```
In [ ]: import matplotlib.pyplot as plt
         import numpy as np
         import os
         import tensorflow as tf
         import keras_tuner as kt
         /usr/local/lib/python3.8/dist-packages/scipy/__init__.py:146: UserWarning: A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy
         (detected version 1.23.0
           warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"</pre>
In [ ]: PATH = "Dataset"
         MODEL = "ResNet50"
In [ ]: train_dir = os.path.join(PATH, 'train')
validation_dir = os.path.join(PATH, 'val')
In [ ]: BATCH SIZE = 32
         IMG_SIZE = (224, 224)
         train_dataset = tf.keras.utils.image_dataset_from_directory(train_dir,
                                                                            shuffle=True
                                                                            batch_size=BATCH_SIZE,
                                                                            image_size=IMG_SIZE)
         Found 7560 files belonging to 4 classes.
In [ ]: validation_dataset = tf.keras.utils.image_dataset_from_directory(validation_dir,
                                                                                 batch_size=BATCH_SIZE,
                                                                                 image_size=IMG_SIZE)
         Found 3240 files belonging to 4 classes.
In [ ]: class names = train dataset.class names
         plt.figure(figsize=(10, 10))
         \begin{tabular}{ll} for images, labels in train\_dataset.take(1): \\ \end{tabular}
           for i in range(9):
             ax = plt.subplot(3, 3, i + 1)
              plt.imshow(images[i].numpy().astype("uint8"))
              plt.title(class_names[labels[i]])
              plt.axis("off")
                                               intersection
                                                                              intersection
                  overpass
                 intersection
                                                crosswalk
                                                                              roundabout
In [ ]: val_batches = tf.data.experimental.cardinality(validation_dataset)
         test dataset = validation dataset.take(val batches // 3)
         validation_dataset = validation_dataset.skip(val_batches // 3)
In [ ]: print('Number of validation batches: %d' % tf.data.experimental.cardinality(validation_dataset))
print('Number of test batches: %d' % tf.data.experimental.cardinality(test_dataset))
         Number of validation batches: 68
         Number of test batches: 34
In [ ]: AUTOTUNE = tf.data.AUTOTUNE
         train_dataset = train_dataset.prefetch(buffer_size=AUTOTUNE)
         validation_dataset = validation_dataset.prefetch(buffer_size=AUTOTUNE)
         test_dataset = test_dataset.prefetch(buffer_size=AUTOTUNE)
In [ ]: data_augmentation = tf.keras.Sequential([
           tf.keras.layers.RandomFlip('horizontal'),
           tf.keras.layers.RandomRotation(0.2),
In [ ]: for image, _ in train_dataset.take(1):
    plt.figure(figsize=(10, 10))
```

```
ax = plt.subplot(3, 3, i + 1)
             augmented_image = data_augmentation(tf.expand_dims(first_image, 0))
            plt.imshow(augmented_image[0] / 255)
            plt.axis('off')
In [ ]: preprocess_input = tf.keras.applications.resnet50.preprocess_input
In [ ]: from tensorflow.keras.applications import ResNet50
        base_model = ResNet50(input_shape=(224,224,3),
                              include_top=False,
                              weights="imagenet")
        base_model.trainable=True
        # trainable_at = int(len(base_model.layers)/3)
        # # freeze 1/3 of the Layers
        # for layer in base_model.layers[:trainable_at]:
             layer.trainable=False
        Downloading \ data \ from \ https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50\_weights\_tf\_dim\_ordering\_tf\_kernels\_notop.h5
        94773248/94765736 [===========] - 5s @us/step
        94781440/94765736 [=========== ] - 5s @us/step
In [ ]: image_batch, label_batch = next(iter(train_dataset))
        feature_batch = base_model(image_batch)
        print(feature_batch.shape)
        (32, 7, 7, 2048)
In [ ]: global_average_layer = tf.keras.layers.GlobalAveragePooling2D()
         feature_batch_average = global_average_layer(feature_batch)
        print(feature_batch_average.shape)
        (32, 2048)
In [ ]: # Create ResNet50 model
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Rescaling, Conv2D, MaxPooling2D, BatchNormalization, Dropout, Dense, Flatten, InputLayer
        \textbf{from} \ \texttt{tensorflow}. \texttt{keras.applications} \ \textbf{import} \ \texttt{ResNet50}
        class ResNet50HyperModel(kt.HyperModel):
             # Create class for hyperparameter tuning
            def build(self, hp):
                base model = ResNet50(input shape=(224,224,3).
                                      include_top=False,
                                       weights="imagenet")
                base model.trainable=True
                inputs = tf.keras.Input(shape=(224, 224, 3))
                x = preprocess_input(inputs)
                x = base model(x, training=True)
                 x = global_average_layer(x)
                 x = Dense(256, activation="relu", kernel_initializer='he_uniform')(x)
                x = Dropout(0.2)(x)
                x = Dense(32, activation="relu", kernel_initializer='he_uniform')(x)
                 x = Dropout(0.2)(x)
                \verb"outputs = Dense(4, activation='softmax', kernel\_initializer='he\_uniform')(x)
                model = tf.keras.Model(inputs, outputs)
                 # Tune the learning rate for the optimizer
                 # Choose an optimal value from 0.01, 0.001, or 0.0001
```

hp\_learning\_rate = hp.Choice('learning\_rate', values=[1e-2, 1e-3, 1e-4])

first\_image = image[0]
for i in range(9):

```
# compile model
                model.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=hp_learning_rate),
                             loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=False),
                             metrics=['accuracy'])
            def fit(self, hp, model, *args, **kwargs):
    return model.fit(
                     batch_size=hp.Choice("batch_size", [32, 64]),
                     **kwargs
In [ ]: tuner = kt.GridSearch(
            ResNet50HyperModel(),
            objective="val_accuracy",
             seed=0.
            directory=MODEL,
            project_name=MODEL
In [ ]: # Logging to tensorboard
        from datetime import datetime
        log\_dir = f"logs/fit/\{MODEL\}\_" + datetime.now().strftime("%Y%m%d-%H%M%S")
        tensorboard\_callback = tf.keras.callbacks.TensorBoard(log\_dir=log\_dir, \ histogram\_freq=1)
In [ ]: tuner.search(train_dataset,
                      epochs=50,
                      validation data=validation dataset,
                      callbacks=[tensorboard_callback],
                      use_multiprocessing=True)
        Trial 6 Complete [00h 15m 18s]
        val_accuracy: 0.9865241646766663
        Best val_accuracy So Far: 0.9865241646766663
Total elapsed time: 01h 31m 29s
        INFO:tensorflow:Oracle triggered exit
In [ ]: best_model = tuner.get_best_models(1)[0]
        best_model.build(input_shape=(224,224,3))
        best_model.summary()
        Model: "model"
         Layer (type)
                                     Output Shape
                                                                 Param #
         input_2 (InputLayer)
                                    [(None, 224, 224, 3)]
         tf.__operators__.getitem (S (None, 224, 224, 3)
licingOpLambda)
         tf.nn.bias_add (TFOpLambda) (None, 224, 224, 3)
                                                                 23587712
         resnet50 (Functional)
                                     (None, 7, 7, 2048)
          global_average_pooling2d (G (None, 2048)
                                                                 0
         lobalAveragePooling2D)
         dense (Dense)
                                     (None, 256)
                                                                 524544
         dropout (Dropout)
                                     (None, 256)
         dense_1 (Dense)
                                      (None, 32)
                                                                 8224
         dropout_1 (Dropout)
                                      (None, 32)
         dense 2 (Dense)
                                      (None, 4)
                                                                 132
        Total params: 24,120,612
        Trainable params: 24,067,492
        Non-trainable params: 53,120
In [ ]: tuner.results_summary()
```

```
Results summary
         Results in ResNet50/ResNet50
         Showing 10 best trials
         Objective(name="val_accuracy", direction="max")
         Trial 0005 summary
         Hyperparameters:
         learning_rate: 0.0001
         batch size: 64
         Score: 0.9865241646766663
         Trial 0004 summary
         Hyperparameters:
         learning_rate: 0.0001
         batch_size: 32
         Score: 0.9851301312446594
         Trial 0002 summary
         Hyperparameters:
         learning_rate: 0.001
         batch_size: 32
         Score: 0.9609665274620056
         Trial 0003 summarv
         Hyperparameters:
         learning_rate: 0.001
         batch_size: 64
Score: 0.9595724940299988
         Trial 0001 summary
         Hyperparameters:
         learning rate: 0.01
         batch_size: 64
         Score: 0.32016727328300476
         Trial 0000 summary
         Hyperparameters:
         learning_rate: 0.01
         batch_size: 32
         Score: 0.25464683771133423
In [ ]: loss, accuracy = best_model.evaluate(test_dataset)
         print('Test accuracy :', accuracy)
         34/34 [===========] - 2s 26ms/step - loss: 0.0714 - accuracy: 0.9853
         Test accuracy : 0.9852941036224365
In [ ]: loss, accuracy = best_model.evaluate(test_dataset)
print('Test accuracy :', accuracy)
# Retrieve a batch of images from the test set
         image_batch, label_batch = test_dataset.as_numpy_iterator().next()
         predictions = best_model.predict_on_batch(image_batch)
         prediction_label = []
         for prediction in predictions:
             pred = max(prediction)
             prediction_label.append(prediction.tolist().index(pred))
         # # Apply a sigmoid since our model returns logits
# predictions = tf.nn.sigmoid(predictions)
# predictions = tf.where(predictions < 0.5, 0, 1)</pre>
         print('Predictions:\n', prediction_label)
print('Labels:\n', label_batch)
         plt.figure(figsize=(10, 10))
         for i in range(9):
           ax = plt.subplot(3, 3, i + 1)
           plt.imshow(image_batch[i].astype("uint8"))
           plt.title(class_names[prediction_label[i]])
plt.axis("off")
         34/34 [======
                                     =======] - 1s 25ms/step - loss: 0.0613 - accuracy: 0.9871
         Test accuracy : 0.9871323704719543
         Predictions:
          [2, 1, 3, 2, 1, 2, 2, 2, 1, 2, 1, 3, 0, 0, 0, 1, 2, 0, 1, 2, 0, 0, 2, 0, 3, 1, 1, 0, 1, 0, 0, 0]
         Labels:
          [2 1 3 2 1 2 2 2 1 2 1 3 0 0 0 1 2 0 1 2 0 0 2 0 3 1 1 0 1 0 0 0]
```









