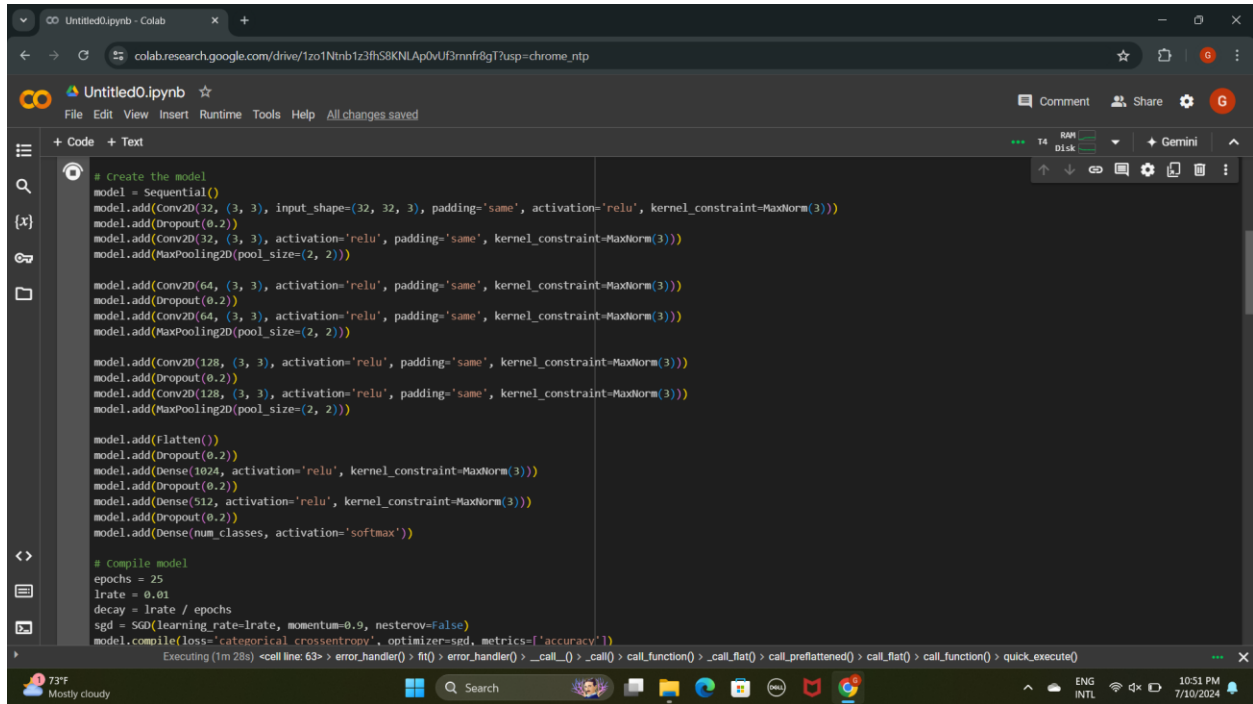


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Github Link: [GaneshKumarKorra/ICP4 \(github.com\)](https://github.com/GaneshKumarKorra/ICP4)



The screenshot shows a Google Colab notebook titled 'Untitled0.ipynb'. The code defines a sequential model with the following layers: Conv2D(32, 3, 3), Dropout(0.2), Conv2D(32, 3, 3), MaxPooling2D(2, 2), Conv2D(64, 3, 3), Dropout(0.2), Conv2D(64, 3, 3), MaxPooling2D(2, 2), Conv2D(128, 3, 3), Dropout(0.2), Conv2D(128, 3, 3), MaxPooling2D(2, 2), Flatten, Dropout(0.2), Dense(1024), Dropout(0.2), Dense(512), Dropout(0.2), and Dense(num_classes) with a softmax activation. The model is compiled with categorical crossentropy loss, SGD optimizer, and accuracy metric. The execution status at the bottom indicates it is running for 1m 28s.

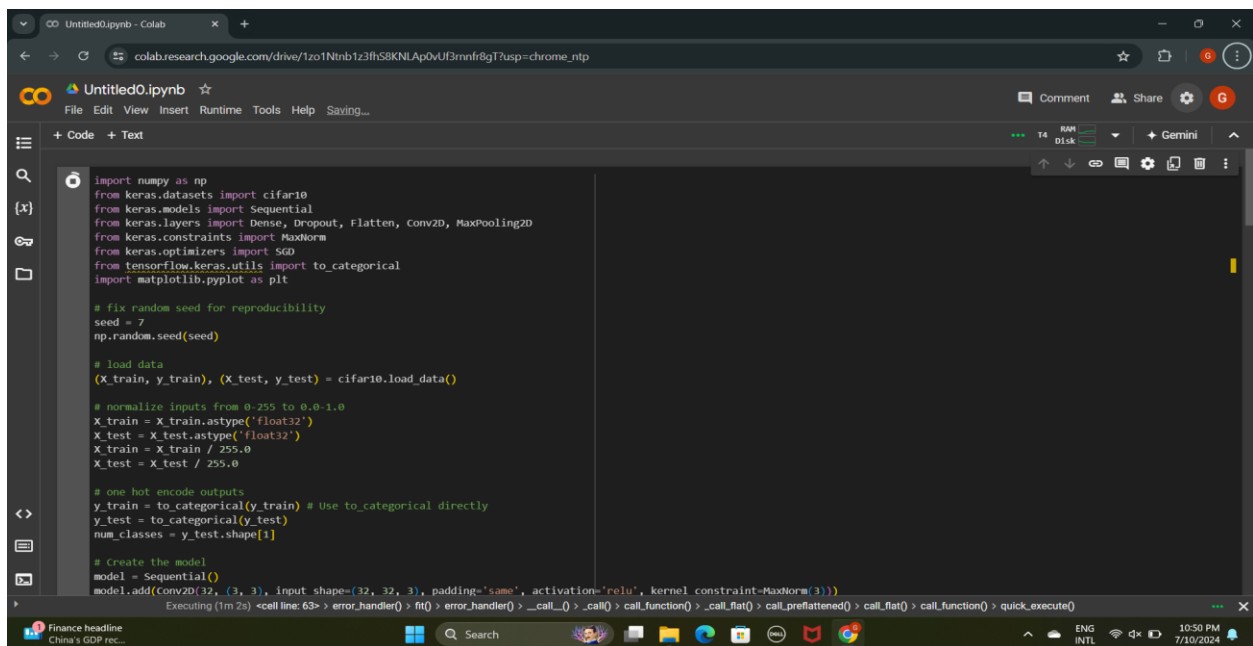
```
# Create the model
model = Sequential()
model.add(Conv2D(32, (3, 3), input_shape=(32, 32, 3), padding='same', activation='relu', kernel_constraint=MaxNorm(3)))
model.add(Dropout(0.2))
model.add(Conv2D(32, (3, 3), activation='relu', padding='same', kernel_constraint=MaxNorm(3)))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Conv2D(64, (3, 3), activation='relu', padding='same', kernel_constraint=MaxNorm(3)))
model.add(Dropout(0.2))
model.add(Conv2D(64, (3, 3), activation='relu', padding='same', kernel_constraint=MaxNorm(3)))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Conv2D(128, (3, 3), activation='relu', padding='same', kernel_constraint=MaxNorm(3)))
model.add(Dropout(0.2))
model.add(Conv2D(128, (3, 3), activation='relu', padding='same', kernel_constraint=MaxNorm(3)))
model.add(MaxPooling2D(pool_size=(2, 2)))

model.add(Flatten())
model.add(Dropout(0.2))
model.add(Dense(1024, activation='relu', kernel_constraint=MaxNorm(3)))
model.add(Dropout(0.2))
model.add(Dense(512, activation='relu', kernel_constraint=MaxNorm(3)))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))

# Compile model
epochs = 25
lr_rate = 0.01
decay = lr_rate / epochs
sgd = SGD(learning_rate=lr_rate, momentum=0.9, nesterov=False)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])
```



The screenshot shows a Google Colab notebook titled 'Untitled0.ipynb'. The code imports necessary libraries (numpy, keras, tensorflow, matplotlib) and sets a random seed for reproducibility. It then loads the CIFAR-10 dataset, normalizes the input data from 0-255 to 0.0-1.0, and converts the labels to one-hot encoded vectors. The execution status at the bottom indicates it is running for 1m 2s.

```
import numpy as np
from keras.datasets import cifar10
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Conv2D, MaxPooling2D
from keras.constraints import MaxNorm
from keras.optimizers import SGD
from tensorflow.keras.utils import to_categorical
import matplotlib.pyplot as plt

# fix random seed for reproducibility
seed = 7
np.random.seed(seed)

# load data
(X_train, y_train), (X_test, y_test) = cifar10.load_data()

# normalize inputs from 0-255 to 0.0-1.0
X_train = X_train.astype('float32')
X_test = X_test.astype('float32')
X_train = X_train / 255.0
X_test = X_test / 255.0

# one hot encode outputs
y_train = to_categorical(y_train) # Use to_categorical directly
y_test = to_categorical(y_test)
num_classes = y_test.shape[1]

# Create the model
model = Sequential()
model.add(Conv2D(32, (3, 3), input_shape=(32, 32, 3), padding='same', activation='relu', kernel_constraint=MaxNorm(3)))
```

```
CO Untitled0.ipynb - Colab
colab.research.google.com/drive/1zo1Ntnb1z3fhS8KNLAp0vU3mnfr8gT?usp=chrome_ntp

File Edit View Insert Runtime Tools Help All changes saved

+ Code + Text

# Compile model
epochs = 25
lr_rate = 0.01
decay = lr_rate / epochs
sgd = SGD(learning_rate=lr_rate, momentum=0.9, nesterov=False)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])

print(model.summary())

# Fit the model
history = model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=epochs, batch_size=32)

# Final evaluation of the model
scores = model.evaluate(X_test, y_test, verbose=0)
print("Accuracy: %.2f%%" % (scores[1] * 100))

# Predict the first 4 images of the test data
predictions = model.predict(X_test[:4])
predicted_classes = np.argmax(predictions, axis=1)
actual_classes = np.argmax(y_test[:4], axis=1)

for i in range(4):
    print(f"Image {i+1}: Predicted: {predicted_classes[i]}, Actual: {actual_classes[i]}")

# Visualize the first 4 test images
for i in range(4):
    plt.imshow(X_test[i])
    plt.title(f"Predicted: {predicted_classes[i]}, Actual: {actual_classes[i]}")
    plt.show()

# Visualize loss and Accuracy
plt.figure(figsize=(12, 4))

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

plt.subplot(1, 2, 2)
plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()

plt.show()
```

Executing (1m 57s) <cell line: 63> error_handler() > fit() > error_handler() > __call__() > _call() > call_function() > _call_flat() > call_prepared() > call_flat() > call_function() > quick_execute()

73°F Mostly cloudy

```
CO Untitled0.ipynb - Colab
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+ Code + Text

# Visualize loss and Accuracy
plt.figure(figsize=(12, 4))

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

# Plot accuracy
plt.subplot(1, 2, 2)
plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()

plt.show()
```

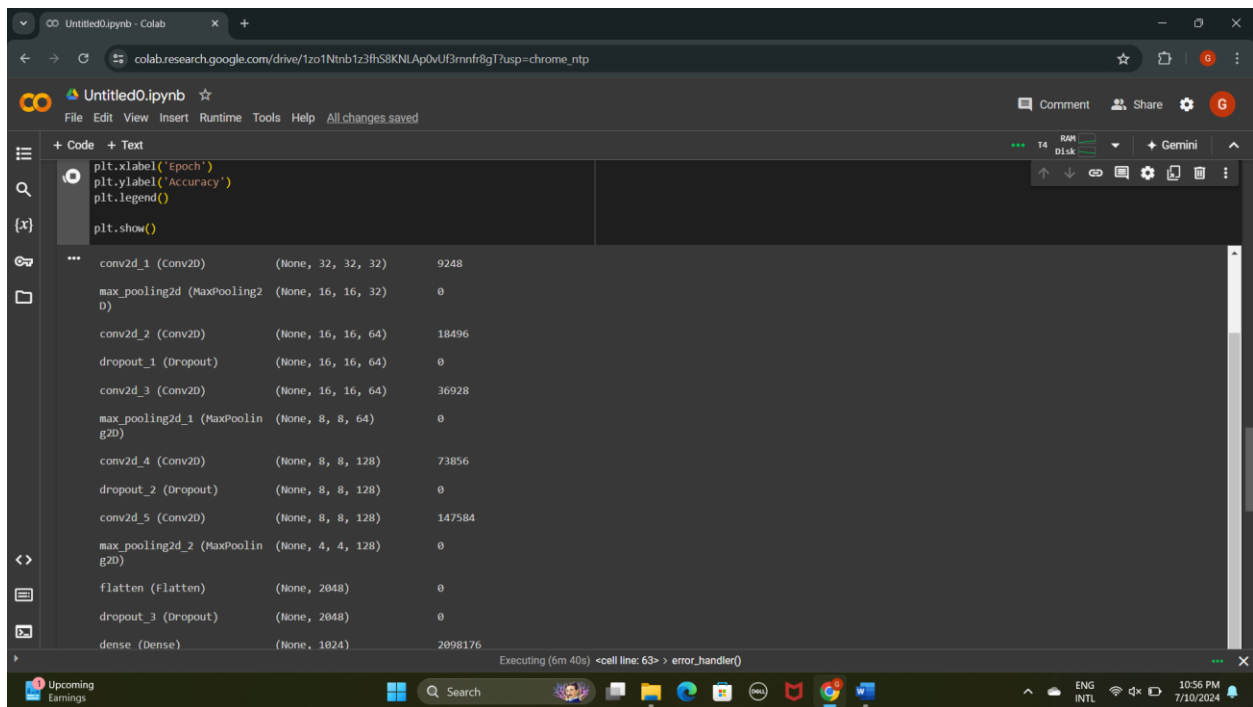
Downloading data from <https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz>
170498071/170498071 [=====] - 65.0us/step
Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 32, 32, 32)	896
dropout (Dropout)	(None, 32, 32, 32)	0

Executing (2m 39s) <cell line: 63> error_handler() > fit() > error_handler() > __call__() > _call() > call_function() > _call_flat() > call_prepared() > call_flat() > call_function() > quick_execute()

Rain coming 10:52 PM

output



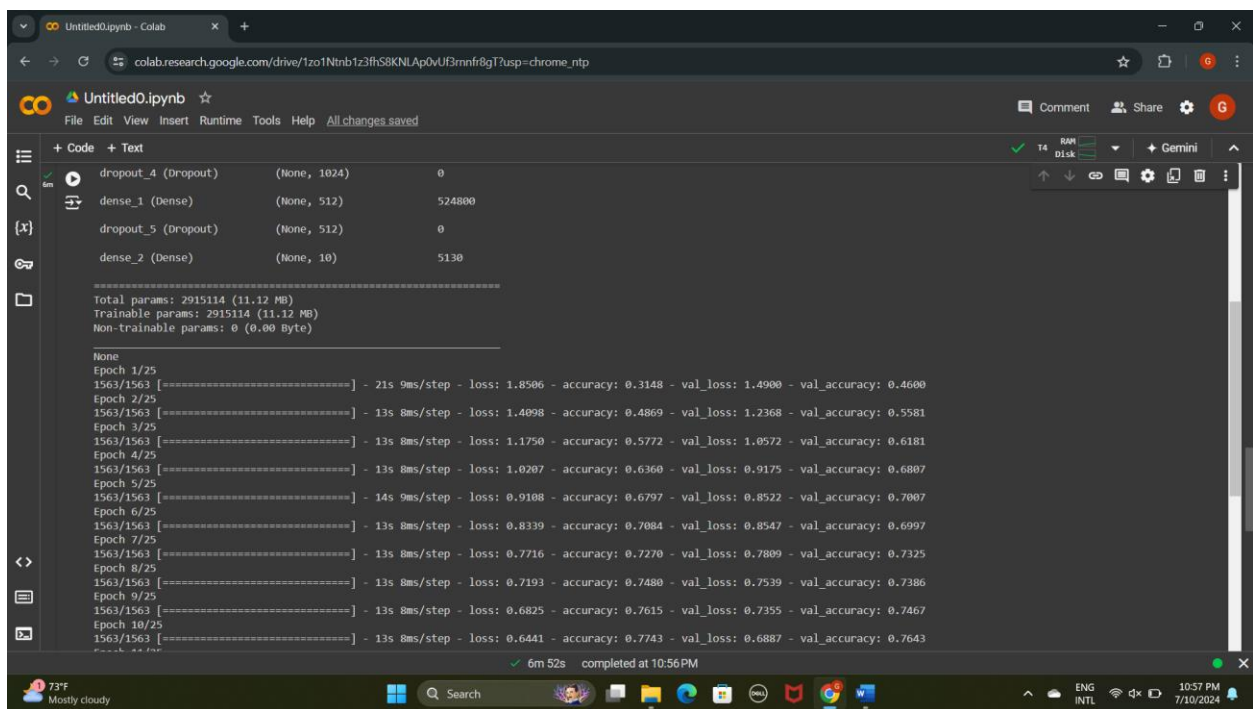
Colab notebook titled 'Untitled0.ipynb' showing a code cell with the following Python code:

```
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```

The output of the code is a summary of the model architecture, showing the number of parameters for each layer:

Layer	Shape	Parameters
conv2d_1 (conv2D)	(None, 32, 32, 32)	9248
max_pooling2d (MaxPooling2D)	(None, 16, 16, 32)	0
conv2d_2 (conv2D)	(None, 16, 16, 64)	18496
dropout_1 (Dropout)	(None, 16, 16, 64)	0
conv2d_3 (conv2D)	(None, 16, 16, 64)	36928
max_pooling2d_1 (MaxPooling2D)	(None, 8, 8, 64)	0
conv2d_4 (conv2D)	(None, 8, 8, 128)	73856
dropout_2 (Dropout)	(None, 8, 8, 128)	0
conv2d_5 (conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_2 (MaxPooling2D)	(None, 4, 4, 128)	0
flatten (Flatten)	(None, 2048)	0
dropout_3 (Dropout)	(None, 2048)	0
dense (Dense)	(None, 1024)	2098176

Executing (6m 40s) <cell line: 63> error_handler()



Colab notebook titled 'Untitled0.ipynb' showing a code cell with the following Python code:

```
dropout_4 (Dropout) (None, 1024) 0
dense_1 (Dense) (None, 512) 524800
dropout_5 (Dropout) (None, 512) 0
dense_2 (Dense) (None, 10) 5130

Total params: 2915114 (11.12 MB)
Trainable params: 2915114 (11.12 MB)
Non-trainable params: 0 (0.00 Byte)

None
Epoch 1/25
1563/1563 [=====] - 21s 9ms/step - loss: 1.8506 - accuracy: 0.3148 - val_loss: 1.4900 - val_accuracy: 0.4600
Epoch 2/25
1563/1563 [=====] - 13s 8ms/step - loss: 1.4098 - accuracy: 0.4869 - val_loss: 1.2368 - val_accuracy: 0.5581
Epoch 3/25
1563/1563 [=====] - 13s 8ms/step - loss: 1.1750 - accuracy: 0.5772 - val_loss: 1.0572 - val_accuracy: 0.6181
Epoch 4/25
1563/1563 [=====] - 13s 8ms/step - loss: 1.0207 - accuracy: 0.6360 - val_loss: 0.9175 - val_accuracy: 0.6807
Epoch 5/25
1563/1563 [=====] - 14s 9ms/step - loss: 0.9108 - accuracy: 0.6797 - val_loss: 0.8522 - val_accuracy: 0.7007
Epoch 6/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.8339 - accuracy: 0.7084 - val_loss: 0.8547 - val_accuracy: 0.6997
Epoch 7/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.7716 - accuracy: 0.7270 - val_loss: 0.7809 - val_accuracy: 0.7325
Epoch 8/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.7193 - accuracy: 0.7480 - val_loss: 0.7539 - val_accuracy: 0.7386
Epoch 9/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.6825 - accuracy: 0.7615 - val_loss: 0.7355 - val_accuracy: 0.7467
Epoch 10/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.6441 - accuracy: 0.7743 - val_loss: 0.6887 - val_accuracy: 0.7643
```

6m 52s completed at 10:56 PM

```
Epoch 11/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.6213 - accuracy: 0.7842 - val_loss: 0.7006 - val_accuracy: 0.7589
Epoch 12/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5988 - accuracy: 0.7918 - val_loss: 0.6669 - val_accuracy: 0.7723
Epoch 13/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5846 - accuracy: 0.7958 - val_loss: 0.6776 - val_accuracy: 0.7721
Epoch 14/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5637 - accuracy: 0.8046 - val_loss: 0.7321 - val_accuracy: 0.7631
Epoch 15/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5563 - accuracy: 0.8059 - val_loss: 0.7261 - val_accuracy: 0.7539
Epoch 16/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5550 - accuracy: 0.8086 - val_loss: 0.6751 - val_accuracy: 0.7790
Epoch 17/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5330 - accuracy: 0.8160 - val_loss: 0.6932 - val_accuracy: 0.7678
Epoch 18/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5228 - accuracy: 0.8175 - val_loss: 0.7228 - val_accuracy: 0.7562
Epoch 19/25
1563/1563 [=====] - 13s 9ms/step - loss: 0.5299 - accuracy: 0.8174 - val_loss: 0.6796 - val_accuracy: 0.7719
Epoch 20/25
1563/1563 [=====] - 13s 9ms/step - loss: 0.5215 - accuracy: 0.8215 - val_loss: 0.7126 - val_accuracy: 0.7599
Epoch 21/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5290 - accuracy: 0.8172 - val_loss: 0.7075 - val_accuracy: 0.7573
Epoch 22/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5256 - accuracy: 0.8201 - val_loss: 0.7205 - val_accuracy: 0.7592
Epoch 23/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5343 - accuracy: 0.8174 - val_loss: 0.7392 - val_accuracy: 0.7589
Epoch 24/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5409 - accuracy: 0.8161 - val_loss: 0.7242 - val_accuracy: 0.7591
Epoch 25/25
1563/1563 [=====] - 13s 8ms/step - loss: 0.5453 - accuracy: 0.8148 - val_loss: 0.6899 - val_accuracy: 0.7755
Accuracy: 77.55%
1/1 [=====] - 0s 457ms/step
Image 1: Predicted: 3, Actual: 3
Image 2: Predicted: 8, Actual: 8
Image 3: Predicted: 8, Actual: 8
```

