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Code Samples:

Sample diabetes.

```
import pandas as pd
   data = pd.read_csv('diabetes.csv')
 ✓ 0.0s
   path_to_csv = 'diabetes.csv'
 ✓ 0.0s
   import keras
   import pandas

→ om keras.models import Sequential

   from keras.layers.core import Dense, Activation
   from sklearn.model_selection import train_test_split
   import pandas as pd
   import numpy as np
   dataset = pd.read_csv(path_to_csv, header=None).values
   X_train, X_test, Y_train, Y_test = train_test_split(dataset[:,0:8], dataset[:,8],
                                                    test_size=0.25, random_state=87)
   np.random.seed(155)
   my_first_nn = Sequential() # create model
   my_first_nn.add(Dense(20, input_dim=8, activation='relu')) # hidden layer
   my_first_nn.add(Dense(4, activation='relu')) # hidden layer
   my_first_nn.add(Dense(1, activation='sigmoid')) # output layer
   my_first_nn.compile(loss='binary_crossentropy', optimizer='adam', metrics=['acc'])
   my_first_nn_fitted = my_first_nn.fit(X_train, Y_train, epochs=100,
                                      initial epoch=0)
   print(my_first_nn.summary())
   print(my_first_nn.evaluate(X_test, Y_test))
✓ 2.2s
Output exceeds the size limit. Open the full output data in a text editor
Epoch 1/100
18/18 [=====
                 ============= ] - 0s 919us/step - loss: 0.6920 - acc: 0.6597
Epoch 2/100
18/18 [==============] - 0s 882us/step - loss: 0.6887 - acc: 0.6615
Epoch 3/100
                 18/18 [=====
Epoch 4/100
```

Dataset of breast cancer

```
data = pd.read csv('breastcancer.csv')
44] 🗸 0.0s
      path to csv = 'sample data/breastcancer.csv'
      import keras
      import pandas as pd
      import numpy as np
      from keras.models import Sequential
      from keras.layers.core import Dense, Activation
      from sklearn.datasets import load_breast_cancer
      from sklearn.model_selection import train_test_split
      cancer data = load breast cancer()
      X_train, X_test, Y_train, Y_test = train_test_split(cancer_data.data, cancer_data.target,
                                                      test_size=0.25, random_state=87)
      np.random.seed(155)
      my_nn = Sequential() # create model
      my_nn.add(Dense(20, input_dim=30, activation='relu')) # hidden layer 1
      my_nn.add(Dense(1, activation='sigmoid')) # output layer
      my_nn.compile(loss='binary_crossentropy', optimizer='adam', metrics=['acc'])
      my_nn_fitted = my_nn.fit(X_train, Y_train, epochs=100,
                             initial_epoch=0)
      print(my_nn.summary())
      print(my_nn.evaluate(X_test, Y_test))
46]
   Output exceeds the size limit. Open the full output data in a text editor
   Epoch 1/100
                     -----] - 0s 901us/step - loss: 18.6157 - acc: 0.2629
   14/14 [=====
   Epoch 2/100
   14/14 [=============== ] - 0s 714us/step - loss: 15.5858 - acc: 0.2066
   Epoch 3/100
   Epoch 4/100
   14/14 [================ ] - 0s 824us/step - loss: 11.7932 - acc: 0.1432
   Epoch 5/100
```

```
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
14/14 [================= ] - 0s 775us/step - loss: 1.0605 - acc: 0.7300
Epoch 16/100
Epoch 17/100
Epoch 18/100
14/14 [============== ] - 0s 782us/step - loss: 0.6475 - acc: 0.8239
Epoch 19/100
14/14 [===============] - 0s 787us/step - loss: 0.5848 - acc: 0.8169
Epoch 20/100
Epoch 21/100
14/14 [================== ] - 0s 707us/step - loss: 0.5227 - acc: 0.8427
Epoch 22/100
Epoch 23/100
None
5/5 [============= ] - 0s 1ms/step - loss: 0.1808 - acc: 0.9580
[0.1807614266872406, 0.9580419659614563]
```

```
data = pd.read_csv('breastcancer.csv')

v 0.0s
```

```
path_to_csv = 'sample_data/breastcancer.csv'
 ✓ 0.0s
   import keras
   import pandas as pd
   import numpy as np
   from keras.models import Sequential
   from keras.layers.core import Dense, Activation
   from sklearn.datasets import load_breast_cancer
   from sklearn.model_selection import train_test_split
   cancer_data = load_breast_cancer()
   X_train, X_test, Y_train, Y_test = train_test_split(cancer_data.data, cancer_data.target,
                                                   test_size=0.25, random_state=87)
   np.random.seed(155)
   my_nn = Sequential() # create model
   my_nn.add(Dense(20, input_dim=30, activation='relu')) # hidden layer 1
   my_nn.add(Dense(1, activation='sigmoid')) # output layer
   my_nn.compile(loss='binary_crossentropy', optimizer='adam', metrics=['acc'])
   my_nn_fitted = my_nn.fit(X_train, Y_train, epochs=100,
                          initial_epoch=0)
   print(my_nn.summary())
   print(my_nn.evaluate(X_test, Y_test))
 ✓ 1.7s
Output exceeds the size limit. Open the full output data in a text editor
14/14 [=============== ] - 0s 901us/step - loss: 18.6157 - acc: 0.2629
Epoch 2/100
14/14 [------- 15.5858 - acc: 0.2066
Epoch 3/100
14/14 [------] - 0s 739us/step - loss: 13.5255 - acc: 0.1549
Epoch 4/100
14/14 [=========================== ] - 0s 824us/sten - loss: 11.7932 - acc: 0.1432
```

Output of Breast cancer dataset

```
14/14 [=====
               ======== ] - 0s 768us/step - loss: 1.0570 - acc: 0.8146
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
None
5/5 [========] - 0s 1ms/step - loss: 0.2498 - acc: 0.8951
[0.24975551664829254, 0.8951048851013184]
  import keras
  from keras.datasets import mnist
  from keras.models import Sequential
  from keras.layers import Dense, Dropout
  import matplotlib.pyplot as plt
  (x_train, y_train), (x_test, y_test) = mnist.load_data()
  x_train = x_train.astype('float32') / 255
  x_test = x_test.astype('float32') / 255
  num_classes = 10
  y_train = keras.utils.to_categorical(y_train, num_classes)
  y_test = keras.utils.to_categorical(y_test, num_classes)
  model = Sequential()
  model.add(Dense(512, activation='relu', input_shape=(784,)))
  model.add(Dropout(0.2))
  model.add(Dense(512, activation='relu'))
  model.add(Dropout(0.2))
  model.add(Dense(num_classes, activation='softmax'))
  model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
  history = model.fit(x_train.reshape(-1, 784), y_train, validation_data=(x_test.reshape(-1, 784), y_test),
```

```
epochs=20, batch_size=128)
           plt.figure(figsize=(10, 5))
           plt.subplot(1, 2, 1)
           plt.plot(history.history['accuracy'])
           plt.plot(history.history['val_accuracy'])
           plt.title('Model Accuracy')
           plt.ylabel('Accuracy')
           plt.xlabel('Epoch')
           plt.legend(['Train', 'Validation'], loc='lower right')
          plt.subplot(1, 2, 2)
           plt.plot(history.history['loss'])
           plt.plot(history.history['val_loss'])
           plt.title('Model Loss')
          plt.ylabel('Loss')
           plt.xlabel('Epoch')
           plt.legend(['Train', 'Validation'], loc='upper right')
          plt.show()

√ 1m 31.3s

Epoch 1/20
469/469 [=================== ] - 4s 8ms/step - loss: 0.2473 - accuracy: 0.9254 - val_loss: 0.1246 - v
Epoch 2/20
469/469 [================== ] - 4s 8ms/step - loss: 0.1000 - accuracy: 0.9686 - val_loss: 0.0744 - v
Epoch 3/20
469/469 [================== ] - 4s 8ms/step - loss: 0.0730 - accuracy: 0.9772 - val_loss: 0.0690 - v
Epoch 4/20
469/469 [=================== ] - 4s 8ms/step - loss: 0.0557 - accuracy: 0.9825 - val_loss: 0.0729 - v
Epoch 5/20
469/469 [================== ] - 4s 9ms/step - loss: 0.0452 - accuracy: 0.9849 - val_loss: 0.0794 - v
Epoch 6/20
469/469 [================== ] - 4s 8ms/step - loss: 0.0406 - accuracy: 0.9869 - val loss: 0.0684 - v
Epoch 7/20
469/469 [============== ] - 4s 8ms/step - loss: 0.0357 - accuracy: 0.9880 - val loss: 0.0639 - val loss: 0.0
Epoch 8/20
469/469 [============== ] - 4s 8ms/step - loss: 0.0278 - accuracy: 0.9909 - val loss: 0.0672 - val loss: 0.0
Epoch 9/20
469/469 [=================== ] - 4s 9ms/step - loss: 0.0267 - accuracy: 0.9908 - val_loss: 0.0693 - v
Epoch 10/20
469/469 [================== ] - 5s 10ms/step - loss: 0.0258 - accuracy: 0.9915 - val_loss: 0.0680 -
```

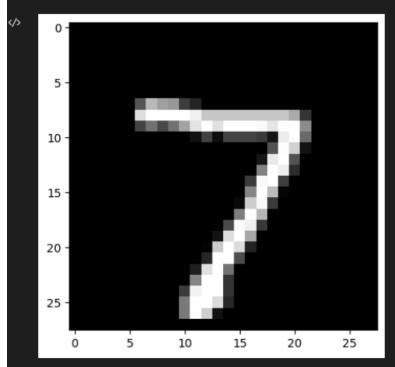
Epoch 11/20

```
469/469 [=
                                   ==] - 4s 9ms/step - loss: 0.0216 - accuracy: 0.9928 - val_loss: 0.0667 -
Epoch 12/20
469/469 [==
                                    =] - 5s 11ms/step - loss: 0.0233 - accuracy: 0.9922 - val_loss: 0.0701
Epoch 13/20
469/469 [==
                                    =] - 5s 10ms/step - loss: 0.0213 - accuracy: 0.9925 - val_loss: 0.0712
Epoch 14/20
                                469/469 [====
Epoch 15/20
469/469 [===
                               =====] - 5s 10ms/step - loss: 0.0172 - accuracy: 0.9944 - val_loss: 0.0812
Epoch 16/20
469/469 [===
                                ====] - 5s 10ms/step - loss: 0.0190 - accuracy: 0.9936 - val_loss: 0.0692
Epoch 17/20
426/469 [====
                               ==>...] - ETA: 0s - loss: 0.0165 - accuracy: 0.9948
                      Model Accuracy
                                                                            Model Loss
                                                       0.25
                                                                                             Train
                                                                                             Validation
   0.99
                                                       0.20
   0.98
   0.97
                                                       0.15
 Accuracy
                                                     Loss
   0.96
                                                       0.10
   0.95
   0.94
                                                       0.05
                                         Train
   0.93
                                         Validation
          0
                    5
                              10
                                        15
                                                             0
                                                                        5
                                                                                  10
                                                                                            15
                           Epoch
                                                                               Epoch
```

Model accuracy using mnist with breast cancer dataset

```
import keras
   from keras.datasets import mnist
   from keras.models import Sequential
   from keras.layers import Dense, Dropout
   import matplotlib.pyplot as plt
   import numpy as np
   (x_train, y_train), (x_test, y_test) = mnist.load_data()
   x_train = x_train.astype('float32') / 255
   x_test = x_test.astype('float32') / 255
   num_classes = 10
   y_train = keras.utils.to_categorical(y_train, num_classes)
   y_test = keras.utils.to_categorical(y_test, num_classes)
   model = Sequential()
   model.add(Dense(512, activation='relu', input_shape=(784,)))
   model.add(Dropout(0.2))
   model.add(Dense(512, activation='relu'))
   model.add(Dropout(0.2))
   model.add(Dense(num_classes, activation='softmax'))
   model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
   model.fit(x_train.reshape(-1, 784), y_train, validation_data=(x_test.reshape(-1, 784), y_test),
             epochs=20, batch size=128)
   plt.imshow(x_test[0], cmap='gray')
   plt.show()
   prediction = model.predict(x_test[0].reshape(1, -1))
   print('Model prediction:', np.argmax(prediction))
 √ 1m 55.4s
Epoch 1/20
469/469 [================= ] - 6s 13ms/step - loss: 0.2495 - accuracy: 0.9251 - val_loss: 0.1098
Epoch 2/20
```

```
Epoch 1/20
469/469 [=============] - 6s 13ms/step - loss: 0.2495 - accuracy: 0.9251 - val_loss: 0.1098
Epoch 2/20
469/469 [============] - 6s 12ms/step - loss: 0.0994 - accuracy: 0.9690 - val_loss: 0.0777
Epoch 3/20
469/469 [============] - 6s 12ms/step - loss: 0.0721 - accuracy: 0.9773 - val_loss: 0.0776
Epoch 4/20
469/469 [============] - 6s 12ms/step - loss: 0.0555 - accuracy: 0.9822 - val_loss: 0.0701
Epoch 5/20
469/469 [============] - 6s 12ms/step - loss: 0.0454 - accuracy: 0.9851 - val_loss: 0.0658
Epoch 6/20
469/469 [============] - 5s 11ms/step - loss: 0.0401 - accuracy: 0.9861 - val_loss: 0.0678
Epoch 7/20
469/469 [===========] - 6s 12ms/step - loss: 0.0349 - accuracy: 0.9887 - val loss: 0.0694
Epoch 8/20
469/469 [=============== ] - 5s 12ms/step - loss: 0.0300 - accuracy: 0.9901 - val_loss: 0.0640
Epoch 9/20
469/469 [==============] - 6s 12ms/step - loss: 0.0275 - accuracy: 0.9908 - val_loss: 0.0666
Epoch 10/20
469/469 [============] - 6s 12ms/step - loss: 0.0253 - accuracy: 0.9917 - val loss: 0.0736
Epoch 11/20
469/469 [=============] - 6s 12ms/step - loss: 0.0225 - accuracy: 0.9926 - val_loss: 0.0822
Epoch 12/20
469/469 [=============] - 5s 12ms/step - loss: 0.0218 - accuracy: 0.9928 - val_loss: 0.0702
Epoch 13/20
469/469 [============] - 6s 12ms/step - loss: 0.0225 - accuracy: 0.9924 - val_loss: 0.0781
Epoch 14/20
469/469 [============] - 6s 13ms/step - loss: 0.0164 - accuracy: 0.9947 - val_loss: 0.0696
Epoch 15/20
469/469 [============] - 6s 13ms/step - loss: 0.0209 - accuracy: 0.9928 - val_loss: 0.0820
Epoch 16/20
469/469 [===========] - 6s 13ms/step - loss: 0.0198 - accuracy: 0.9934 - val_loss: 0.0747
Epoch 17/20
Epoch 18/20
469/469 [============] - 6s 12ms/step - loss: 0.0165 - accuracy: 0.9947 - val_loss: 0.0797
Epoch 19/20
469/469 [===========] - 6s 12ms/step - loss: 0.0151 - accuracy: 0.9952 - val loss: 0.0699
Epoch 20/20
469/469 [=
                          =======] - 6s 12ms/step - loss: 0.0157 - accuracy: 0.9949 - val loss: 0.0811
```



1/1 [======] - 0s 82ms/step Model prediction: 7

```
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout
import matplotlib.pyplot as plt
import numpy as np

(x_train, y_train), (x_test, y_test) = mnist.load_data()

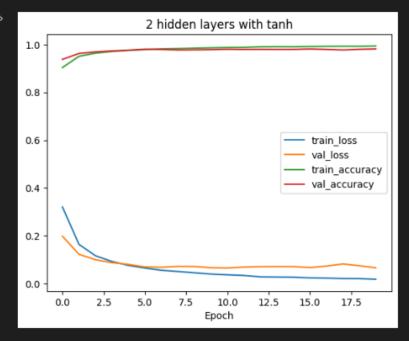
x_train = x_train.astype('float32') / 255
x_test = x_test.astype('float32') / 255
```

```
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout
import matplotlib.pyplot as plt
import numpy as np
(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_train = x_train.astype('float32') / 255
x test = x test.astype('float32') / 255
num classes = 10
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)
models = []
model = Sequential()
model.add(Dense(512, activation='tanh', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('1 hidden layer with tanh', model))
model = Sequential()
model.add(Dense(512, activation='sigmoid', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('1 hidden layer with sigmoid', model))
model = Sequential()
model.add(Dense(512, activation='tanh', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(512, activation='tanh'))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('2 hidden layers with tanh', model))
model = Sequential()
model.add(Dense(512, activation='sigmoid', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(512, activation='sigmoid'))
model.add(Dropout(0.2))
model add/Dence/num classes activation-'coftmay'))
```

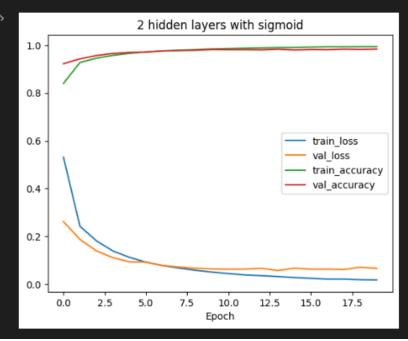
```
models.append(('2 hidden layers with tanh', model))
      model = Sequential()
     model.add(Dense(512, activation='sigmoid', input_shape=(784,)))
      model.add(Dropout(0.2))
     model.add(Dense(512, activation='sigmoid'))
      model.add(Dropout(0.2))
      model.add(Dense(num_classes, activation='softmax'))
      models.append(('2 hidden layers with sigmoid', model))
      for name, model in models:
                 model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
                 history = model.fit(x\_train.reshape(-1, 784), y\_train, validation\_data = (x\_test.reshape(-1, 784), y\_test.reshape(-1, 7
                                                                            epochs=20, batch_size=128, verbose=0)
                 plt.plot(history.history['loss'], label='train_loss')
                 plt.plot(history.history['val_loss'], label='val_loss')
                 plt.plot(history.history['accuracy'], label='train_accuracy')
                 plt.plot(history.history['val_accuracy'], label='val_accuracy')
                 plt.title(name)
                 plt.xlabel('Epoch')
                 plt.legend()
                 plt.show()
                 loss, accuracy = model.evaluate(x_test.reshape(-1, 784), y_test, verbose=0)
                 print('{} - Test loss: {:.4f}, Test accuracy: {:.4f}'.format(name, loss, accuracy))

√ 6m 11.4s
```



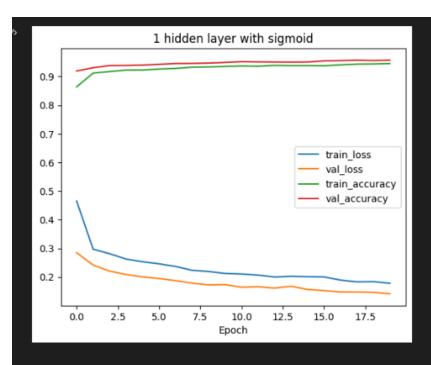


2 hidden layers with tanh - Test loss: 0.0658, Test accuracy: 0.9817

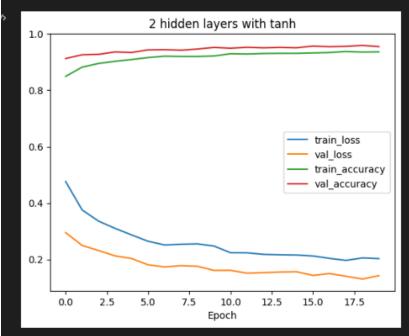


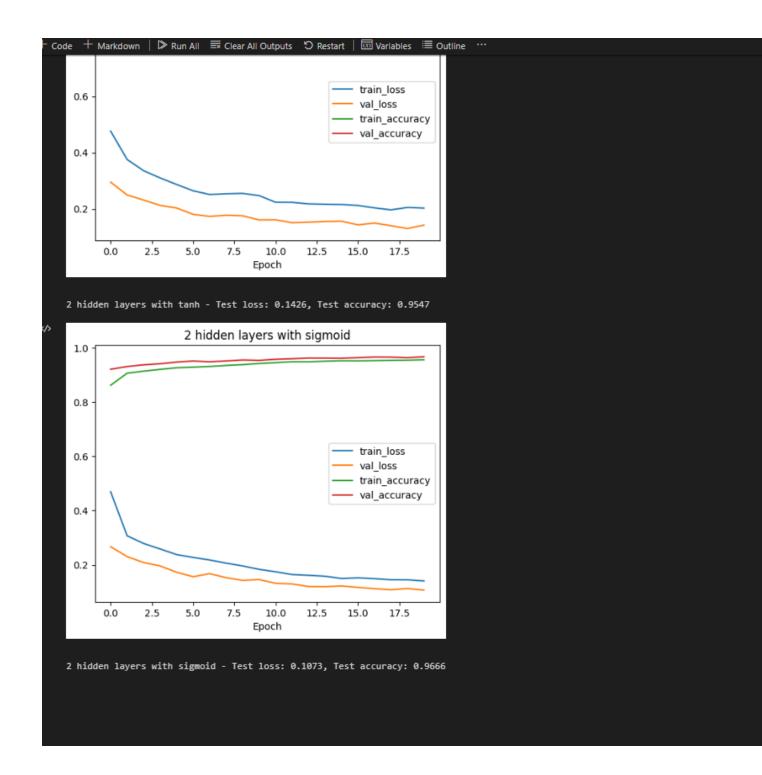
```
import keras
from keras.datasets import mnist
import matplotlib.pyplot as plt
(x_train, y_train), (x_test, y_test) = mnist.load_data()
num_classes = 10
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)
models = []
model = Sequential()
model.add(Dense(512, activation='tanh', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('1 hidden layer with tanh', model))
model = Sequential()
model.add(Dense(512, activation='sigmoid', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('1 hidden layer with sigmoid', model))
model = Sequential()
model.add(Dense(512, activation='tanh', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(512, activation='tanh'))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('2 hidden layers with tanh', model))
model = Sequential()
model.add(Dense(512, activation='sigmoid', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(512, activation='sigmoid'))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('2 hidden layers with sigmoid', model))
for name, model in models:
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    history = model.fit(x_train.reshape(-1, 784), y_train, validation_data=(x_test.reshape(-1, 784), y_test),
                        epochs=20, batch_size=128, verbose=0)
    plt.plot(history.history['loss'], label='train_loss')
    plt.plot(history.history['val_loss'], label='val_loss')
    plt.plot(history.history['accuracy'], label='train_accuracy')
    plt.plot(history.history['val_accuracy'], label='val_accuracy')
```

```
plt.plot(history.history['val_loss'], label='val_loss')
       plt.plot(history.history['accuracy'], label='train_accuracy')
       plt.plot(history.history['val_accuracy'], label='val_accuracy')
       plt.title(name)
       plt.xlabel('Epoch')
       plt.legend()
       plt.show()
       loss, accuracy = model.evaluate(x_test.reshape(-1, 784), y_test, verbose=0)
       print('{} - Test loss: {:.4f}, Test accuracy: {:.4f}'.format(name, loss, accuracy))
() 5m 38.2s
                         1 hidden layer with tanh
 0.9
 0.8
 0.7
                                                        train_loss
 0.6
                                                        val_loss
                                                        train_accuracy
 0.5
                                                        val_accuracy
 0.4
 0.3
 0.2
       0.0
              2.5
                      5.0
                              7.5
                                      10.0
                                             12.5
                                                     15.0
                                                             17.5
                                   Epoch
1 hidden layer with tanh - Test loss: 0.1896, Test accuracy: 0.9442
                       1 hidden layer with sigmoid
 0.9
 0.8
 0.7
                                                        train_loss
 0.6
                                                        val_loss
                                                        train_accuracy
 0.5
                                                        val_accuracy
```



1 hidden layer with sigmoid - Test loss: 0.1414, Test accuracy: 0.9570





Git Hub: https://github.com/Tejasawimarri/NN-ASSIGNMENT-6/