc (Conv2D) | (None, 1, 1, 1248) | 66,144 | |
|---|-----------------------|---------------|--|
| <pre> </pre> | (None, 4, 4, 1248) | 0 | |
| block6f_project_co block6f_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |
| block6f_project_bn block6f_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
| block6f_drop block6f_project (Dropout) | (None, 4, 4, 208) | 0 | |
| | (None, 4, 4, 208) | 0 | |
| block6g_expand_conv block6f_add[0][0] (Conv2D) | (None, 4, 4, 1248) | 259,584 | |
| block6g_expand_bn block6g_expand_c (BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6g_expand_act | (None, 4, 4, | 0 | |

| <pre>block6g_expand_b (Activation)</pre> | 1248) | | |
|--|-----------------------|-------------|---------------|
| block6g_dwconv2 block6g_expand_a (DepthwiseConv2D) | (None, 4, 4, 1248) | 11,232 | |
| block6g_bn block6g_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6g_activation [0] (Activation) | (None, 4, 4, 1248) | 0 | block6g_bn[0] |
| block6g_se_squeeze block6g_activati (GlobalAveragePool | (None, 1248) | 0 | |
| block6g_se_reshape block6g_se_squee (Reshape) | (None, 1, 1, 1248) | 0 | |
| block6g_se_reduce block6g_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
| block6g_se_expand block6g_se_reduc (Conv2D) | (None, 1, 1, 1248) | 66,144 | |
| block6g_se_excite block6g_activati | (None, 4, 4, | 0 | |

| (Multiply) block6g_se_expan | 1248) | | |
|--|-----------------------|-------------------|--|
| block6g_project_co block6g_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |
| block6g_project_bn block6g_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
| block6g_drop block6g_project (Dropout) | (None, 4, 4, 208) | 0 | |
| block6g_add (Add) block6g_drop[0][block6f_add[0][0] | (None, 4, 4, 208) | 0 | |
| | (None, 4, 4, 1248) | 259,584 | |
| block6h_expand_bn block6h_expand_c (BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6h_expand_act block6h_expand_b (Activation) | (None, 4, 4, 1248) | 0 | |
| block6h_dwconv2 block6h_expand_a (DepthwiseConv2D) | (None, 4, 4, | 11,232 | |

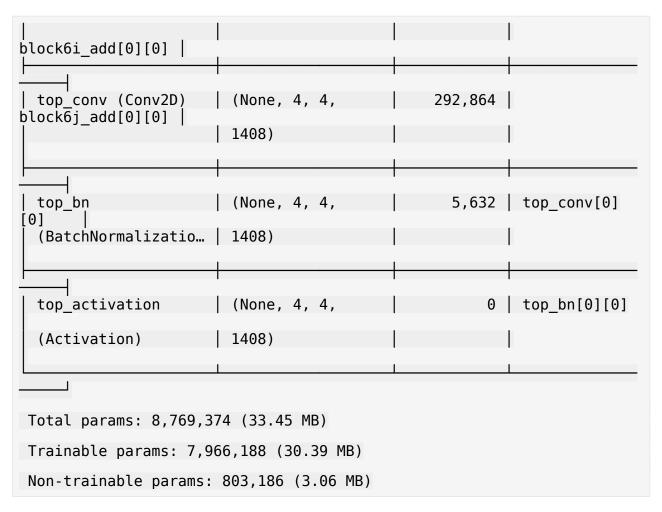
| block6h_bn block6h_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
|--|-----------------------|-------------|---------------|
| block6h_activation [0] (Activation) | (None, 4, 4, 1248) | 0 | block6h_bn[0] |
| block6h_se_squeeze block6h_activati (GlobalAveragePool | (None, 1248) | 0 | |
| block6h_se_reshape block6h_se_squee (Reshape) | (None, 1, 1, 1248) | 0 | |
| block6h_se_reduce block6h_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
| block6h_se_expand block6h_se_reduc (Conv2D) | (None, 1, 1, 1, 1248) | 66,144 | |
| block6h_se_excite block6h_activati (Multiply) block6h_se_expan | (None, 4, 4, 1248) | 0 | |
| block6h_project_co block6h_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |

| block6h_project_bn block6h_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
|---|-----------------------|---------------------|--|
| block6h_drop block6h_project (Dropout) | (None, 4, 4, 208) | 0 | |
| block6h_add (Add) block6h_drop[0][| (None, 4, 4, 208) | 0 | |
| block6g_add[0][0] | | | |
| block6i_expand_conv block6h_add[0][0] (Conv2D) | (None, 4, 4, 1248) | 259,584 | |
| block6i_expand_bn block6i_expand_c (BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |
| block6i_expand_act block6i_expand_b (Activation) | (None, 4, 4, 1248) | 0 | |
| block6i_dwconv2 block6i_expand_a (DepthwiseConv2D) | (None, 4, 4, 1248) | 11,232 | |
| block6i_bn block6i_dwconv2[(BatchNormalizatio | (None, 4, 4, 1248) | 4,992 | |

| block6i_activation [0] | (None, 4, 4, | 0 | block6i_bn[0] |
|--|-----------------------|---------|---------------|
| | 1248) | | |
| block6i_se_squeeze block6i_activati (GlobalAveragePool | (None, 1248) | 0 | |
| block6i_se_reshape block6i_se_squee (Reshape) | (None, 1, 1, 1248) | 0 | |
| block6i_se_reduce block6i_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
| block6i_se_expand block6i_se_reduc (Conv2D) | (None, 1, 1, 1248) | 66,144 | |
| block6i_se_excite block6i_activati (Multiply) block6i_se_expan | (None, 4, 4, 1248) | 0 | |
| block6i_project_co block6i_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |
| block6i_project_bn block6i_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
| | | | |

| block6i_drop block6i_project (Dropout) | (None, 4, 4, 208) | 0 | |
|--|-----------------------|-------------------|---------------|
| block6i_add (Add) block6i_drop[0][| (None, 4, 4, 208) | 0 | |
| block6h_add[0][0] | | | |
| block6j_expand_conv block6i_add[0][0] (Conv2D) | (None, 4, 4, 1248) | 259,584 | |
| block6j_expand_bnblock6j_expand_c (BatchNormalizatio | (None, 4, 4, | 4,992 | |
| block6j_expand_act block6j_expand_b (Activation) | (None, 4, 4, | 0 | |
| block6j_dwconv2 block6j_expand_a (DepthwiseConv2D) | (None, 4, 4, 1248) | 11,232 | |
| | (None, 4, 4, 1248) | 4,992 | |
| | (None, 4, 4, 1248) | 0 | block6j_bn[0] |
| | (None, 1248) | 0 | |

| <pre>block6j_activati (GlobalAveragePool </pre> | | | |
|---|-----------------------|-------------|--|
| block6j_se_squee | (None, 1, 1, 1, 1248) | 0 | |
| block6j_se_reduce block6j_se_resha (Conv2D) | (None, 1, 1, 52) | 64,948 | |
| block6j_se_expand block6j_se_reduc (Conv2D) | (None, 1, 1, 1248) | 66,144 | |
| block6j_se_excite block6j_activati (Multiply) block6j_se_expan | (None, 4, 4, 1248) | 0 | |
| block6j_project_co block6j_se_excit (Conv2D) | (None, 4, 4, 208) | 259,584 | |
| block6j_project_bn block6j_project (BatchNormalizatio | (None, 4, 4, 208) | 832 | |
| block6j_drop block6j_project (Dropout) | (None, 4, 4, 208) | 0 | |
| | (None, 4, 4, 208) | 0 | |



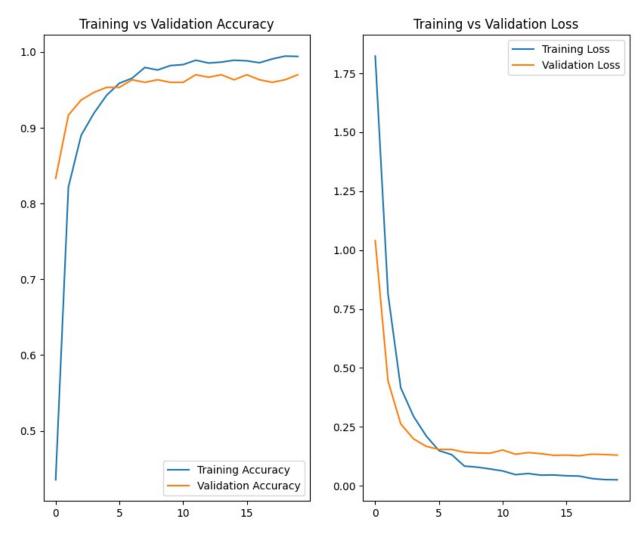
Model Performance Visualization

```
# Plotting Training and Validation Accuracy and Loss Over Epochs
acc = history.history['Accuracy']
                                           # Training accuracy
val_acc = history.history['val_Accuracy'] # Validation accuracy
loss = history.history['loss']
                                          # Training loss
val loss = history.history['val loss']
                                          # Validation loss
epochs range = range(len(acc))
                                           # X-axis range based on
number of epochs
plt.figure(figsize=(10, 8))
plt.subplot(1, 2, 1)
                                           # 1 row, 2 columns,
position 1
plt.plot(epochs range, acc, label='Training Accuracy')
training accuracy
plt.plot(epochs range, val acc, label='Validation Accuracy') #
validation accuracy
plt.legend(loc='lower right')
```

```
plt.title('Training vs Validation Accuracy')

plt.subplot(1, 2, 2)  # 1 row, 2 columns,
    position 2
plt.plot(epochs_range, loss, label='Training Loss')  #
    training loss
plt.plot(epochs_range, val_loss, label='Validation Loss')  #
    validation loss
plt.legend(loc='upper right')
plt.title('Training vs Validation Loss')

plt.show()
```

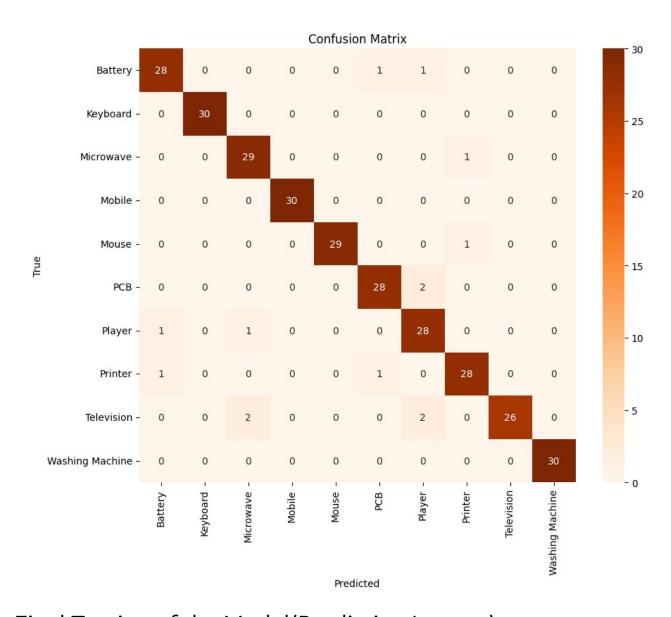


Model Evaluation (EfficientNetV2B2)

```
loss, accuracy = model.evaluate(datatest)
print(f'Test accuracy is{accuracy:.4f}, Test loss is {loss:.4f}')
```

```
10/10 -
                         -- 5s 500ms/step - Accuracy: 0.9591 - loss:
0.1493
Test accuracy is0.9533, Test loss is 0.1532
# Evaluate Model Performance on Test Data using Confusion Matrix and
Classification Report
# Extract true labels from all batches
y true = np.concatenate([y.numpy() for x, y in datatest], axis=\frac{0}{0}) #
Ground truth labels
# Get predictions as probabilities and then predicted classes
y pred probs = model.predict(datatest)
y_pred = np.argmax(y_pred_probs, axis=1)
print(confusion matrix(y true, y pred))
print(classification report(y true, y pred))
10/10
                            10s 756ms/step
[[28 0
         0
            0
               0
                   1
                      1
                         0
                            0
                                01
 [ 0 30
         0
            0
               0
                  0
                      0
                         0
                            0
                                01
 0
      0 29
            0
               0
                  0
                     0
                         1 0
                                0]
         0 30
                  0
                      0
  0
      0
               0
                         0
                            0
                                0]
            0 29
                  0
 0 1
      0
                      0
                         1
                            0
         0
                                01
   0
      0
            0
              0 28
                      2
                         0
                            0
                                01
         0
                  0 28
                         0 0
 [ 1
      0
         1
           0
               0
                                0]
 [ 1
      0
         0
            0
               0
                   1
                      0 28
                            0
                                0]
         2
            0
                      2
 [ 0
      0
               0
                   0
                         0 26
                                0]
 [ 0
      0
         0
            0
               0
                   0
                      0
                         0
                            0 30]]
               precision
                            recall
                                    f1-score
                                                 support
           0
                    0.93
                              0.93
                                         0.93
                                                      30
           1
                    1.00
                               1.00
                                         1.00
                                                      30
           2
                    0.91
                              0.97
                                         0.94
                                                      30
           3
                    1.00
                               1.00
                                         1.00
                                                      30
           4
                              0.97
                    1.00
                                         0.98
                                                      30
           5
                    0.93
                              0.93
                                         0.93
                                                      30
           6
                    0.85
                              0.93
                                         0.89
                                                      30
           7
                    0.93
                              0.93
                                         0.93
                                                      30
           8
                    1.00
                               0.87
                                         0.93
                                                      30
           9
                    1.00
                               1.00
                                         1.00
                                                      30
    accuracy
                                         0.95
                                                     300
                    0.96
                               0.95
                                         0.95
                                                     300
   macro avg
weighted avg
                    0.96
                               0.95
                                         0.95
                                                     300
```

```
print(len(datatrain.class names))
class names = datatrain.class names
print(class_names)
10
['Battery', 'Keyboard', 'Microwave', 'Mobile', 'Mouse', 'PCB',
'Player', 'Printer', 'Television', 'Washing Machine']
### Plot Confusion Matrix as Heatmap for Visualization
cm = confusion_matrix(y_true, y_pred)
plt.figure(figsize=(10, 8))
sns.heatmap(cm, annot=True, fmt='d',
            xticklabels=class names,
            yticklabels=class names,
            cmap='Oranges')
plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('Confusion Matrix')
plt.show()
```



Final Testing of the Model(Predicting Images)

```
### Display Sample Predictions: True Labels vs Predicted Labels

class_names = datatest.class_names
# Get class names from test dataset

for images, labels in datatest.take(3):
# Take 2 batch from test data
    predictions = model.predict(images)
# Predicting class probabilities
    pred_labels = tf.argmax(predictions, axis=1)
# Get predicted class indices

    for i in range(3):
# Display first 3 images from batch
```

```
plt.imshow(images[i].numpy().astype("uint8"))
# Convert tensor to image
        plt.title(f"True: {class_names[labels[i]]}, Pred:
{class_names[pred_labels[i]]}")
        plt.axis("off")

        plt.show()

1/1 _______ 2s 2s/step
```

True: Battery, Pred: Battery



True: Battery, Pred: Battery



True: Battery, Pred: Battery



True: Keyboard, Pred: Keyboard



True: Keyboard, Pred: Keyboard



True: Keyboard, Pred: Keyboard



1/1 — 0s 229ms/step

True: Microwave, Pred: Microwave



True: Microwave, Pred: Microwave



True: Microwave, Pred: Microwave



Using Resnet50

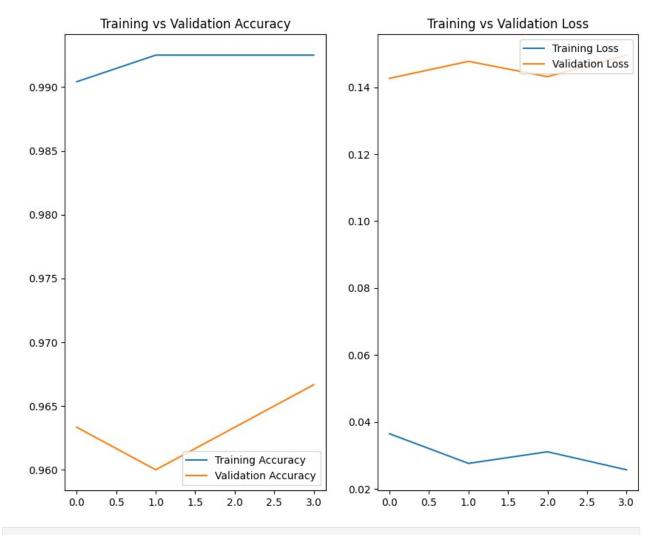
```
Resnet50 model = tf.keras.applications.ResNet50(
    input shape=(128, 128, 3),
    include top=False,
   weights='imagenet'
)
base model.trainable = True # Allow training
for layer in Resnet50 model.layers[:100]:
   layer.trainable = False
Downloading data from https://storage.googleapis.com/tensorflow/keras-
applications/resnet/
resnet50 weights tf dim ordering tf kernels notop.h5
94765736/94765736 ——
                                ---- 7s Ous/step
model 2 = tf.keras.Sequential([
   tf.keras.layers.Input(shape=(128, 128, 3)),
   base model,
   tf.keras.layers.GlobalAveragePooling2D(),
   tf.keras.layers.Dropout(0.4),
   tf.keras.layers.Dense(10, activation='softmax') # for 10 classes
])
model 2.compile(optimizer=tf.keras.optimizers.Adam(learning rate=0.000
1),loss = tf.keras.losses.SparseCategoricalCrossentropy(),
metrics=['Accuracy'])
early = tf.keras.callbacks.EarlyStopping(
   monitor='val loss',
   patience=5,
    restore best weights=True
)
# Set the number of training epochs
epochs = 20
# Train the model using the training and validation datasets
history = model.fit(
   datatrain,
                                   # Training dataset
   validation data=datavalid,
                                 # Validation dataset to monitor
performance
   epochs=epochs,
                                  # Train for up to 20 epochs
                                   # Number of samples per training
   batch size=30,
   callbacks=[early],
                                   # Stop early if validation loss
doesn't improve
```

```
verbose=1
                            # Print training progress
)
Epoch 1/20
              37s 498ms/step - Accuracy: 0.9949 - loss:
75/75 ——
0.0200 - val Accuracy: 0.9633 - val loss: 0.1378
Epoch 2/20
              38s 512ms/step - Accuracy: 0.9962 - loss:
75/75 ———
0.0190 - val Accuracy: 0.9700 - val loss: 0.1384
Epoch 3/20
                  ---- 38s 504ms/step - Accuracy: 0.9961 - loss:
0.0176 - val_Accuracy: 0.9667 - val_loss: 0.1532
Epoch 4/20
                  ---- 38s 513ms/step - Accuracy: 0.9947 - loss:
75/75 —
0.0204 - val Accuracy: 0.9700 - val loss: 0.1353
Epoch 5/20

39s 519ms/step - Accuracy: 0.9943 - loss:
0.0237 - val Accuracy: 0.9667 - val loss: 0.1397
0.0140 - val Accuracy: 0.9633 - val loss: 0.1444
Epoch 7/20
          40s 537ms/step - Accuracy: 0.9953 - loss:
75/75 ———
0.0196 - val Accuracy: 0.9600 - val loss: 0.1387
Epoch 8/20
           40s 536ms/step - Accuracy: 0.9934 - loss:
0.0196 - val Accuracy: 0.9700 - val loss: 0.1369
Epoch 9/20
                  —— 41s 541ms/step - Accuracy: 0.9942 - loss:
75/75 —
0.0230 - val Accuracy: 0.9667 - val loss: 0.1228
Epoch 10/20
              41s 540ms/step - Accuracy: 0.9970 - loss:
0.0169 - val Accuracy: 0.9733 - val loss: 0.1321
Epoch 11/20

41s 548ms/step - Accuracy: 0.9963 - loss:
0.0148 - val Accuracy: 0.9667 - val_loss: 0.1413
0.0222 - val Accuracy: 0.9633 - val_loss: 0.1352
0.0167 - val Accuracy: 0.9633 - val_loss: 0.1367
Epoch 14/20 42s 557ms/step - Accuracy: 0.9968 - loss:
0.0145 - val_Accuracy: 0.9667 - val_loss: 0.1442
# Plotting Training and Validation Accuracy and Loss Over Epochs
acc = history.history['Accuracy'] # Training accuracy
val_acc = history.history['val_Accuracy'] # Validation accuracy
```

```
loss = history.history['loss']
                                           # Training loss
val_loss = history.history['val_loss'] # Validation loss
epochs range = range(len(acc))
                                          # X-axis range based on
number of epochs
plt.figure(figsize=(10, 8))
plt.subplot(1, 2, 1)
                                           # 1 row, 2 columns,
position 1
plt.plot(epochs range, acc, label='Training Accuracy')
training accuracy
plt.plot(epochs range, val acc, label='Validation Accuracy') #
validation accuracy
plt.legend(loc='lower right')
plt.title('Training vs Validation Accuracy')
                                           # 1 row, 2 columns,
plt.subplot(1, 2, 2)
position 2
plt.plot(epochs range, loss, label='Training Loss')
training loss
plt.plot(epochs_range, val loss, label='Validation Loss')
validation loss
plt.legend(loc='upper right')
plt.title('Training vs Validation Loss')
plt.show()
```



model.save('Best_model.keras')