```
In [1]: # Plot ad hoc CIFAR10 instances
        from tensorflow.keras.datasets import cifar10
        from matplotlib import pyplot
In [2]: # Load data
        (X_train, y_train), (X_test, y_test) = cifar10.load_data()
        Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
         (https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz)
        In [3]: # Simple CNN model for the CIFAR-10 Dataset
        import numpy
        from keras.datasets import cifar10
        from keras.models import Sequential
        from keras.layers import Dense
        from keras.layers import Dropout
        from keras.layers import Flatten
        from keras.constraints import maxnorm
        from keras.optimizers import SGD
        from keras.layers.convolutional import Convolution2D
        from keras.layers.convolutional import MaxPooling2D
        from keras.utils import np utils
        from keras import backend as K
        #K.set image dim ordering( 'th' )
In [4]: # fix random seed for reproducibility
        seed = 7
        numpy.random.seed(seed)
In [5]: # Load data
        (X_train, y_train), (X_test, y_test) = cifar10.load_data()
In [7]: # 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 \text{ test} = X \text{ test} / 255.0
In [8]: # one hot encode outputs
        y_train = np_utils.to_categorical(y_train)
        y_test = np_utils.to_categorical(y_test)
        num_classes = y_test.shape[1]
```

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In [9]: # Create the model
    model = Sequential()
    model.add(Convolution2D(32,(3, 3), input_shape=(32, 32, 3), activation= 'relu'))
    model.add(Dropout(0.2))
    model.add(Convolution2D(32, 3, 3, activation= 'relu',))
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Flatten())
    model.add(Dense(512, activation= 'relu' , ))
    model.add(Dropout(0.5))
    model.add(Dense(num_classes, activation= 'softmax' ))
```


Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 30, 30, 32)	896
dropout (Dropout)	(None, 30, 30, 32)	0
conv2d_1 (Conv2D)	(None, 10, 10, 32)	9248
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 5, 5, 32)	0
flatten (Flatten)	(None, 800)	0
dense (Dense)	(None, 512)	410112
dropout_1 (Dropout)	(None, 512)	0
dense 1 (Dense)	(None, 10)	5130

Total params: 425,386 Trainable params: 425,386 Non-trainable params: 0

None

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D:\anaconda3\lib\site-packages\keras\optimizers\optimizer_v2\gradient_descent.p
y:111: UserWarning: The `lr` argument is deprecated, use `learning_rate` instea
d.
   super().__init__(name, **kwargs)
```

```
Epoch 1/25
1563/1563 [============== ] - 73s 45ms/step - loss: 1.7763 - acc
uracy: 0.3566 - val_loss: 1.4066 - val_accuracy: 0.5038
Epoch 2/25
1563/1563 [================ ] - 69s 44ms/step - loss: 1.4216 - acc
uracy: 0.4882 - val_loss: 1.2681 - val_accuracy: 0.5529
Epoch 3/25
uracy: 0.5302 - val_loss: 1.1992 - val_accuracy: 0.5732
Epoch 4/25
uracy: 0.5622 - val_loss: 1.1907 - val_accuracy: 0.5831
Epoch 5/25
1563/1563 [============== ] - 70s 45ms/step - loss: 1.1719 - acc
uracy: 0.5830 - val loss: 1.1069 - val accuracy: 0.6130
Epoch 6/25
1563/1563 [=============== ] - 70s 45ms/step - loss: 1.1261 - acc
uracy: 0.5997 - val_loss: 1.0979 - val_accuracy: 0.6091
Epoch 7/25
1563/1563 [============== ] - 69s 44ms/step - loss: 1.0894 - acc
uracy: 0.6151 - val loss: 1.0565 - val accuracy: 0.6286
Epoch 8/25
1563/1563 [=============== ] - 69s 44ms/step - loss: 1.0559 - acc
uracy: 0.6228 - val loss: 1.0409 - val accuracy: 0.6299
Epoch 9/25
1563/1563 [=============== ] - 70s 45ms/step - loss: 1.0313 - acc
uracy: 0.6340 - val loss: 1.0329 - val accuracy: 0.6365
Epoch 10/25
uracy: 0.6441 - val_loss: 1.0177 - val_accuracy: 0.6391
Epoch 11/25
1563/1563 [============= ] - 71s 45ms/step - loss: 0.9793 - acc
uracy: 0.6532 - val loss: 1.0007 - val accuracy: 0.6476
Epoch 12/25
1563/1563 [================ ] - 71s 45ms/step - loss: 0.9591 - acc
uracy: 0.6608 - val loss: 0.9891 - val accuracy: 0.6517
Epoch 13/25
1563/1563 [=============== ] - 71s 46ms/step - loss: 0.9362 - acc
uracy: 0.6684 - val loss: 0.9896 - val accuracy: 0.6528
Epoch 14/25
1563/1563 [================ ] - 71s 45ms/step - loss: 0.9219 - acc
uracy: 0.6727 - val_loss: 0.9775 - val_accuracy: 0.6575
Epoch 15/25
1563/1563 [================ ] - 71s 45ms/step - loss: 0.9049 - acc
uracy: 0.6783 - val_loss: 0.9692 - val_accuracy: 0.6612
Epoch 16/25
1563/1563 [================ ] - 79s 50ms/step - loss: 0.8927 - acc
uracy: 0.6826 - val loss: 0.9627 - val accuracy: 0.6623
Epoch 17/25
1563/1563 [=============== ] - 69s 44ms/step - loss: 0.8749 - acc
uracy: 0.6892 - val_loss: 0.9549 - val_accuracy: 0.6640
Epoch 18/25
```

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uracy: 0.6953 - val_loss: 0.9561 - val_accuracy: 0.6650
      Epoch 19/25
      1563/1563 [=============== ] - 75s 48ms/step - loss: 0.8519 - acc
      uracy: 0.6986 - val loss: 0.9448 - val accuracy: 0.6665
      Epoch 20/25
      uracy: 0.7039 - val_loss: 0.9443 - val_accuracy: 0.6723
      Epoch 21/25
      uracy: 0.7063 - val_loss: 0.9432 - val_accuracy: 0.6680
      Epoch 22/25
      uracy: 0.7113 - val_loss: 0.9334 - val_accuracy: 0.6745
      Epoch 23/25
      1563/1563 [=============== ] - 70s 45ms/step - loss: 0.8060 - acc
      uracy: 0.7151 - val_loss: 0.9375 - val_accuracy: 0.6745
      Epoch 24/25
      uracy: 0.7169 - val loss: 0.9354 - val accuracy: 0.6742
      Epoch 25/25
      uracy: 0.7193 - val_loss: 0.9308 - val_accuracy: 0.6752
Out[11]: <keras.callbacks.History at 0x10ea76ac9a0>
In [12]: # Final evaluation of the model
      scores = model.evaluate(X_test, y_test)
      print("Accuracy: %.2f%%" % (scores[1]*100))
      313/313 [============== ] - 3s 8ms/step - loss: 0.9308 - accurac
      y: 0.6752
      Accuracy: 67.52%
In [ ]:
```