MLP on MNIST

```
In [1]: from keras.utils import np_utils
    from keras.datasets import mnist
    from keras.initializers import RandomNormal
```

Using TensorFlow backend.

This MNIST dataset contains 70,000 hand-written digits in 28*28 pixels image. It has 10 class labels from 0 to 9

```
t data of mnist dataset
print("total number of data points in train data",len(x_train))
print("total number of data points in test data",len(x_test))

total number of data points in train data 60000
total number of data points in test data 10000

In [3]: print(x_train.shape)
print(x_test.shape)

(60000, 28, 28)
(10000, 28, 28)
```

In [2]: (x_train,y_train),(x_test,y_test)=mnist.load_data() #loading the train and tes

we have 6000 image of 28*28

```
In [0]: x_train=x_train.reshape(x_train.shape[0],x_train.shape[1]*x_train.shape[2])
    x_test=x_test.reshape(x_test.shape[0],x_test.shape[1]*x_test.shape[2])

In [5]: print(x_train.shape)
    print(x_test.shape)

    (60000, 784)
    (10000, 784)
```

now we convert 3-dimention(60000,28,28) matrix into 2-dimention(60000,784) matrix

Data Normalization

here we normalize the data using $X \Rightarrow (X - Xmin)/(Xmax-Xmin) = X/256$

```
In [0]: x_train=x_train/255
x_test=x_test/255
```

```
In [0]: y_train=np_utils.to_categorical(y_train,10)
y_test=np_utils.to_categorical(y_test,10)
```

Building model

1. Two layers MNP with activation- ReLU and optimizer-Adam

Case 1-using with only BatchNormalization

```
In [8]: from keras.models import Sequential
    from keras.layers import Activation
    from keras.layers import Dense
    from keras.layers.normalization import BatchNormalization

model_1=Sequential()
    model_1.add(Dense(128,input_dim=784,activation="relu",kernel_initializer='he_n
        ormal'))
    model_1.add(BatchNormalization()) #batchNormalization
    model_1.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_1.add(Dense(10,activation='softmax'))
    model_1.compile(optimizer='adam',loss='categorical_crossentropy',metrics=['accuracy'])
    history1=model_1.fit(x_train,y_train,epochs=30,verbose=1,batch_size=128,validation_data=(x_test,y_test))
```

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
acc: 0.9195 - val loss: 0.1487 - val acc: 0.9536
Epoch 2/30
acc: 0.9695 - val loss: 0.1081 - val acc: 0.9676
Epoch 3/30
acc: 0.9792 - val loss: 0.0957 - val acc: 0.9694
Epoch 4/30
acc: 0.9843 - val loss: 0.1031 - val acc: 0.9668
acc: 0.9870 - val_loss: 0.1061 - val_acc: 0.9679
Epoch 6/30
acc: 0.9902 - val_loss: 0.0868 - val_acc: 0.9759
Epoch 7/30
acc: 0.9916 - val_loss: 0.0928 - val_acc: 0.9734
Epoch 8/30
acc: 0.9930 - val loss: 0.0798 - val acc: 0.9771
Epoch 9/30
acc: 0.9937 - val_loss: 0.0854 - val_acc: 0.9761
Epoch 10/30
acc: 0.9950 - val_loss: 0.0913 - val_acc: 0.9760
acc: 0.9958 - val loss: 0.0909 - val acc: 0.9763
Epoch 12/30
acc: 0.9958 - val loss: 0.0948 - val acc: 0.9759
Epoch 13/30
acc: 0.9954 - val loss: 0.0945 - val acc: 0.9755
Epoch 14/30
acc: 0.9960 - val_loss: 0.1026 - val_acc: 0.9740
Epoch 15/30
acc: 0.9967 - val loss: 0.0931 - val acc: 0.9774
Epoch 16/30
acc: 0.9974 - val loss: 0.0881 - val acc: 0.9785
Epoch 17/30
acc: 0.9965 - val loss: 0.1058 - val acc: 0.9764
Epoch 18/30
acc: 0.9971 - val loss: 0.1026 - val acc: 0.9760
Epoch 19/30
60000/60000 [============= ] - 3s 52us/step - loss: 0.0086 -
```

```
acc: 0.9973 - val loss: 0.1037 - val acc: 0.9785
Epoch 20/30
acc: 0.9980 - val loss: 0.0971 - val acc: 0.9796
Epoch 21/30
60000/60000 [============== ] - 3s 52us/step - loss: 0.0077 -
acc: 0.9972 - val loss: 0.1007 - val acc: 0.9770
Epoch 22/30
60000/60000 [============ ] - 3s 52us/step - loss: 0.0087 -
acc: 0.9969 - val loss: 0.1086 - val acc: 0.9788
Epoch 23/30
acc: 0.9976 - val loss: 0.0947 - val acc: 0.9790
Epoch 24/30
acc: 0.9992 - val loss: 0.0954 - val acc: 0.9799
Epoch 25/30
acc: 0.9983 - val loss: 0.1095 - val acc: 0.9765
Epoch 26/30
60000/60000 [============== ] - 3s 50us/step - loss: 0.0080 -
acc: 0.9972 - val loss: 0.1121 - val acc: 0.9778
Epoch 27/30
acc: 0.9977 - val_loss: 0.1011 - val_acc: 0.9807
Epoch 28/30
60000/60000 [============ ] - 3s 50us/step - loss: 0.0075 -
acc: 0.9974 - val_loss: 0.1085 - val_acc: 0.9774
Epoch 29/30
acc: 0.9979 - val_loss: 0.1082 - val_acc: 0.9776
Epoch 30/30
acc: 0.9980 - val loss: 0.1017 - val acc: 0.9793
```

```
In [9]: import matplotlib.pyplot as plt

score = model_1.evaluate(x_test, y_test, verbose=1)
print('Test error:', score[0])
print('Test accuracy:', score[1])

x=list(range(1,31))

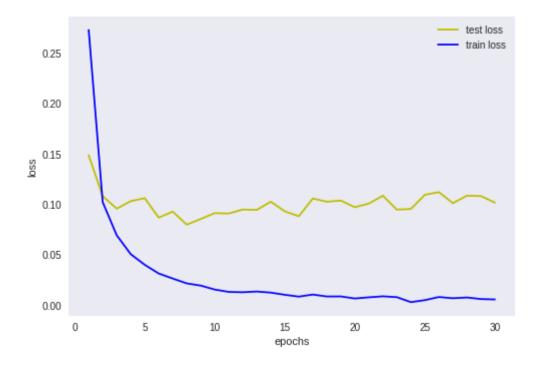
vl1=history1.history['val_loss']
l1=history1.history['loss']
plt.plot(x,vl1,color='y',label='test loss')
plt.plot(x,l1,color='b',label='train loss')
plt.legend()
plt.xlabel('epochs')
plt.ylabel('loss')

plt.grid()
plt.show()
```

10000/10000 [==========] - 0s 29us/step

Test error: 0.1016737612550397

Test accuracy: 0.9793



case 2-with Dropout only

```
In [10]: from keras.layers import Dropout
    model_2=Sequential()
    model_2.add(Dense(128,input_dim=784,activation="relu",kernel_initializer='he_n
        ormal'))
    model_2.add(Dropout(0.5))
    model_2.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_2.add(Dense(10,activation='softmax'))
    model_2.compile(optimizer='adam',loss='categorical_crossentropy',metrics=['acc
        uracy'])
    history2=model_2.fit(x_train,y_train,epochs=30,verbose=1,batch_size=128,valida
        tion_data=(x_test,y_test))
```

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
acc: 0.8447 - val loss: 0.1883 - val acc: 0.9442
Epoch 2/30
acc: 0.9263 - val loss: 0.1432 - val acc: 0.9566
Epoch 3/30
acc: 0.9400 - val loss: 0.1164 - val acc: 0.9636
Epoch 4/30
acc: 0.9462 - val_loss: 0.1091 - val_acc: 0.9685
acc: 0.9504 - val_loss: 0.1028 - val_acc: 0.9696
Epoch 6/30
acc: 0.9549 - val_loss: 0.0915 - val_acc: 0.9729
Epoch 7/30
acc: 0.9567 - val_loss: 0.0853 - val_acc: 0.9735
Epoch 8/30
acc: 0.9599 - val loss: 0.0880 - val acc: 0.9745
Epoch 9/30
acc: 0.9607 - val_loss: 0.0894 - val_acc: 0.9738
Epoch 10/30
acc: 0.9634 - val_loss: 0.0798 - val_acc: 0.9768
acc: 0.9640 - val loss: 0.0809 - val acc: 0.9763
Epoch 12/30
acc: 0.9654 - val loss: 0.0793 - val acc: 0.9776
Epoch 13/30
acc: 0.9661 - val loss: 0.0787 - val acc: 0.9762
Epoch 14/30
acc: 0.9671 - val_loss: 0.0800 - val_acc: 0.9776
Epoch 15/30
acc: 0.9673 - val loss: 0.0775 - val acc: 0.9775
Epoch 16/30
acc: 0.9690 - val loss: 0.0769 - val acc: 0.9777
Epoch 17/30
acc: 0.9697 - val loss: 0.0747 - val acc: 0.9789
Epoch 18/30
acc: 0.9700 - val loss: 0.0748 - val acc: 0.9799
Epoch 19/30
60000/60000 [============= ] - 3s 44us/step - loss: 0.0925 -
```

```
acc: 0.9701 - val loss: 0.0756 - val acc: 0.9786
Epoch 20/30
acc: 0.9711 - val loss: 0.0775 - val acc: 0.9779
Epoch 21/30
60000/60000 [=============== ] - 3s 44us/step - loss: 0.0872 -
acc: 0.9708 - val loss: 0.0754 - val acc: 0.9792
Epoch 22/30
acc: 0.9718 - val loss: 0.0781 - val acc: 0.9786
Epoch 23/30
acc: 0.9726 - val loss: 0.0748 - val acc: 0.9798
Epoch 24/30
acc: 0.9734 - val loss: 0.0735 - val acc: 0.9804
Epoch 25/30
acc: 0.9740 - val loss: 0.0818 - val acc: 0.9788
Epoch 26/30
60000/60000 [============== ] - 3s 45us/step - loss: 0.0806 -
acc: 0.9726 - val loss: 0.0752 - val acc: 0.9786
Epoch 27/30
acc: 0.9737 - val_loss: 0.0761 - val_acc: 0.9794
Epoch 28/30
acc: 0.9748 - val_loss: 0.0775 - val_acc: 0.9800
Epoch 29/30
acc: 0.9743 - val_loss: 0.0731 - val_acc: 0.9805
Epoch 30/30
acc: 0.9754 - val loss: 0.0773 - val acc: 0.9788
```

```
In [11]:
         score=model_2.evaluate(x_test,y_test,verbose=1)
         print("test error",score[0])
         print("test accuracy",score[1])
         x=list(range(1,31))
         vl2=history2.history['val_loss']
         12=history2.history['loss']
         plt.plot(x,vl2,'y',label='validation loss')
         plt.plot(x,12,'b',label='training loss')
         plt.legend()
         plt.xlabel("epochs")
         plt.ylabel("losses")
         plt.grid()
         plt.show()
         10000/10000 [========= ] - 0s 26us/step
         test error 0.07725040358993283
         test accuracy 0.9788
                                                                 validation loss
            0.5

    training loss

            0.4
            0.3
            0.2
            0.1
```

case 3- with both Dropout and BatchNormalization

5

10

15

epochs

20

25

30

0

```
In [12]: model_3=Sequential()
    model_3.add(Dense(128,input_dim=784,activation='relu',kernel_initializer='he_n
        ormal'))
    model_3.add(BatchNormalization())
    model_3.add(Dropout(0.5))
    model_3.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_3.add(Dropout(0.5))
    model_3.add(Dense(10,activation='softmax'))
    model_3.compile(loss='categorical_crossentropy',metrics=['accuracy'],optimizer
    ='adam')
    history3=model_3.fit(x_train,y_train,epochs=30,batch_size=128,verbose=1,valida
    tion_data=(x_test,y_test))
```

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
60000/60000 [============== ] - 4s 60us/step - loss: 0.8147 -
acc: 0.7492 - val loss: 0.2388 - val acc: 0.9277
Epoch 2/30
acc: 0.8805 - val loss: 0.1842 - val acc: 0.9441
Epoch 3/30
60000/60000 [============== ] - 3s 51us/step - loss: 0.3288 -
acc: 0.9053 - val loss: 0.1606 - val acc: 0.9517
Epoch 4/30
acc: 0.9182 - val loss: 0.1385 - val acc: 0.9578
acc: 0.9236 - val_loss: 0.1298 - val_acc: 0.9605
Epoch 6/30
acc: 0.9320 - val_loss: 0.1162 - val_acc: 0.9644
Epoch 7/30
acc: 0.9351 - val_loss: 0.1161 - val_acc: 0.9641
Epoch 8/30
acc: 0.9392 - val_loss: 0.1160 - val_acc: 0.9668
Epoch 9/30
acc: 0.9417 - val_loss: 0.1068 - val_acc: 0.9680
Epoch 10/30
acc: 0.9450 - val_loss: 0.0994 - val_acc: 0.9716
acc: 0.9459 - val loss: 0.0990 - val acc: 0.9710
Epoch 12/30
acc: 0.9464 - val loss: 0.0978 - val acc: 0.9698
Epoch 13/30
acc: 0.9490 - val loss: 0.0984 - val acc: 0.9713
Epoch 14/30
acc: 0.9507 - val_loss: 0.0915 - val_acc: 0.9727
Epoch 15/30
acc: 0.9514 - val loss: 0.0942 - val acc: 0.9739
Epoch 16/30
acc: 0.9525 - val loss: 0.0925 - val acc: 0.9734
Epoch 17/30
acc: 0.9521 - val loss: 0.0928 - val acc: 0.9732
Epoch 18/30
acc: 0.9541 - val_loss: 0.0887 - val_acc: 0.9741
Epoch 19/30
60000/60000 [============= ] - 3s 52us/step - loss: 0.1555 -
```

```
acc: 0.9553 - val loss: 0.0919 - val acc: 0.9732
Epoch 20/30
acc: 0.9565 - val_loss: 0.0930 - val_acc: 0.9733
Epoch 21/30
60000/60000 [============== ] - 3s 52us/step - loss: 0.1532 -
acc: 0.9559 - val loss: 0.0930 - val acc: 0.9732
Epoch 22/30
acc: 0.9573 - val loss: 0.0902 - val acc: 0.9739
Epoch 23/30
60000/60000 [=============== ] - 3s 51us/step - loss: 0.1455 -
acc: 0.9584 - val loss: 0.0921 - val acc: 0.9727
Epoch 24/30
acc: 0.9579 - val loss: 0.0956 - val acc: 0.9727
Epoch 25/30
60000/60000 [=============== ] - 3s 53us/step - loss: 0.1408 -
acc: 0.9578 - val loss: 0.0922 - val acc: 0.9734
Epoch 26/30
60000/60000 [============ ] - 3s 53us/step - loss: 0.1425 -
acc: 0.9592 - val loss: 0.0970 - val acc: 0.9722
Epoch 27/30
acc: 0.9597 - val_loss: 0.0912 - val_acc: 0.9729
Epoch 28/30
acc: 0.9601 - val_loss: 0.0885 - val_acc: 0.9744
Epoch 29/30
acc: 0.9604 - val_loss: 0.0844 - val_acc: 0.9747
Epoch 30/30
acc: 0.9595 - val loss: 0.0942 - val acc: 0.9740
```

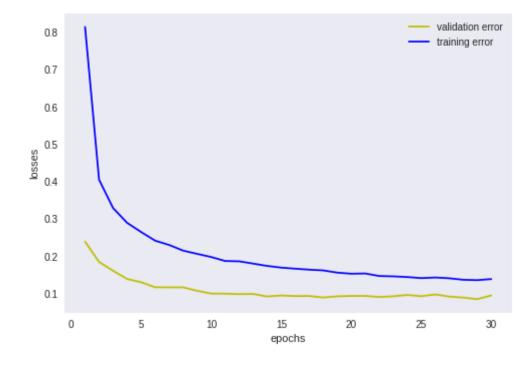
```
In [13]: score = model_3.evaluate(x_test, y_test, verbose=1)
    print('Test error:', score[0])
    print('Test accuracy:', score[1])

    x=list(range(1,31))
    vl3=history3.history['val_loss']
    l3=history3.history['loss']
    plt.plot(x,vl3,'y',label='validation error')
    plt.plot(x,l3,'b',label='training error')
    plt.legend()
    plt.grid()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.show()
```

10000/10000 [============] - 0s 31us/step

Test error: 0.09424658504130203

Test accuracy: 0.974



case 4- without Dropout and BatchNormalization

```
In [14]: model_4=Sequential()
    model_4.add(Dense(128,input_dim=784,activation='relu',kernel_initializer='he_n
    ormal'))
    model_4.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_4.add(Dropout(0.5))
    model_4.add(Dense(10,activation='softmax'))
    model_4.compile(loss='categorical_crossentropy',metrics=['accuracy'],optimizer
    ='adam')
    history4=model_4.fit(x_train,y_train,epochs=50,batch_size=128,verbose=1,valida
    tion_data=(x_test,y_test))
```

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
acc: 0.8470 - val loss: 0.1695 - val acc: 0.9464
Epoch 2/50
acc: 0.9395 - val loss: 0.1285 - val acc: 0.9594
Epoch 3/50
acc: 0.9567 - val loss: 0.1014 - val acc: 0.9687
Epoch 4/50
acc: 0.9651 - val loss: 0.0929 - val acc: 0.9720
acc: 0.9710 - val_loss: 0.0852 - val_acc: 0.9754
Epoch 6/50
acc: 0.9745 - val_loss: 0.0796 - val_acc: 0.9757
Epoch 7/50
acc: 0.9773 - val_loss: 0.0783 - val_acc: 0.9770
Epoch 8/50
acc: 0.9796 - val loss: 0.0772 - val acc: 0.9789
Epoch 9/50
acc: 0.9820 - val_loss: 0.0805 - val_acc: 0.9758
Epoch 10/50
acc: 0.9836 - val_loss: 0.0753 - val_acc: 0.9787
acc: 0.9847 - val loss: 0.0821 - val acc: 0.9779
Epoch 12/50
acc: 0.9868 - val loss: 0.0770 - val acc: 0.9798
Epoch 13/50
acc: 0.9873 - val loss: 0.0801 - val acc: 0.9794
Epoch 14/50
acc: 0.9883 - val_loss: 0.0853 - val_acc: 0.9788
Epoch 15/50
60000/60000 [=========== ] - 2s 41us/step - loss: 0.0347 -
acc: 0.9892 - val loss: 0.0797 - val acc: 0.9799
Epoch 16/50
acc: 0.9897 - val loss: 0.0890 - val acc: 0.9797
Epoch 17/50
acc: 0.9896 - val loss: 0.0837 - val acc: 0.9814
Epoch 18/50
acc: 0.9914 - val loss: 0.0918 - val acc: 0.9807
Epoch 19/50
60000/60000 [============= ] - 2s 41us/step - loss: 0.0263 -
```

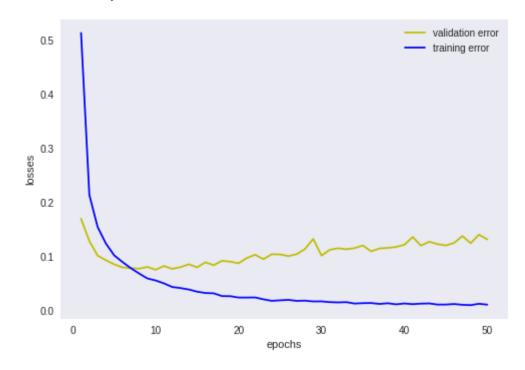
```
acc: 0.9917 - val loss: 0.0902 - val acc: 0.9810
Epoch 20/50
60000/60000 [============== ] - 2s 42us/step - loss: 0.0238 -
acc: 0.9921 - val loss: 0.0873 - val acc: 0.9808
Epoch 21/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0238 -
acc: 0.9922 - val loss: 0.0969 - val acc: 0.9797
Epoch 22/50
60000/60000 [============ ] - 2s 41us/step - loss: 0.0240 -
acc: 0.9922 - val loss: 0.1030 - val acc: 0.9789
Epoch 23/50
60000/60000 [=============== ] - 3s 42us/step - loss: 0.0205 -
acc: 0.9935 - val loss: 0.0948 - val acc: 0.9804
Epoch 24/50
acc: 0.9941 - val loss: 0.1039 - val acc: 0.9794
Epoch 25/50
acc: 0.9938 - val loss: 0.1036 - val acc: 0.9813
Epoch 26/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0195 -
acc: 0.9936 - val loss: 0.1005 - val acc: 0.9803
Epoch 27/50
acc: 0.9941 - val_loss: 0.1040 - val_acc: 0.9788
Epoch 28/50
acc: 0.9940 - val_loss: 0.1134 - val_acc: 0.9796
Epoch 29/50
acc: 0.9945 - val_loss: 0.1324 - val_acc: 0.9758
Epoch 30/50
acc: 0.9945 - val loss: 0.1015 - val acc: 0.9809
Epoch 31/50
60000/60000 [============== ] - 3s 42us/step - loss: 0.0153 -
acc: 0.9949 - val_loss: 0.1121 - val_acc: 0.9799
Epoch 32/50
60000/60000 [=========== ] - 2s 41us/step - loss: 0.0147 -
acc: 0.9954 - val_loss: 0.1150 - val_acc: 0.9793
Epoch 33/50
acc: 0.9948 - val_loss: 0.1131 - val_acc: 0.9790
acc: 0.9959 - val loss: 0.1153 - val acc: 0.9794
Epoch 35/50
60000/60000 [============== ] - 2s 42us/step - loss: 0.0134 -
acc: 0.9957 - val loss: 0.1201 - val acc: 0.9786
Epoch 36/50
acc: 0.9953 - val loss: 0.1092 - val acc: 0.9804
Epoch 37/50
60000/60000 [=========== ] - 2s 41us/step - loss: 0.0120 -
acc: 0.9959 - val_loss: 0.1147 - val_acc: 0.9828
Epoch 38/50
```

```
acc: 0.9957 - val loss: 0.1156 - val acc: 0.9796
Epoch 39/50
acc: 0.9961 - val_loss: 0.1174 - val_acc: 0.9814
Epoch 40/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0128 -
acc: 0.9958 - val loss: 0.1214 - val acc: 0.9796
Epoch 41/50
acc: 0.9959 - val loss: 0.1359 - val acc: 0.9784
Epoch 42/50
60000/60000 [=============== ] - 2s 41us/step - loss: 0.0125 -
acc: 0.9959 - val loss: 0.1199 - val acc: 0.9814
Epoch 43/50
acc: 0.9957 - val loss: 0.1271 - val acc: 0.9788
Epoch 44/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0108 -
acc: 0.9962 - val loss: 0.1226 - val acc: 0.9803
Epoch 45/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0108 -
acc: 0.9967 - val loss: 0.1202 - val acc: 0.9801
Epoch 46/50
acc: 0.9962 - val_loss: 0.1249 - val_acc: 0.9814
Epoch 47/50
acc: 0.9966 - val_loss: 0.1377 - val_acc: 0.9784
Epoch 48/50
acc: 0.9967 - val_loss: 0.1244 - val_acc: 0.9802
Epoch 49/50
acc: 0.9961 - val loss: 0.1401 - val acc: 0.9789
Epoch 50/50
acc: 0.9968 - val_loss: 0.1314 - val_acc: 0.9802
```

```
In [15]: score=model_4.evaluate(x_test,y_test)
    print("test error",score[0])
    print("test accuracy",score[1])

x1=list(range(1,51))
    v14=history4.history['val_loss']
    l4=history4.history['loss']
    plt.plot(x1,v14,'y',label='validation error')
    plt.plot(x1,14,'b',label='training error')
    plt.legend()
    plt.grid()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.show()
```

10000/10000 [============] - 0s 27us/step test error 0.13143735942197668 test accuracy 0.9802



2. Three layers MNP with activation- ReLU and optimizer-Adam

case 1-with only BatchNormalization

```
In [16]: model_5=Sequential()
    model_5.add(Dense(128,input_dim=784,activation='relu',kernel_initializer='he_n
        ormal'))
    model_5.add(BatchNormalization())
    model_5.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_5.add(BatchNormalization())
    model_5.add(Dense(32,activation='relu',kernel_initializer='he_normal'))
    model_5.add(Dense(10,activation='softmax'))
    model_5.compile(optimizer='adam',loss='categorical_crossentropy',metrics=['accuracy'])
    history5=model_5.fit(x_train,y_train,epochs=30,batch_size=128,verbose=1,validation_data=(x_test,y_test))
```

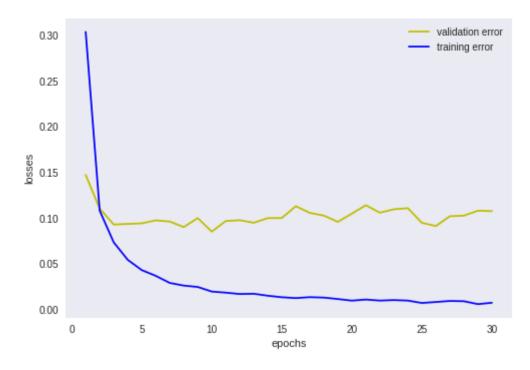
```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
60000/60000 [============== ] - 4s 69us/step - loss: 0.3039 -
acc: 0.9120 - val loss: 0.1472 - val acc: 0.9565
Epoch 2/30
60000/60000 [=============== ] - 3s 55us/step - loss: 0.1079 -
acc: 0.9666 - val loss: 0.1103 - val acc: 0.9647
Epoch 3/30
acc: 0.9767 - val loss: 0.0928 - val acc: 0.9708
Epoch 4/30
acc: 0.9835 - val loss: 0.0935 - val acc: 0.9717
acc: 0.9860 - val_loss: 0.0941 - val_acc: 0.9727
Epoch 6/30
acc: 0.9885 - val_loss: 0.0974 - val_acc: 0.9715
Epoch 7/30
acc: 0.9906 - val_loss: 0.0961 - val_acc: 0.9735
Epoch 8/30
acc: 0.9919 - val loss: 0.0900 - val acc: 0.9737
Epoch 9/30
acc: 0.9917 - val_loss: 0.0999 - val_acc: 0.9730
Epoch 10/30
acc: 0.9937 - val_loss: 0.0850 - val_acc: 0.9771
acc: 0.9941 - val loss: 0.0966 - val acc: 0.9742
Epoch 12/30
acc: 0.9944 - val loss: 0.0975 - val acc: 0.9736
Epoch 13/30
acc: 0.9943 - val loss: 0.0947 - val acc: 0.9758
Epoch 14/30
acc: 0.9949 - val_loss: 0.0998 - val_acc: 0.9768
Epoch 15/30
acc: 0.9958 - val loss: 0.0999 - val acc: 0.9764
Epoch 16/30
acc: 0.9956 - val loss: 0.1129 - val acc: 0.9734
Epoch 17/30
acc: 0.9955 - val loss: 0.1056 - val acc: 0.9745
Epoch 18/30
acc: 0.9952 - val loss: 0.1026 - val acc: 0.9754
Epoch 19/30
60000/60000 [============= ] - 3s 56us/step - loss: 0.0111 -
```

```
acc: 0.9962 - val loss: 0.0957 - val acc: 0.9771
Epoch 20/30
acc: 0.9966 - val_loss: 0.1047 - val_acc: 0.9757
Epoch 21/30
acc: 0.9965 - val loss: 0.1140 - val acc: 0.9742
Epoch 22/30
acc: 0.9966 - val loss: 0.1058 - val acc: 0.9771
Epoch 23/30
60000/60000 [=============== ] - 3s 54us/step - loss: 0.0100 -
acc: 0.9966 - val loss: 0.1095 - val acc: 0.9748
Epoch 24/30
acc: 0.9967 - val loss: 0.1107 - val acc: 0.9757
Epoch 25/30
acc: 0.9975 - val loss: 0.0947 - val acc: 0.9785
Epoch 26/30
60000/60000 [============== ] - 3s 54us/step - loss: 0.0079 -
acc: 0.9974 - val loss: 0.0911 - val acc: 0.9793
Epoch 27/30
acc: 0.9969 - val_loss: 0.1019 - val acc: 0.9778
Epoch 28/30
acc: 0.9971 - val_loss: 0.1025 - val_acc: 0.9784
Epoch 29/30
acc: 0.9981 - val_loss: 0.1080 - val_acc: 0.9775
Epoch 30/30
acc: 0.9977 - val loss: 0.1075 - val acc: 0.9781
```

```
In [17]: score=model_5.evaluate(x_test,y_test,verbose=0)
    print('test error',score[0])
    print('test accuracy',score[1])

x=list(range(1,31))
    vl5=history5.history['val_loss']
    l5=history5.history['loss']
    plt.plot(x,vl5,'y',label='validation error')
    plt.plot(x,15,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

test error 0.10749240423332085 test accuracy 0.9781



case 2-with Dropout

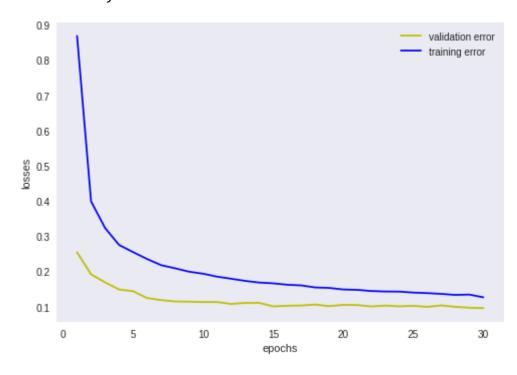
```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
acc: 0.7177 - val loss: 0.2545 - val acc: 0.9264
Epoch 2/30
acc: 0.8862 - val loss: 0.1918 - val acc: 0.9437
Epoch 3/30
60000/60000 [============== ] - 3s 46us/step - loss: 0.3235 -
acc: 0.9096 - val loss: 0.1693 - val acc: 0.9517
Epoch 4/30
acc: 0.9242 - val loss: 0.1488 - val acc: 0.9565
acc: 0.9282 - val_loss: 0.1440 - val_acc: 0.9589
Epoch 6/30
acc: 0.9335 - val_loss: 0.1246 - val_acc: 0.9639
Epoch 7/30
acc: 0.9389 - val_loss: 0.1187 - val_acc: 0.9663
Epoch 8/30
acc: 0.9422 - val loss: 0.1149 - val acc: 0.9649
Epoch 9/30
acc: 0.9439 - val_loss: 0.1143 - val_acc: 0.9676
Epoch 10/30
acc: 0.9451 - val_loss: 0.1133 - val_acc: 0.9671
acc: 0.9472 - val_loss: 0.1131 - val_acc: 0.9686
Epoch 12/30
acc: 0.9493 - val loss: 0.1078 - val acc: 0.9693
Epoch 13/30
acc: 0.9501 - val loss: 0.1108 - val acc: 0.9685
Epoch 14/30
acc: 0.9516 - val_loss: 0.1109 - val_acc: 0.9696
Epoch 15/30
acc: 0.9533 - val loss: 0.1009 - val acc: 0.9715
Epoch 16/30
acc: 0.9537 - val loss: 0.1026 - val acc: 0.9713
Epoch 17/30
acc: 0.9537 - val loss: 0.1033 - val acc: 0.9718
Epoch 18/30
acc: 0.9548 - val loss: 0.1061 - val acc: 0.9707
Epoch 19/30
60000/60000 [============= ] - 3s 46us/step - loss: 0.1533 -
```

```
acc: 0.9568 - val loss: 0.1015 - val acc: 0.9723
Epoch 20/30
acc: 0.9567 - val_loss: 0.1053 - val_acc: 0.9698
Epoch 21/30
60000/60000 [============== ] - 3s 46us/step - loss: 0.1478 -
acc: 0.9578 - val loss: 0.1047 - val acc: 0.9725
Epoch 22/30
acc: 0.9583 - val loss: 0.1007 - val acc: 0.9725
Epoch 23/30
60000/60000 [============== ] - 3s 46us/step - loss: 0.1431 -
acc: 0.9589 - val loss: 0.1032 - val acc: 0.9719
Epoch 24/30
acc: 0.9588 - val loss: 0.1010 - val acc: 0.9734
Epoch 25/30
acc: 0.9596 - val loss: 0.1026 - val acc: 0.9720
Epoch 26/30
60000/60000 [============== ] - 3s 47us/step - loss: 0.1385 -
acc: 0.9602 - val loss: 0.0995 - val acc: 0.9719
Epoch 27/30
acc: 0.9607 - val_loss: 0.1036 - val_acc: 0.9724
Epoch 28/30
acc: 0.9611 - val_loss: 0.0997 - val_acc: 0.9720
Epoch 29/30
acc: 0.9613 - val_loss: 0.0970 - val_acc: 0.9738
Epoch 30/30
60000/60000 [============ ] - 3s 46us/step - loss: 0.1267 -
acc: 0.9636 - val loss: 0.0959 - val acc: 0.9741
```

```
In [19]: score=model_6.evaluate(x_test,y_test,verbose=0)
    print('test error',score[0])
    print('test accuracy',score[1])

x=list(range(1,31))
    vl6=history6.history['val_loss']
    l6=history6.history['loss']
    plt.plot(x,vl6,'y',label='validation error')
    plt.plot(x,l6,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

test error 0.09591007846567082 test accuracy 0.9741



case 3-with both Dropout and BatchNormalization

```
In [20]: model_7=Sequential()
    model_7.add(Dense(128,input_dim=784,activation='relu',kernel_initializer='he_n
        ormal'))
    model_7.add(Dropout(0.5))
    model_7.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_7.add(BatchNormalization())
    model_7.add(Dropout(0.5))
    model_7.add(Dense(32,activation='relu',kernel_initializer='he_normal'))
    model_7.add(Dense(10,activation='softmax'))
    model_7.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
    history7=model_7.fit(x_train,y_train,epochs=30,batch_size=127,verbose=1,validation_data=(x_test,y_test))
```

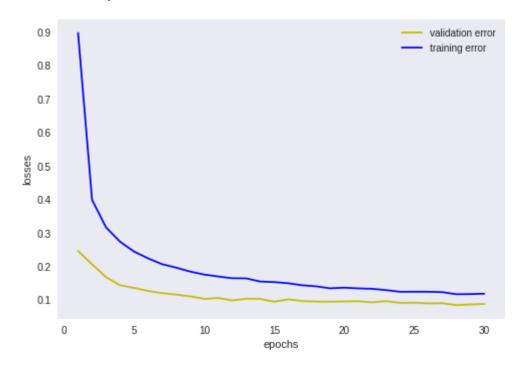
```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
60000/60000 [============== ] - 4s 67us/step - loss: 0.8978 -
acc: 0.7159 - val loss: 0.2460 - val acc: 0.9272
Epoch 2/30
60000/60000 [=============== ] - 3s 52us/step - loss: 0.3989 -
acc: 0.8838 - val loss: 0.2064 - val acc: 0.9396
Epoch 3/30
acc: 0.9104 - val loss: 0.1678 - val acc: 0.9512
Epoch 4/30
acc: 0.9217 - val loss: 0.1440 - val acc: 0.9579
acc: 0.9308 - val_loss: 0.1357 - val_acc: 0.9625
Epoch 6/30
acc: 0.9355 - val_loss: 0.1267 - val_acc: 0.9629
Epoch 7/30
acc: 0.9419 - val_loss: 0.1202 - val_acc: 0.9656
Epoch 8/30
acc: 0.9449 - val_loss: 0.1157 - val_acc: 0.9636
Epoch 9/30
acc: 0.9470 - val_loss: 0.1106 - val_acc: 0.9686
Epoch 10/30
acc: 0.9502 - val_loss: 0.1029 - val_acc: 0.9705
acc: 0.9514 - val loss: 0.1056 - val acc: 0.9699
Epoch 12/30
acc: 0.9535 - val loss: 0.0983 - val acc: 0.9714
Epoch 13/30
acc: 0.9526 - val loss: 0.1033 - val acc: 0.9708
Epoch 14/30
acc: 0.9561 - val_loss: 0.1029 - val_acc: 0.9700
Epoch 15/30
acc: 0.9573 - val loss: 0.0946 - val acc: 0.9720
Epoch 16/30
acc: 0.9574 - val loss: 0.1015 - val acc: 0.9703
Epoch 17/30
acc: 0.9586 - val loss: 0.0964 - val acc: 0.9724
Epoch 18/30
acc: 0.9598 - val loss: 0.0950 - val acc: 0.9735
Epoch 19/30
60000/60000 [============= ] - 3s 53us/step - loss: 0.1345 -
```

```
acc: 0.9616 - val loss: 0.0945 - val acc: 0.9736
Epoch 20/30
acc: 0.9603 - val_loss: 0.0953 - val_acc: 0.9747
Epoch 21/30
60000/60000 [============== ] - 3s 52us/step - loss: 0.1343 -
acc: 0.9607 - val loss: 0.0961 - val acc: 0.9722
Epoch 22/30
60000/60000 [============ ] - 3s 52us/step - loss: 0.1331 -
acc: 0.9618 - val loss: 0.0926 - val acc: 0.9744
Epoch 23/30
60000/60000 [============== ] - 3s 52us/step - loss: 0.1291 -
acc: 0.9627 - val loss: 0.0963 - val acc: 0.9722
Epoch 24/30
acc: 0.9637 - val loss: 0.0909 - val acc: 0.9729
Epoch 25/30
acc: 0.9640 - val loss: 0.0916 - val acc: 0.9732
Epoch 26/30
60000/60000 [============ ] - 3s 51us/step - loss: 0.1242 -
acc: 0.9638 - val loss: 0.0895 - val acc: 0.9738
Epoch 27/30
acc: 0.9649 - val_loss: 0.0902 - val_acc: 0.9750
Epoch 28/30
acc: 0.9661 - val_loss: 0.0843 - val_acc: 0.9767
Epoch 29/30
acc: 0.9667 - val_loss: 0.0864 - val_acc: 0.9758
Epoch 30/30
acc: 0.9657 - val loss: 0.0878 - val acc: 0.9751
```

```
In [21]: score=model_7.evaluate(x_test,y_test,verbose=0)
    print("test error",score[0])
    print("test accuracy",score[1])

x = list(range(1,31))
    vl7=history7.history['val_loss']
    l7=history7.history['loss']
    plt.plot(x,vl7,'y',label='validation error')
    plt.plot(x,l7,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

test error 0.08784910060963594 test accuracy 0.9751



case 4-without Dropout and BatchNormalization

```
In [22]: model_8=Sequential()
    model_8.add(Dense(128,input_dim=784,activation='relu',kernel_initializer='he_n
    ormal'))
    model_8.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_8.add(Dense(32,activation='relu',kernel_initializer='he_normal'))
    model_8.add(Dense(10,activation='softmax'))
    model_8.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['acc uracy'])
    history8=model_8.fit(x_train,y_train,epochs=50,batch_size=128,verbose=1,valida tion_data=(x_test,y_test))
```

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
acc: 0.8962 - val loss: 0.1468 - val acc: 0.9577
Epoch 2/50
acc: 0.9610 - val loss: 0.1308 - val acc: 0.9605
Epoch 3/50
60000/60000 [============== ] - 3s 42us/step - loss: 0.0914 -
acc: 0.9724 - val loss: 0.1100 - val acc: 0.9692
Epoch 4/50
acc: 0.9788 - val loss: 0.0895 - val acc: 0.9716
acc: 0.9829 - val_loss: 0.0933 - val_acc: 0.9703
Epoch 6/50
acc: 0.9863 - val_loss: 0.0885 - val_acc: 0.9724
Epoch 7/50
acc: 0.9889 - val_loss: 0.0843 - val_acc: 0.9756
Epoch 8/50
acc: 0.9902 - val loss: 0.0934 - val acc: 0.9748
Epoch 9/50
acc: 0.9918 - val_loss: 0.0843 - val_acc: 0.9769
Epoch 10/50
acc: 0.9931 - val_loss: 0.0829 - val_acc: 0.9766
acc: 0.9933 - val loss: 0.0963 - val acc: 0.9741
Epoch 12/50
acc: 0.9946 - val loss: 0.0954 - val acc: 0.9758
Epoch 13/50
acc: 0.9946 - val loss: 0.1005 - val acc: 0.9768
Epoch 14/50
acc: 0.9956 - val_loss: 0.0941 - val_acc: 0.9771
Epoch 15/50
60000/60000 [=========== ] - 2s 41us/step - loss: 0.0120 -
acc: 0.9959 - val loss: 0.1115 - val acc: 0.9758
Epoch 16/50
acc: 0.9959 - val loss: 0.1346 - val acc: 0.9717
Epoch 17/50
acc: 0.9950 - val loss: 0.1149 - val acc: 0.9749
Epoch 18/50
acc: 0.9956 - val loss: 0.1119 - val acc: 0.9772
Epoch 19/50
60000/60000 [============= ] - 2s 41us/step - loss: 0.0052 -
```

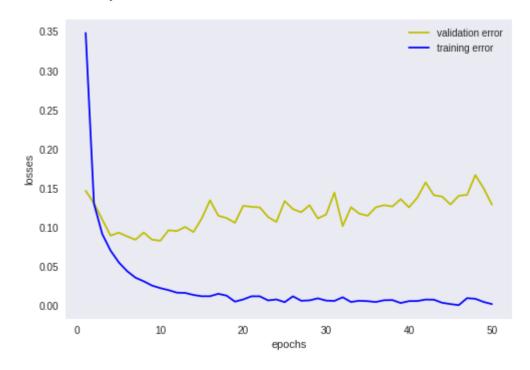
```
acc: 0.9983 - val loss: 0.1059 - val acc: 0.9782
Epoch 20/50
acc: 0.9973 - val loss: 0.1276 - val acc: 0.9744
Epoch 21/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0120 -
acc: 0.9959 - val loss: 0.1262 - val acc: 0.9747
Epoch 22/50
60000/60000 [============ ] - 2s 41us/step - loss: 0.0120 -
acc: 0.9959 - val loss: 0.1256 - val acc: 0.9733
Epoch 23/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0067 -
acc: 0.9975 - val loss: 0.1133 - val acc: 0.9779
Epoch 24/50
acc: 0.9973 - val loss: 0.1070 - val acc: 0.9795
Epoch 25/50
acc: 0.9986 - val loss: 0.1336 - val acc: 0.9754
Epoch 26/50
60000/60000 [============== ] - 2s 41us/step - loss: 0.0119 -
acc: 0.9960 - val loss: 0.1234 - val acc: 0.9778
Epoch 27/50
acc: 0.9979 - val_loss: 0.1194 - val_acc: 0.9791
Epoch 28/50
acc: 0.9979 - val_loss: 0.1283 - val_acc: 0.9777
Epoch 29/50
acc: 0.9972 - val_loss: 0.1114 - val_acc: 0.9789
Epoch 30/50
60000/60000 [=========== ] - 3s 43us/step - loss: 0.0065 -
acc: 0.9980 - val loss: 0.1164 - val acc: 0.9785
Epoch 31/50
acc: 0.9980 - val_loss: 0.1444 - val_acc: 0.9763
Epoch 32/50
60000/60000 [============= ] - 3s 42us/step - loss: 0.0107 -
acc: 0.9965 - val_loss: 0.1017 - val_acc: 0.9812
Epoch 33/50
acc: 0.9985 - val_loss: 0.1257 - val_acc: 0.9792
acc: 0.9978 - val loss: 0.1176 - val acc: 0.9799
Epoch 35/50
acc: 0.9982 - val loss: 0.1149 - val acc: 0.9798
Epoch 36/50
acc: 0.9984 - val loss: 0.1257 - val acc: 0.9787
Epoch 37/50
acc: 0.9978 - val_loss: 0.1284 - val_acc: 0.9776
Epoch 38/50
```

```
acc: 0.9978 - val loss: 0.1267 - val acc: 0.9801
Epoch 39/50
acc: 0.9989 - val loss: 0.1361 - val acc: 0.9788
Epoch 40/50
60000/60000 [============== ] - 3s 42us/step - loss: 0.0059 -
acc: 0.9981 - val loss: 0.1256 - val acc: 0.9798
Epoch 41/50
60000/60000 [============ ] - 3s 42us/step - loss: 0.0059 -
acc: 0.9981 - val loss: 0.1380 - val acc: 0.9794
Epoch 42/50
60000/60000 [=============== ] - 3s 42us/step - loss: 0.0078 -
acc: 0.9974 - val loss: 0.1575 - val acc: 0.9753
Epoch 43/50
acc: 0.9976 - val loss: 0.1411 - val acc: 0.9773
Epoch 44/50
acc: 0.9987 - val loss: 0.1394 - val acc: 0.9760
Epoch 45/50
60000/60000 [============== ] - 3s 42us/step - loss: 0.0021 -
acc: 0.9994 - val_loss: 0.1292 - val acc: 0.9807
Epoch 46/50
60000/60000 [================ ] - 3s 42us/step - loss: 6.9593e-0
4 - acc: 0.9999 - val_loss: 0.1405 - val_acc: 0.9793
Epoch 47/50
acc: 0.9967 - val_loss: 0.1415 - val_acc: 0.9771
Epoch 48/50
acc: 0.9974 - val_loss: 0.1668 - val_acc: 0.9743
Epoch 49/50
acc: 0.9983 - val loss: 0.1501 - val acc: 0.9762
Epoch 50/50
acc: 0.9995 - val_loss: 0.1290 - val_acc: 0.9816
```

```
In [23]: score=model_8.evaluate(x_test,y_test,verbose=0)
    print('test error',score[0])
    print('test accuracy',score[1])

x=list(range(1,51))
    v18=history8.history['val_loss']
    18=history8.history['loss']
    plt.plot(x1,v18,'y',label='validation error')
    plt.plot(x1,18,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

test error 0.12899938047406814 test accuracy 0.9816



3. Five layers MNPs with activation 'relu' and optimizer 'adam'

case 1-with only BatchNormalization

```
In [24]: model_9=Sequential()
    model_9.add(Dense(512,input_dim=784,activation='relu',kernel_initializer='he_n
        ormal'))
    model_9.add(Dense(256,activation='relu',kernel_initializer='he_normal'))
    model_9.add(Dense(128,activation='relu',kernel_initializer='he_normal'))
    model_9.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
    model_9.add(Dense(32,activation='relu',kernel_initializer='he_normal'))
    model_9.add(Dense(32,activation='relu',kernel_initializer='he_normal'))
    model_9.add(Dense(10,activation='softmax'))
    model_9.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
    history9=model_9.fit(x_train,y_train,epochs=30,batch_size=128,verbose=1,validation_data=(x_test,y_test))
```

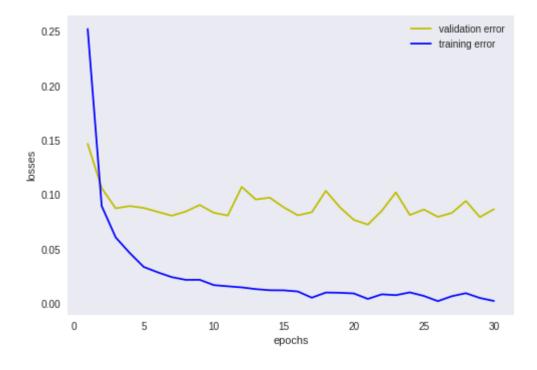
```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
acc: 0.9286 - val loss: 0.1468 - val acc: 0.9559
Epoch 2/30
acc: 0.9725 - val loss: 0.1062 - val acc: 0.9673
Epoch 3/30
acc: 0.9810 - val loss: 0.0877 - val acc: 0.9715
Epoch 4/30
acc: 0.9852 - val loss: 0.0898 - val acc: 0.9724
acc: 0.9887 - val_loss: 0.0880 - val_acc: 0.9742
Epoch 6/30
acc: 0.9904 - val_loss: 0.0845 - val_acc: 0.9768
Epoch 7/30
60000/60000 [================ ] - 8s 134us/step - loss: 0.0245 -
acc: 0.9922 - val loss: 0.0809 - val acc: 0.9777
Epoch 8/30
60000/60000 [================ ] - 8s 136us/step - loss: 0.0220 -
acc: 0.9929 - val loss: 0.0848 - val acc: 0.9773
Epoch 9/30
60000/60000 [============ ] - 8s 139us/step - loss: 0.0221 -
acc: 0.9927 - val_loss: 0.0907 - val_acc: 0.9761
Epoch 10/30
acc: 0.9943 - val_loss: 0.0836 - val_acc: 0.9776
60000/60000 [================ ] - 8s 137us/step - loss: 0.0161 -
acc: 0.9948 - val_loss: 0.0811 - val_acc: 0.9780
Epoch 12/30
60000/60000 [=============== ] - 8s 138us/step - loss: 0.0151 -
acc: 0.9948 - val loss: 0.1074 - val acc: 0.9752
Epoch 13/30
60000/60000 [================ ] - 8s 137us/step - loss: 0.0135 -
acc: 0.9957 - val loss: 0.0958 - val acc: 0.9783
Epoch 14/30
60000/60000 [=============== ] - 8s 138us/step - loss: 0.0125 -
acc: 0.9959 - val_loss: 0.0975 - val_acc: 0.9778
Epoch 15/30
60000/60000 [================ ] - 8s 137us/step - loss: 0.0124 -
acc: 0.9957 - val loss: 0.0885 - val acc: 0.9794
Epoch 16/30
60000/60000 [=============== ] - 8s 142us/step - loss: 0.0113 -
acc: 0.9963 - val loss: 0.0813 - val acc: 0.9806
Epoch 17/30
acc: 0.9980 - val loss: 0.0841 - val acc: 0.9798
Epoch 18/30
60000/60000 [=============== ] - 8s 138us/step - loss: 0.0103 -
acc: 0.9965 - val_loss: 0.1037 - val_acc: 0.9783
Epoch 19/30
```

```
acc: 0.9965 - val loss: 0.0887 - val acc: 0.9793
Epoch 20/30
60000/60000 [================= ] - 8s 141us/step - loss: 0.0095 -
acc: 0.9971 - val loss: 0.0771 - val acc: 0.9811
Epoch 21/30
60000/60000 [============= ] - 8s 139us/step - loss: 0.0045 -
acc: 0.9988 - val loss: 0.0728 - val acc: 0.9836
Epoch 22/30
60000/60000 [================= ] - 8s 137us/step - loss: 0.0087 -
acc: 0.9971 - val loss: 0.0854 - val acc: 0.9793
Epoch 23/30
60000/60000 [============= ] - 8s 141us/step - loss: 0.0079 -
acc: 0.9975 - val loss: 0.1023 - val acc: 0.9780
Epoch 24/30
60000/60000 [================ ] - 8s 139us/step - loss: 0.0105 -
acc: 0.9966 - val loss: 0.0816 - val acc: 0.9812
Epoch 25/30
acc: 0.9978 - val loss: 0.0866 - val acc: 0.9797
Epoch 26/30
60000/60000 [============= ] - 8s 135us/step - loss: 0.0025 -
acc: 0.9992 - val loss: 0.0798 - val acc: 0.9803
Epoch 27/30
acc: 0.9979 - val_loss: 0.0835 - val_acc: 0.9820
Epoch 28/30
acc: 0.9969 - val_loss: 0.0944 - val_acc: 0.9807
Epoch 29/30
acc: 0.9982 - val_loss: 0.0796 - val_acc: 0.9829
Epoch 30/30
60000/60000 [================ ] - 8s 140us/step - loss: 0.0026 -
acc: 0.9992 - val loss: 0.0869 - val acc: 0.9833
```

```
In [32]: score=model_9.evaluate(x_test,y_test,verbose=1)
    print('test error',score[0])
    print('test accuracy',score[1])

x=list(range(1,31))
    v19=history9.history['val_loss']
    l9=history9.history['loss']
    plt.plot(x,v19,'y',label='validation error')
    plt.plot(x,19,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

```
10000/10000 [============] - 1s 84us/step test error 0.08685781101842004 test accuracy 0.9833
```



case 2-with Dropout only

```
In [26]:
         model 10=Sequential()
         model_10.add(Dense(512,input_dim=784,activation='relu',kernel_initializer='he_
         normal'))
         model 10.add(Dropout(0.5))
         model_10.add(Dense(256,activation='relu',kernel_initializer='he_normal'))
         model 10.add(Dropout(0.5))
         model 10.add(Dense(128,activation='relu',kernel initializer='he normal'))
         model 10.add(Dropout(0.5))
         model_10.add(Dense(64,activation='relu',kernel_initializer='he_normal'))
         model_10.add(Dropout(0.5))
         model 10.add(Dense(32,activation='relu',kernel initializer='he normal'))
         model_10.add(Dense(10,activation='softmax'))
         model_10.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['ac
         curacy'])
         history10=model_10.fit(x_train,y_train,epochs=30,batch_size=108,verbose=1,vali
         dation_data=(x_test,y_test))
```

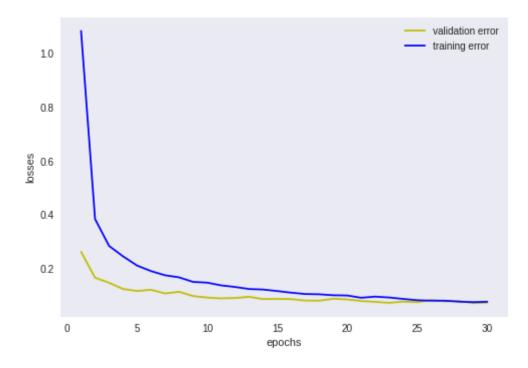
```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
- acc: 0.6271 - val loss: 0.2629 - val acc: 0.9330
Epoch 2/30
60000/60000 [==================== ] - 10s 166us/step - loss: 0.3844
- acc: 0.8969 - val loss: 0.1667 - val acc: 0.9544
Epoch 3/30
- acc: 0.9281 - val loss: 0.1480 - val acc: 0.9615
Epoch 4/30
- acc: 0.9408 - val loss: 0.1246 - val acc: 0.9679
- acc: 0.9477 - val_loss: 0.1174 - val_acc: 0.9681
Epoch 6/30
60000/60000 [================== ] - 10s 169us/step - loss: 0.1911
- acc: 0.9535 - val_loss: 0.1217 - val_acc: 0.9701
Epoch 7/30
60000/60000 [============ ] - 10s 170us/step - loss: 0.1757
- acc: 0.9573 - val_loss: 0.1082 - val_acc: 0.9742
Epoch 8/30
- acc: 0.9588 - val loss: 0.1144 - val acc: 0.9714
Epoch 9/30
60000/60000 [=================== ] - 11s 180us/step - loss: 0.1511
- acc: 0.9629 - val_loss: 0.0985 - val_acc: 0.9760
Epoch 10/30
60000/60000 [================ ] - 10s 169us/step - loss: 0.1483
- acc: 0.9637 - val_loss: 0.0930 - val_acc: 0.9757
Epoch 11/30
- acc: 0.9659 - val loss: 0.0901 - val acc: 0.9780
Epoch 12/30
60000/60000 [============= ] - 10s 170us/step - loss: 0.1322
- acc: 0.9678 - val loss: 0.0913 - val acc: 0.9781
Epoch 13/30
60000/60000 [==================== ] - 10s 170us/step - loss: 0.1247
- acc: 0.9685 - val loss: 0.0957 - val acc: 0.9777
Epoch 14/30
60000/60000 [=================== ] - 10s 168us/step - loss: 0.1228
- acc: 0.9698 - val_loss: 0.0873 - val_acc: 0.9787
Epoch 15/30
60000/60000 [================== ] - 10s 166us/step - loss: 0.1172
- acc: 0.9711 - val loss: 0.0881 - val acc: 0.9792
Epoch 16/30
60000/60000 [================== ] - 10s 166us/step - loss: 0.1111
- acc: 0.9726 - val loss: 0.0871 - val acc: 0.9803
Epoch 17/30
60000/60000 [============= ] - 10s 168us/step - loss: 0.1062
- acc: 0.9734 - val loss: 0.0819 - val acc: 0.9818
Epoch 18/30
60000/60000 [============= ] - 10s 171us/step - loss: 0.1051
- acc: 0.9739 - val loss: 0.0815 - val acc: 0.9820
Epoch 19/30
60000/60000 [======================== ] - 10s 171us/step - loss: 0.1016
```

```
- acc: 0.9753 - val loss: 0.0886 - val acc: 0.9802
Epoch 20/30
60000/60000 [============== ] - 10s 168us/step - loss: 0.1006
- acc: 0.9754 - val loss: 0.0860 - val acc: 0.9798
Epoch 21/30
60000/60000 [============= ] - 10s 168us/step - loss: 0.0923
- acc: 0.9771 - val loss: 0.0803 - val acc: 0.9811
Epoch 22/30
60000/60000 [============== ] - 10s 167us/step - loss: 0.0964
- acc: 0.9759 - val loss: 0.0770 - val acc: 0.9815
Epoch 23/30
60000/60000 [============= ] - 10s 168us/step - loss: 0.0932
- acc: 0.9765 - val loss: 0.0734 - val acc: 0.9830
Epoch 24/30
- acc: 0.9781 - val loss: 0.0779 - val acc: 0.9820
Epoch 25/30
60000/60000 [============ ] - 10s 169us/step - loss: 0.0831
- acc: 0.9797 - val loss: 0.0758 - val acc: 0.9837
Epoch 26/30
60000/60000 [============ ] - 10s 167us/step - loss: 0.0815
- acc: 0.9795 - val loss: 0.0836 - val acc: 0.9797
Epoch 27/30
60000/60000 [============ ] - 10s 165us/step - loss: 0.0814
- acc: 0.9797 - val loss: 0.0789 - val acc: 0.9824
Epoch 28/30
60000/60000 [=========== ] - 10s 167us/step - loss: 0.0777
- acc: 0.9804 - val_loss: 0.0802 - val_acc: 0.9814
Epoch 29/30
- acc: 0.9814 - val_loss: 0.0723 - val_acc: 0.9828
Epoch 30/30
- acc: 0.9810 - val loss: 0.0747 - val acc: 0.9830
```

```
In [27]: score=model_10.evaluate(x_test,y_test,verbose=0)
    print('test error',score[0])
    print('test accuracy',score[1])

vl10=history10.history['val_loss']
    l10=history10.history['loss']
    plt.plot(x,vl10,'y',label='validation error')
    plt.plot(x,l10,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

test error 0.0746986644620818 test accuracy 0.983



case 3-with both Dropout and BatchNormalization

```
In [28]:
         model 11=Sequential()
         model_11.add(Dense(512,input_dim=784,activation='relu',kernel_initializer='he_
         normal'))
         model 11.add(Dropout(0.5))
         model_11.add(Dense(256,activation='relu',kernel_initializer='he_normal'))
         model 11.add(Dropout(0.5))
         model 11.add(Dense(128,activation='relu',kernel initializer='he normal'))
         model 11.add(Dropout(0.5))
         model 11.add(Dense(64,activation='relu',kernel initializer='he normal'))
         model_11.add(BatchNormalization())
         model 11.add(Dropout(0.5))
         model_11.add(Dense(32,activation='relu',kernel_initializer='he_normal'))
         model 11.add(Dense(10,activation='softmax'))
         model_11.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['ac
         curacy'])
         history11=model_11.fit(x_train,y_train,epochs=30,batch_size=128,verbose=1,vali
         dation_data=(x_test,y_test))
```

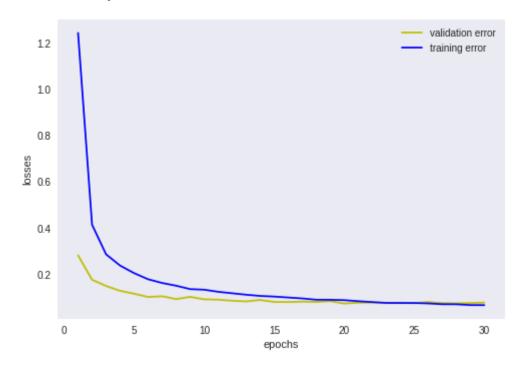
```
Train on 60000 samples, validate on 10000 samples
Epoch 1/30
- acc: 0.5913 - val loss: 0.2819 - val acc: 0.9219
Epoch 2/30
- acc: 0.8829 - val loss: 0.1760 - val acc: 0.9506
Epoch 3/30
- acc: 0.9228 - val loss: 0.1490 - val acc: 0.9606
Epoch 4/30
- acc: 0.9393 - val loss: 0.1280 - val acc: 0.9651
- acc: 0.9468 - val_loss: 0.1155 - val_acc: 0.9694
Epoch 6/30
- acc: 0.9542 - val_loss: 0.1014 - val_acc: 0.9735
Epoch 7/30
60000/60000 [============ ] - 9s 158us/step - loss: 0.1620 -
acc: 0.9591 - val_loss: 0.1045 - val_acc: 0.9718
Epoch 8/30
60000/60000 [================== ] - 10s 162us/step - loss: 0.1501
- acc: 0.9626 - val loss: 0.0922 - val acc: 0.9754
Epoch 9/30
60000/60000 [==================== ] - 10s 161us/step - loss: 0.1354
- acc: 0.9656 - val loss: 0.1019 - val acc: 0.9732
Epoch 10/30
60000/60000 [=================== ] - 10s 166us/step - loss: 0.1329
- acc: 0.9670 - val_loss: 0.0912 - val_acc: 0.9769
Epoch 11/30
acc: 0.9688 - val_loss: 0.0897 - val_acc: 0.9775
Epoch 12/30
60000/60000 [============= ] - 10s 162us/step - loss: 0.1174
- acc: 0.9696 - val loss: 0.0851 - val acc: 0.9777
Epoch 13/30
60000/60000 [=================== ] - 10s 159us/step - loss: 0.1111
- acc: 0.9716 - val loss: 0.0820 - val acc: 0.9784
Epoch 14/30
acc: 0.9734 - val_loss: 0.0884 - val_acc: 0.9782
Epoch 15/30
60000/60000 [============ ] - 9s 156us/step - loss: 0.1031 -
acc: 0.9739 - val loss: 0.0798 - val acc: 0.9800
Epoch 16/30
acc: 0.9749 - val loss: 0.0795 - val acc: 0.9794
Epoch 17/30
acc: 0.9762 - val loss: 0.0812 - val acc: 0.9798
Epoch 18/30
60000/60000 [============ ] - 10s 161us/step - loss: 0.0894
- acc: 0.9775 - val loss: 0.0798 - val acc: 0.9800
Epoch 19/30
60000/60000 [======================== ] - 10s 160us/step - loss: 0.0892
```

```
- acc: 0.9766 - val loss: 0.0835 - val acc: 0.9797
Epoch 20/30
60000/60000 [============== ] - 10s 160us/step - loss: 0.0876
- acc: 0.9780 - val loss: 0.0728 - val acc: 0.9823
Epoch 21/30
60000/60000 [============== ] - 10s 162us/step - loss: 0.0831
- acc: 0.9784 - val loss: 0.0766 - val acc: 0.9816
Epoch 22/30
60000/60000 [============== ] - 10s 162us/step - loss: 0.0792
- acc: 0.9798 - val loss: 0.0772 - val acc: 0.9824
60000/60000 [============== ] - 9s 158us/step - loss: 0.0757 -
acc: 0.9816 - val loss: 0.0755 - val acc: 0.9817
Epoch 24/30
- acc: 0.9808 - val loss: 0.0768 - val acc: 0.9825
Epoch 25/30
60000/60000 [============ ] - 10s 162us/step - loss: 0.0753
- acc: 0.9812 - val loss: 0.0743 - val acc: 0.9828
Epoch 26/30
60000/60000 [============ ] - 10s 162us/step - loss: 0.0734
- acc: 0.9813 - val_loss: 0.0796 - val acc: 0.9811
Epoch 27/30
60000/60000 [============= ] - 10s 161us/step - loss: 0.0700
- acc: 0.9822 - val loss: 0.0747 - val acc: 0.9813
Epoch 28/30
60000/60000 [============ ] - 10s 160us/step - loss: 0.0698
- acc: 0.9829 - val_loss: 0.0736 - val_acc: 0.9829
Epoch 29/30
acc: 0.9835 - val_loss: 0.0754 - val_acc: 0.9819
Epoch 30/30
60000/60000 [============ ] - 10s 159us/step - loss: 0.0660
- acc: 0.9831 - val loss: 0.0769 - val acc: 0.9814
```

```
In [29]: score=model_11.evaluate(x_test,y_test,verbose=0)
    print('test error',score[0])
    print('test accuracy',score[1])

x=list(range(1,31))
    vl11=history11.history['val_loss']
    l11=history11.history['loss']
    plt.plot(x,vl11,'y',label='validation error')
    plt.plot(x,l11,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

test error 0.07691529653224861 test accuracy 0.9814



case 4-without Dropout and BatchNormalization

```
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
- acc: 0.9261 - val loss: 0.1293 - val acc: 0.9584
Epoch 2/50
acc: 0.9728 - val loss: 0.0848 - val acc: 0.9760
Epoch 3/50
60000/60000 [================= ] - 8s 134us/step - loss: 0.0609 -
acc: 0.9805 - val loss: 0.0846 - val acc: 0.9726
Epoch 4/50
acc: 0.9857 - val loss: 0.0952 - val acc: 0.9730
acc: 0.9891 - val_loss: 0.0833 - val_acc: 0.9778
Epoch 6/50
acc: 0.9897 - val_loss: 0.0902 - val_acc: 0.9739
Epoch 7/50
60000/60000 [============ ] - 8s 136us/step - loss: 0.0241 -
acc: 0.9923 - val_loss: 0.0857 - val_acc: 0.9777
Epoch 8/50
acc: 0.9923 - val loss: 0.0883 - val acc: 0.9771
Epoch 9/50
acc: 0.9936 - val_loss: 0.0757 - val_acc: 0.9820
Epoch 10/50
60000/60000 [================ ] - 8s 136us/step - loss: 0.0143 -
acc: 0.9955 - val_loss: 0.0914 - val_acc: 0.9804
60000/60000 [================ ] - 8s 140us/step - loss: 0.0208 -
acc: 0.9939 - val loss: 0.0866 - val acc: 0.9807
Epoch 12/50
60000/60000 [================ ] - 8s 142us/step - loss: 0.0159 -
acc: 0.9946 - val loss: 0.0857 - val acc: 0.9809
Epoch 13/50
acc: 0.9952 - val loss: 0.0848 - val acc: 0.9799
Epoch 14/50
acc: 0.9961 - val_loss: 0.0834 - val_acc: 0.9787
Epoch 15/50
60000/60000 [================ ] - 8s 135us/step - loss: 0.0138 -
acc: 0.9959 - val loss: 0.0763 - val acc: 0.9823
Epoch 16/50
60000/60000 [================ ] - 8s 132us/step - loss: 0.0106 -
acc: 0.9966 - val loss: 0.0950 - val acc: 0.9787
Epoch 17/50
acc: 0.9956 - val loss: 0.0870 - val acc: 0.9803
Epoch 18/50
60000/60000 [=============== ] - 8s 133us/step - loss: 0.0093 -
acc: 0.9970 - val loss: 0.0816 - val acc: 0.9835
Epoch 19/50
```

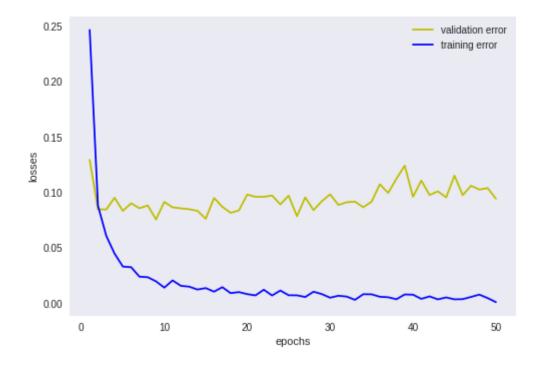
```
acc: 0.9969 - val loss: 0.0838 - val acc: 0.9819
Epoch 20/50
acc: 0.9973 - val loss: 0.0980 - val acc: 0.9797
Epoch 21/50
60000/60000 [============== ] - 8s 134us/step - loss: 0.0073 -
acc: 0.9979 - val loss: 0.0961 - val acc: 0.9831
Epoch 22/50
60000/60000 [================= ] - 8s 135us/step - loss: 0.0124 -
acc: 0.9962 - val loss: 0.0960 - val acc: 0.9811
Epoch 23/50
60000/60000 [============== ] - 8s 138us/step - loss: 0.0072 -
acc: 0.9979 - val loss: 0.0971 - val acc: 0.9814
Epoch 24/50
acc: 0.9968 - val loss: 0.0892 - val acc: 0.9813
Epoch 25/50
60000/60000 [================ ] - 8s 132us/step - loss: 0.0074 -
acc: 0.9980 - val loss: 0.0970 - val acc: 0.9814
Epoch 26/50
60000/60000 [============= ] - 8s 140us/step - loss: 0.0073 -
acc: 0.9981 - val loss: 0.0786 - val acc: 0.9843
Epoch 27/50
acc: 0.9983 - val_loss: 0.0955 - val_acc: 0.9813
Epoch 28/50
acc: 0.9971 - val_loss: 0.0840 - val_acc: 0.9832
Epoch 29/50
acc: 0.9975 - val_loss: 0.0920 - val_acc: 0.9816
Epoch 30/50
60000/60000 [================ ] - 8s 140us/step - loss: 0.0052 -
acc: 0.9988 - val loss: 0.0982 - val acc: 0.9839
Epoch 31/50
acc: 0.9982 - val_loss: 0.0887 - val_acc: 0.9825
Epoch 32/50
acc: 0.9981 - val_loss: 0.0911 - val_acc: 0.9839
Epoch 33/50
60000/60000 [================ ] - 8s 136us/step - loss: 0.0033 -
acc: 0.9992 - val_loss: 0.0917 - val_acc: 0.9851
60000/60000 [============= ] - 8s 132us/step - loss: 0.0083 -
acc: 0.9977 - val loss: 0.0866 - val acc: 0.9839
Epoch 35/50
acc: 0.9977 - val loss: 0.0916 - val acc: 0.9835
Epoch 36/50
60000/60000 [================ ] - 8s 133us/step - loss: 0.0060 -
acc: 0.9984 - val loss: 0.1073 - val acc: 0.9789
Epoch 37/50
60000/60000 [================ ] - 8s 140us/step - loss: 0.0056 -
acc: 0.9986 - val_loss: 0.0995 - val_acc: 0.9816
Epoch 38/50
60000/60000 [================ ] - 8s 138us/step - loss: 0.0038 -
```

```
acc: 0.9990 - val loss: 0.1122 - val acc: 0.9811
Epoch 39/50
acc: 0.9978 - val loss: 0.1241 - val acc: 0.9784
Epoch 40/50
60000/60000 [============== ] - 8s 135us/step - loss: 0.0079 -
acc: 0.9981 - val loss: 0.0961 - val acc: 0.9842
Epoch 41/50
60000/60000 [================ ] - 8s 134us/step - loss: 0.0041 -
acc: 0.9990 - val loss: 0.1106 - val acc: 0.9817
Epoch 42/50
60000/60000 [============== ] - 8s 133us/step - loss: 0.0064 -
acc: 0.9984 - val loss: 0.0977 - val acc: 0.9838
Epoch 43/50
acc: 0.9992 - val loss: 0.1009 - val acc: 0.9828
Epoch 44/50
60000/60000 [================ ] - 8s 136us/step - loss: 0.0054 -
acc: 0.9985 - val loss: 0.0955 - val acc: 0.9826
Epoch 45/50
60000/60000 [============= ] - 8s 137us/step - loss: 0.0037 -
acc: 0.9991 - val loss: 0.1151 - val acc: 0.9801
Epoch 46/50
60000/60000 [================= ] - 8s 134us/step - loss: 0.0039 -
acc: 0.9989 - val_loss: 0.0975 - val acc: 0.9848
Epoch 47/50
acc: 0.9983 - val_loss: 0.1060 - val_acc: 0.9810
Epoch 48/50
acc: 0.9980 - val_loss: 0.1025 - val_acc: 0.9829
Epoch 49/50
acc: 0.9988 - val loss: 0.1039 - val acc: 0.9826
Epoch 50/50
60000/60000 [================= ] - 8s 138us/step - loss: 0.0012 -
acc: 0.9997 - val_loss: 0.0944 - val_acc: 0.9851
```

```
In [31]: score=model_12.evaluate(x_test,y_test,verbose=0)
    print('test error',score[0])
    print('test accuracy',score[1])

x1=list(range(1,51))
    vl12=history12.history['val_loss']
    l12=history12.history['loss']
    plt.plot(x1,vl12,'y',label='validation error')
    plt.plot(x1,l12,'b',label='training error')
    plt.legend()
    plt.xlabel('epochs')
    plt.ylabel('losses')
    plt.grid()
    plt.show()
```

test error 0.0943650556378368 test accuracy 0.9851



2-layer	Test Error	Test Accuracy
Batch Normal	0.1016	0.9793
DropOut	0.7725	0.9788
with Batch-Normal & DropOut	0.0942	0.974
without Batch-Normal & DropOut	0.1314	0.9802

3-layer	Test Error	Test Accuracy
Batch Normal	0.1074	0.9781
DropOut	0.0959	0.9741
with Batch-Normal & DropOut	0.0878	0.9751
without Batch-Normal & DropOut	0.1289	0.9816

5-layer	Test Error	Test Accuracy
Batch Normal	0.0868	0.9833
DropOut	0.0746	0.983
with Batch-Normal & DropOut	0.0769	0.98114
without Batch-Normal & DropOut	0.0943	0.9851

Conclusion-

- 1.As we increase number of layer it may increase accuracy.
- 2.As we increase number of neurons in layer it may increase accuracy.
- 3.without Dropout and BatchNormalization model goes overfit.
- 4.with Dropout or BatchNormalization or both it give appox same accuracy.
- 5.The model with five layers, activation-relu and optimizer-adam, with dropout and Batch normalization gives the best with accuracy 0.9814