Import module and datasets.

Train a regular NN model.

```
x_train, x_test = x_train / 255.0, x_test / 255.0
   model = models.Sequential([
     layers.Flatten(input_shape=(32, 32, 3)),
     layers.Dense(64, activation='relu'),
     layers.Dense(32, activation='relu'),
     layers.Dense(10, activation='softmax')
   model.compile(optimizer='adam',
         loss='sparse_categorical_crossentropy',
         metrics=['accuracy'])
   model.fit(x_train, y_train, epochs=5, validation_data=(x_test, y_test), batch_size=128)
   391/391 [=
         Epoch 2/5
         Epoch 3/5
   <keras.callbacks.History at 0x7f62ea84ebd0>
```

Add more convolutional layers and pooling layers, and get a summary of the model.

```
x_train, x_test = x_train / 255.0, x_test / 255.0
    model = models.Sequential([
        layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32, 32, 3)),
        layers.MaxPooling2D((2, 2)),
        layers.Conv2D(64, (3, 3), activation='relu'),
        layers.MaxPooling2D((2, 2)),
        layers.Conv2D(128, (3, 3), activation='relu'),
        layers.MaxPooling2D((2, 2)),
        layers.Flatten(),
        layers.Dense(64, activation='relu'),
        layers.Dense(32, activation='relu'),
        layers.Dense(10, activation='softmax')
    ])
    model.summary()
    model.summary()
F→ Model: "sequential_3"
    Layer (type)
                           Output Shape
                                               Param #
    _____
    conv2d_1 (Conv2D)
                          (None, 30, 30, 32)
    max_pooling2d_1 (MaxPooling (None, 15, 15, 32)
                                               18496
    conv2d_2 (Conv2D)
                           (None, 13, 13, 64)
    max_pooling2d_2 (MaxPooling (None, 6, 6, 64)
    conv2d_3 (Conv2D)
                           (None, 4, 4, 128)
                                               73856
    max_pooling2d_3 (MaxPooling (None, 2, 2, 128)
    flatten_3 (Flatten)
                           (None, 512)
    dense_9 (Dense)
                           (None, 64)
                                               32832
    dense_10 (Dense)
                           (None, 32)
                                                2080
    dense_11 (Dense)
                           (None, 10)
                                                330
    _____
```

Train the above model.

Run it longer and visualize the accuracy.

```
391/391 [==============] - 61s 157ms/step - loss: 0.8543 - accuracy: 0.7041 - val_loss: 0.9813 - val_accur ↑ ↓ ↔ 🖪 🕻
Epoch 3/10
  391/391 [==
           ============================= ] - 60s 153ms/step - loss: 0.8104 - accuracy: 0.7195 - val_loss: 0.9133 - val_accuracy: 0.6879
₽
  Epoch 4/10
                 ========] - 59s 152ms/step - loss: 0.7613 - accuracy: 0.7349 - val_loss: 0.9004 - val_accuracy: 0.6898
  391/391 [==
               391/391 [==
  Epoch 6/10
               =========] - 59s 151ms/step - loss: 0.6918 - accuracy: 0.7605 - val_loss: 0.8680 - val_accuracy: 0.7024
  Epoch 7/10
   391/391 [==
             ==========] - 60s 154ms/step - loss: 0.6541 - accuracy: 0.7729 - val_loss: 0.8675 - val_accuracy: 0.7072
  Epoch 8/10
   391/391 [==
              Epoch 9/10
                391/391 [==
  1.0
    0.8
    0.6
   Q 0.4
    0.2
                          val accuracy
```

Set more arguments in convolutional layers and pool layers.

```
k_train, x_test = x_train / 255.0, x_test / 255.0
   model = models.Sequential([
      layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32, 32, 3)),
      layers.MaxPooling2D((2, 2)),
      layers.Conv2D(32, (3, 3), padding='valid', dilation_rate=(1,1), groups=1, use_bias=True,
                 kernel_initializer='glorot_uniform', bias_initializer='zeros', input_shape=(32, 32, 3)),
      layers.MaxPooling2D(pool_size=(2, 2), strides=None, padding='valid'),
      layers.Conv2D(64, (3, 3), activation='relu'),
      layers.MaxPooling2D((2, 2)),
      layers.Flatten(),
      layers.Dense(64, activation='relu'),
      layers.Dense(32, activation='relu'),
      layers.Dense(10, activation='softmax')
   model.compile(optimizer='adam',
             loss='sparse_categorical_crossentropy',
             metrics=['accuracy'])
   model.fit(x train, y train, epochs=5, validation data=(x test, y test), batch size=128)
                kernel_initializer='glorot_uniform', bias_initializer='zeros'),
                                                                                    ↑ ↓ ⊖ ■
 0
       layers.MaxPooling2D(pool_size=(2, 2), strides=None, padding='valid'),
       layers.Conv2D(64, (3, 3), activation='relu'),
       layers.MaxPooling2D((2, 2)),
       layers.Flatten(),
       layers.Dense(64, activation='relu'),
       layers.Dense(32, activation='relu'),
       layers.Dense(10, activation='softmax')
    model.compile(optimizer='adam',
             loss='sparse_categorical_crossentropy',
             metrics=['accuracy'])
    model.fit(x_train, y_train, epochs=5, validation_data=(x_test, y_test), batch_size=128)
 Epoch 1/5
    391/391 [===
            Epoch 2/5
    Epoch 3/5
             Epoch 4/5
    Epoch 5/5
    <keras.callbacks.History at 0x7f3d0bdcebd0>
```

Use AveragePooling2D method.

```
x_train, x_test = x_train / 255.0, x_test / 255.0
    model = models.Sequential([
        layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32, 32, 3)),
        layers.AveragePooling2D((2, 2)),
        layers.Conv2D(32, (3, 3), padding='valid', dilation_rate=(1,1), groups=1, use_bias=True,
                      kernel_initializer='glorot_uniform', bias_initializer='zeros'),
        layers.AveragePooling2D(pool_size=(2, 2), strides=None, padding='valid'),
        layers.Conv2D(64, (3, 3), activation='relu'),
        layers.AveragePooling2D((2, 2)),
        layers.Flatten(),
        layers.Dense(64, activation='relu'),
        layers.Dense(32, activation='relu'),
        layers.Dense(10, activation='softmax')
    ])
    model.compile(optimizer='adam',
                  loss='sparse_categorical_crossentropy',
                  metrics=['accuracy'])
    model.fit(x_train, y_train, epochs=5, validation_data=(x_test, y_test), batch_size=128)
```