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Follow the instruction below and then report how the performance changed.  
(apply all at once)

- Convolutional input layer, 32 feature maps with a size of  $3 \times 3$  and a rectifier activation function.
- Dropout layer at 20%.
- Convolutional layer, 32 feature maps with a size of  $3 \times 3$  and a rectifier activation function.
- Max Pool layer with size  $2 \times 2$ .
- Convolutional layer, 64 feature maps with a size of  $3 \times 3$  and a rectifier activation function.
- Dropout layer at 20%.
- Convolutional layer, 64 feature maps with a size of  $3 \times 3$  and a rectifier activation function.
- Max Pool layer with size  $2 \times 2$ .
- Convolutional layer, 128 feature maps with a size of  $3 \times 3$  and a rectifier activation function.
- Dropout layer at 20%.
- Convolutional layer, 128 feature maps with a size of  $3 \times 3$  and a rectifier activation function.
- Max Pool layer with size  $2 \times 2$ .
- Flatten layer.
- Dropout layer at 20%.
- Fully connected layer with 1024 units and a rectifier activation function.

- Dropout layer at 20%.
  - Fully connected layer with 512 units and a rectifier activation function.
  - Dropout layer at 20%.
  - Fully connected output layer with 10 units and a Softmax activation function
- Did the performance change?

✓ [3] Model: "sequential\_2"

Layer (type)	Output Shape	Param #
conv2d_4 (Conv2D)	(None, 32, 32, 32)	896
dropout_4 (Dropout)	(None, 32, 32, 32)	0
conv2d_5 (Conv2D)	(None, 32, 32, 32)	9248
max_pooling2d_2 (MaxPooling 2D)	(None, 16, 16, 32)	0
conv2d_6 (Conv2D)	(None, 16, 16, 64)	18496
dropout_5 (Dropout)	(None, 16, 16, 64)	0
conv2d_7 (Conv2D)	(None, 16, 16, 64)	36928
max_pooling2d_3 (MaxPooling 2D)	(None, 8, 8, 64)	0
conv2d_8 (Conv2D)	(None, 8, 8, 128)	73856
dropout_6 (Dropout)	(None, 8, 8, 128)	0
conv2d_9 (Conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_4 (MaxPooling 2D)	(None, 4, 4, 128)	0
flatten_2 (Flatten)	(None, 2048)	0
dropout_7 (Dropout)	(None, 2048)	0
dense_4 (Dense)	(None, 1024)	2098176
dropout_8 (Dropout)	(None, 1024)	0
dense_5 (Dense)	(None, 512)	524800
dropout_9 (Dropout)	(None, 512)	0
dense_6 (Dense)	(None, 10)	5130

```

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Total params: 2,915,114
Trainable params: 2,915,114
Non-trainable params: 0
None
Epoch 1/5
1563/1563 [=====] - 529s 338ms/step - loss: 1.9325 - accuracy: 0.2845 - val_loss: 1.6837 - val_accuracy: 0.3913
Epoch 2/5
1563/1563 [=====] - 534s 342ms/step - loss: 1.5569 - accuracy: 0.4308 - val_loss: 1.4765 - val_accuracy: 0.4626
Epoch 3/5
1563/1563 [=====] - 509s 325ms/step - loss: 1.4112 - accuracy: 0.4849 - val_loss: 1.3202 - val_accuracy: 0.5198
Epoch 4/5
1563/1563 [=====] - 508s 325ms/step - loss: 1.3260 - accuracy: 0.5191 - val_loss: 1.2744 - val_accuracy: 0.5386
Epoch 5/5
1563/1563 [=====] - 509s 326ms/step - loss: 1.2652 - accuracy: 0.5406 - val_loss: 1.1948 - val_accuracy: 0.5695
Accuracv: 56.95%

```

2. Predict the first 4 images of the test data using the above model. Then, compare with the actual label for those 4 images to check whether or not the model has predicted correctly.

```

✓ [4]
js
# 2. Predict the first 4 images of the test data
predictions = model.predict(X_test[:4])
# Convert the predictions to class labels
predicted_labels = numpy.argmax(predictions, axis=1)
# Convert the actual labels to class labels
actual_labels = numpy.argmax(y_test[:4], axis=1)

# Print the predicted and actual labels for the first 4 images
print("Predicted labels:", predicted_labels)
print("Actual labels:   ", actual_labels)

1/1 [=====] - 0s 159ms/step
Predicted labels: [3 8 8 8]
Actual labels:   [3 8 8 0]

```

### 3. Visualize Loss and Accuracy using the history object

