

ASSIGNMENT-2

NEURAL NETWORKS & DEEP LEARNING

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Github link:

<https://github.com/GayathriKeshamoni/ASSIGNMENT-2-NEURAL-NETWORKS-DEEP-LEARNING/upload/main>

Video Link: <https://youtu.be/wAPfIdFs4LQ>

Description: Predicting the diabetes disease.

Programming elements: Keras Basics.

In class programming:

1. Use the use case in the class: a. Add more Dense layers to the existing code and check how the accuracy changes.
2. Change the data source to Breast Cancer dataset *available in the source code folder and make required changes. Report accuracy of the model.
3. Normalize the data before feeding the data to the model and check how the normalization changes your accuracy (code given below).

```
from sklearn.preprocessing  
import StandardScaler  
sc=StandardScaler()
```

Breast Cancer dataset is designated to predict if a patient has Malignant (M) or Benign (B) cancer.

In class programming:

Use Image Classification on the handwritten digits dataset (mnist).

1. Plot the loss and accuracy for both training data and validation data using the history object in the source code.
2. Plot one of the images in the test data, and then do inference to check what is the prediction of the model on

that single image.

3. We had used 2 hidden layers and Relu activation. Try to change the number of hidden layer and the activation to tanh or sigmoid and see what happens.
4. Run the same code without scaling the images and check the performance?

```
In [58]: #read the data
data = pd.read_csv('sample_data/diabetes.csv')
```

```
In [59]: path_to_csv = 'sample_data/diabetes.csv'
```

```
In [63]: import keras
import pandas
from keras.models import Sequential
from keras.layers.core import Dense, Activation

# Load dataset
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))
```

```
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))
```

```
Epoch 1/100
18/18 [=====] - 1s 2ms/step - loss: 18.2141 - acc: 0.3385
Epoch 2/100
18/18 [=====] - 0s 2ms/step - loss: 8.1899 - acc: 0.3438
Epoch 3/100
18/18 [=====] - 0s 3ms/step - loss: 1.7616 - acc: 0.3924
Epoch 4/100
18/18 [=====] - 0s 2ms/step - loss: 0.8124 - acc: 0.5278
Epoch 5/100
18/18 [=====] - 0s 3ms/step - loss: 0.7466 - acc: 0.5972
Epoch 6/100
18/18 [=====] - 0s 2ms/step - loss: 0.7242 - acc: 0.6181
Epoch 7/100
18/18 [=====] - 0s 3ms/step - loss: 0.7203 - acc: 0.6319
Epoch 8/100
18/18 [=====] - 0s 2ms/step - loss: 0.7132 - acc: 0.6458
Epoch 9/100
18/18 [=====] - 0s 3ms/step - loss: 0.7066 - acc: 0.6458
Epoch 10/100
```

```
In [72]: #read the data
data = pd.read_csv('sample_data/breastcancer.csv')
```

```
In [73]: path_to_csv = 'sample_data/breastcancer.csv'
```

```
In [75]: 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

# Load dataset
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=8)

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))
```

```
Epoch 1/100
14/14 [=====] - 1s 5ms/step - loss: 67.9584 - acc: 0.3803
Epoch 2/100
14/14 [=====] - 0s 3ms/step - loss: 20.8848 - acc: 0.3897
Epoch 3/100
14/14 [=====] - 0s 3ms/step - loss: 5.6956 - acc: 0.6901
Epoch 4/100
14/14 [=====] - 0s 4ms/step - loss: 1.8838 - acc: 0.6643
Epoch 5/100
14/14 [=====] - 0s 3ms/step - loss: 1.0273 - acc: 0.8732
Epoch 6/100
14/14 [=====] - 0s 3ms/step - loss: 0.7197 - acc: 0.8498
Epoch 7/100
14/14 [=====] - 0s 4ms/step - loss: 0.6906 - acc: 0.8920
Epoch 8/100
14/14 [=====] - 0s 4ms/step - loss: 0.6208 - acc: 0.8685
Epoch 9/100
14/14 [=====] - 0s 3ms/step - loss: 0.6000 - acc: 0.8800
```

```
In [76]: #read the data
data = pd.read_csv('sample_data/breastcancer.csv')
```

```
In [77]: path_to_csv = 'sample_data/breastcancer.csv'
```

```
In [81]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
```

```
In [82]: 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

# Load dataset
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=8)

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))
```

```
print(my_nn.evaluate(X_test, Y_test))
```

```
Epoch 1/100
14/14 [=====] - 1s 2ms/step - loss: 173.1653 - acc: 0.6197
Epoch 2/100
14/14 [=====] - 0s 2ms/step - loss: 98.1999 - acc: 0.6197
Epoch 3/100
14/14 [=====] - 0s 2ms/step - loss: 25.2683 - acc: 0.6174
Epoch 4/100
14/14 [=====] - 0s 3ms/step - loss: 11.1987 - acc: 0.4061
Epoch 5/100
14/14 [=====] - 0s 2ms/step - loss: 4.9497 - acc: 0.7324
Epoch 6/100
14/14 [=====] - 0s 3ms/step - loss: 4.4129 - acc: 0.7606
Epoch 7/100
14/14 [=====] - 0s 3ms/step - loss: 4.2134 - acc: 0.6808
Epoch 8/100
14/14 [=====] - 0s 2ms/step - loss: 3.7800 - acc: 0.7324
```

```

In [84]: import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout
import matplotlib.pyplot as plt

# Load MNIST dataset
(x_train, y_train), (x_test, y_test) = mnist.load_data()

# normalize pixel values to range [0, 1]
x_train = x_train.astype('float32') / 255
x_test = x_test.astype('float32') / 255

# convert class labels to binary class matrices
num_classes = 10
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)

# create a simple neural network model
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'])

# train the model and record the training history
history = model.fit(x_train.reshape(-1, 784), y_train, validation_data=(x_test.reshape(-1, 784), y_test),
                    epochs=20, batch_size=128)

# plot the training and validation accuracy and loss curves
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])

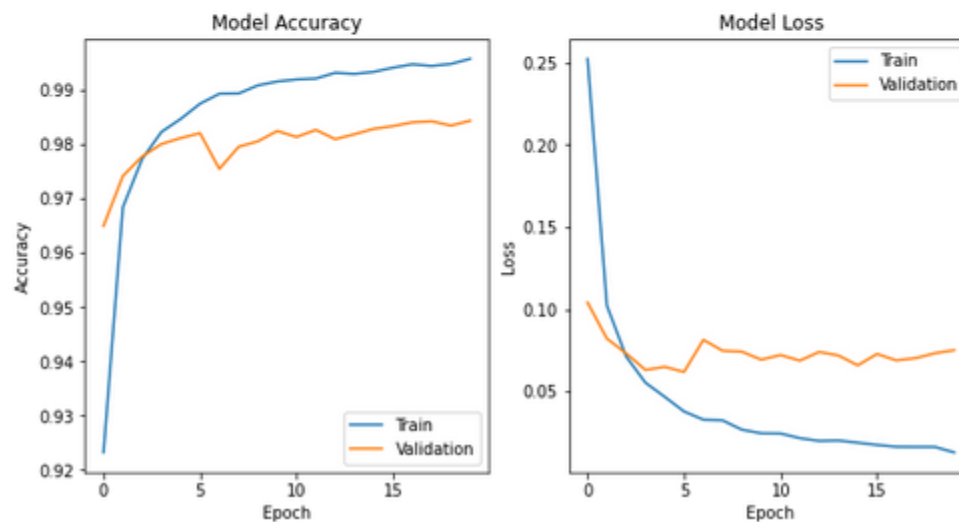
```

```
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()
```

```
Epoch 1/20
469/469 [=====] - 16s 27ms/step - loss: 0.2524 - accurac
y: 0.9232 - val_loss: 0.1042 - val_accuracy: 0.9650
Epoch 2/20
469/469 [=====] - 17s 36ms/step - loss: 0.1024 - accurac
y: 0.9684 - val_loss: 0.0823 - val_accuracy: 0.9742
Epoch 3/20
469/469 [=====] - 14s 29ms/step - loss: 0.0713 - accurac
y: 0.9773 - val_loss: 0.0733 - val_accuracy: 0.9778
Epoch 4/20
469/469 [=====] - 13s 28ms/step - loss: 0.0554 - accurac
y: 0.9823 - val_loss: 0.0632 - val_accuracy: 0.9801
Epoch 5/20
469/469 [=====] - 12s 25ms/step - loss: 0.0468 - accurac
y: 0.9847 - val_loss: 0.0651 - val_accuracy: 0.9812
Epoch 6/20
469/469 [=====] - 12s 25ms/step - loss: 0.0379 - accurac
y: 0.9875 - val_loss: 0.0620 - val_accuracy: 0.9821
Epoch 7/20
469/469 [=====] - 13s 28ms/step - loss: 0.0330 - accurac
y: 0.9894 - val_loss: 0.0815 - val_accuracy: 0.9755
Epoch 8/20
469/469 [=====] - 12s 25ms/step - loss: 0.0325 - accurac
y: 0.9894 - val_loss: 0.0749 - val_accuracy: 0.9796
Epoch 9/20
469/469 [=====] - 15s 31ms/step - loss: 0.0269 - accurac
y: 0.9909 - val_loss: 0.0743 - val_accuracy: 0.9806
Epoch 10/20
469/469 [=====] - 12s 27ms/step - loss: 0.0247 - accurac
v: 0.9916 - val loss: 0.0694 - val accuracv: 0.9825
```

Epoch 15/20
 469/469 [=====] - 11s 24ms/step - loss: 0.0190 - accuracy: 0.9934 - val_loss: 0.0660 - val_accuracy: 0.9829
 Epoch 16/20
 469/469 [=====] - 11s 23ms/step - loss: 0.0176 - accuracy: 0.9942 - val_loss: 0.0729 - val_accuracy: 0.9834
 Epoch 17/20
 469/469 [=====] - 11s 23ms/step - loss: 0.0165 - accuracy: 0.9948 - val_loss: 0.0690 - val_accuracy: 0.9841
 Epoch 18/20
 469/469 [=====] - 11s 24ms/step - loss: 0.0164 - accuracy: 0.9945 - val_loss: 0.0704 - val_accuracy: 0.9843
 Epoch 19/20
 469/469 [=====] - 12s 26ms/step - loss: 0.0164 - accuracy: 0.9949 - val_loss: 0.0734 - val_accuracy: 0.9835
 Epoch 20/20
 469/469 [=====] - 11s 24ms/step - loss: 0.0131 - accuracy: 0.9958 - val_loss: 0.0752 - val_accuracy: 0.9844




```
In [85]: 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

# Load MNIST dataset
(x_train, y_train), (x_test, y_test) = mnist.load_data()

# normalize pixel values to range [0, 1]
x_train = x_train.astype('float32') / 255
x_test = x_test.astype('float32') / 255

# convert class labels to binary class matrices
num_classes = 10
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)

# create a simple neural network model
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'])

# train the model
model.fit(x_train.reshape(-1, 784), y_train, validation_data=(x_test.reshape(-1, 784), y_test),
        epochs=20, batch_size=128)

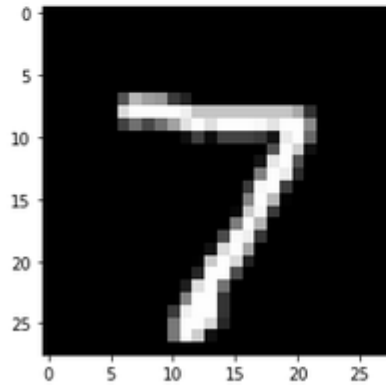
# plot one of the images in the test data
plt.imshow(x_test[0], cmap='gray')
plt.show()

# make a prediction on the image using the trained model
```

```

y: 0.9949 - val_loss: 0.0783 - val_accuracy: 0.9883
Epoch 14/20
469/469 [=====] - 11s 24ms/step - loss: 0.0168 - accurac
y: 0.9945 - val_loss: 0.0694 - val_accuracy: 0.9838
Epoch 15/20
469/469 [=====] - 11s 24ms/step - loss: 0.0192 - accurac
y: 0.9934 - val_loss: 0.0786 - val_accuracy: 0.9820
Epoch 16/20
469/469 [=====] - 11s 23ms/step - loss: 0.0184 - accurac
y: 0.9938 - val_loss: 0.0768 - val_accuracy: 0.9827
Epoch 17/20
469/469 [=====] - 11s 23ms/step - loss: 0.0164 - accurac
y: 0.9948 - val_loss: 0.0775 - val_accuracy: 0.9823
Epoch 18/20
469/469 [=====] - 10s 22ms/step - loss: 0.0162 - accurac
y: 0.9948 - val_loss: 0.0800 - val_accuracy: 0.9822
Epoch 19/20
469/469 [=====] - 11s 24ms/step - loss: 0.0145 - accurac
y: 0.9951 - val_loss: 0.0873 - val_accuracy: 0.9820
Epoch 20/20
469/469 [=====] - 11s 24ms/step - loss: 0.0140 - accurac
y: 0.9957 - val_loss: 0.0807 - val_accuracy: 0.9841

```



```

1/1 [=====] - 0s 120ms/step
Model prediction: 7

```

```

In [88]: 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

        # Load MNIST dataset
        (x_train, y_train), (x_test, y_test) = mnist.load_data()

        # normalize pixel values to range [0, 1]
        x_train = x_train.astype('float32') / 255
        x_test = x_test.astype('float32') / 255

        # convert class labels to binary class matrices
        num_classes = 10
        y_train = keras.utils.to_categorical(y_train, num_classes)
        y_test = keras.utils.to_categorical(y_test, num_classes)

        # create a list of models to train
        models = []

        # model with 1 hidden layer and tanh activation
        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 with 1 hidden layer and sigmoid activation
        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 with 2 hidden layers and tanh activation
        model = Sequential()
        model.add(Dense(512, activation='tanh', input_shape=(784,)))
        model.add(Dropout(0.2))

```

```

models.append(('1 hidden layer with sigmoid', model))

# model with 2 hidden layers and tanh activation
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))

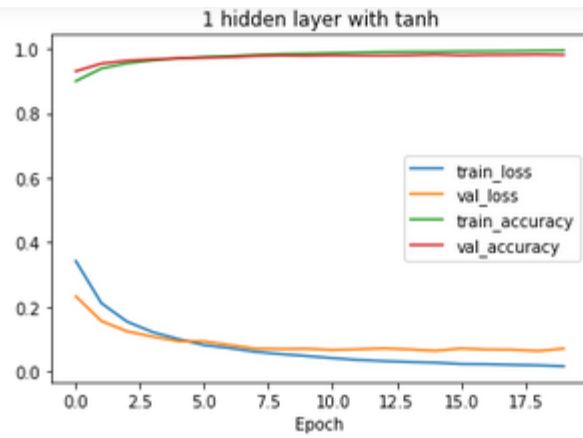
# model with 2 hidden layers and sigmoid activation
model = Sequential()
model.add(Dense(512, activation='sigmoid', input_shape=(784,)))
model.add(Dropout(0.2))
model.add(Dense(512, activation='sigmoid'))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))
models.append(('2 hidden layers with sigmoid', model))

# train each model and plot Loss and accuracy curves
for name, model in models:
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['acc'])
    history = model.fit(x_train.reshape(-1, 784), y_train, validation_data=(x_test, y_test),
                        epochs=20, batch_size=128, verbose=0)

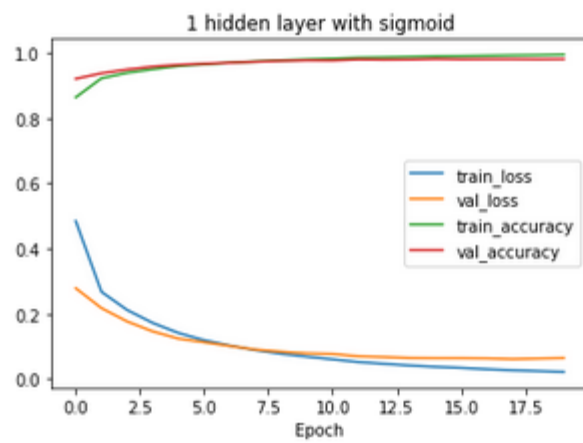
    # plot loss and accuracy curves
    plt.plot(history.history['loss'], label='train_loss')
    plt.plot(history.history['val_loss'], label='val_loss')
    plt.plot(history.history['accuracy'], label='train_accuracy')
    plt.plot(history.history['val_accuracy'], label='val_accuracy')
    plt.title(name)
    plt.xlabel('Epoch')
    plt.legend()
    plt.show()

# evaluate the model on test data
loss, accuracy = model.evaluate(x_test.reshape(-1, 784), y_test, verbose=0)
print('{} - Test loss: {:.4f}, Test accuracy: {:.4f}'.format(name, loss, accuracy))

```

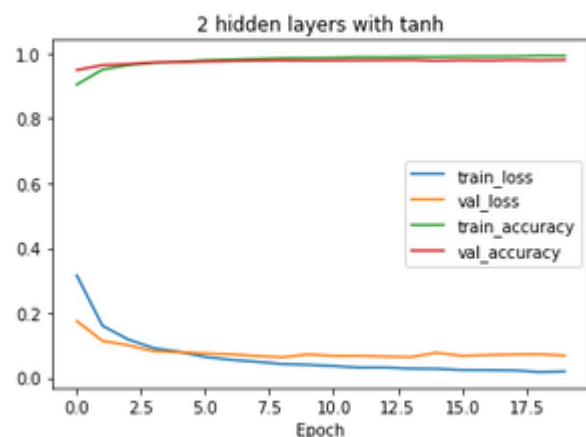


1 hidden layer with tanh - Test loss: 0.0716, Test accuracy: 0.9809

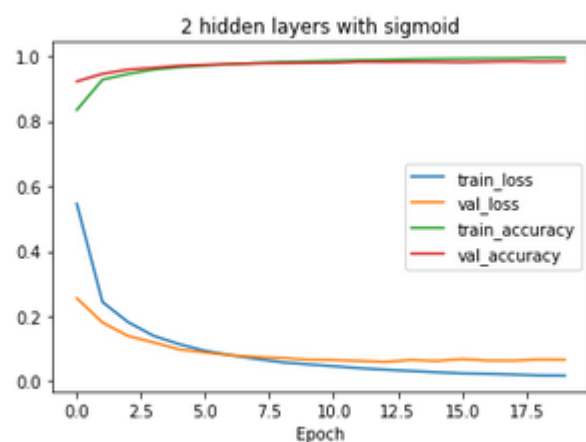


1 hidden layer with sigmoid - Test loss: 0.0642, Test accuracy: 0.9809

1 hidden layer with sigmoid - Test loss: 0.0642, Test accuracy: 0.9809



2 hidden layers with tanh - Test loss: 0.0686, Test accuracy: 0.9808



2 hidden layers with sigmoid - Test loss: 0.0663, Test accuracy: 0.9830

```
In [89]: 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

# Load MNIST dataset
(x_train, y_train), (x_test, y_test) = mnist.load_data()

# convert class labels to binary class matrices
num_classes = 10
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)

# create a list of models to train
models = []

# model with 1 hidden layer and tanh activation
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 with 1 hidden layer and sigmoid activation
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 with 2 hidden layers and tanh activation
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'))
```

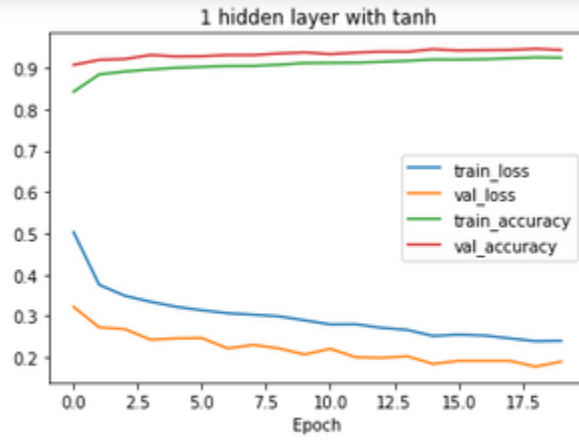
```

# model with 2 hidden layers and sigmoid activation
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))

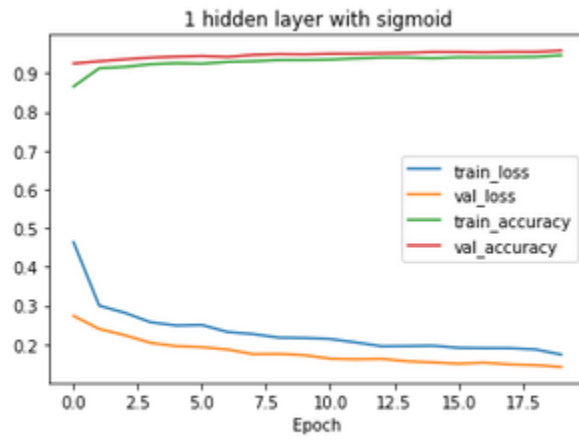
# train each model and plot loss and accuracy curves
for name, model in models:
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['acc'])
    history = model.fit(x_train.reshape(-1, 784), y_train, validation_data=(x_test, y_test),
                        epochs=20, batch_size=128, verbose=0)
    # plot loss and accuracy curves
    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()

# evaluate the model on test data
loss, accuracy = model.evaluate(x_test.reshape(-1, 784), y_test, verbose=0)
print('{} - Test loss: {:.4f}, Test accuracy: {:.4f}'.format(name, loss, accuracy))

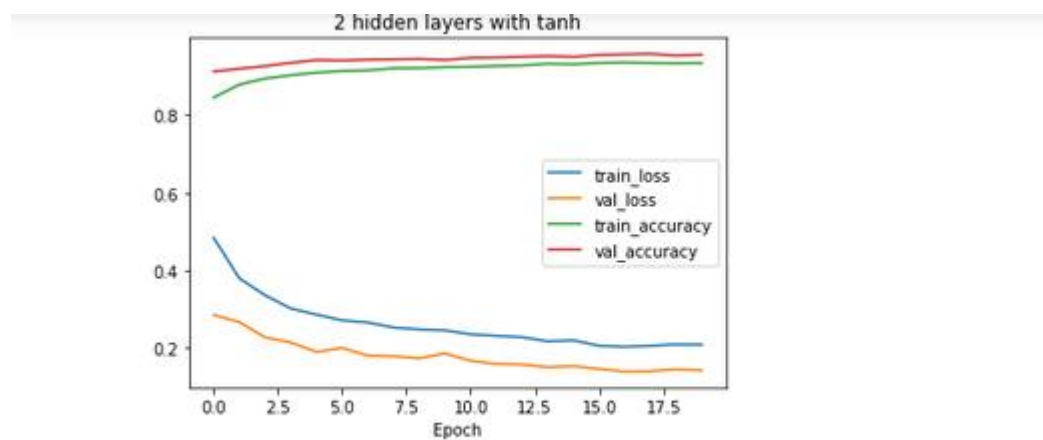
```

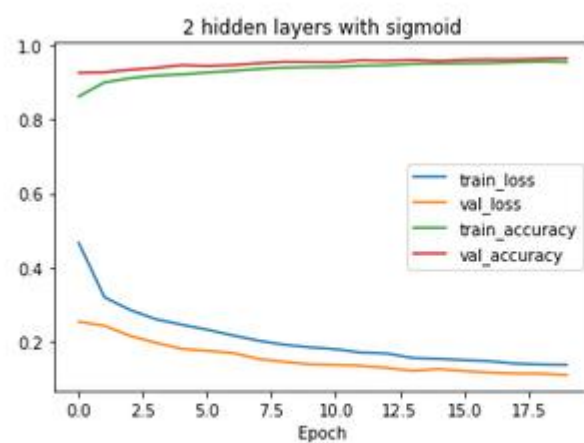
1 hidden layer with tanh - Test loss: 0.1895, Test accuracy: 0.9439



1 hidden layer with sigmoid - Test loss: 0.1420, Test accuracy: 0.9582



2 hidden layers with tanh - Test loss: 0.1422, Test accuracy: 0.9563



2 hidden layers with sigmoid - Test loss: 0.1095, Test accuracy: 0.9652