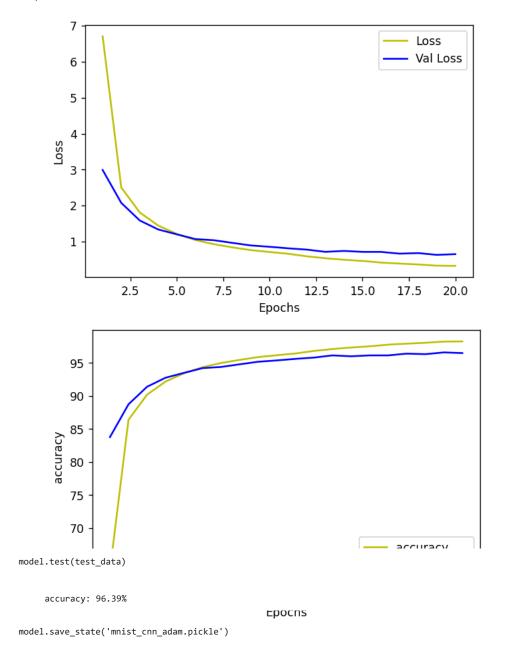
```
from google.colab import drive
drive.mount('/content/drive')
     Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
%cd /content/drive/MyDrive/ml_projects/my_model
import sys
import time
import pickle
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from keras.datasets import mnist, cifar10
import sklearn.datasets
from tqdm import tqdm
from dataset import Dataset
from layers import Dense, Conv2D, Flatten, MaxPooling, AvgPooling, Dropout
from model import Model
 ┌⇒ /content/drive/MyDrive/ml_projects/my_model
(train_x, train_y), (test_x, test_y) = mnist.load_data()
# (train_x, train_y), (test_x, test_y) = cifar10.load_data()
# train x, train y = sklearn.datasets.make circles(n samples=100, shuffle=False, factor=0.3, noise=0.1)
# train_x, train_y = sklearn.datasets.load_iris(return_X_y=True)
# train_x = train_x[:5000]
# train y = train y[:5000]
# test_x = test_x[:500]
# test_y = test_y[:500]
train_data = Dataset(train_x, train_y)
train data.one hot()
# min_val, max_val = train_data.normalize()
test data = Dataset(test x, test y)
test_data.one_hot()
# test_data.normalize(min_val, max_val)
layers = [Conv2D(16, 3, activation='relu',input shape=train data.get shape()),
         MaxPooling(),
         Conv2D(32, 3, activation='relu'),
         MaxPooling(),
         Conv2D(64, 3, activation='relu'),
          Flatten(),
          # Dropout(0.2),
          Dense(64, activation='relu'),
          # Dropout(0.2),
```

```
Dense(train data.num classes, activation='softmax')]
# layers = [Flatten(input shape=train data.get shape()),
            # Dense(256, activation='relu'),
            Dense(128, activation='relu'),
            Dense(train_data.num_classes, activation='softmax')]
model = Model(layers)
model.compile(optim='adam',metrics='accuracy', loss='crossentropy')
model.summary()
     Model Summary
     conv2d_0: 16 x 3 x 3 Activation: relu Parameters:160
     max_pool_0: Activation:None Parameters:0
     conv2d_1: 32 x 3 x 3 Activation: relu Parameters:4640
     max pool 1: Activation: None Parameters:0
     conv2d 2: 64 x 3 x 3 Activation: relu Parameters:18496
     flatten 0: Activation:None Parameters:0
     dense_0: 576 x 64 Activation: relu Parameters:36928
     dense_1: 64 x 10 Activation: softmax Parameters:650
     Total Parameters:60874
results = model.train(train_data,
                        batch size = 256,
                        1r = 1e-3,
                        epochs = 20,
                        val data = test data)
     Epoch 1: 235 batches [07:31, 1.92s/ batches, loss=6.71, val_loss=2.99, acc=63.6, val_acc=83.8]
     Epoch 2: 235 batches [07:29, 1.91s/ batches, loss=2.51, val_loss=2.08, acc=86.4, val_acc=88.7]
     Epoch 3: 235 batches [07:26, 1.90s/ batches, loss=1.81, val loss=1.59, acc=90.2, val acc=91.4]
     Epoch 4: 235 batches [07:28, 1.91s/ batches, loss=1.44, val loss=1.33, acc=92.2, val acc=92.7]
     Epoch 5: 235 batches [07:27, 1.90s/ batches, loss=1.21, val_loss=1.2, acc=93.4, val_acc=93.5]
     Epoch 6: 235 batches [07:26, 1.90s/ batches, loss=1.04, val_loss=1.07, acc=94.3, val_acc=94.2]
     Epoch 7: 235 batches [07:29, 1.91s/ batches, loss=0.926, val_loss=1.04, acc=95, val_acc=94.4]
     Epoch 8: 235 batches [07:30, 1.92s/ batches, loss=0.837, val loss=0.962, acc=95.4, val acc=94.8]
     Epoch 9: 235 batches [07:29, 1.91s/ batches, loss=0.761, val_loss=0.89, acc=95.9, val_acc=95.2]
     Epoch 10: 235 batches [07:28, 1.91s/ batches, loss=0.709, val loss=0.854, acc=96.1, val acc=95.4]
     Epoch 11: 235 batches [07:26, 1.90s/ batches, loss=0.66, val_loss=0.811, acc=96.4, val_acc=95.6]
     Epoch 12: 235 batches [07:28, 1.91s/ batches, loss=0.589, val loss=0.775, acc=96.8, val acc=95.8]
     Epoch 13: 235 batches [07:29, 1.91s/ batches, loss=0.535, val loss=0.713, acc=97.1, val acc=96.1]
     Epoch 14: 235 batches [07:30, 1.91s/ batches, loss=0.493, val_loss=0.739, acc=97.3, val_acc=96]
     Epoch 15: 235 batches [07:26, 1.90s/ batches, loss=0.461, val_loss=0.713, acc=97.5, val_acc=96.1]
     Epoch 16: 235 batches [07:32, 1.93s/ batches, loss=0.414, val_loss=0.712, acc=97.7, val_acc=96.1]
     Epoch 17: 235 batches [07:30, 1.91s/ batches, loss=0.387, val loss=0.663, acc=97.9, val acc=96.4]
     Epoch 18: 235 batches [07:29, 1.91s/ batches, loss=0.362, val_loss=0.678, acc=98, val_acc=96.3]
     Epoch 19: 235 batches [07:32, 1.93s/ batches, loss=0.329, val loss=0.63, acc=98.2, val acc=96.6]
     Epoch 20: 235 batches [07:27, 1.90s/ batches, loss=0.324, val loss=0.649, acc=98.2, val acc=96.5]
model.plot_results()
```



✓ 0s completed at 5:34PM

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