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```
In [1]: import keras
        from keras.datasets import cifar10
        from keras.preprocessing.image import ImageDataGenerator
        from keras.models import Sequential
        from keras.layers import Dense, Dropout, Activation, Flatten
        from keras.layers import Conv2D, MaxPooling2D
        import os
```

```
Using TensorFlow backend.
In [5]: batch size = 32
        num classes = 10
        epochs = 10
        data augmentation = True
        num predictions = 20
        save dir = os.path.join(os.getcwd(), 'saved models')
        model_name = 'keras_cifar10_trained_model.h5'
        # The data, split between train and test sets:
         (x_train, y_train), (x_test, y_test) = cifar10.load_data()
        print('x_train shape:', x_train.shape)
        print(x_train.shape[0], 'train samples')
        print(x_test.shape[0], 'test samples')
        # Convert class vectors to binary class matrices.
        y train = keras.utils.to categorical(y train, num classes)
        y test = keras.utils.to categorical(y test, num classes)
        x_train shape: (50000, 32, 32, 3)
        50000 train samples
        10000 test samples
In [3]: | model = Sequential()
        model.add(Conv2D(32, (3, 3), padding='same',
                          input_shape=x_train.shape[1:]))
        model.add(Activation('relu'))
        model.add(Conv2D(32, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Conv2D(64, (3, 3), padding='same'))
        model.add(Activation('relu'))
        model.add(Conv2D(64, (3, 3)))
        model.add(Activation('relu'))
        model.add(MaxPooling2D(pool size=(2, 2)))
        model.add(Dropout(0.25))
        model.add(Flatten())
        model.add(Dense(512))
        model.add(Activation('relu'))
        model.add(Dropout(0.5))
        model.add(Dense(num classes))
        model.add(Activation('softmax'))
```

```
In [6]: # initialize adam optimizer
        opt = keras.optimizers.adam(lr=0.0001, decay=1e-6)
        # Let's train the model using Adam
        model.compile(loss='categorical crossentropy',
                      optimizer=opt,
                      metrics=['accuracy'])
        x train = x train.astype('float32')
        x_test = x_test.astype('float32')
        x train /= 255
        x_test /= 255
        if not data augmentation:
            print('Not using data augmentation.')
            model.fit(x_train, y_train,
                      batch size=batch size,
                      epochs=epochs,
                      validation_data=(x_test, y_test),
                      shuffle=True)
        else:
            print('Using real-time data augmentation.')
            # This will do preprocessing and realtime data augmentation:
            datagen = ImageDataGenerator(
                featurewise_center=False, # set input mean to 0 over the dataset
                samplewise_center=False, # set each sample mean to 0
                featurewise std normalization=False, # divide inputs by std of the da
        taset
                samplewise std normalization=False, # divide each input by its std
                zca whitening=False, # apply ZCA whitening
                zca_epsilon=1e-06, # epsilon for ZCA whitening
                rotation range=0, # randomly rotate images in the range (degrees, 0 t
        o 180)
                # randomly shift images horizontally (fraction of total width)
                width_shift_range=0.1,
                # randomly shift images vertically (fraction of total height)
                height shift range=0.1,
                shear_range=0., # set range for random shear
                zoom range=0., # set range for random zoom
                channel_shift_range=0., # set range for random channel shifts
                # set mode for filling points outside the input boundaries
                fill mode='nearest',
                cval=0., # value used for fill_mode = "constant"
                horizontal_flip=True, # randomly flip images
                vertical flip=False, # randomly flip images
                # set rescaling factor (applied before any other transformation)
                rescale=None,
                # set function that will be applied on each input
                preprocessing function=None,
                # image data format, either "channels first" or "channels last"
                data format=None,
                # fraction of images reserved for validation (strictly between 0 and
         1)
                validation split=0.0)
            # Compute quantities required for feature-wise normalization
```

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```
Using real-time data augmentation.
Epoch 1/10
1563/1563 [============== ] - 1386s 887ms/step - loss: 1.8046
- acc: 0.3312 - val_loss: 1.5402 - val_acc: 0.4480
Epoch 2/10
- acc: 0.4421 - val_loss: 1.3277 - val_acc: 0.5256
Epoch 3/10
1563/1563 [============== ] - 1209s 773ms/step - loss: 1.4208
- acc: 0.4843 - val_loss: 1.3159 - val_acc: 0.5288
Epoch 4/10
1563/1563 [=================== ] - 1261s 807ms/step - loss: 1.3435
- acc: 0.5181 - val_loss: 1.1970 - val_acc: 0.5703
Epoch 5/10
1563/1563 [============== ] - 1263s 808ms/step - loss: 1.2825
- acc: 0.5400 - val loss: 1.1581 - val acc: 0.5885
Epoch 6/10
- acc: 0.5651 - val_loss: 1.1002 - val_acc: 0.6056
Epoch 7/10
1563/1563 [============== ] - 1247s 798ms/step - loss: 1.1771
- acc: 0.5823 - val_loss: 1.0436 - val_acc: 0.6302
1563/1563 [============== ] - 1449s 927ms/step - loss: 1.1322
- acc: 0.5985 - val_loss: 1.0214 - val_acc: 0.6376
Epoch 9/10
1563/1563 [============== ] - 1217s 779ms/step - loss: 1.1036
- acc: 0.6096 - val_loss: 1.0035 - val_acc: 0.6468
Epoch 10/10
1563/1563 [============== ] - 1198s 767ms/step - loss: 1.0701
- acc: 0.6200 - val loss: 1.0250 - val acc: 0.6365
```

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```
In [7]: # Save model and weights
        if not os.path.isdir(save dir):
            os.makedirs(save_dir)
        model_path = os.path.join(save_dir, model_name)
        model.save(model path)
        print('Saved trained model at %s ' % model_path)
        # Score trained model.
        scores = model.evaluate(x_test, y_test, verbose=1)
        print('Test loss:', scores[0])
        print('Test accuracy:', scores[1])
        Saved trained model at /Users/devikasubramanian/Dropbox/devika/Documents/Teac
        hing/ml_short_course/code_2019/lab2/MLBootCamp/Python3/lab4/saved_models/kera
        s cifar10 trained model.h5
        10000/10000 [========= ] - 74s 7ms/step
        Test loss: 1.0249755365371704
        Test accuracy: 0.6365
In [ ]:
```