Garbage Classification Week2

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1 Garbage Classification with EfficientNetV2B2

1.1 Developed by Sai Pranesh

1.2 Project Description

In this project, the aim is to develop a sophisticated **garbage classification system** leveraging the **EfficientNetV2B2** architecture. Our primary dataset serves as a foundation for building models that can eventually automate waste segregation, a critical step in optimizing recycling and waste management, ultimately aiding in environmental conservation.

Goal: To develop an accurate and efficient garbage classification model using EfficientNetV2B2 and transfer learning for automated waste sorting.

1.3 Challenges and Scope

Key Challenge: A notable challenge encountered is the inherent **class imbalance** within the dataset.

1.3.1 Core Libraries

- tensorflow: For deep learning model building and training.
- $\bullet\,$ numpy: For numerical operations and array manipulation.
- matplotlib.pyplot: For plotting training curves and results.

```
import numpy as np # Importing NumPy for numerical operations and array_

manipulations
import matplotlib.pyplot as plt # Importing Matplotlib for plotting graphs and_
visualizations
import seaborn as sns # Importing Seaborn for statistical data visualization,_
built on top of Matplotlib
import tensorflow as tf # Importing TensorFlow for building and training_
machine learning models
from tensorflow import keras # Importing Keras, a high-level API for_
TensorFlow, to simplify model building
from tensorflow.keras import Layer # Importing Layer class for creating custom_
alayers in Keras
```

```
from tensorflow.keras.models import Sequential # Importing Sequential model_
for building neural networks layer-by-layer

from tensorflow.keras.layers import Rescaling , GlobalAveragePooling2D

from tensorflow.keras import layers, optimizers, callbacks # Importing various_
modules for layers, optimizers, and callbacks in Keras

from sklearn.utils.class_weight import compute_class_weight # Importing_
function to compute class weights for imbalanced datasets

from tensorflow.keras.applications import EfficientNetV2B2 # Importing_
from tensorflow.keras.applications import EfficientNetV2B2 # Importing_
from sklearn.metrics import confusion_matrix, classification_report #_
Importing functions to evaluate model performance
import gradio as gr # Importing Gradio for creating interactive web interfaces_
for machine learning models
```

1.4 1. Exploring and Understanding the Data.

- Loading image dataset using tools like image_dataset_from_directory.
- Visualizing sample images from each class.
- Check the number of images per class to ensure balance.
- Understand image dimensions, color channels, and class labels.

```
[2]: dataset_dir= r"C:\Users\Pranesh\OneDrive\Desktop\LMS\AICTE_\( \) \cdot Internship\TrashType_Image_Dataset"

image_size = (124, 124)

batch_size = 32

seed = 42
```

Found 2527 files belonging to 6 classes. Using 2022 files for training.

```
val_class= val_ds.class_names
```

Found 2527 files belonging to 6 classes. Using 505 files for validation.

```
[5]: # Get the total number of batches in the validation dataset
val_batches = tf.data.experimental.cardinality(val_ds)

# Split the validation dataset into two equal parts:
# First half becomes the test dataset
test_ds = val_ds.take(val_batches // 2)

# Second half remains as the validation dataset
val_dat = val_ds.skip(val_batches // 2)

# Optimize test dataset by caching and prefetching to improve performance
test_ds_eval = test_ds.cache().prefetch(tf.data.AUTOTUNE)
```

```
[6]: print(train_ds.class_names)
    print(val_class)
    print(len(train_ds.class_names))

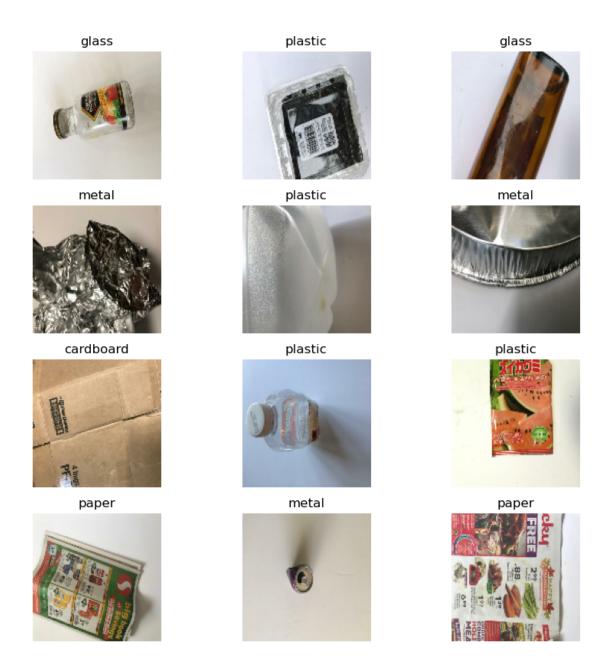
['cardboard', 'glass', 'metal', 'paper', 'plastic', 'trash']
    ['cardboard', 'glass', 'metal', 'paper', 'plastic', 'trash']
```

1.4.1 Visualization of sample images from each class.

6

```
[7]: import matplotlib.pyplot as plt

plt.figure(figsize=(10, 10))
for images, labels in train_ds.take(1):
    for i in range(12):
        ax = plt.subplot(4, 3, i + 1)
        plt.imshow(images[i].numpy().astype("uint8"))
        plt.title(train_ds.class_names[labels[i]])
        plt.axis("off")
```



1.5 Checking the number of images per class to ensure balance.

•

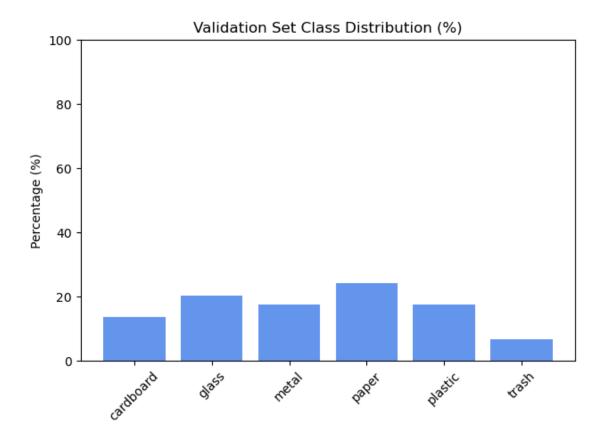
1.6 Understanding image properties like Image dimensions, Class labels.

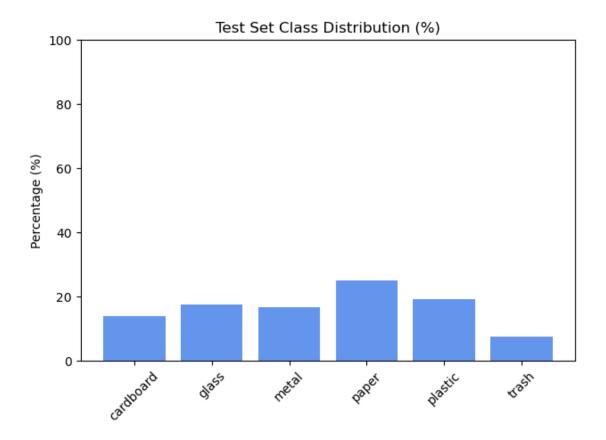
```
[8]: def count_distribution(dataset, class_names):
          total = 0
          counts = {name: 0 for name in class_names}
          for _, labels in dataset:
              for label in labels.numpy():
                  class_name = class_names[label]
                  counts[class_name] += 1
                  total += 1
          for k in counts:
              counts[k] = round((counts[k] / total) * 100, 2) # Convert to percentage
          return counts
 [9]: # Plot class distribution
      def simple_bar_plot(dist, title):
          plt.bar(dist.keys(), dist.values(), color='cornflowerblue')
          plt.title(title)
          plt.ylabel('Percentage (%)')
          plt.xticks(rotation=45)
          plt.ylim(0, 100)
          plt.tight_layout()
          plt.show()
[10]: class_names = train_ds.class_names
      # Get class distributions
      train_dist = count_distribution(train_ds, class_names)
      val_dist = count_distribution(val_ds, class_names)
      test_dist = count_distribution(test_ds, class_names)
      overall_dist = {}
      for k in class_names:
          overall_dist[k] = round((train_dist[k] + val_dist[k]) / 2, 2)
      print(train_dist)
      print(val_dist)
      print(test_dist)
      print(overall_dist)
     {'cardboard': 16.52, 'glass': 19.73, 'metal': 15.92, 'paper': 23.29, 'plastic':
     19.44, 'trash': 5.09}
     {'cardboard': 13.66, 'glass': 20.2, 'metal': 17.43, 'paper': 24.36, 'plastic':
     17.62, 'trash': 6.73}
     {'cardboard': 14.06, 'glass': 17.58, 'metal': 16.8, 'paper': 25.0, 'plastic':
     19.14, 'trash': 7.42}
     {'cardboard': 15.09, 'glass': 19.96, 'metal': 16.68, 'paper': 23.82, 'plastic':
```

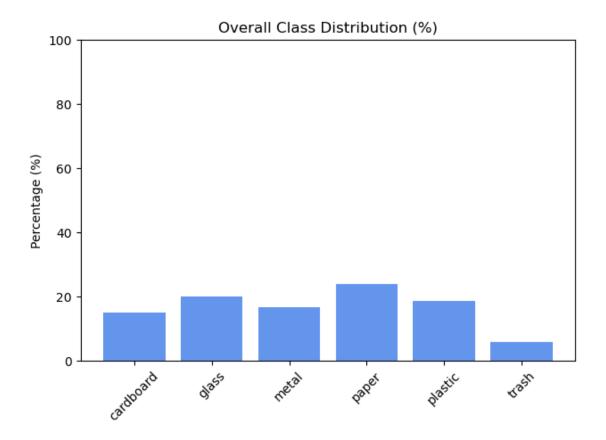
18.53, 'trash': 5.91}

```
[11]: # Show visualizations
simple_bar_plot(train_dist, "Training Set Class Distribution (%)")
simple_bar_plot(val_dist, "Validation Set Class Distribution (%)")
simple_bar_plot(test_dist, "Test Set Class Distribution (%)")
simple_bar_plot(overall_dist, "Overall Class Distribution (%)")
```









2 Inference on Class Imbalance

The "Garbage Image Dataset" reveals a noticeable imbalance in the distribution of its image categories:

Category	Image Count	Updated Distribution
Cardboard	403	15.09
Glass	501	19.96
Metal	410	16.68
Paper	594	23.82
Plastic	482	18.53
Trash	137	5.91

$2.0.1\;$ Key Problems Caused by Class Imbalance:

1 Bias

• The model may overpredict common classes like "paper" and underpredict rare ones like "trash".

2 Generalization Issues

• If the real-world distribution is more balanced, the model may **fail to generalize** and **misclassify rare classes**.

3 Accuracy Deception

layers.RandomZoom(0.1),
layers.RandomContrast(0.1),

])

• The model might appear to have **high overall accuracy** just by **predicting the majority class**, while **failing** on underrepresented ones.

```
[12]: # Count class occurrences and prepare label list
      class_counts = {i: 0 for i in range(len(class_names))}
      all_labels = []
      for images, labels in train_ds:
          for label in labels.numpy():
              class_counts[label] += 1
              all_labels.append(label)
      # Compute class weights (index aligned)
      class_weights_array = compute_class_weight(
          class_weight='balanced',
          classes=np.arange(len(class_names)),
          y=all_labels
      )
      # Create dictionary mapping class index to weight
      class_weights = {i: w for i, w in enumerate(class_weights_array)}
[13]: # Optional: print results
      print("Class Counts:", class_counts)
      print("Class Weights:", class_weights)
     Class Counts: {0: 334, 1: 399, 2: 322, 3: 471, 4: 393, 5: 103}
     Class Weights: {0: 1.0089820359281436, 1: 0.8446115288220551, 2:
     1.046583850931677, 3: 0.7154989384288747, 4: 0.8575063613231552, 5:
     3.2718446601941746}
     2.1 Data Preprocessing
[14]: data_augmentation = Sequential([
          layers.RandomFlip("horizontal"),
          layers.RandomRotation(0.1),
```

```
[16]: base model = EfficientNetV2B2(include top=False, input_shape=(124, 124, __
       →3),include_preprocessing=True, weights='imagenet')
      # Freeze early layers (to retain general pretrained features)
      base_model.trainable = True
      for layer in base_model.layers[:100]: # You can adjust this number
          layer.trainable = False
[17]: # Building the final model
      model = Sequential([
          layers.Input(shape=(124, 124, 3)),
          data_augmentation,
          base_model,
          GlobalAveragePooling2D(),
          layers.Dropout(0.3),
          layers.Dense(6, activation='softmax') # Change to your number of classes
      ])
[19]: # Compiling the model
      model.compile(
          optimizer=optimizers.Adam(learning rate=1e-4),
          loss='sparse_categorical_crossentropy',
          metrics=['accuracy']
      )
[20]: early = tf.keras.callbacks.EarlyStopping(
          monitor='val_loss',
                                         # Metric to monitor (validation loss here)
          patience=3,
                                        # Number of epochs to wait after last_
       ⇔improvement before stopping
          restore best weights=True
                                       # After stopping, restore the model weights.
       → from the epoch with the best val_loss
      )
     Train the model using .fit() with appropriate epochs, batch_size, and callbacks like EarlyStopping.
[22]: epochs = 15 # Number of times the model will go through the entire dataset
```

```
# Implements early stopping to prevent unnecessary
          callbacks=[early]
       \hookrightarrow training
      )
     Epoch 1/15
     64/64
                       42s 658ms/step -
     accuracy: 0.9553 - loss: 0.1309 - val_accuracy: 0.8931 - val_loss: 0.2921
     Epoch 2/15
     64/64
                       40s 631ms/step -
     accuracy: 0.9640 - loss: 0.1169 - val_accuracy: 0.8891 - val_loss: 0.3070
     Epoch 3/15
     64/64
                       40s 628ms/step -
     accuracy: 0.9726 - loss: 0.0885 - val accuracy: 0.8891 - val loss: 0.2985
     Epoch 4/15
     64/64
                       40s 630ms/step -
     accuracy: 0.9747 - loss: 0.0853 - val_accuracy: 0.9089 - val_loss: 0.2650
     Epoch 5/15
     64/64
                       41s 633ms/step -
     accuracy: 0.9785 - loss: 0.0714 - val accuracy: 0.9228 - val loss: 0.2434
     Epoch 6/15
     64/64
                       41s 639ms/step -
     accuracy: 0.9872 - loss: 0.0548 - val_accuracy: 0.9188 - val_loss: 0.2385
     Epoch 7/15
     64/64
                       41s 635ms/step -
     accuracy: 0.9798 - loss: 0.0637 - val_accuracy: 0.9030 - val_loss: 0.2512
     Epoch 8/15
     64/64
                       41s 639ms/step -
     accuracy: 0.9835 - loss: 0.0567 - val_accuracy: 0.9168 - val_loss: 0.2507
     Epoch 9/15
     64/64
                       41s 632ms/step -
     accuracy: 0.9850 - loss: 0.0484 - val_accuracy: 0.9149 - val_loss: 0.2586
[23]: model.summary()
     Model: "sequential_1"
      Layer (type)
                                              Output Shape
                                                                                   Ш
      →Param #
      sequential (Sequential)
                                             (None, 124, 124, 3)
                                                                                       H
      → 0
      efficientnetv2-b2 (Functional) (None, 4, 4, 1408)
      48,769,374
```

Number of samples processed in each training step

batch_size=32,

global_average_pooling2d (None, 1408)

O (GlobalAveragePooling2D)

dropout (Dropout) (None, 1408)

dense (Dense) (None, 6)

-8,454

Total params: 24,727,114 (94.33 MB)

Trainable params: 7,974,642 (30.42 MB)

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

Optimizer params: 15,949,286 (60.84 MB)

[24]: base_model.summary() # Print the architecture summary of the base model

Model: "efficientnetv2-b2"

Layer (type) Connected to	Output Shape	Param #	Ш
<pre>input_layer_1 (InputLayer) </pre>	(None, 124, 124, 3)	0	- ⊔
rescaling_1 (Rescaling) input_layer_1[0][0]	(None, 124, 124, 3)	0	Ш
normalization_1 →rescaling_1[0][0] (Normalization)	(None, 124, 124, 3)	0	u u
stem_conv (Conv2D) onormalization_1[0][0]	(None, 62, 62, 32)	864	ш
<pre>stem_bn (BatchNormalization) stem_conv[0][0]</pre>	(None, 62, 62, 32)	128	Ш

<pre>stem_activation (Activation) stem_bn[0][0]</pre>	(None,	62,	62,	32)	О ц	
block1a_project_conv (Conv2D) stem_activation[0][0]	(None,	62,	62,	16)	4,608 _L	
block1a_project_bn block1a_project_conv[0][0] (BatchNormalization)	(None,	62,	62,	16)	64 ц	Ш
block1a_project_activation ⇒block1a_project_bn[0][0] (Activation)	(None,	62,	62,	16)	О ц	Ш
block1b_project_conv (Conv2D) ⇔block1a_project_activatio	(None,	62,	62,	16)	2,304 ப	
block1b_project_bn block1b_project_conv[0][0] (BatchNormalization)	(None,	62,	62,	16)	64 ц	ш
block1b_project_activation ⇒block1b_project_bn[0][0] (Activation)	(None,	62,	62,	16)	0	Ш
block1b_drop (Dropout) block1b_project_activatio	(None,	62,	62,	16)	О ц	
block1b_add (Add) oblock1b_drop[0][0],	(None,	62,	62,	16)	О ц	
⇔block1a_project_activatio					Ц	
block2a_expand_conv (Conv2D) shlock1b_add[0][0]	(None,	31,	31,	64)	9,216 ப	
block2a_expand_bn ⇔block2a_expand_conv[0][0] (BatchNormalization)	(None,	31,	31,	64)	256 ப	Ш

block2a_expand_activation block2a_expand_bn[0][0] (Activation)	(None, 31, 31, 64)	0
block2a_project_conv (Conv2D) →block2a_expand_activation	(None, 31, 31, 32)	2,048 🔟
block2a_project_bn ⇒block2a_project_conv[0][0] (BatchNormalization) ↔	(None, 31, 31, 32)	128 ப
block2b_expand_conv (Conv2D) block2a_project_bn[0][0]	(None, 31, 31, 128)	36,864 ⊔
block2b_expand_bn ⇒block2b_expand_conv[0][0] (BatchNormalization)	(None, 31, 31, 128)	512 ப
block2b_expand_activation block2b_expand_bn[0][0] (Activation)	(None, 31, 31, 128)	0
block2b_project_conv (Conv2D) block2b_expand_activation	(None, 31, 31, 32)	4,096 _⊔
block2b_project_bn block2b_project_conv[0][0] (BatchNormalization)	(None, 31, 31, 32)	128 ப
block2b_drop (Dropout) block2b_project_bn[0][0]	(None, 31, 31, 32)	О ц
block2b_add (Add) ⇔block2b_drop[0][0],	(None, 31, 31, 32)	О ц
⇒block2a_project_bn[0][0]		П
block2c_expand_conv (Conv2D) block2b_add[0][0]	(None, 31, 31, 128)	36,864 ⊔

block2c_expand_bn ⇔block2c_expand_conv[0][0] (BatchNormalization)	(None,	31,	31,	128)	512	Ш	Ш
block2c_expand_activation ⇒block2c_expand_bn[0][0] (Activation)	(None,	31,	31,	128)	0	Ш	Ш
block2c_project_conv (Conv2D) block2c_expand_activation	(None,	31,	31,	32)	4,096	Ш	
block2c_project_bn ⇔block2c_project_conv[0][0] (BatchNormalization)	(None,	31,	31,	32)	128	Ш	Ш
block2c_drop (Dropout) block2c_project_bn[0][0]	(None,	31,	31,	32)	0	Ш	
block2c_add (Add) oblock2c_drop[0][0],	(None,	31,	31,	32)	0	Ш	
⇒block2b_add[0][0]						Ш	
block3a_expand_conv (Conv2D) block2c_add[0][0]	(None,	16,	16,	128)	36,864	Ш	
block3a_expand_bn ⇔block3a_expand_conv[0][0] (BatchNormalization)	(None,	16,	16,	128)	512	Ш	Ш
block3a_expand_activation ⇔block3a_expand_bn[0][0] (Activation)	(None,	16,	16,	128)	0	Ш	Ш
block3a_project_conv (Conv2D) ⇔block3a_expand_activation	(None,	16,	16,	56)	7,168	Ш	
block3a_project_bn block3a_project_conv[0][0] (BatchNormalization)	(None,	16,	16,	56)	224	Ш	Ш

block3b_expand_conv (Conv2D) block3a_project_bn[0][0]	(None,	16,	16,	224)	112,896	ш	
block3b_expand_bn ⇔block3b_expand_conv[0][0] (BatchNormalization)	(None,	16,	16,	224)	896	Ш	Ш
block3b_expand_activation block3b_expand_bn[0][0] (Activation)	(None,	16,	16,	224)	0	Ш	Ш
block3b_project_conv (Conv2D) block3b_expand_activation	(None,	16,	16,	56)	12,544	ш	
block3b_project_bn ⇔block3b_project_conv[0][0] (BatchNormalization)	(None,	16,	16,	56)	224	Ш	Ш
block3b_drop (Dropout) block3b_project_bn[0][0]	(None,	16,	16,	56)	0	ш	
block3b_add (Add) ⇔block3b_drop[0][0],	(None,	16,	16,	56)	0	ш	
⇔block3a_project_bn[0][0]						П	
block3c_expand_conv (Conv2D) block3b_add[0][0]	(None,	16,	16,	224)	112,896	ш	
block3c_expand_bn ⇔block3c_expand_conv[0][0] (BatchNormalization)	(None,	16,	16,	224)	896	Ш	ш
block3c_expand_activation block3c_expand_bn[0][0] (Activation)	(None,	16,	16,	224)	0	Ш	ш
block3c_project_conv (Conv2D) block3c_expand_activation	(None,	16,	16,	56)	12,544	Ш	

block3c_project_bn block3c_project_conv[0][0] (BatchNormalization)	(None, 16, 16, 56)	224 ப
block3c_drop (Dropout) ⇔block3c_project_bn[0][0]	(None, 16, 16, 56)	О ц
block3c_add (Add) shlock3c_drop[0][0],	(None, 16, 16, 56)	О ц
⊖block3b_add[0][0]		ш
block4a_expand_conv (Conv2D) ⇒block3c_add[0][0]	(None, 16, 16, 224)	12,544 _⊔
block4a_expand_bn ⇔block4a_expand_conv[0][0] (BatchNormalization)	(None, 16, 16, 224)	896 ப
block4a_expand_activation ⇔block4a_expand_bn[0][0] (Activation)	(None, 16, 16, 224)	О ц
block4a_dwconv2 ⇔block4a_expand_activation (DepthwiseConv2D) ↔	(None, 8, 8, 224)	2,016 ப
block4a_bn ⇔block4a_dwconv2[0][0] (BatchNormalization)	(None, 8, 8, 224)	896 ц
block4a_activation ⇔block4a_bn[0][0] (Activation)	(None, 8, 8, 224)	О ц
block4a_se_squeeze ⇒block4a_activation[0][0] (GlobalAveragePooling2D)	(None, 224)	О п

block4a_se_reshape (Reshape) shlock4a_se_squeeze[0][0]	(None,	1,	1,	224)	0	Ш	
block4a_se_reduce (Conv2D) oblock4a_se_reshape[0][0]	(None,	1,	1,	14)	3,150	Ш	
block4a_se_expand (Conv2D) ⇒block4a_se_reduce[0][0]	(None,	1,	1,	224)	3,360	Ш	
block4a_se_excite (Multiply)	(None,	8,	8,	224)	0	Ш	
⇒block4a_se_expand[0][0]						Ш	
block4a_project_conv (Conv2D) →block4a_se_excite[0][0]	(None,	8,	8,	104)	23,296	Ш	
block4a_project_bn ⇒block4a_project_conv[0][0] (BatchNormalization) ↔	(None,	8,	8,	104)	416	Ш	Ш
block4b_expand_conv (Conv2D) shlock4a_project_bn[0][0]	(None,	8,	8,	416)	43,264	Ш	
block4b_expand_bn ⇔block4b_expand_conv[0][0] (BatchNormalization) ↔	(None,	8,	8,	416)	1,664	Ш	Ш
block4b_expand_activation ⇔block4b_expand_bn[0][0] (Activation) ↔	(None,	8,	8,	416)	0	Ш	Ш
block4b_dwconv2 ⇒block4b_expand_activation (DepthwiseConv2D) ↔	(None,	8,	8,	416)	3,744	Ш	Ш
block4b_bn ⇒block4b_dwconv2[0][0] (BatchNormalization)	(None,	8,	8,	416)	1,664	ш	Ш

block4b_activation ⇒block4b_bn[0][0] (Activation) ↔	(None, 8, 8, 416)	0	Ш	Ш
block4b_se_squeeze block4b_activation[0][0] (GlobalAveragePooling2D)	(None, 416)	0	ш	Ш
block4b_se_reshape (Reshape) ⇒block4b_se_squeeze[0][0]	(None, 1, 1, 416)	0	Ш	
block4b_se_reduce (Conv2D) block4b_se_reshape[0][0]	(None, 1, 1, 26)	10,842	Ш	
block4b_se_expand (Conv2D) ⇒block4b_se_reduce[0][0]	(None, 1, 1, 416)	11,232	Ш	
block4b_se_excite (Multiply) block4b_activation[0][0],	(None, 8, 8, 416)	0	Ш	
⇒block4b_se_expand[0][0]			П	
block4b_project_conv (Conv2D) block4b_se_excite[0][0]	(None, 8, 8, 104)	43,264	Ш	
block4b_project_bn ⇔block4b_project_conv[0][0] (BatchNormalization) ↔	(None, 8, 8, 104)	416	Ш	Ш
block4b_drop (Dropout) ⇔block4b_project_bn[0][0]	(None, 8, 8, 104)	0	Ш	
block4b_add (Add) ⇔block4b_drop[0][0],	(None, 8, 8, 104)	0	Ш	
⇔block4a_project_bn[0][0]			П	
block4c_expand_conv (Conv2D) shlock4b_add[0][0]	(None, 8, 8, 416)	43,264	Ш	
block4c_expand_bn	(None, 8, 8, 416)	1,664	Ш	

```
(BatchNormalization)
                                                                            0 🔟
block4c_expand_activation
                                 (None, 8, 8, 416)
⇒block4c_expand_bn[0][0]
(Activation)
                                                                                  Ш
block4c_dwconv2
                                 (None, 8, 8, 416)
                                                                        3,744 🔲
⇔block4c_expand_activation...
(DepthwiseConv2D)
                                                                                  Ш
                                 (None, 8, 8, 416)
block4c_bn
                                                                        1,664
⇔block4c_dwconv2[0][0]
(BatchNormalization)
                                                                                  Ш
                                                                            0 🔟
block4c_activation
                                 (None, 8, 8, 416)
\hookrightarrowblock4c_bn[0][0]
(Activation)
                                                                                  Ш
                                 (None, 416)
block4c_se_squeeze
                                                                            0 🔟
⇔block4c_activation[0][0]
(GlobalAveragePooling2D)
                                                                                  Ш
block4c_se_reshape (Reshape)
                                 (None, 1, 1, 416)
                                                                            0 🔟
⇔block4c_se_squeeze[0][0]
block4c_se_reduce (Conv2D)
                                 (None, 1, 1, 26)
                                                                       10,842 🔲
⇔block4c_se_reshape[0][0]
block4c_se_expand (Conv2D)
                                 (None, 1, 1, 416)
                                                                       11,232
⇔block4c_se_reduce[0][0]
block4c_se_excite (Multiply)
                                 (None, 8, 8, 416)
                                                                            0 ц
⇔block4c_activation[0][0],
                                                                               Ш
⇔block4c_se_expand[0][0]
block4c_project_conv (Conv2D)
                                 (None, 8, 8, 104)
                                                                       43,264 \square
⇔block4c_se_excite[0][0]
```

block4c_project_bn ⇔block4c_project_conv[0][0] (BatchNormalization) ↔	(None, 8, 8, 104)	416 ப
block4c_drop (Dropout) block4c_project_bn[0][0]	(None, 8, 8, 104)	О ц
block4c_add (Add) shlock4c_drop[0][0],	(None, 8, 8, 104)	О ц
⇒block4b_add[0][0]		Ц
block4d_expand_conv (Conv2D) block4c_add[0][0]	(None, 8, 8, 416)	43,264 ⊔
block4d_expand_bn ⇔block4d_expand_conv[0][0] (BatchNormalization) ↔	(None, 8, 8, 416)	1,664 ப
block4d_expand_activation ⇔block4d_expand_bn[0][0] (Activation)	(None, 8, 8, 416)	О ц
block4d_dwconv2 ⇔block4d_expand_activation (DepthwiseConv2D) ↔	(None, 8, 8, 416)	3,744 ப
block4d_bn ⇔block4d_dwconv2[0][0] (BatchNormalization)	(None, 8, 8, 416)	1,664 ப
block4d_activation ⇔block4d_bn[0][0] (Activation)	(None, 8, 8, 416)	О ц
block4d_se_squeeze ⇔block4d_activation[0][0] (GlobalAveragePooling2D)	(None, 416)	О ц

block4d_se_reshape (Reshape) block4d_se_squeeze[0][0]	(None, 1, 1, 416)	0	Ш	
block4d_se_reduce (Conv2D) block4d_se_reshape[0][0]	(None, 1, 1, 26)	10,842	Ш	
block4d_se_expand (Conv2D) block4d_se_reduce[0][0]	(None, 1, 1, 416)	11,232	Ш	
block4d_se_excite (Multiply) shlock4d_activation[0][0],	(None, 8, 8, 416)	0	ш	
⇒block4d_se_expand[0][0]			П	
block4d_project_conv (Conv2D) block4d_se_excite[0][0]	(None, 8, 8, 104)	43,264	Ш	
block4d_project_bn ⇔block4d_project_conv[0][0] (BatchNormalization)	(None, 8, 8, 104)	416	Ш	Ш
block4d_drop (Dropout)	(None, 8, 8, 104)	0	Ш	
oblock4d_project_bn[0][0]	(1010, 0, 0, 101)	· ·		
block4d_add (Add) ⇔block4d_drop[0][0],	(None, 8, 8, 104)	0	Ш	
⇔block4c_add[0][0]			Ш	
block5a_expand_conv (Conv2D) block4d_add[0][0]	(None, 8, 8, 624)	64,896	ш	
block5a_expand_bn ⇒block5a_expand_conv[0][0] (BatchNormalization)	(None, 8, 8, 624)	2,496	Ш	
÷				_
block5a_expand_activation block5a_expand_bn[0][0] (Activation)	(None, 8, 8, 624)	0	Ш	
\(\)				J
block5a_dwconv2	(None, 8, 8, 624)	5,616	Ш	

```
(DepthwiseConv2D)
block5a_bn
                                (None, 8, 8, 624)
                                                                       2,496
⇔block5a_dwconv2[0][0]
(BatchNormalization)
                                                                                 Ш
                                                                           0 🔟
block5a_activation
                                (None, 8, 8, 624)
\rightarrowblock5a_bn[0][0]
(Activation)
                                                                                 Ш
block5a se squeeze
                                (None, 624)
                                                                           0 🔟
⇔block5a_activation[0][0]
(GlobalAveragePooling2D)
                                                                                 Ш
block5a_se_reshape (Reshape)
                                (None, 1, 1, 624)
                                                                           0 🔟
⇔block5a_se_squeeze[0][0]
block5a_se_reduce (Conv2D)
                                (None, 1, 1, 26)
                                                                      16,250 🔲
⇔block5a_se_reshape[0][0]
block5a se expand (Conv2D)
                                (None, 1, 1, 624)
                                                                      16,848
⇔block5a_se_reduce[0][0]
                                                                           0 🔟
block5a_se_excite (Multiply)
                                (None, 8, 8, 624)
⇔block5a_activation[0][0],
⇔block5a_se_expand[0][0]
block5a_project_conv (Conv2D)
                                (None, 8, 8, 120)
                                                                      74,880
⇔block5a_se_excite[0][0]
block5a_project_bn
                                (None, 8, 8, 120)
                                                                         480 🔲
⇔block5a_project_conv[0][0]
(BatchNormalization)
                                                                                 Ш
block5b_expand_conv (Conv2D)
                                (None, 8, 8, 720)
                                                                      86,400
⇔block5a_project_bn[0][0]
block5b_expand_bn
                                (None, 8, 8, 720)
                                                                       2,880
→block5b_expand_conv[0][0]
```

```
(BatchNormalization)
                                                                             0 🔟
block5b_expand_activation
                                 (None, 8, 8, 720)
⇔block5b_expand_bn[0][0]
(Activation)
                                                                                   Ш
block5b_dwconv2
                                 (None, 8, 8, 720)
                                                                         6,480 <sub>⊔</sub>
⇒block5b_expand_activation...
(DepthwiseConv2D)
                                                                                   Ш
block5b_bn
                                 (None, 8, 8, 720)
                                                                         2,880 11
⇔block5b_dwconv2[0][0]
(BatchNormalization)
                                                                                   Ш
block5b_activation
                                 (None, 8, 8, 720)
                                                                             0 🔟
\hookrightarrowblock5b_bn[0][0]
(Activation)
                                                                                   Ш
                                 (None, 720)
                                                                             0 🔟
block5b_se_squeeze
⇔block5b_activation[0][0]
(GlobalAveragePooling2D)
                                                                                   Ш
block5b_se_reshape (Reshape)
                                 (None, 1, 1, 720)
                                                                             0 🔟
→block5b_se_squeeze[0][0]
block5b_se_reduce (Conv2D)
                                 (None, 1, 1, 30)
                                                                        21,630 🔲
⇔block5b_se_reshape[0][0]
block5b_se_expand (Conv2D)
                                 (None, 1, 1, 720)
                                                                        22,320
⇔block5b_se_reduce[0][0]
block5b_se_excite (Multiply)
                                 (None, 8, 8, 720)
                                                                             0 🔟
⇔block5b_activation[0][0],
                                                                                Ш
⇔block5b_se_expand[0][0]
block5b_project_conv (Conv2D)
                                 (None, 8, 8, 120)
                                                                        86,400 🔲
⇔block5b_se_excite[0][0]
```

block5b_project_bn ⇔block5b_project_conv[0][0] (BatchNormalization)	(None, 8, 8, 120)	480 ப
block5b_drop (Dropout) sblock5b_project_bn[0][0]	(None, 8, 8, 120)	О ц
block5b_add (Add) sblock5b_drop[0][0],	(None, 8, 8, 120)	О ц
⇒block5a_project_bn[0][0]		ш
block5c_expand_conv (Conv2D) block5b_add[0][0]	(None, 8, 8, 720)	86,400 _⊔
block5c_expand_bn ⇔block5c_expand_conv[0][0] (BatchNormalization) ↔	(None, 8, 8, 720)	2,880 ப
block5c_expand_activation ⇔block5c_expand_bn[0][0] (Activation)	(None, 8, 8, 720)	0 ப
block5c_dwconv2 ⇔block5c_expand_activation (DepthwiseConv2D) ↔	(None, 8, 8, 720)	6,480 ப
block5c_bn ⇒block5c_dwconv2[0][0] (BatchNormalization)	(None, 8, 8, 720)	2,880 ப
block5c_activation ⇔block5c_bn[0][0] (Activation)	(None, 8, 8, 720)	0 ப
block5c_se_squeeze ⇔block5c_activation[0][0] (GlobalAveragePooling2D)	(None, 720)	0

block5c_se_reshape (Reshape) block5c_se_squeeze[0][0]	(None, 1, 1, 720)	0	П	
block5c_se_reduce (Conv2D) block5c_se_reshape[0][0]	(None, 1, 1, 30)	21,630	Ш	
block5c_se_expand (Conv2D) block5c_se_reduce[0][0]	(None, 1, 1, 720)	22,320	Ш	
block5c_se_excite (Multiply) shlock5c_activation[0][0],	(None, 8, 8, 720)	0	ш	
⇒block5c_se_expand[0][0]			П	
block5c_project_conv (Conv2D) block5c_se_excite[0][0]	(None, 8, 8, 120)	86,400	Ш	
block5c_project_bn block5c_project_conv[0][0] (BatchNormalization)	(None, 8, 8, 120)	480	u L	J
\(\)				
block5c_drop (Dropout) block5c_project_bn[0][0]	(None, 8, 8, 120)	0	Ш	
block5c_add (Add)	(None, 8, 8, 120)	0	ш	
⇔block5b_add[0][0]			П	
block5d_expand_conv (Conv2D) block5c_add[0][0]	(None, 8, 8, 720)	86,400	ш	
block5d_expand_bn block5d_expand_conv[0][0] (BatchNormalization)	(None, 8, 8, 720)	2,880		J
4				
block5d_expand_activation block5d_expand_bn[0][0] (Activation)	(None, 8, 8, 720)	0		J
· · · · · · · · · · · · · · · · · · ·				
block5d_dwconv2	(None, 8, 8, 720)	6,480	Ш	

```
(DepthwiseConv2D)
block5d_bn
                                 (None, 8, 8, 720)
                                                                        2,880 🔲
⇔block5d_dwconv2[0][0]
(BatchNormalization)
                                                                                  Ш
                                                                            0 🔟
block5d_activation
                                 (None, 8, 8, 720)
\hookrightarrowblock5d_bn[0][0]
(Activation)
                                                                                  Ш
                                                                            0 🔟
block5d se squeeze
                                 (None, 720)
⇔block5d_activation[0][0]
(GlobalAveragePooling2D)
                                                                                  Ш
block5d_se_reshape (Reshape)
                                 (None, 1, 1, 720)
                                                                            0 🔟
⇔block5d_se_squeeze[0][0]
block5d_se_reduce (Conv2D)
                                 (None, 1, 1, 30)
                                                                       21,630 🗓
⇔block5d_se_reshape[0][0]
block5d se expand (Conv2D)
                                 (None, 1, 1, 720)
                                                                       22,320 🔲
⇔block5d_se_reduce[0][0]
                                                                            0 🔟
block5d_se_excite (Multiply)
                                 (None, 8, 8, 720)
⇔block5d_activation[0][0],
⇔block5d_se_expand[0][0]
block5d_project_conv (Conv2D)
                                 (None, 8, 8, 120)
                                                                       86,400 🔟
⇔block5d_se_excite[0][0]
block5d_project_bn
                                 (None, 8, 8, 120)
                                                                          480 🔲
⇔block5d_project_conv[0][0]
(BatchNormalization)
                                                                                  Ш
block5d_drop (Dropout)
                                                                            0 🔟
                                 (None, 8, 8, 120)
⇔block5d_project_bn[0][0]
block5d_add (Add)
                                                                            0 ц
                                 (None, 8, 8, 120)
⇔block5d_drop[0][0],
```

⇔block5c_add[0][0]		П
block5e_expand_conv (Conv2D) block5d_add[0][0]	(None, 8, 8, 720)	86,400 ⊔
block5e_expand_bn ⇔block5e_expand_conv[0][0] (BatchNormalization)	(None, 8, 8, 720)	2,880 ப
block5e_expand_activation block5e_expand_bn[0][0] (Activation)	(None, 8, 8, 720)	0 ப
block5e_dwconv2 block5e_expand_activation (DepthwiseConv2D)	(None, 8, 8, 720)	6,480 ப
block5e_bn block5e_dwconv2[0][0] (BatchNormalization)	(None, 8, 8, 720)	2,880 ப
block5e_activation block5e_bn[0][0] (Activation)	(None, 8, 8, 720)	О п
block5e_se_squeeze ⇒block5e_activation[0][0] (GlobalAveragePooling2D) →	(None, 720)	0 ப
block5e_se_reshape (Reshape) block5e_se_squeeze[0][0]	(None, 1, 1, 720)	О ц
block5e_se_reduce (Conv2D) block5e_se_reshape[0][0]	(None, 1, 1, 30)	21,630 ц
block5e_se_expand (Conv2D) ⇔block5e_se_reduce[0][0]	(None, 1, 1, 720)	22,320 ப

block5e_se_excite (Multiply) block5e_activation[0][0],	(None,	8,	8,	720)	0	Ш	
<pre>oblock5e_se_expand[0][0]</pre>						П	
block5e_project_conv (Conv2D) block5e_se_excite[0][0]	(None,	8,	8,	120)	86,400	Ш	
block5e_project_bn block5e_project_conv[0][0] (BatchNormalization)	(None,	8,	8,	120)	480	Ш	Ш
block5e_drop (Dropout) block5e_project_bn[0][0]	(None,	8,	8,	120)	0	Ш	
block5e_add (Add) block5e_drop[0][0],	(None,	8,	8,	120)	0	Ш	
⊖block5d_add[0][0]						П	
block5f_expand_conv (Conv2D) block5e_add[0][0]	(None,	8,	8,	720)	86,400	Ш	
block5f_expand_bn ⇔block5f_expand_conv[0][0] (BatchNormalization)	(None,	8,	8,	720)	2,880	Ш	ш
block5f_expand_activation ⇔block5f_expand_bn[0][0] (Activation)	(None,	8,	8,	720)	0	Ш	Ш
block5f_dwconv2 ⇒block5f_expand_activation (DepthwiseConv2D)	(None,	8,	8,	720)	6,480	Ш	Ш
block5f_bn ⇔block5f_dwconv2[0][0] (BatchNormalization)	(None,	8,	8,	720)	2,880	Ш	Ш
block5f_activation shlock5f_bn[0][0]	(None,	8,	8,	720)	0	Ш	

```
(Activation)
                                 (None, 720)
                                                                            0 🔟
block5f_se_squeeze
⇔block5f_activation[0][0]
(GlobalAveragePooling2D)
                                                                                  Ш
                                                                            0 🔟
block5f_se_reshape (Reshape)
                                 (None, 1, 1, 720)
⇔block5f_se_squeeze[0][0]
block5f_se_reduce (Conv2D)
                                 (None, 1, 1, 30)
                                                                       21,630 🔲
⇔block5f_se_reshape[0][0]
block5f_se_expand (Conv2D)
                                 (None, 1, 1, 720)
                                                                       22,320 🔲
⇔block5f_se_reduce[0][0]
block5f_se_excite (Multiply)
                                 (None, 8, 8, 720)
                                                                            0 🔟
⇔block5f_activation[0][0],
                                                                               Ш
⇔block5f_se_expand[0][0]
block5f_project_conv (Conv2D)
                                 (None, 8, 8, 120)
                                                                       86,400 🗆
⇔block5f_se_excite[0][0]
block5f_project_bn
                                 (None, 8, 8, 120)
                                                                          480 🔲
→block5f_project_conv[0][0]
(BatchNormalization)
                                                                                  Ш
\hookrightarrow
                                                                            0 🔟
block5f_drop (Dropout)
                                 (None, 8, 8, 120)
⇔block5f_project_bn[0][0]
block5f_add (Add)
                                 (None, 8, 8, 120)
                                                                            0 🔟
⇔block5f_drop[0][0],
                                                                               Ш
→block5e_add[0][0]
                                                                       86,400 🔲
block6a_expand_conv (Conv2D)
                                 (None, 8, 8, 720)
⇔block5f_add[0][0]
block6a_expand_bn
                                 (None, 8, 8, 720)
                                                                        2,880 11
→block6a_expand_conv[0][0]
(BatchNormalization)
                                                                                  Ш
```

block6a_expand_activation ⇔block6a_expand_bn[0][0] (Activation)	(None, 8, 8, 720)	0 ц	Ш
block6a_dwconv2 ⇔block6a_expand_activation (DepthwiseConv2D)	(None, 4, 4, 720)	6,480 _⊔	Ш
block6a_bn ⇔block6a_dwconv2[0][0] (BatchNormalization)	(None, 4, 4, 720)	2,880 ப	Ш
block6a_activation ⇒block6a_bn[0][0] (Activation)	(None, 4, 4, 720)	О ц	Ш
block6a_se_squeeze ⇔block6a_activation[0][0] (GlobalAveragePooling2D)	(None, 720)	0 ц	Ш
block6a_se_reshape (Reshape) ⇒block6a_se_squeeze[0][0]	(None, 1, 1, 720)	0 ц	
block6a_se_reduce (Conv2D) block6a_se_reshape[0][0]	(None, 1, 1, 30)	21,630 ц	
block6a_se_expand (Conv2D) block6a_se_reduce[0][0]	(None, 1, 1, 720)	22,320 ц	
block6a_se_excite (Multiply) oblock6a_activation[0][0],	(None, 4, 4, 720)	0 ц	
⇒block6a_se_expand[0][0]		ш	
block6a_project_conv (Conv2D) block6a_se_excite[0][0]	(None, 4, 4, 208)	149,760 ப	
block6a_project_bn block6a_project_conv[0][0] (BatchNormalization)	(None, 4, 4, 208)	832 ц	Ш

block6b_expand_conv (Conv2D) block6a_project_bn[0][0]	(None, 4, 4, 1248)	259,584 ப
block6b_expand_bn ⇔block6b_expand_conv[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992 ப
block6b_expand_activation block6b_expand_bn[0][0] (Activation)	(None, 4, 4, 1248)	0 ப
block6b_dwconv2 ⇔block6b_expand_activation (DepthwiseConv2D) ↔	(None, 4, 4, 1248)	11,232 ப
block6b_bn ⇔block6b_dwconv2[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992 ں
block6b_activation block6b_bn[0][0] (Activation)	(None, 4, 4, 1248)	0 ப
block6b_se_squeeze block6b_activation[0][0] (GlobalAveragePooling2D)	(None, 1248)	О п
block6b_se_reshape (Reshape) block6b_se_squeeze[0][0]	(None, 1, 1, 1248)	0 ц
block6b_se_reduce (Conv2D) block6b_se_reshape[0][0]	(None, 1, 1, 52)	64,948 ц
block6b_se_expand (Conv2D) block6b_se_reduce[0][0]	(None, 1, 1, 1248)	66,144 ц
block6b_se_excite (Multiply) block6b_activation[0][0],	(None, 4, 4, 1248)	О ц

<pre>→block6b_se_expand[0][0]</pre>		ı	П	
block6b_project_conv (Conv2D) →block6b_se_excite[0][0]	(None, 4, 4, 208)	259,584	Ц	
<pre>block6b_project_bn</pre>	(None, 4, 4, 208)	832		
block6b_drop (Dropout) →block6b_project_bn[0][0]	(None, 4, 4, 208)	0	Ц	
block6b_add (Add) block6b_drop[0][0],	(None, 4, 4, 208)	0	Ш	
<pre>⇔block6a_project_bn[0][0]</pre>		l	П	
block6c_expand_conv (Conv2D)	(None, 4, 4, 1248)	259,584	Ш	
block6c_expand_bn ⇒block6c_expand_conv[0][0] (BatchNormalization) ↔	(None, 4, 4, 1248)	4,992		٦
block6c_expand_activation ⇒block6c_expand_bn[0][0] (Activation)	(None, 4, 4, 1248)	0		
block6c_dwconv2 →block6c_expand_activation (DepthwiseConv2D) →	(None, 4, 4, 1248)	11,232		_
block6c_bn ⇒block6c_dwconv2[0][0] (BatchNormalization) ↔	(None, 4, 4, 1248)	4,992		٦
block6c_activation ⇒block6c_bn[0][0] (Activation)	(None, 4, 4, 1248)	0	LI L	_

block6c_se_squeeze ⇒block6c_activation[0][0] (GlobalAveragePooling2D) →	(None, 1248)) ப	Ш
block6c_se_reshape (Reshape) block6c_se_squeeze[0][0]	(None, 1, 1, 12	248)) п	
block6c_se_reduce (Conv2D) block6c_se_reshape[0][0]	(None, 1, 1, 52	2) 64,948	3 ⊔	
block6c_se_expand (Conv2D) block6c_se_reduce[0][0]	(None, 1, 1, 12	248) 66,144	l u	
block6c_se_excite (Multiply) block6c_activation[0][0],	(None, 4, 4, 12	248) (
⇒block6c_se_expand[0][0]			П	
block6c_project_conv (Conv2D) block6c_se_excite[0][0]	(None, 4, 4, 20	08) 259,584	l u	
<pre>block6c_project_bn</pre>	(None, 4, 4, 20	08) 832	2 ப	Ш
block6c_drop (Dropout) block6c_project_bn[0][0]	(None, 4, 4, 20	08)) п	
block6c_add (Add)	(None, 4, 4, 20	08)) п	
⇒block6b_add[0][0]			П	
block6d_expand_conv (Conv2D) block6c_add[0][0]	(None, 4, 4, 12	248) 259,584	l u	
block6d_expand_bn ⇒block6d_expand_conv[0][0] (BatchNormalization) →	(None, 4, 4, 12	248) 4,992	2 ப	Ш
block6d_expand_activation block6d_expand_bn[0][0]	(None, 4, 4, 12	248)) п	

```
(Activation)
block6d_dwconv2
                                 (None, 4, 4, 1248)
                                                                       11,232
→block6d_expand_activation...
(DepthwiseConv2D)
                                                                                 Ш
block6d_bn
                                 (None, 4, 4, 1248)
                                                                       4,992 🔲
⇔block6d_dwconv2[0][0]
(BatchNormalization)
                                                                                 Ш
block6d_activation
                                 (None, 4, 4, 1248)
                                                                           0 🔟
\hookrightarrowblock6d_bn[0][0]
(Activation)
                                                                                 Ш
block6d_se_squeeze
                                 (None, 1248)
                                                                            0 🔟
⇔block6d_activation[0][0]
(GlobalAveragePooling2D)
                                                                                 Ш
block6d_se_reshape (Reshape)
                                 (None, 1, 1, 1248)
                                                                            0 🔟
⇔block6d_se_squeeze[0][0]
block6d_se_reduce (Conv2D)
                                 (None, 1, 1, 52)
                                                                      64,948
→block6d_se_reshape[0][0]
block6d_se_expand (Conv2D)
                                 (None, 1, 1, 1248)
                                                                      66,144
⇔block6d_se_reduce[0][0]
block6d_se_excite (Multiply)
                                 (None, 4, 4, 1248)
                                                                            0 🔟
⇔block6d_activation[0][0],
                                                                              Ш
⇔block6d_se_expand[0][0]
block6d_project_conv (Conv2D)
                                 (None, 4, 4, 208)
                                                                     259,584
⇔block6d_se_excite[0][0]
                                 (None, 4, 4, 208)
                                                                          832 🔲
block6d_project_bn
⇒block6d_project_conv[0][0]
(BatchNormalization)
                                                                                 Ш
```

block6d_drop (Dropout) shlock6d_project_bn[0][0]	(None,	4,	4,	208)	0	П	
block6d_add (Add)	(None,	4,	4,	208)	0	Ш	
→block6c_add[0][0]							
block6e_expand_conv (Conv2D) block6d_add[0][0]	(None,	4,	4,	1248)	259,584	Ш	
block6e_expand_bn ⇒block6e_expand_conv[0][0] (BatchNormalization) →	(None,	4,	4,	1248)	4,992	Ш	Ш
block6e_expand_activation ⇒block6e_expand_bn[0][0] (Activation) ↔	(None,	4,	4,	1248)	0	Ш	Ш
block6e_dwconv2 ⇔block6e_expand_activation (DepthwiseConv2D) ↔	(None,	4,	4,	1248)	11,232	Ш	ш
block6e_bn ⇒block6e_dwconv2[0][0] (BatchNormalization) ↔	(None,	4,	4,	1248)	4,992	Ш	Ш
block6e_activation ⇒block6e_bn[0][0] (Activation)	(None,	4,	4,	1248)	0	Ш	Ш
block6e_se_squeeze ⇒block6e_activation[0][0] (GlobalAveragePooling2D) →	(None,	124	48)		0	Ш	Ш
block6e_se_reshape (Reshape) \$\text{oblock6e_se_squeeze[0][0]}\$	(None,	1,	1,	1248)	0	Ш	
block6e_se_reduce (Conv2D) oblock6e_se_reshape[0][0]	(None,	1,	1,	52)	64,948	П	

block6e_se_expand (Conv2D) block6e_se_reduce[0][0]	(None, 1, 1, 1248)	66,144 _⊔
block6e_se_excite (Multiply) block6e_activation[0][0],	(None, 4, 4, 1248)	О ц
⇒block6e_se_expand[0][0]		П
block6e_project_conv (Conv2D)	(None, 4, 4, 208)	259,584 ⊔
block6e_project_bn block6e_project_conv[0][0] (BatchNormalization)	(None, 4, 4, 208)	832 ப
block6e_drop (Dropout) block6e_project_bn[0][0]	(None, 4, 4, 208)	О ц
block6e_add (Add)	(None, 4, 4, 208)	О ц
⇒block6d_add[0][0]		П
block6f_expand_conv (Conv2D) block6e_add[0][0]	(None, 4, 4, 1248)	259,584 ⊔
block6f_expand_bn ⇒block6f_expand_conv[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992 ப
block6f_expand_activation ⇒block6f_expand_bn[0][0] (Activation)	(None, 4, 4, 1248)	0 ப
block6f_dwconv2 ⇒block6f_expand_activation (DepthwiseConv2D) →	(None, 4, 4, 1248)	11,232 ப
block6f_bn	(None, 4, 4, 1248)	4,992 ⊔

```
(BatchNormalization)
                                                                          0 🔟
block6f_activation
                                (None, 4, 4, 1248)
⇔block6f_bn[0][0]
(Activation)
                                                                                Ш
                                                                          0 🔟
block6f_se_squeeze
                                (None, 1248)
⇔block6f_activation[0][0]
(GlobalAveragePooling2D)
                                                                                Ш
block6f_se_reshape (Reshape)
                                (None, 1, 1, 1248)
                                                                           0 🔟
⇔block6f_se_squeeze[0][0]
block6f_se_reduce (Conv2D)
                                (None, 1, 1, 52)
                                                                     64,948 🔲
⇔block6f_se_reshape[0][0]
block6f_se_expand (Conv2D)
                                (None, 1, 1, 1248)
                                                                     66,144
→block6f_se_reduce[0][0]
block6f_se_excite (Multiply)
                                (None, 4, 4, 1248)
                                                                           0 🔟
⇔block6f_activation[0][0],
                                                                             Ш
→block6f_se_expand[0][0]
block6f_project_conv (Conv2D)
                                (None, 4, 4, 208)
                                                                    259,584 🔲
⇔block6f_se_excite[0][0]
                                                                        832 🔟
block6f_project_bn
                                (None, 4, 4, 208)
⇒block6f_project_conv[0][0]
(BatchNormalization)
                                                                                Ш
block6f_drop (Dropout)
                                (None, 4, 4, 208)
                                                                           0 🔟
⇔block6f_project_bn[0][0]
                                                                           0 🔟
block6f_add (Add)
                                (None, 4, 4, 208)
⇔block6f_drop[0][0],
→block6e_add[0][0]
block6g_expand_conv (Conv2D)
                                (None, 4, 4, 1248)
                                                                    259,584
⇔block6f_add[0][0]
```

block6g_expand_bn ⇔block6g_expand_conv[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992	u u
block6g_expand_activation block6g_expand_bn[0][0] (Activation)	(None, 4, 4, 1248)	0	u u
block6g_dwconv2 ⇔block6g_expand_activation (DepthwiseConv2D)	(None, 4, 4, 1248)	11,232	u u
block6g_bn block6g_dwconv2[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992	u u
block6g_activation block6g_bn[0][0] (Activation)	(None, 4, 4, 1248)	0	u u
block6g_se_squeeze ⇒block6g_activation[0][0] (GlobalAveragePooling2D) →	(None, 1248)	0	u u
block6g_se_reshape (Reshape) block6g_se_squeeze[0][0]	(None, 1, 1, 1248)	0	Ц
block6g_se_reduce (Conv2D) block6g_se_reshape[0][0]	(None, 1, 1, 52)	64,948	П
block6g_se_expand (Conv2D) block6g_se_reduce[0][0]	(None, 1, 1, 1248)	66,144	П
block6g_se_excite (Multiply) block6g_activation[0][0],	(None, 4, 4, 1248)	0	ш
⇔block6g_se_expand[0][0]			П
block6g_project_conv (Conv2D) block6g_se_excite[0][0]	(None, 4, 4, 208)	259,584	П

block6g_project_bn ⇔block6g_project_conv[0][0] (BatchNormalization)	(None, 4, 4, 208)	832 ப
block6g_drop (Dropout) block6g_project_bn[0][0]	(None, 4, 4, 208)	О ц
block6g_add (Add) hlock6g_drop[0][0],	(None, 4, 4, 208)	О ц
⇒block6f_add[0][0]		Ц
block6h_expand_conv (Conv2D) block6g_add[0][0]	(None, 4, 4, 1248)	259,584 ப
block6h_expand_bn block6h_expand_conv[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992 ப
block6h_expand_activation ⇔block6h_expand_bn[0][0] (Activation)	(None, 4, 4, 1248)	0 ப
block6h_dwconv2 ⇔block6h_expand_activation (DepthwiseConv2D) ↔	(None, 4, 4, 1248)	11,232 ப
block6h_bn ⇔block6h_dwconv2[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992 ں
block6h_activation ⇔block6h_bn[0][0] (Activation)	(None, 4, 4, 1248)	0
block6h_se_squeeze block6h_activation[0][0] (GlobalAveragePooling2D)	(None, 1248)	0 ப

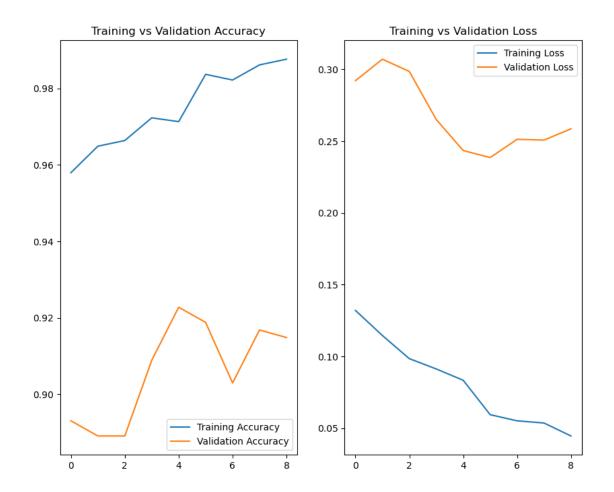
block6h_se_reshape (Reshape) →block6h_se_squeeze[0][0]	(None, 1, 1, 1248)	0 ∟	I
block6h_se_reduce (Conv2D) →block6h_se_reshape[0][0]	(None, 1, 1, 52)	64,948 _	I
block6h_se_expand (Conv2D) block6h_se_reduce[0][0]	(None, 1, 1, 1248)	66,144 _	I
block6h_se_excite (Multiply) block6h_activation[0][0],	(None, 4, 4, 1248)	0 L	
<pre>⇔block6h_se_expand[0][0]</pre>		L	I
block6h_project_conv (Conv2D) block6h_se_excite[0][0]	(None, 4, 4, 208)	259,584 _	I
block6h_project_bn ⇒block6h_project_conv[0][0] (BatchNormalization)	(None, 4, 4, 208)	832 ∟	ı
· · · · · · · · · · · · · · · · · · ·	(7		
block6h_drop (Dropout)	(None, 4, 4, 208)	О _	I
block6h_add (Add)	(None, 4, 4, 208)	0 _	
<pre>⇔block6g_add[0][0]</pre>		L	I
block6i_expand_conv (Conv2D) oblock6h_add[0][0]	(None, 4, 4, 1248)	259,584 _	I
block6i_expand_bn ⇒block6i_expand_conv[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992 ∟	ı
↔			
block6i_expand_activation ⇒block6i_expand_bn[0][0] (Activation)	(None, 4, 4, 1248)	О L	ı
block6i_dwconv2	(None, 4, 4, 1248)	11,232	
⇒block6i_expand_activation	(10110)	11,202	I

```
(DepthwiseConv2D)
block6i_bn
                                (None, 4, 4, 1248)
                                                                      4,992 🔲
⇔block6i_dwconv2[0][0]
(BatchNormalization)
                                                                                Ш
                                                                          0 🔟
block6i_activation
                                (None, 4, 4, 1248)
→block6i_bn[0][0]
(Activation)
                                                                                Ш
block6i_se_squeeze
                                (None, 1248)
                                                                          0 🔟
⇔block6i_activation[0][0]
(GlobalAveragePooling2D)
                                                                                Ш
block6i_se_reshape (Reshape)
                                (None, 1, 1, 1248)
                                                                          0 🔟
⇔block6i_se_squeeze[0][0]
block6i_se_reduce (Conv2D)
                                (None, 1, 1, 52)
                                                                     64,948
→block6i_se_reshape[0][0]
block6i se expand (Conv2D)
                                (None, 1, 1, 1248)
                                                                     66,144
⇔block6i_se_reduce[0][0]
                                                                          0 🔟
block6i_se_excite (Multiply)
                                (None, 4, 4, 1248)
⇔block6i_activation[0][0],
⇔block6i_se_expand[0][0]
block6i_project_conv (Conv2D)
                                (None, 4, 4, 208)
                                                                    259,584
⇔block6i_se_excite[0][0]
block6i_project_bn
                                (None, 4, 4, 208)
                                                                        832 🔲
⇒block6i_project_conv[0][0]
(BatchNormalization)
                                                                                Ш
block6i_drop (Dropout)
                                                                          0 🔟
                                (None, 4, 4, 208)
⇔block6i_project_bn[0][0]
                                (None, 4, 4, 208)
                                                                           0 ц
block6i_add (Add)
→block6i_drop[0][0],
```

⇒block6h_add[0][0]			П
block6j_expand_conv (Conv2D) block6i_add[0][0]	(None, 4, 4, 1248)	259,584	ш
block6j_expand_bn ⇔block6j_expand_conv[0][0] (BatchNormalization) ↔	(None, 4, 4, 1248)	4,992	u
block6j_expand_activation block6j_expand_bn[0][0] (Activation)	(None, 4, 4, 1248)	0	u
block6j_dwconv2 ⇔block6j_expand_activation (DepthwiseConv2D)	(None, 4, 4, 1248)	11,232	u u
block6j_bn ⇔block6j_dwconv2[0][0] (BatchNormalization)	(None, 4, 4, 1248)	4,992	u u
block6j_activation ⇔block6j_bn[0][0] (Activation)	(None, 4, 4, 1248)	0	u u
block6j_se_squeeze ⇔block6j_activation[0][0] (GlobalAveragePooling2D) ↔	(None, 1248)	0	u u
block6j_se_reshape (Reshape) block6j_se_squeeze[0][0]	(None, 1, 1, 1248)	0	Ш
block6j_se_reduce (Conv2D) block6j_se_reshape[0][0]	(None, 1, 1, 52)	64,948	ш
block6j_se_expand (Conv2D) block6j_se_reduce[0][0]	(None, 1, 1, 1248)	66,144	ш

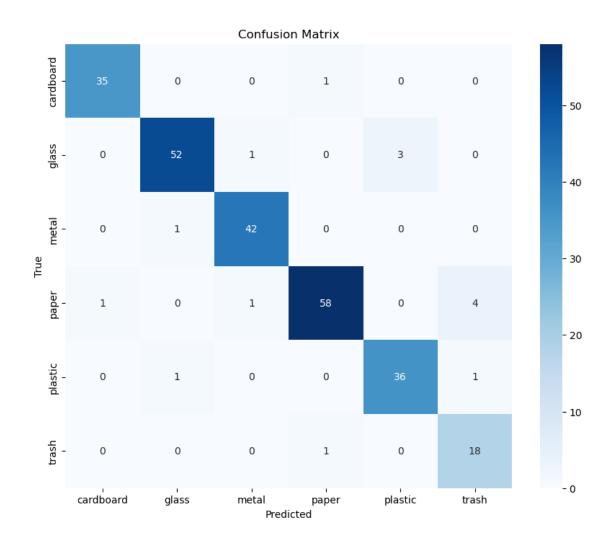
```
0 🔟
       block6j_se_excite (Multiply)
                                        (None, 4, 4, 1248)
       ⇔block6j_activation[0][0],
                                                                                      Ш
       ⇔block6j_se_expand[0][0]
       block6j_project_conv (Conv2D)
                                        (None, 4, 4, 208)
                                                                             259,584
       ⇔block6j_se_excite[0][0]
       block6j_project_bn
                                        (None, 4, 4, 208)
                                                                                  832 🔟
       ⇔block6j_project_conv[0][0]
       (BatchNormalization)
                                                                                          Ш
       block6j drop (Dropout)
                                        (None, 4, 4, 208)
                                                                                    0 🔟
       ⇔block6j_project_bn[0][0]
       block6j_add (Add)
                                        (None, 4, 4, 208)
                                                                                    0 🔟
       ⇔block6j_drop[0][0],
       →block6i_add[0][0]
       top_conv (Conv2D)
                                        (None, 4, 4, 1408)
                                                                             292,864
       ⇔block6j_add[0][0]
       top_bn (BatchNormalization)
                                        (None, 4, 4, 1408)
                                                                                5,632 11
       →top_conv[0][0]
                                                                                    0 🔟
       top_activation (Activation)
                                        (None, 4, 4, 1408)
       \rightarrowtop_bn[0][0]
      Total params: 8,769,374 (33.45 MB)
      Trainable params: 7,966,188 (30.39 MB)
      Non-trainable params: 803,186 (3.06 MB)
[25]: acc = history.history['accuracy']
                                                 # Extract training accuracy from
       \hookrightarrowhistory
      val_acc = history.history['val_accuracy'] # Extract validation accuracy from_
      \hookrightarrowhistory
      loss = history.history['loss']
                                                   # Extract training loss from history
      val_loss = history.history['val_loss']
                                                   # Extract validation loss from
       \hookrightarrowhistory
```

```
epochs_range = range(len(acc))
                                           # Define range for epochs based on_
 ⇔accuracy length
plt.figure(figsize=(10,8))
                                           # Set overall figure size for
 \hookrightarrow visualization
                                            # Create first subplot (1 row, 2_{\square}
plt.subplot(1,2,1)
 ⇔columns, position 1)
plt.plot(epochs_range, acc, label='Training Accuracy') # Plot training_
 \rightarrowaccuracy
plt.plot(epochs_range, val_acc, label='Validation Accuracy') # Plot validation_
 \hookrightarrowaccuracy
plt.legend(loc='lower right')
                                            # Place legend in lower-right corner
plt.title('Training vs Validation Accuracy') # Add title for accuracy plot
                                            # Create second subplot (1 row, 2
plt.subplot(1,2,2)
 ⇔columns, position 2)
plt.plot(epochs_range, loss, label='Training Loss')  # Plot training loss
plt.plot(epochs_range, val_loss, label='Validation Loss') # Plot validation_
 ⇔loss
plt.legend(loc='upper right')
                                            # Place legend in upper-right corner
plt.title('Training vs Validation Loss') # Add title for loss plot
plt.show()
                                            # Display the plots
```



2.2 Model Evaluation

```
[ 0 52 1 0 3 0]
 [ 0 1 42 0 0 0]
 [ 1 0 1 58 0 4]
 [ 0 1 0 0 36 1]
 [0001018]]
             precision
                         recall f1-score
                                           support
          0
                  0.97
                           0.97
                                     0.97
                                                36
                           0.93
          1
                  0.96
                                     0.95
                                                56
          2
                           0.98
                  0.95
                                     0.97
                                                43
          3
                  0.97
                           0.91
                                     0.94
                                                64
          4
                  0.92
                           0.95
                                     0.94
                                                38
          5
                  0.78
                           0.95
                                     0.86
                                                19
                                     0.94
                                               256
   accuracy
                                     0.94
                                               256
  macro avg
                  0.93
                           0.95
weighted avg
                  0.94
                           0.94
                                     0.94
                                               256
```



2.3 7. Final Testing and Save the Model

```
[31]: class_names = train_ds.class_names

for images, labels in test_ds_eval.take(1):

    predictions = model.predict(images)

    pred_labels = tf.argmax(predictions, axis=1)

    for i in range(8):
        plt.imshow(images[i].numpy().astype("uint8"))
```

1/1 4s 4s/step

True: glass, Pred: glass



True: trash, Pred: trash



True: glass, Pred: glass





True: plastic, Pred: plastic





True: glass, Pred: glass

True: metal, Pred: metal



2.4 Save the trained model using model.save() or save_model() for future inference.

```
[32]: model.save('Effiicientnetv2b2.keras')

model = tf.keras.models.load_model('Effiicientnetv2b2.keras')
```

2.4.1 Week II Submission

[]: