### **Question 1**

# Minimization of Negative Log Likelihood with Softmax outputs

We are given the negative log likelihood function that we want to minimize:

$$J(w) = -\log(\prod_{n} \prod_{m=0}^{9} p(t_n = m | x_n; w))$$

where there are 0 to 9 classes.

Now let us define that  $p(t_m|x;w)$  is a categorical probability distribution that is :

$$p(t_m|x;w) = y_m(x;w)^{t_m}$$

following the constraints that  $t_m \in \{0, 1\}$  and  $\sum_m t_m = 1$ 

As we know that we can interpret the output of the softmax as the probabilities that a certain set of features belongs to a certain class. Let us use the probability nodes as softmax output nodes

$$y_i(x; w) = \frac{\exp^{y_i(x;w)}}{\sum_{i=0}^m \exp^{y_j(x;w)}}$$

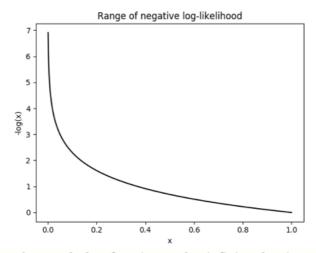
This ensures that sum of probability of each feature sample belonging to one of the 10 classes equals to 1

The negative log likelihood can be reduced to:

$$J(w) = -\sum_{n} \sum_{m=0}^{9} \log p(t_n = m | x_n; w)$$

$$J(w) = -\sum_{n} \sum_{m=0}^{9} t_{m}^{(n)} \log(y_{m}^{(n)}(x^{(n)}; w))$$

Now the nature of curve of negative log likelihood is:



**Figure:** The loss function reaches infinity when input is 0, and reaches 0 when input is 1.

This means for the curve to reach its minimum we have to generate the likelihood of the correct label as close to 1 as possible.

Now noting that  $t_m$  is one-hot encoded vector thus reducing all terms of J(w) to 0 other than the target(true) class say 'k' for every feature vector i.e.

$$J(w) = -\sum_{n} 1 * \log(y_k^{(n)}(x^{(n)}; w)) = -\log(\prod_{n} (y_k^{(n)}(x^{(n)}; w)))$$

Thus the weights must be trained in such a way that the likelihood softmax(probability) of the target label is maximum for each feature vector (i.e. close to 1).

Hence it is proved that a neural network to maximize the log likelihood of label is one that has softmax output nodes and minimizes the criterion function of the negative log probability of training data set

## Interpretation of L2 regularization

Let us assume that our input is linearly correlated with our output using the following equation

$$y_i = wx_i + \epsilon$$

where w is the set of weights and  $\epsilon$  is the random noise that also follows normal distribution with 0 mean and constant variance  $\sigma^2$ .

This gives us the likelihood that we want to maximize:

$$\prod_{i=1}^{N} \mathcal{N}(y_i; x_i, w, \sigma^2)$$

Now,let us assume that w has the Gaussian prior  $\mathcal{N}(0,\alpha^{-1})$ 

From Bayes Rule we know:

$$P(w|D) = \frac{P(D|w)P(w)}{P(D)}$$

$$P(w|D) \approx P(D|w)P(w)$$

Thus the posterior probability is:

$$P(w|D) = \prod_{i=1}^{N} \mathcal{N}(y_i; x_i, w, \sigma^2) \mathcal{N}(0, \alpha^{-1})$$

Taking logarithm of the posterior:

$$\log P(w|D) = \sum_{i=1}^{N} \log \mathcal{N}(y_i; x_i, w, \sigma^2) + \log \mathcal{N}(0, \alpha^{-1}) + constant$$

Ignoring the constant we need to maximize over w

$$-\frac{1}{\sigma^2}\sum_{i=1}^{N}(y_i-wx_i)^2-\alpha w^2$$

This is the same as minimizing the cost function with L2 regularization with  $\alpha$  being the regularization constant.

Thus we have proved optimizing model weights to minimize loss function with L2 regularization is equivalent to finding the weights that are most likely under a posterior distribution evaluated using Bayes rule, with a zero-mean independent Gaussian weights prior

Using TensorFlow backend.

```
In [3]: x_train.shape,y_train.shape
Out[3]: ((60000, 28, 28), (60000,))
In [4]: x_test.shape,y_test.shape
Out[4]: ((10000, 28, 28), (10000,))
```

#### Reshaping and normalizing the data

### Generating 1000 images with 100 images per class

```
In [8]: def generate_minidata(x,y):
    mini=[]

    for i in range(10):
        count=0
        for j in range(len(x)):
            if y[j]==i:
                 mini.append(np.array([i]+list(x[j])))
                 count+=1
            if count==100:
                 break

    mini=np.array(mini)
    np.random.shuffle(mini)
    #return(mini[:,0].reshape((len(mini),1)),mini[:,1:])
    return(mini[:,0],mini[:,1:])
```

```
In [9]: mini_trainlabels,mini_train=generate_minidata(x_train,y_train)
```

```
In [10]: mini_testlabels,mini_test=generate_minidata(x_test,y_test)
```

```
In [11]: mini_train.shape,mini_trainlabels.shape
Out[11]: ((1000, 784), (1000,))
In [12]: mini_test.shape,mini_testlabels.shape
Out[12]: ((1000, 784), (1000,))
```

#### One hot encoding

```
In [15]: from sklearn.preprocessing import OneHotEncoder
    ohe = OneHotEncoder()
    train_labels = ohe.fit_transform(mini_trainlabels.reshape((1000,1))
    ).toarray()
    test_labels = ohe.fit_transform(mini_testlabels.reshape((1000,1))).
    toarray()

In [16]: print(mini_test.shape,test_labels.shape)
    (1000, 784) (1000, 10)

In [17]: print(mini_train.shape,train_labels.shape)
    (1000, 784) (1000, 10)
```

## Question 2)a) 1 Hidden Layer

#### One hidden layer without regularization

```
In [19]: model=neural_net_1hidden(False)
    history = model.fit(mini_train, train_labels, epochs=30, batch_size
    =10,validation_data=(mini_test,test_labels))
```

Model: "sequential"

Layer	(type)	Output	Shape	Param #
dense	(Dense)	(None,	30)	23550
dense_	1 (Dense)	(None,	10)	310
=====		======		========
$m_{0}+1$	narame. 23 860			

Total params: 23,860 Trainable params: 23,860 Non-trainable params: 0

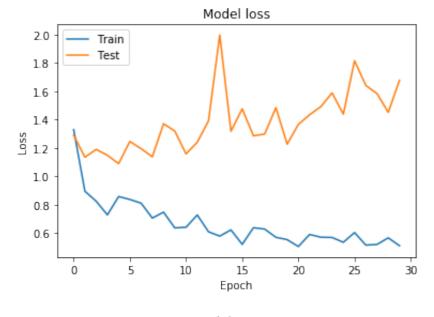
```
s: 0.8239 - accuracy: 0.7640 - val loss: 1.1899 - val accuracy: 0.
6460
Epoch 4/30
1000/1000 [=========== ] - 0s 312us/sample - los
s: 0.7272 - accuracy: 0.7710 - val loss: 1.1487 - val accuracy: 0.
6830
Epoch 5/30
s: 0.8568 - accuracy: 0.7420 - val loss: 1.0889 - val accuracy: 0.
6760
Epoch 6/30
1000/1000 [============ ] - 0s 327us/sample - los
s: 0.8361 - accuracy: 0.7670 - val loss: 1.2460 - val accuracy: 0.
6770
Epoch 7/30
s: 0.8100 - accuracy: 0.7710 - val loss: 1.1949 - val accuracy: 0.
6740
Epoch 8/30
s: 0.7046 - accuracy: 0.8110 - val loss: 1.1376 - val accuracy: 0.
6910
Epoch 9/30
s: 0.7463 - accuracy: 0.7980 - val_loss: 1.3716 - val_accuracy: 0.
6540
Epoch 10/30
s: 0.6351 - accuracy: 0.8250 - val loss: 1.3194 - val accuracy: 0.
6680
Epoch 11/30
s: 0.6398 - accuracy: 0.8300 - val loss: 1.1581 - val accuracy: 0.
6810
Epoch 12/30
1000/1000 [============ ] - 0s 313us/sample - los
s: 0.7264 - accuracy: 0.8040 - val loss: 1.2404 - val accuracy: 0.
6870
Epoch 13/30
s: 0.6077 - accuracy: 0.8230 - val loss: 1.3915 - val accuracy: 0.
6760
Epoch 14/30
s: 0.5773 - accuracy: 0.8400 - val loss: 1.9983 - val accuracy: 0.
6000
Epoch 15/30
s: 0.6211 - accuracy: 0.8330 - val loss: 1.3181 - val accuracy: 0.
7010
Epoch 16/30
s: 0.5186 - accuracy: 0.8460 - val loss: 1.4768 - val accuracy: 0.
```

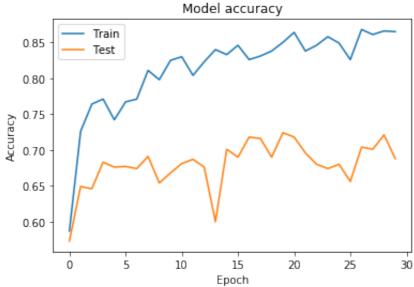
```
6900
Epoch 17/30
s: 0.6365 - accuracy: 0.8260 - val loss: 1.2859 - val accuracy: 0.
7180
Epoch 18/30
s: 0.6268 - accuracy: 0.8310 - val loss: 1.2983 - val accuracy: 0.
7160
Epoch 19/30
1000/1000 [============] - 0s 301us/sample - los
s: 0.5686 - accuracy: 0.8380 - val loss: 1.4851 - val accuracy: 0.
6900
Epoch 20/30
s: 0.5521 - accuracy: 0.8500 - val loss: 1.2269 - val accuracy: 0.
7240
Epoch 21/30
s: 0.5034 - accuracy: 0.8640 - val_loss: 1.3665 - val_accuracy: 0.
7180
Epoch 22/30
s: 0.5890 - accuracy: 0.8380 - val loss: 1.4347 - val accuracy: 0.
6960
Epoch 23/30
s: 0.5691 - accuracy: 0.8460 - val loss: 1.4923 - val accuracy: 0.
6800
Epoch 24/30
s: 0.5675 - accuracy: 0.8580 - val loss: 1.5891 - val accuracy: 0.
6740
Epoch 25/30
s: 0.5334 - accuracy: 0.8490 - val loss: 1.4392 - val accuracy: 0.
6800
Epoch 26/30
s: 0.6023 - accuracy: 0.8260 - val loss: 1.8166 - val accuracy: 0.
6560
Epoch 27/30
s: 0.5137 - accuracy: 0.8680 - val loss: 1.6423 - val accuracy: 0.
7040
Epoch 28/30
s: 0.5182 - accuracy: 0.8610 - val loss: 1.5842 - val accuracy: 0.
7010
Epoch 29/30
1000/1000 [============ ] - 0s 318us/sample - los
s: 0.5649 - accuracy: 0.8660 - val loss: 1.4517 - val accuracy: 0.
7210
```

```
In [20]: import matplotlib.pyplot as plt
         def model training plot():
             # Plot training & validation loss values
             plt.plot(history.history['loss'])
             plt.plot(history.history['val loss'])
             plt.title('Model loss')
             plt.ylabel('Loss')
             plt.xlabel('Epoch')
             plt.legend(['Train', 'Test'], loc='upper left')
             plt.show()
             # Plot training & validation accuracy values
             plt.plot(history.history['accuracy'])
             plt.plot(history.history['val accuracy'])
             plt.title('Model accuracy')
             plt.ylabel('Accuracy')
             plt.xlabel('Epoch')
             plt.legend(['Train', 'Test'], loc='upper left')
             plt.show()
```

#### Plot of criterion categorical cross entropy loss function

```
In [21]: model_training_plot()
```





```
In [22]: def for error plot(model):
             train errors=[]
             test errors=[]
             train accuracy=[]
             test accuracy=[]
             for i in range(30):
                 model.fit(mini train, train labels, epochs=1, batch size=10
         )
                 predictions_train=model.predict_classes(mini_train)
                 predictions test=model.predict classes(mini test)
                 te=0
                 tr=0
                 for i in range(1000):
                      if train labels[i][predictions train[i]]==1:
                      if test labels[i][predictions test[i]]==1:
                          te+=1
                 train errors.append(1-tr/1000)
                 test errors.append(1-te/1000)
                 train accuracy.append(tr/1000)
                 test accuracy.append(te/1000)
             return train_errors,test_errors,train_accuracy,test_accuracy
```

## In [23]: model=neural\_net\_1hidden(False) traine,teste,traina,testa=for\_error\_plot(model)

Model: "sequential 1"

Layer (type)	Output Shape	Param #	
dense_2 (Dense)	(None, 30)	23550	
dense_3 (Dense)	(None, 10)	310	
Total params: 23,860 Trainable params: 23,860 Non-trainable params: 0			
Train on 1000 samples			

s: 1.3550 - accuracy: 0.6010

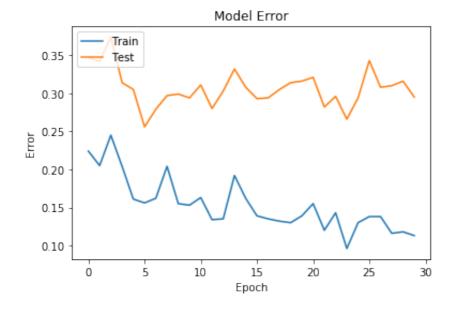
```
Train on 1000 samples
s: 0.8118 - accuracy: 0.7450
Train on 1000 samples
1000/1000 [============ ] - 0s 188us/sample - los
s: 0.8000 - accuracy: 0.7720
Train on 1000 samples
s: 0.7834 - accuracy: 0.7830
Train on 1000 samples
1000/1000 [============] - 0s 204us/sample - los
s: 0.6142 - accuracy: 0.8240
Train on 1000 samples
s: 0.6590 - accuracy: 0.8080
Train on 1000 samples
1000/1000 [============ ] - 0s 213us/sample - los
s: 0.5990 - accuracy: 0.8320
Train on 1000 samples
1000/1000 [============ ] - 0s 188us/sample - los
s: 0.6900 - accuracy: 0.8000
Train on 1000 samples
s: 0.6767 - accuracy: 0.8140
Train on 1000 samples
s: 0.6917 - accuracy: 0.8270
Train on 1000 samples
s: 0.6282 - accuracy: 0.8360
Train on 1000 samples
1000/1000 [============ ] - 0s 180us/sample - los
s: 0.5449 - accuracy: 0.8400
Train on 1000 samples
s: 0.5866 - accuracy: 0.8440
Train on 1000 samples
s: 0.5944 - accuracy: 0.8400
Train on 1000 samples
s: 0.6462 - accuracy: 0.8310
Train on 1000 samples
1000/1000 [============ ] - 0s 183us/sample - los
s: 0.7162 - accuracy: 0.8270
Train on 1000 samples
s: 0.7007 - accuracy: 0.8120
Train on 1000 samples
s: 0.6469 - accuracy: 0.8470
Train on 1000 samples
```

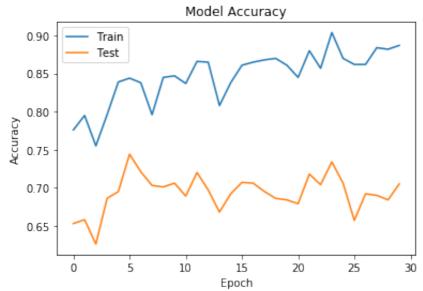
```
s: 0.6453 - accuracy: 0.8280
Train on 1000 samples
s: 0.6324 - accuracy: 0.8440
Train on 1000 samples
s: 0.6496 - accuracy: 0.8350
Train on 1000 samples
1000/1000 [============= ] - 0s 243us/sample - los
s: 0.6711 - accuracy: 0.8280
Train on 1000 samples
s: 0.5601 - accuracy: 0.8580
Train on 1000 samples
s: 0.5569 - accuracy: 0.8570
Train on 1000 samples
1000/1000 [============= ] - 0s 194us/sample - los
s: 0.5587 - accuracy: 0.8500
Train on 1000 samples
s: 0.5770 - accuracy: 0.8680
Train on 1000 samples
s: 0.5083 - accuracy: 0.8690
Train on 1000 samples
s: 0.5445 - accuracy: 0.8620
Train on 1000 samples
s: 0.5288 - accuracy: 0.8670
Train on 1000 samples
s: 0.4683 - accuracy: 0.8690
```

```
In [24]: import matplotlib.pyplot as plt
         def zero one error plot(x,y,a,b):
             # Plot training & validation loss values
             plt.plot(x)
             plt.plot(y)
             plt.title('Model Error')
             plt.ylabel('Error')
             plt.xlabel('Epoch')
             plt.legend(['Train', 'Test'], loc='upper left')
             plt.show()
             plt.plot(a)
             plt.plot(b)
             plt.title('Model Accuracy')
             plt.ylabel('Accuracy')
             plt.xlabel('Epoch')
             plt.legend(['Train', 'Test'], loc='upper left')
             plt.show()
```

#### Plot of zero-one error

```
In [25]: zero_one_error_plot(traine,teste,traina,testa)
```





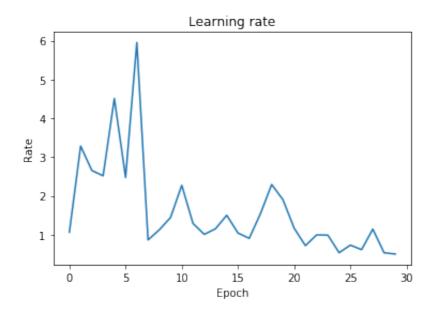
```
In [26]: def learning rate plot(params, model):
             learning rate=[]
              for i in range(30):
                  old weights=model.get weights()
                 model.fit(mini train, train labels, epochs=1, batch size=10
         )
                 new weights=model.get weights()
                 summation=0
                  for j in range(len(new weights)):
                      diff=np.sum(np.absolute((old weights[j]-new weights[j])
         /new_weights[j]))
                      summation+=diff
                  learning rate.append(summation/params)
             # Plot learning rate
             plt.plot(learning rate)
             plt.title('Learning rate')
             plt.ylabel('Rate')
             plt.xlabel('Epoch')
             plt.show()
```

#### Learning speed of hidden layer

```
In [28]: model=neural net 1hidden(False)
     learning rate plot(23860, model)
     Model: "sequential 2"
     Layer (type)
                       Output Shape
                                       Param #
     ______
     dense 4 (Dense)
                       (None, 30)
                                       23550
     dense 5 (Dense)
                       (None, 10)
                                       310
     ______
     Total params: 23,860
     Trainable params: 23,860
     Non-trainable params: 0
     Train on 1000 samples
     s: 1.3328 - accuracy: 0.5860
     Train on 1000 samples
     s: 0.8527 - accuracy: 0.7260
```

```
Train on 1000 samples
s: 0.7924 - accuracy: 0.7580
Train on 1000 samples
1000/1000 [============ ] - 0s 178us/sample - los
s: 0.7943 - accuracy: 0.7660
Train on 1000 samples
s: 0.7105 - accuracy: 0.8000
Train on 1000 samples
1000/1000 [============] - 0s 177us/sample - los
s: 0.6744 - accuracy: 0.8120
Train on 1000 samples
s: 0.7136 - accuracy: 0.7940
Train on 1000 samples
1000/1000 [============ ] - 0s 179us/sample - los
s: 0.5927 - accuracy: 0.8290
Train on 1000 samples
s: 0.6010 - accuracy: 0.8250
Train on 1000 samples
1000/1000 [============= ] - 0s 182us/sample - los
s: 0.5567 - accuracy: 0.8310
Train on 1000 samples
s: 0.6206 - accuracy: 0.8300
Train on 1000 samples
s: 0.6232 - accuracy: 0.8190
Train on 1000 samples
1000/1000 [============ ] - 0s 184us/sample - los
s: 0.6614 - accuracy: 0.8210
Train on 1000 samples
s: 0.5912 - accuracy: 0.8470
Train on 1000 samples
s: 0.6023 - accuracy: 0.8310
Train on 1000 samples
s: 0.5788 - accuracy: 0.8460
Train on 1000 samples
1000/1000 [============ ] - 0s 181us/sample - los
s: 0.4843 - accuracy: 0.8600
Train on 1000 samples
1000/1000 [============ ] - 0s 190us/sample - los
s: 0.5590 - accuracy: 0.8540
Train on 1000 samples
s: 0.5239 - accuracy: 0.8590
Train on 1000 samples
```

```
s: 0.6009 - accuracy: 0.8370
Train on 1000 samples
s: 0.5834 - accuracy: 0.8400
Train on 1000 samples
s: 0.5561 - accuracy: 0.8540
Train on 1000 samples
s: 0.5381 - accuracy: 0.8490
Train on 1000 samples
s: 0.5531 - accuracy: 0.8320
Train on 1000 samples
s: 0.5594 - accuracy: 0.8340
Train on 1000 samples
1000/1000 [============ ] - 0s 180us/sample - los
s: 0.4907 - accuracy: 0.8630
Train on 1000 samples
s: 0.5608 - accuracy: 0.8510
Train on 1000 samples
s: 0.5077 - accuracy: 0.8570
Train on 1000 samples
s: 0.4772 - accuracy: 0.8670
Train on 1000 samples
s: 0.4610 - accuracy: 0.8750
```



#### One hidden layer with regularization

#### With criterion function loss

```
In [32]: model=neural_net_1hidden(True)
    history = model.fit(mini_train, train_labels, epochs=30, batch_size
    =10,validation_data=(mini_test,test_labels))
    model_training_plot()
```

Model: "sequential_5"		
Layer (type)	Output Shape	Param #
dense_9 (Dense)	(None, 30)	23550
dense_10 (Dense)	(None, 10)	310
Total params: 23,860 Trainable params: 23,860 Non-trainable params: 0		=======================================
Train on 1000 samples, va	lidate on 1000 samples	
Epoch 1/30 1000/1000 [==================================	<del>-</del>	<del>-</del>
Epoch 2/30 1000/1000 [==================================		<del>-</del>
0910 Epoch 3/30 1000/1000 [==================================		<del>-</del>
s: 4.5490 - accuracy: 0.1 1130 Epoch 4/30	_	
1000/1000 [==================================	<del>-</del>	<del>-</del>
Epoch 5/30 1000/1000 [==================================		<del>-</del>
1910 Epoch 6/30 1000/1000 [==================================	=====] - 0s	328us/sample - los
s: 6.0101 - accuracy: 0.20 1310 Epoch 7/30	060 - val_loss: 6.7398	- val_accuracy: 0.
1000/1000 [========	======] - 0s	387us/sample - los

s: 6.3535 - accuracy: 0.2180 - val\_loss: 5.8183 - val\_accuracy: 0.

1000/1000 [==============] - 0s 352us/sample - los s: 5.4076 - accuracy: 0.1910 - val loss: 6.0906 - val accuracy: 0.

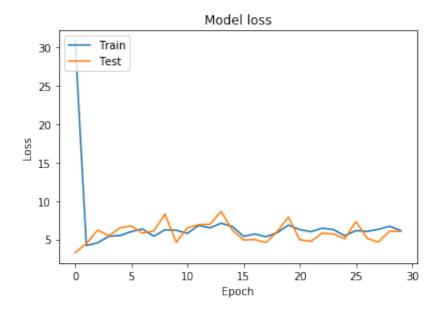
1170

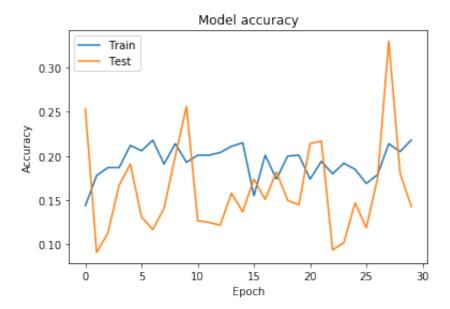
1410

Epoch 8/30

```
Epoch 9/30
s: 6.2441 - accuracy: 0.2140 - val loss: 8.2925 - val accuracy: 0.
2000
Epoch 10/30
s: 6.1735 - accuracy: 0.1930 - val loss: 4.6071 - val accuracy: 0.
2560
Epoch 11/30
s: 5.7798 - accuracy: 0.2010 - val loss: 6.4946 - val accuracy: 0.
1270
Epoch 12/30
s: 6.8121 - accuracy: 0.2010 - val_loss: 6.9213 - val accuracy: 0.
1250
Epoch 13/30
1000/1000 [============ ] - 0s 370us/sample - los
s: 6.4988 - accuracy: 0.2040 - val loss: 6.9691 - val accuracy: 0.
1220
Epoch 14/30
s: 7.1023 - accuracy: 0.2110 - val_loss: 8.6227 - val accuracy: 0.
1580
Epoch 15/30
s: 6.6713 - accuracy: 0.2150 - val loss: 6.2033 - val accuracy: 0.
1370
Epoch 16/30
s: 5.3896 - accuracy: 0.1550 - val loss: 4.9290 - val accuracy: 0.
1740
Epoch 17/30
s: 5.6952 - accuracy: 0.2010 - val loss: 4.9898 - val accuracy: 0.
1510
Epoch 18/30
s: 5.3386 - accuracy: 0.1740 - val loss: 4.5789 - val accuracy: 0.
1820
Epoch 19/30
s: 5.8689 - accuracy: 0.2000 - val loss: 6.0154 - val accuracy: 0.
1500
Epoch 20/30
s: 6.8612 - accuracy: 0.2010 - val loss: 7.9184 - val accuracy: 0.
1450
Epoch 21/30
s: 6.2756 - accuracy: 0.1740 - val loss: 4.9567 - val accuracy: 0.
2140
Epoch 22/30
```

```
s: 5.9909 - accuracy: 0.1940 - val loss: 4.7433 - val accuracy: 0.
2170
Epoch 23/30
s: 6.4548 - accuracy: 0.1800 - val loss: 5.8111 - val_accuracy: 0.
0940
Epoch 24/30
s: 6.2695 - accuracy: 0.1920 - val loss: 5.7078 - val accuracy: 0.
1020
Epoch 25/30
1000/1000 [============ ] - 0s 387us/sample - los
s: 5.4674 - accuracy: 0.1850 - val loss: 5.0711 - val accuracy: 0.
1470
Epoch 26/30
1000/1000 [============ ] - 0s 408us/sample - los
s: 6.1370 - accuracy: 0.1690 - val loss: 7.3002 - val accuracy: 0.
1190
Epoch 27/30
s: 6.0303 - accuracy: 0.1790 - val loss: 5.1454 - val accuracy: 0.
1750
Epoch 28/30
1000/1000 [============ ] - 0s 313us/sample - los
s: 6.3046 - accuracy: 0.2140 - val loss: 4.6457 - val accuracy: 0.
3300
Epoch 29/30
s: 6.7093 - accuracy: 0.2050 - val loss: 6.0724 - val accuracy: 0.
1810
Epoch 30/30
s: 6.1461 - accuracy: 0.2180 - val loss: 6.0274 - val accuracy: 0.
1430
```





#### **Zero-one Error**

In [33]: model=neural\_net\_1hidden(True)
 traine,teste,traina,testa=for\_error\_plot(model)
 zero\_one\_error\_plot(traine,teste,traina,testa)

Model: "sequential\_6"

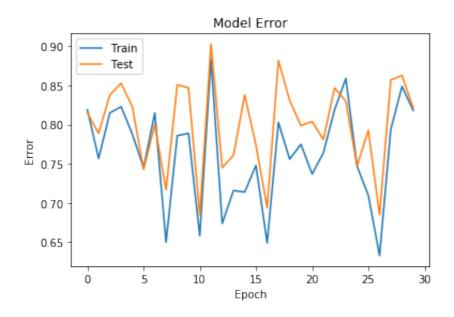
Layer (type)	Output Shape	Param #
dense_11 (Dense)	(None, 30)	23550
dense_12 (Dense)	(None, 10)	310

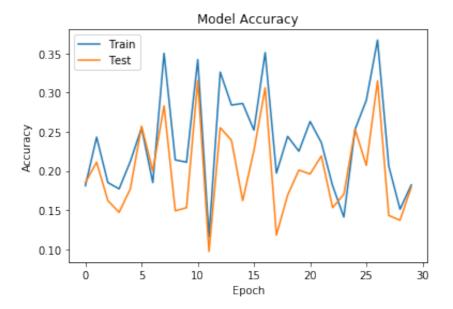
Total params: 23,860 Trainable params: 23,860 Non-trainable params: 0

```
Train on 1000 samples
s: 30.8283 - accuracy: 0.1350
Train on 1000 samples
s: 4.3015 - accuracy: 0.1640
Train on 1000 samples
s: 5.2170 - accuracy: 0.2010
Train on 1000 samples
1000/1000 [============ ] - 0s 186us/sample - los
s: 4.8569 - accuracy: 0.1790
Train on 1000 samples
1000/1000 [=======
                   s: 5.3937 - accuracy: 0.2070
Train on 1000 samples
```

```
1000/1000 [============ ] - 0s 185us/sample - los
s: 5.8270 - accuracy: 0.1980
Train on 1000 samples
1000/1000 [============ ] - 0s 183us/sample - los
s: 5.7690 - accuracy: 0.2310
Train on 1000 samples
s: 6.7065 - accuracy: 0.2510
Train on 1000 samples
1000/1000 [============= ] - 0s 187us/sample - los
s: 6.5117 - accuracy: 0.2100
Train on 1000 samples
1000/1000 [============ ] - 0s 197us/sample - los
s: 5.9477 - accuracy: 0.2310
Train on 1000 samples
s: 7.1525 - accuracy: 0.2470
Train on 1000 samples
s: 7.4980 - accuracy: 0.2010
Train on 1000 samples
1000/1000 [============= ] - 0s 199us/sample - los
s: 5.6581 - accuracy: 0.2220
Train on 1000 samples
1000/1000 [============ ] - 0s 204us/sample - los
s: 5.8502 - accuracy: 0.1720
Train on 1000 samples
s: 6.3494 - accuracy: 0.2100
Train on 1000 samples
1000/1000 [============ ] - 0s 237us/sample - los
s: 6.4638 - accuracy: 0.1990
Train on 1000 samples
s: 5.8281 - accuracy: 0.2160
Train on 1000 samples
s: 6.0723 - accuracy: 0.1920
Train on 1000 samples
s: 5.7845 - accuracy: 0.2230
Train on 1000 samples
s: 6.9948 - accuracy: 0.2010
Train on 1000 samples
1000/1000 [============ ] - 0s 231us/sample - los
s: 6.9080 - accuracy: 0.1870
Train on 1000 samples
s: 5.6190 - accuracy: 0.2100
Train on 1000 samples
s: 5.6348 - accuracy: 0.1820
```

Train on 1000 samples s: 5.9712 - accuracy: 0.1790 Train on 1000 samples s: 6.0057 - accuracy: 0.1990 Train on 1000 samples 5 - accuracy: 0.21 - 0s 210us/sample - loss: 5.8027 - accuracy: 0. 2150 Train on 1000 samples 1000/1000 [=============== ] - 0s 221us/sample - los s: 6.6944 - accuracy: 0.1860 Train on 1000 samples 1000/1000 [======= ======= ] - Os 186us/sample - los s: 5.9219 - accuracy: 0.2030 Train on 1000 samples 1000/1000 [============ ] - 0s 247us/sample - los s: 5.9162 - accuracy: 0.2150 Train on 1000 samples 1000/1000 [============ ] - 0s 220us/sample - los s: 5.8164 - accuracy: 0.2100





#### Learning rate plot

In [34]: model=neural\_net\_1hidden(True)
learning rate plot(23860,model)

Model: "sequential 7"

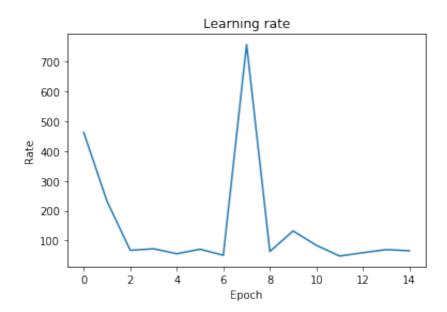
Layer (type)	Output Shape	Param #
dense_13 (Dense)	(None, 30)	23550
dense_14 (Dense)	(None, 10)	310

Total params: 23,860 Trainable params: 23,860 Non-trainable params: 0

```
Train on 1000 samples
1000/1000 [=========
                     s: 30.8666 - accuracy: 0.1610
Train on 1000 samples
1000/1000 [=======
                  ======== ] - 0s 246us/sample - los
s: 4.2672 - accuracy: 0.1750
Train on 1000 samples
1000/1000 [========
                   =======] - 0s 184us/sample - los
s: 5.1633 - accuracy: 0.2090
Train on 1000 samples
s: 5.4903 - accuracy: 0.2090
Train on 1000 samples
s: 5.6340 - accuracy: 0.1900
Train on 1000 samples
```

```
s: 5.4936 - accuracy: 0.2080
Train on 1000 samples
s: 7.0238 - accuracy: 0.2390
Train on 1000 samples
s: 6.9018 - accuracy: 0.2250
Train on 1000 samples
1000/1000 [============= ] - 0s 185us/sample - los
s: 6.2076 - accuracy: 0.2170
Train on 1000 samples
s: 6.7087 - accuracy: 0.2140
Train on 1000 samples
s: 5.8844 - accuracy: 0.2360
Train on 1000 samples
1000/1000 [============= ] - 0s 186us/sample - los
s: 5.8784 - accuracy: 0.2300
Train on 1000 samples
s: 6.5338 - accuracy: 0.2330
Train on 1000 samples
s: 6.4696 - accuracy: 0.2190
Train on 1000 samples
s: 6.2606 - accuracy: 0.2100
Train on 1000 samples
s: 7.1551 - accuracy: 0.2230
Train on 1000 samples
s: 6.6390 - accuracy: 0.1950
Train on 1000 samples
1000/1000 [============ ] - 0s 239us/sample - los
s: 7.7188 - accuracy: 0.2360
Train on 1000 samples
s: 6.6714 - accuracy: 0.2460
Train on 1000 samples
1000/1000 [============= ] - 0s 256us/sample - los
s: 6.9349 - accuracy: 0.2280
Train on 1000 samples
s: 6.7569 - accuracy: 0.1880
Train on 1000 samples
s: 7.3080 - accuracy: 0.2300
Train on 1000 samples
1000/1000 [============ ] - 0s 236us/sample - los
s: 6.8204 - accuracy: 0.2140
Train on 1000 samples
```

```
s: 6.7142 - accuracy: 0.2160
Train on 1000 samples
s: 7.0781 - accuracy: 0.2330
Train on 1000 samples
1000/1000 [=============== ] - 0s 213us/sample - los
s: 6.5930 - accuracy: 0.2090
Train on 1000 samples
s: 6.4572 - accuracy: 0.2260
Train on 1000 samples
1000/1000 [============ ] - 0s 182us/sample - los
s: 6.8373 - accuracy: 0.2180
Train on 1000 samples
s: 6.1812 - accuracy: 0.1930
Train on 1000 samples
s: 6.0630 - accuracy: 0.1970
```



## Question 2)b) 2 and 3 hidden Layers

```
In [35]: def neural net 2hidden(12):
             model = tf.keras.Sequential()
             if 12==True:
                 model.add(tf.keras.layers.Dense(30, input dim=784,activatio
         n='sigmoid', kernel regularizer=tf.keras.regularizers.12(5)))
                 model.add(tf.keras.layers.Dense(30, input dim=30,activation
         ='sigmoid',kernel regularizer=tf.keras.regularizers.12(5)))
             else:
                 model.add(tf.keras.layers.Dense(30, input dim=784,activatio
         n='sigmoid'))
                 model.add(tf.keras.layers.Dense(30, input dim=30,activation
         ='sigmoid'))
             model.add(tf.keras.layers.Dense(10, activation='softmax'))
             model.summary()
             adam=tf.keras.optimizers.Adam(learning rate=0.1)
             model.compile(loss='categorical crossentropy', optimizer=adam,
         metrics=['accuracy'])
             return model
```

#### Two hidden layers without regularization

```
In [36]: model=neural_net_2hidden(False)
    history = model.fit(mini_train, train_labels, epochs=30, batch_size
    =10,validation_data=(mini_test,test_labels))
```

Model: "sequential 8"

Layer (type)	Output Shape	Param #
dense_15 (Dense)	(None, 30)	23550
dense_16 (Dense)	(None, 30)	930
dense_17 (Dense)	(None, 10)	310

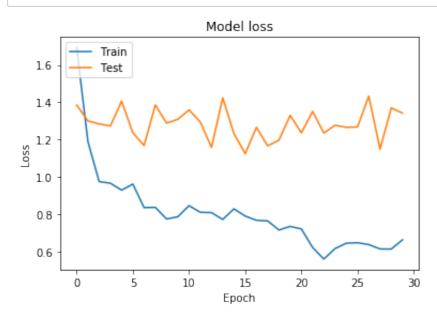
Total params: 24,790 Trainable params: 24,790 Non-trainable params: 0

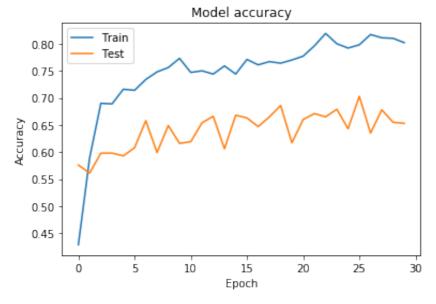
```
s: 1.1888 - accuracy: 0.5890 - val loss: 1.2996 - val accuracy: 0.
5610
Epoch 3/30
1000/1000 [=========== ] - 0s 316us/sample - los
s: 0.9741 - accuracy: 0.6900 - val loss: 1.2834 - val accuracy: 0.
5980
Epoch 4/30
s: 0.9653 - accuracy: 0.6890 - val loss: 1.2727 - val accuracy: 0.
5980
Epoch 5/30
1000/1000 [============ ] - 0s 373us/sample - los
s: 0.9281 - accuracy: 0.7160 - val loss: 1.4058 - val accuracy: 0.
5930
Epoch 6/30
s: 0.9608 - accuracy: 0.7140 - val loss: 1.2359 - val accuracy: 0.
Epoch 7/30
s: 0.8343 - accuracy: 0.7340 - val loss: 1.1676 - val accuracy: 0.
6580
Epoch 8/30
s: 0.8357 - accuracy: 0.7480 - val_loss: 1.3857 - val_accuracy: 0.
5990
Epoch 9/30
s: 0.7733 - accuracy: 0.7560 - val loss: 1.2879 - val accuracy: 0.
6490
Epoch 10/30
s: 0.7853 - accuracy: 0.7730 - val loss: 1.3084 - val accuracy: 0.
6160
Epoch 11/30
1000/1000 [=========== ] - 0s 332us/sample - los
s: 0.8449 - accuracy: 0.7470 - val loss: 1.3589 - val accuracy: 0.
6190
Epoch 12/30
s: 0.8092 - accuracy: 0.7500 - val loss: 1.2941 - val accuracy: 0.
6540
Epoch 13/30
s: 0.8074 - accuracy: 0.7440 - val loss: 1.1574 - val accuracy: 0.
6660
Epoch 14/30
s: 0.7708 - accuracy: 0.7590 - val loss: 1.4231 - val accuracy: 0.
6060
Epoch 15/30
s: 0.8277 - accuracy: 0.7440 - val loss: 1.2309 - val accuracy: 0.
```

```
6680
Epoch 16/30
s: 0.7890 - accuracy: 0.7710 - val loss: 1.1239 - val accuracy: 0.
6630
Epoch 17/30
s: 0.7663 - accuracy: 0.7610 - val loss: 1.2652 - val accuracy: 0.
6470
Epoch 18/30
s: 0.7632 - accuracy: 0.7670 - val loss: 1.1659 - val accuracy: 0.
6650
Epoch 19/30
s: 0.7140 - accuracy: 0.7640 - val loss: 1.1964 - val accuracy: 0.
6860
Epoch 20/30
s: 0.7335 - accuracy: 0.7700 - val_loss: 1.3293 - val_accuracy: 0.
6170
Epoch 21/30
s: 0.7198 - accuracy: 0.7770 - val loss: 1.2356 - val accuracy: 0.
6600
Epoch 22/30
s: 0.6204 - accuracy: 0.7960 - val loss: 1.3508 - val accuracy: 0.
6710
Epoch 23/30
s: 0.5583 - accuracy: 0.8190 - val_loss: 1.2347 - val_accuracy: 0.
6650
Epoch 24/30
s: 0.6145 - accuracy: 0.8000 - val loss: 1.2766 - val accuracy: 0.
6790
Epoch 25/30
s: 0.6433 - accuracy: 0.7920 - val loss: 1.2652 - val accuracy: 0.
6430
Epoch 26/30
s: 0.6458 - accuracy: 0.7980 - val loss: 1.2674 - val accuracy: 0.
7030
Epoch 27/30
s: 0.6359 - accuracy: 0.8170 - val loss: 1.4330 - val accuracy: 0.
6350
Epoch 28/30
1000/1000 [============ ] - 0s 383us/sample - los
s: 0.6125 - accuracy: 0.8110 - val loss: 1.1482 - val accuracy: 0.
6780
```

#### Plot for criterion function loss

In [37]: model\_training\_plot()





#### Plot for zero-one error

17/04/20, 4:35 PM PR\_NeuralNetworks

In [38]: model=neural\_net\_2hidden(False) traine,teste,traina,testa=for error plot(model) zero\_one\_error\_plot(traine,teste,traina,testa)

Model: "sequential\_9"

Layer (type)	Output	Shape			Param	#
dense_18 (Dense)	(None,		===	-==	23550	====
dense_19 (Dense)	(None,	30)			930	
dense_20 (Dense)	(None,	10)			310	
Total params: 24,790 Trainable params: 24,790 Non-trainable params: 0	=====	=====	===	===		====
Train on 1000 samples 1000/1000 [==================================	=====	=====]	_	1s	558us/sample	- los
1000/1000 [==================================	=====	=====]	-	0s	188us/sample	- los
1000/1000 [==================================	=====	=====]	-	0s	194us/sample	- los
1000/1000 [==================================		=====]	-	0s	193us/sample	- los
1000/1000 [==================================	=====	=====]	-	0s	188us/sample	- los
1000/1000 [==================================	=====	=====]	-	0s	193us/sample	- los
1000/1000 [==================================	=====	=====]	-	0s	195us/sample	- los
1000/1000 [==================================		=====]	-	0s	225us/sample	- los
Train on 1000 samples  1000/1000 [==================================	=====	=====]	-	0s	194us/sample	- los
Train on 1000 samples 1000/1000 [==================================	=====	=====]	-	0s	195us/sample	- los
Train on 1000 samples 1000/1000 [==================================	=====	=====]	-	0s	192us/sample	- los
Train on 1000 samples 1000/1000 [==================================	======	=====]	_	0s	196us/sample	- los

```
s: 0.7849 - accuracy: 0.7540
Train on 1000 samples
s: 0.7600 - accuracy: 0.7760
Train on 1000 samples
s: 0.7600 - accuracy: 0.7510
Train on 1000 samples
1000/1000 [============= ] - 0s 191us/sample - los
s: 0.7074 - accuracy: 0.7770
Train on 1000 samples
s: 0.7434 - accuracy: 0.7780
Train on 1000 samples
1000/1000 [============= ] - 0s 189us/sample - los
s: 0.7563 - accuracy: 0.7660
Train on 1000 samples
1000/1000 [============= ] - 0s 258us/sample - los
s: 0.8655 - accuracy: 0.7580
Train on 1000 samples
s: 0.8545 - accuracy: 0.7510
Train on 1000 samples
s: 0.7556 - accuracy: 0.7520
Train on 1000 samples
s: 0.7085 - accuracy: 0.7850
Train on 1000 samples
s: 0.6538 - accuracy: 0.7960
Train on 1000 samples
s: 0.6791 - accuracy: 0.7920
Train on 1000 samples
1000/1000 [============ ] - 0s 193us/sample - los
s: 0.6676 - accuracy: 0.8030
Train on 1000 samples
s: 0.6636 - accuracy: 0.7970
Train on 1000 samples
1000/1000 [============ ] - 0s 191us/sample - los
s: 0.6071 - accuracy: 0.8100
Train on 1000 samples
s: 0.5883 - accuracy: 0.8070
Train on 1000 samples
s: 0.6070 - accuracy: 0.8140
Train on 1000 samples
1000/1000 [============ ] - 0s 240us/sample - los
s: 0.6154 - accuracy: 0.8040
Train on 1000 samples
```

1000/1000 [=============] - 0s 196us/sample - los s: 0.6613 - accuracy: 0.7890





#### Plot for learning rate

In [40]: model=neural\_net\_2hidden(False)
learning\_rate\_plot(24790, model)

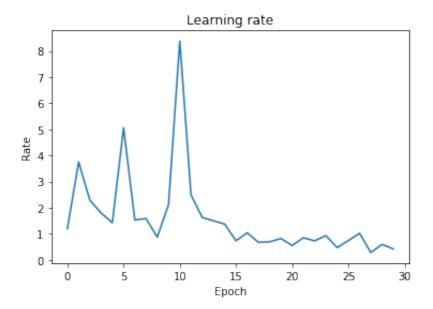
Model: "sequential\_11"

Layer (type)	Output Shape	Param #
dense_24 (Dense)	(None, 30)	23550
dense_25 (Dense)	(None, 30)	930
dense_26 (Dense)	(None, 10)	310

Total params: 24,790 Trainable params: 24,790 Non-trainable params: 0

```
Train on 1000 samples
s: 1.5793 - accuracy: 0.4430
Train on 1000 samples
1000/1000 [============ ] - 0s 208us/sample - los
s: 1.0365 - accuracy: 0.6630
Train on 1000 samples
s: 0.9198 - accuracy: 0.6940
Train on 1000 samples
s: 0.8485 - accuracy: 0.7250
Train on 1000 samples
1000/1000 [============ ] - 0s 212us/sample - los
s: 0.8280 - accuracy: 0.7160
Train on 1000 samples
s: 0.7396 - accuracy: 0.7490
Train on 1000 samples
s: 0.7668 - accuracy: 0.7180
Train on 1000 samples
s: 0.8473 - accuracy: 0.7290
Train on 1000 samples
s: 0.7289 - accuracy: 0.7600
Train on 1000 samples
s: 0.7743 - accuracy: 0.7510
Train on 1000 samples
1000/1000 [============ ] - 0s 189us/sample - los
s: 0.8649 - accuracy: 0.7320
Train on 1000 samples
s: 0.7400 - accuracy: 0.7660
Train on 1000 samples
1000/1000 [============= ] - 0s 191us/sample - los
s: 0.7700 - accuracy: 0.7470
Train on 1000 samples
s: 0.7770 - accuracy: 0.7560
Train on 1000 samples
s: 0.7830 - accuracy: 0.7470
Train on 1000 samples
1000/1000 [=============== ] - 0s 195us/sample - los
s: 0.7971 - accuracy: 0.7480
Train on 1000 samples
```

```
1000/1000 [============ ] - 0s 187us/sample - los
s: 0.7495 - accuracy: 0.7570
Train on 1000 samples
s: 0.7103 - accuracy: 0.7900
Train on 1000 samples
s: 0.7099 - accuracy: 0.7940
Train on 1000 samples
1000/1000 [============= ] - 0s 190us/sample - los
s: 0.6896 - accuracy: 0.7780
Train on 1000 samples
1000/1000 [============ ] - 0s 193us/sample - los
s: 0.6564 - accuracy: 0.7850
Train on 1000 samples
s: 0.7390 - accuracy: 0.7920
Train on 1000 samples
s: 0.6696 - accuracy: 0.8010
Train on 1000 samples
1000/1000 [============= ] - 0s 189us/sample - los
s: 0.6313 - accuracy: 0.8160
Train on 1000 samples
1000/1000 [============ ] - 0s 193us/sample - los
s: 0.6275 - accuracy: 0.8160
Train on 1000 samples
s: 0.6753 - accuracy: 0.7880
Train on 1000 samples
1000/1000 [============= ] - 0s 216us/sample - los
s: 0.7381 - accuracy: 0.7760
Train on 1000 samples
s: 0.6682 - accuracy: 0.7990
Train on 1000 samples
s: 0.6817 - accuracy: 0.7940
Train on 1000 samples
s: 0.5951 - accuracy: 0.8070
```



#### Two hidden layers with regularization

Model: "sequential 12"

Layer (type)	Output Shape	Param #
dense_27 (Dense)	(None, 30)	23550
dense_28 (Dense)	(None, 30)	930
dense_29 (Dense)	(None, 10)	310

Total params: 24,790 Trainable params: 24,790 Non-trainable params: 0

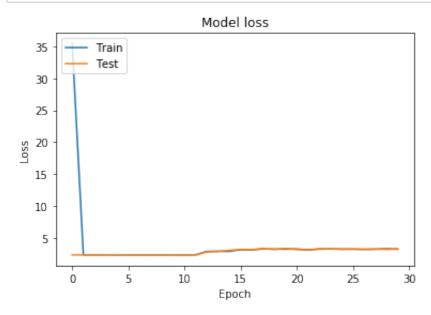
```
s: 2.3368 - accuracy: 0.0850 - val loss: 2.3255 - val accuracy: 0.
1000
Epoch 5/30
s: 2.3283 - accuracy: 0.1210 - val loss: 2.3461 - val accuracy: 0.
1000
Epoch 6/30
s: 2.3304 - accuracy: 0.0960 - val loss: 2.3214 - val accuracy: 0.
1000
Epoch 7/30
1000/1000 [============ ] - 0s 417us/sample - los
s: 2.3339 - accuracy: 0.0820 - val loss: 2.3073 - val accuracy: 0.
1000
Epoch 8/30
1000/1000 [============ ] - 0s 387us/sample - los
s: 2.3242 - accuracy: 0.0900 - val loss: 2.3154 - val accuracy: 0.
1000
Epoch 9/30
s: 2.3284 - accuracy: 0.1040 - val loss: 2.3150 - val accuracy: 0.
1000
Epoch 10/30
1000/1000 [============ ] - 0s 348us/sample - los
s: 2.3276 - accuracy: 0.1070 - val loss: 2.3255 - val accuracy: 0.
1000
Epoch 11/30
s: 2.3235 - accuracy: 0.0970 - val loss: 2.3189 - val accuracy: 0.
1000
Epoch 12/30
s: 2.3323 - accuracy: 0.0860 - val loss: 2.3299 - val accuracy: 0.
1000
Epoch 13/30
s: 2.8774 - accuracy: 0.1000 - val loss: 2.7886 - val accuracy: 0.
1000
Epoch 14/30
s: 2.8972 - accuracy: 0.1100 - val loss: 2.8664 - val accuracy: 0.
1000
Epoch 15/30
s: 2.8798 - accuracy: 0.0860 - val_loss: 3.0443 - val_accuracy: 0.
1000
Epoch 16/30
s: 3.1523 - accuracy: 0.0940 - val loss: 3.1739 - val accuracy: 0.
1000
Epoch 17/30
```

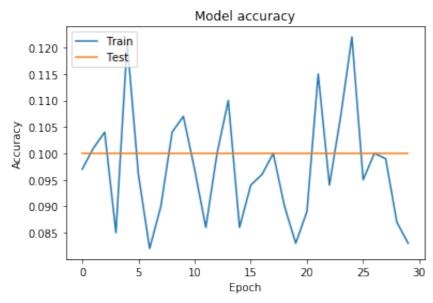
```
s: 3.1223 - accuracy: 0.0960 - val loss: 3.1636 - val accuracy: 0.
1000
Epoch 18/30
1000/1000 [============] - 0s 338us/sample - los
s: 3.2697 - accuracy: 0.1000 - val loss: 3.3040 - val accuracy: 0.
1000
Epoch 19/30
s: 3.2387 - accuracy: 0.0900 - val loss: 3.1925 - val accuracy: 0.
1000
Epoch 20/30
1000/1000 [============ ] - 0s 325us/sample - los
s: 3.2253 - accuracy: 0.0830 - val loss: 3.3379 - val accuracy: 0.
1000
Epoch 21/30
s: 3.2587 - accuracy: 0.0890 - val loss: 3.1757 - val accuracy: 0.
Epoch 22/30
s: 3.1113 - accuracy: 0.1150 - val loss: 3.1259 - val accuracy: 0.
1000
Epoch 23/30
s: 3.2360 - accuracy: 0.0940 - val_loss: 3.2885 - val_accuracy: 0.
1000
Epoch 24/30
s: 3.2744 - accuracy: 0.1070 - val loss: 3.2503 - val accuracy: 0.
1000
Epoch 25/30
s: 3.2355 - accuracy: 0.1220 - val loss: 3.2200 - val accuracy: 0.
1000
Epoch 26/30
1000/1000 [============ ] - 0s 336us/sample - los
s: 3.2211 - accuracy: 0.0950 - val loss: 3.2609 - val accuracy: 0.
1000
Epoch 27/30
s: 3.1975 - accuracy: 0.1000 - val loss: 3.2188 - val accuracy: 0.
1000
Epoch 28/30
s: 3.2180 - accuracy: 0.0990 - val loss: 3.2444 - val accuracy: 0.
1000
Epoch 29/30
s: 3.2976 - accuracy: 0.0870 - val loss: 3.1887 - val accuracy: 0.
1000
Epoch 30/30
s: 3.2200 - accuracy: 0.0830 - val loss: 3.2762 - val accuracy: 0.
```

1000

#### Plot for criterion function loss







#### Plot for zero-one error

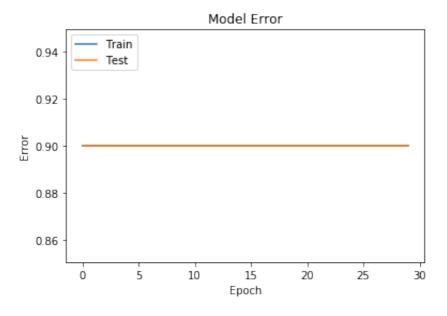
In [43]: model=neural\_net\_2hidden(True)
 traine,teste,traina,testa=for\_error\_plot(model)
 zero\_one\_error\_plot(traine,teste,traina,testa)

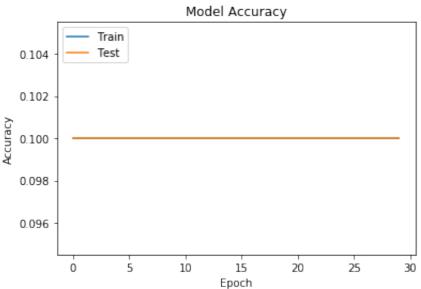
Model: "sequential\_13"

Layer (type) Output Shape Param #

```
(None, 30)
dense 30 (Dense)
                             23550
dense 31 (Dense)
               (None, 30)
                             930
dense 32 (Dense)
               (None, 10)
                             310
______
Total params: 24,790
Trainable params: 24,790
Non-trainable params: 0
Train on 1000 samples
s: 35.3744 - accuracy: 0.1210
Train on 1000 samples
s: 2.3323 - accuracy: 0.0900
Train on 1000 samples
1000/1000 [============ ] - 0s 237us/sample - los
s: 2.3320 - accuracy: 0.0820
Train on 1000 samples
s: 2.3367 - accuracy: 0.0850
Train on 1000 samples
s: 2.3323 - accuracy: 0.0870
Train on 1000 samples
s: 2.3313 - accuracy: 0.0930
Train on 1000 samples
s: 2.3274 - accuracy: 0.0960
Train on 1000 samples
s: 2.3274 - accuracy: 0.0940
Train on 1000 samples
1000/1000 [=============== ] - 0s 195us/sample - los
s: 2.3275 - accuracy: 0.0980
Train on 1000 samples
s: 2.3219 - accuracy: 0.0950
Train on 1000 samples
s: 2.3320 - accuracy: 0.0890
Train on 1000 samples
s: 2.3270 - accuracy: 0.0840
Train on 1000 samples
s: 2.8895 - accuracy: 0.0970
Train on 1000 samples
s: 2.9074 - accuracy: 0.0870
Train on 1000 samples
```

```
1000/1000 [============ ] - 0s 195us/sample - los
s: 2.8571 - accuracy: 0.0920
Train on 1000 samples
1000/1000 [============ ] - 0s 199us/sample - los
s: 3.1341 - accuracy: 0.1060
Train on 1000 samples
s: 3.1350 - accuracy: 0.0900
Train on 1000 samples
1000/1000 [============= ] - 0s 193us/sample - los
s: 3.2785 - accuracy: 0.1000
Train on 1000 samples
1000/1000 [============ ] - 0s 194us/sample - los
s: 3.2268 - accuracy: 0.1010
Train on 1000 samples
s: 3.2006 - accuracy: 0.0960
Train on 1000 samples
s: 3.2948 - accuracy: 0.0890
Train on 1000 samples
1000/1000 [============= ] - 0s 197us/sample - los
s: 3.1292 - accuracy: 0.0840
Train on 1000 samples
1000/1000 [============ ] - 0s 194us/sample - los
s: 3.2618 - accuracy: 0.0920
Train on 1000 samples
s: 3.2727 - accuracy: 0.1060
Train on 1000 samples
1000/1000 [============ ] - 0s 195us/sample - los
s: 3.2283 - accuracy: 0.0890
Train on 1000 samples
s: 3.2242 - accuracy: 0.0990
Train on 1000 samples
s: 3.1738 - accuracy: 0.0950
Train on 1000 samples
s: 3.2659 - accuracy: 0.0820
Train on 1000 samples
s: 3.2749 - accuracy: 0.0780
Train on 1000 samples
s: 3.2193 - accuracy: 0.0930
```





# Plot for learning rate

In [44]: model=neural\_net\_2hidden(True)
learning\_rate\_plot(24790,model)

Model: "sequential\_14"

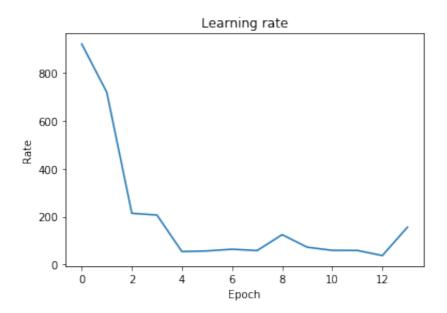
Layer (type)	Output Shape	Param #
dense_33 (Dense)	(None, 30)	23550
dense_34 (Dense)	(None, 30)	930
dense_35 (Dense)	(None, 10)	310

Total params: 24,790
Trainable params: 24,790

#### Non-trainable params: 0

```
Train on 1000 samples
1000/1000 [============ ] - 1s 637us/sample - los
s: 35.4035 - accuracy: 0.1010
Train on 1000 samples
s: 2.3344 - accuracy: 0.0970
Train on 1000 samples
1000/1000 [============= ] - 0s 195us/sample - los
s: 2.3330 - accuracy: 0.1030
Train on 1000 samples
1000/1000 [============ ] - 0s 198us/sample - los
s: 2.3329 - accuracy: 0.0750
Train on 1000 samples
s: 2.3275 - accuracy: 0.1060
Train on 1000 samples
s: 2.3257 - accuracy: 0.1010
Train on 1000 samples
1000/1000 [============= ] - 0s 237us/sample - los
s: 2.3267 - accuracy: 0.0920
Train on 1000 samples
1000/1000 [============ ] - 0s 238us/sample - los
s: 2.3286 - accuracy: 0.0910
Train on 1000 samples
s: 2.3293 - accuracy: 0.1010
Train on 1000 samples
s: 2.3273 - accuracy: 0.1020
Train on 1000 samples
s: 2.3290 - accuracy: 0.0950
Train on 1000 samples
s: 2.3300 - accuracy: 0.1010
Train on 1000 samples
s: 2.8664 - accuracy: 0.1010
Train on 1000 samples
s: 2.8807 - accuracy: 0.0910
Train on 1000 samples
s: 2.8765 - accuracy: 0.0910
Train on 1000 samples
s: 3.1454 - accuracy: 0.0850
Train on 1000 samples
s: 3.1379 - accuracy: 0.0820
```

```
Train on 1000 samples
s: 3.2949 - accuracy: 0.0850
Train on 1000 samples
s: 3.1835 - accuracy: 0.1110
Train on 1000 samples
s: 3.2446 - accuracy: 0.0940
Train on 1000 samples
1000/1000 [============] - 0s 201us/sample - los
s: 3.2777 - accuracy: 0.1080
Train on 1000 samples
s: 3.0962 - accuracy: 0.0820
Train on 1000 samples
1000/1000 [============ ] - 0s 239us/sample - los
s: 3.2774 - accuracy: 0.0750
Train on 1000 samples
1000/1000 [============ ] - 0s 236us/sample - los
s: 3.2792 - accuracy: 0.0900
Train on 1000 samples
s: 3.2315 - accuracy: 0.0980
Train on 1000 samples
s: 3.2029 - accuracy: 0.1020
Train on 1000 samples
s: 3.2055 - accuracy: 0.0930
Train on 1000 samples
1000/1000 [=============== ] - 0s 195us/sample - los
s: 3.2551 - accuracy: 0.0840
Train on 1000 samples
s: 3.2564 - accuracy: 0.0930
Train on 1000 samples
s: 3.2150 - accuracy: 0.0910
```



```
In [45]:
         def neural net 3hidden(12):
             model = tf.keras.Sequential()
             if 12==True:
                 model.add(tf.keras.layers.Dense(30, input dim=784,activatio
         n='sigmoid',kernel regularizer=tf.keras.regularizers.12(5)))
                 model.add(tf.keras.layers.Dense(30, input dim=30,activation
         ='sigmoid',kernel regularizer=tf.keras.regularizers.12(5)))
                 model.add(tf.keras.layers.Dense(30, input dim=30,activation
         ='sigmoid',kernel regularizer=tf.keras.regularizers.12(5)))
             else:
                 model.add(tf.keras.layers.Dense(30, input dim=784,activatio
         n='sigmoid'))
                 model.add(tf.keras.layers.Dense(30, input dim=30,activation
         ='sigmoid'))
                 model.add(tf.keras.layers.Dense(30, input dim=30,activation
         ='sigmoid'))
             model.add(tf.keras.layers.Dense(10, activation='softmax'))
             model.summary()
             adam=tf.keras.optimizers.Adam(learning rate=0.1)
             model.compile(loss='categorical_crossentropy', optimizer=adam,
         metrics=['accuracy'])
             return model
```

#### 3 hidden layers without regularization

# **Plot for Criterion Function Loss**

# In [46]: model=neural\_net\_3hidden(False) history = model.fit(mini\_train, train\_labels, epochs=30, batch\_size =10,validation\_data=(mini\_test,test\_labels)) model\_training\_plot()

Model: "sequential 15"

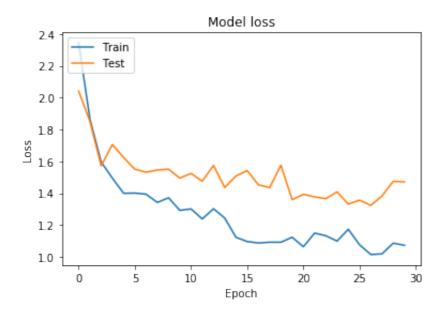
Layer (type)	Output	Shape	Param #
dense_36 (Dense)	(None,	30)	23550
dense_37 (Dense)	(None,	30)	930
dense_38 (Dense)	(None,	30)	930
dense_39 (Dense)	(None,	10)	310
Total params: 25,720			

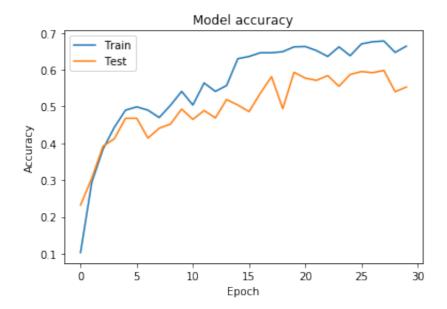
Trainable params: 25,720 Non-trainable params: 0

```
Train on 1000 samples, validate on 1000 samples
Epoch 1/30
s: 2.3430 - accuracy: 0.1030 - val loss: 2.0410 - val accuracy: 0.
2320
Epoch 2/30
s: 1.8662 - accuracy: 0.2940 - val loss: 1.8509 - val accuracy: 0.
3050
Epoch 3/30
s: 1.5957 - accuracy: 0.3840 - val loss: 1.5722 - val accuracy: 0.
3920
Epoch 4/30
s: 1.4944 - accuracy: 0.4430 - val_loss: 1.7053 - val_accuracy: 0.
4120
Epoch 5/30
1000/1000 [============] - 0s 334us/sample - los
s: 1.3993 - accuracy: 0.4900 - val loss: 1.6245 - val accuracy: 0.
4680
Epoch 6/30
1000/1000 [============ ] - 0s 323us/sample - los
s: 1.4007 - accuracy: 0.4990 - val loss: 1.5509 - val accuracy: 0.
4680
Epoch 7/30
1000/1000 [=========== ] - 0s 322us/sample - los
s: 1.3939 - accuracy: 0.4900 - val loss: 1.5316 - val accuracy: 0.
4140
Epoch 8/30
1000/1000 [=========== ] - 0s 322us/sample - los
s: 1.3420 - accuracy: 0.4700 - val loss: 1.5455 - val accuracy: 0.
```

```
4410
Epoch 9/30
s: 1.3711 - accuracy: 0.5030 - val loss: 1.5507 - val accuracy: 0.
4520
Epoch 10/30
s: 1.2922 - accuracy: 0.5410 - val loss: 1.4947 - val accuracy: 0.
4930
Epoch 11/30
1000/1000 [===========] - 0s 322us/sample - los
s: 1.3013 - accuracy: 0.5040 - val loss: 1.5241 - val accuracy: 0.
4650
Epoch 12/30
s: 1.2383 - accuracy: 0.5640 - val loss: 1.4754 - val accuracy: 0.
4890
Epoch 13/30
s: 1.3024 - accuracy: 0.5410 - val_loss: 1.5744 - val_accuracy: 0.
4690
Epoch 14/30
s: 1.2440 - accuracy: 0.5570 - val loss: 1.4361 - val accuracy: 0.
5190
Epoch 15/30
s: 1.1231 - accuracy: 0.6300 - val loss: 1.5081 - val accuracy: 0.
5040
Epoch 16/30
s: 1.0965 - accuracy: 0.6360 - val_loss: 1.5418 - val_accuracy: 0.
4860
Epoch 17/30
s: 1.0873 - accuracy: 0.6460 - val loss: 1.4526 - val accuracy: 0.
5360
Epoch 18/30
s: 1.0923 - accuracy: 0.6460 - val loss: 1.4353 - val accuracy: 0.
5810
Epoch 19/30
s: 1.0920 - accuracy: 0.6490 - val loss: 1.5757 - val accuracy: 0.
4940
Epoch 20/30
s: 1.1236 - accuracy: 0.6620 - val loss: 1.3591 - val accuracy: 0.
5930
Epoch 21/30
1000/1000 [============ ] - 0s 325us/sample - los
s: 1.0635 - accuracy: 0.6630 - val loss: 1.3930 - val accuracy: 0.
5770
```

```
Epoch 22/30
s: 1.1496 - accuracy: 0.6520 - val loss: 1.3767 - val accuracy: 0.
5710
Epoch 23/30
1000/1000 [============] - 0s 330us/sample - los
s: 1.1327 - accuracy: 0.6360 - val loss: 1.3660 - val accuracy: 0.
5840
Epoch 24/30
s: 1.0993 - accuracy: 0.6620 - val loss: 1.4085 - val accuracy: 0.
5550
Epoch 25/30
s: 1.1731 - accuracy: 0.6380 - val loss: 1.3314 - val accuracy: 0.
5870
Epoch 26/30
s: 1.0761 - accuracy: 0.6700 - val loss: 1.3563 - val accuracy: 0.
5950
Epoch 27/30
s: 1.0148 - accuracy: 0.6760 - val loss: 1.3234 - val accuracy: 0.
5920
Epoch 28/30
s: 1.0191 - accuracy: 0.6780 - val loss: 1.3827 - val accuracy: 0.
5980
Epoch 29/30
s: 1.0864 - accuracy: 0.6470 - val loss: 1.4749 - val accuracy: 0.
5400
Epoch 30/30
s: 1.0728 - accuracy: 0.6640 - val loss: 1.4711 - val accuracy: 0.
5530
```





#### Plot for zero-one error

In [47]: model=neural\_net\_3hidden(False)
 traine,teste,traina,testa=for\_error\_plot(model)
 zero\_one\_error\_plot(traine,teste,traina,testa)

Model: "sequential\_16"

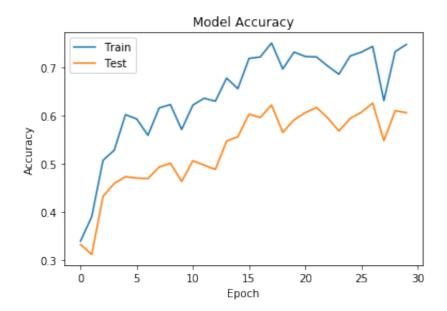
Layer (typ	e)	Output	Shape	Param #
dense_40 (	Dense)	(None,	30)	23550
dense_41 (	Dense)	(None,	30)	930
dense_42 (	Dense)	(None,	30)	930
dense_43 (	Dense)	(None,	10)	310

Total params: 25,720
Trainable params: 25,720
Non-trainable params: 0

```
Train on 1000 samples
s: 1.2865 - accuracy: 0.5560
Train on 1000 samples
s: 1.2003 - accuracy: 0.5720
Train on 1000 samples
s: 1.1844 - accuracy: 0.5950
Train on 1000 samples
1000/1000 [============] - 0s 202us/sample - los
s: 1.2122 - accuracy: 0.5820
Train on 1000 samples
s: 1.2061 - accuracy: 0.5990
Train on 1000 samples
1000/1000 [============ ] - 0s 200us/sample - los
s: 1.2653 - accuracy: 0.5750
Train on 1000 samples
1000/1000 [============ ] - 0s 198us/sample - los
s: 1.2113 - accuracy: 0.5790
Train on 1000 samples
s: 1.1095 - accuracy: 0.6410
Train on 1000 samples
s: 1.0905 - accuracy: 0.6270
Train on 1000 samples
s: 1.0379 - accuracy: 0.6690
Train on 1000 samples
1000/1000 [============ ] - 0s 207us/sample - los
s: 1.0170 - accuracy: 0.6720
Train on 1000 samples
s: 1.0812 - accuracy: 0.6500
Train on 1000 samples
s: 1.0299 - accuracy: 0.6700
Train on 1000 samples
s: 1.0339 - accuracy: 0.6890
Train on 1000 samples
1000/1000 [============ ] - 0s 202us/sample - los
s: 1.0752 - accuracy: 0.6800
Train on 1000 samples
s: 1.0100 - accuracy: 0.7100
Train on 1000 samples
s: 0.9543 - accuracy: 0.7100
Train on 1000 samples
```

s: 0.9570 - accuracy: 0.6930 Train on 1000 samples s: 0.9696 - accuracy: 0.7160 Train on 1000 samples s: 0.9250 - accuracy: 0.7280 Train on 1000 samples 1000/1000 [============ ] - 0s 251us/sample - los s: 0.9687 - accuracy: 0.7160 Train on 1000 samples s: 1.0204 - accuracy: 0.6930 Train on 1000 samples 1000/1000 [======== s: 0.9849 - accuracy: 0.7180 Train on 1000 samples 1000/1000 [============= ] - 0s 198us/sample - los s: 0.9518 - accuracy: 0.7130 Train on 1000 samples s: 1.0698 - accuracy: 0.6690 Train on 1000 samples s: 0.9762 - accuracy: 0.7100





#### Plot for learning rate

In [48]: model=neural\_net\_3hidden(False)
learning rate plot(25720,model)

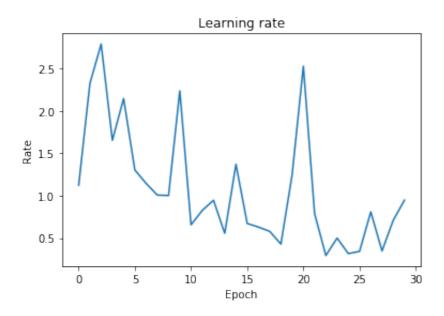
Model: "sequential\_17"

Layer (type)	Output Shape	Param #
dense_44 (Dense)	(None, 30)	23550
dense_45 (Dense)	(None, 30)	930
dense_46 (Dense)	(None, 30)	930
dense_47 (Dense)	(None, 10)	310

Total params: 25,720
Trainable params: 25,720
Non-trainable params: 0

```
1000/1000 [============ ] - 0s 198us/sample - los
s: 1.2775 - accuracy: 0.5270
Train on 1000 samples
1000/1000 [============ ] - 0s 226us/sample - los
s: 1.3131 - accuracy: 0.5260
Train on 1000 samples
s: 1.2109 - accuracy: 0.5840
Train on 1000 samples
1000/1000 [============= ] - 0s 248us/sample - los
s: 1.1113 - accuracy: 0.6450
Train on 1000 samples
1000/1000 [============ ] - 0s 278us/sample - los
s: 1.1134 - accuracy: 0.6090
Train on 1000 samples
s: 1.2456 - accuracy: 0.5790
Train on 1000 samples
s: 1.1719 - accuracy: 0.5980
Train on 1000 samples
1000/1000 [============= ] - 0s 262us/sample - los
s: 1.1489 - accuracy: 0.6250
Train on 1000 samples
1000/1000 [============ ] - 0s 266us/sample - los
s: 1.1796 - accuracy: 0.6010
Train on 1000 samples
s: 1.1008 - accuracy: 0.6580
Train on 1000 samples
1000/1000 [============ ] - 0s 258us/sample - los
s: 1.1574 - accuracy: 0.6280
Train on 1000 samples
s: 1.1407 - accuracy: 0.6300
Train on 1000 samples
1000/1000 [============ ] - 0s 240us/sample - los
s: 1.0512 - accuracy: 0.6790
Train on 1000 samples
s: 0.9877 - accuracy: 0.6960
Train on 1000 samples
s: 0.9734 - accuracy: 0.6920
Train on 1000 samples
1000/1000 [============= ] - 0s 235us/sample - los
s: 1.0383 - accuracy: 0.6700
Train on 1000 samples
s: 1.0989 - accuracy: 0.6620
Train on 1000 samples
s: 1.1810 - accuracy: 0.6100
```

```
Train on 1000 samples
s: 1.0507 - accuracy: 0.6680
Train on 1000 samples
s: 1.0581 - accuracy: 0.6760
Train on 1000 samples
s: 1.0162 - accuracy: 0.6810
Train on 1000 samples
s: 0.9451 - accuracy: 0.7000
Train on 1000 samples
1000/1000 [============= ] - 0s 245us/sample - los
s: 1.0069 - accuracy: 0.6870
Train on 1000 samples
1000/1000 [============ ] - 0s 246us/sample - los
s: 0.9982 - accuracy: 0.6950
Train on 1000 samples
1000/1000 [============ ] - 0s 263us/sample - los
s: 1.0097 - accuracy: 0.6760
Train on 1000 samples
s: 1.0494 - accuracy: 0.6540
```



# 3 hidden layers with regularization

Model: "sequential 18"

Layer	(type)	Output Shape	Param #
		=======================================	

(None, 30)

23550

dense 48 (Dense)

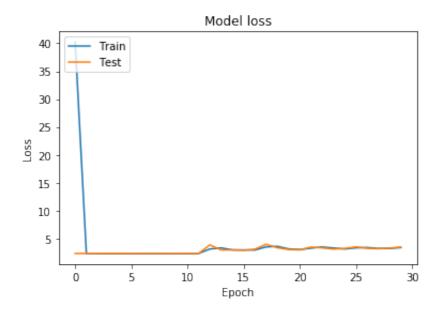
```
dense 49 (Dense)
                  (None, 30)
                                   930
dense 50 (Dense)
                   (None, 30)
                                   930
dense 51 (Dense)
                  (None, 10)
                                   310
Total params: 25,720
Trainable params: 25,720
Non-trainable params: 0
Train on 1000 samples, validate on 1000 samples
Epoch 1/30
40.2314 - accuracy: 0.0870 - val loss: 2.3294 - val accuracy: 0.10
00
Epoch 2/30
s: 2.3410 - accuracy: 0.0870 - val_loss: 2.3274 - val accuracy: 0.
1000
Epoch 3/30
s: 2.3343 - accuracy: 0.0880 - val loss: 2.3309 - val accuracy: 0.
1000
Epoch 4/30
s: 2.3348 - accuracy: 0.0950 - val loss: 2.3154 - val accuracy: 0.
1000
Epoch 5/30
s: 2.3340 - accuracy: 0.1010 - val loss: 2.3240 - val accuracy: 0.
1000
Epoch 6/30
s: 2.3341 - accuracy: 0.0870 - val loss: 2.3132 - val accuracy: 0.
1000
Epoch 7/30
s: 2.3319 - accuracy: 0.0770 - val loss: 2.3103 - val accuracy: 0.
1000
Epoch 8/30
s: 2.3252 - accuracy: 0.0900 - val loss: 2.3129 - val accuracy: 0.
1000
Epoch 9/30
s: 2.3321 - accuracy: 0.0850 - val loss: 2.3105 - val accuracy: 0.
1000
Epoch 10/30
1000/1000 [============ ] - 0s 349us/sample - los
s: 2.3332 - accuracy: 0.1040 - val loss: 2.3180 - val accuracy: 0.
1000
```

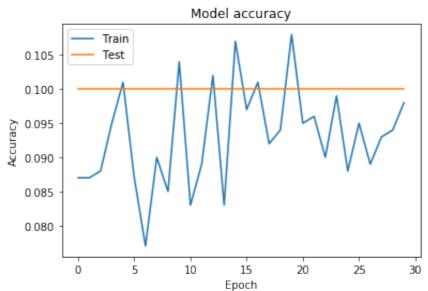
```
Epoch 11/30
s: 2.3272 - accuracy: 0.0830 - val loss: 2.3172 - val accuracy: 0.
1000
Epoch 12/30
1000/1000 [============] - 0s 341us/sample - los
s: 2.3325 - accuracy: 0.0890 - val loss: 2.3182 - val accuracy: 0.
1000
Epoch 13/30
s: 3.1256 - accuracy: 0.1020 - val loss: 3.8704 - val accuracy: 0.
1000
Epoch 14/30
s: 3.3208 - accuracy: 0.0830 - val loss: 2.9531 - val accuracy: 0.
1000
Epoch 15/30
s: 2.9627 - accuracy: 0.1070 - val loss: 2.9554 - val accuracy: 0.
1000
Epoch 16/30
s: 2.9534 - accuracy: 0.0970 - val loss: 2.9018 - val accuracy: 0.
1000
Epoch 17/30
s: 2.9412 - accuracy: 0.1010 - val loss: 3.0722 - val accuracy: 0.
1000
Epoch 18/30
s: 3.4950 - accuracy: 0.0920 - val_loss: 3.9782 - val_accuracy: 0.
1000
Epoch 19/30
s: 3.5994 - accuracy: 0.0940 - val loss: 3.3505 - val accuracy: 0.
1000
Epoch 20/30
s: 3.1450 - accuracy: 0.1080 - val loss: 3.0348 - val accuracy: 0.
1000
Epoch 21/30
s: 3.0183 - accuracy: 0.0950 - val loss: 2.9622 - val accuracy: 0.
1000
Epoch 22/30
s: 3.2708 - accuracy: 0.0960 - val loss: 3.4565 - val accuracy: 0.
1000
Epoch 23/30
s: 3.4836 - accuracy: 0.0900 - val loss: 3.3276 - val accuracy: 0.
1000
Epoch 24/30
```

```
s: 3.2950 - accuracy: 0.0990 - val loss: 3.1120 - val accuracy: 0.
1000
Epoch 25/30
s: 3.1229 - accuracy: 0.0880 - val loss: 3.2092 - val accuracy: 0.
1000
Epoch 26/30
s: 3.3273 - accuracy: 0.0950 - val loss: 3.4942 - val accuracy: 0.
1000
Epoch 27/30
1000/1000 [============ ] - 0s 337us/sample - los
s: 3.3847 - accuracy: 0.0890 - val loss: 3.2494 - val accuracy: 0.
1000
Epoch 28/30
1000/1000 [============ ] - 0s 327us/sample - los
s: 3.2190 - accuracy: 0.0930 - val loss: 3.1648 - val accuracy: 0.
1000
Epoch 29/30
s: 3.2281 - accuracy: 0.0940 - val loss: 3.2706 - val accuracy: 0.
1000
Epoch 30/30
1000/1000 [===========] - 0s 329us/sample - los
s: 3.3710 - accuracy: 0.0980 - val loss: 3.4740 - val accuracy: 0.
1000
```

### Plot for criterion function loss

```
In [50]: model_training_plot()
```





#### Plot for zero-one error

In [51]: model=neural\_net\_3hidden(True)
 traine,teste,traina,testa=for\_error\_plot(model)
 zero\_one\_error\_plot(traine,teste,traina,testa)

Model: "sequential\_19"

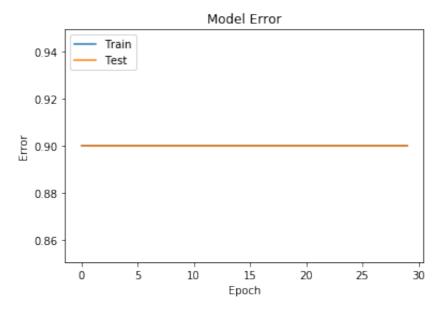
Layer (ty	/pe)	Output	Shape	Param #
dense_52	(Dense)	(None,	30)	23550
dense_53	(Dense)	(None,	30)	930
dense_54	(Dense)	(None,	30)	930
dense_55	(Dense)	(None,	10)	310

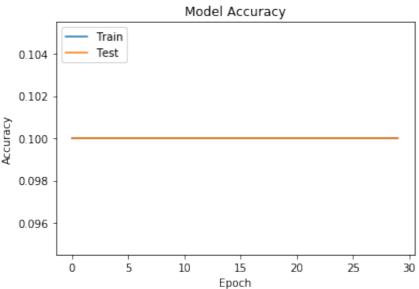
\_\_\_\_\_

```
Total params: 25,720
Trainable params: 25,720
Non-trainable params: 0
```

```
Train on 1000 samples
s: 40.4962 - accuracy: 0.1090
Train on 1000 samples
1000/1000 [============= ] - 0s 226us/sample - los
s: 2.3453 - accuracy: 0.1090
Train on 1000 samples
1000/1000 [============ ] - 0s 280us/sample - los
s: 2.3412 - accuracy: 0.0980
Train on 1000 samples
s: 2.3324 - accuracy: 0.1150
Train on 1000 samples
s: 2.3385 - accuracy: 0.0850
Train on 1000 samples
1000/1000 [============= ] - 0s 272us/sample - los
s: 2.3301 - accuracy: 0.0960
Train on 1000 samples
1000/1000 [============ ] - 0s 246us/sample - los
s: 2.3243 - accuracy: 0.0950
Train on 1000 samples
s: 2.3283 - accuracy: 0.1030
Train on 1000 samples
s: 2.3264 - accuracy: 0.0980
Train on 1000 samples
1000/1000 [============= ] - 0s 233us/sample - los
s: 2.3368 - accuracy: 0.0950
Train on 1000 samples
s: 2.3364 - accuracy: 0.1090
Train on 1000 samples
s: 2.3298 - accuracy: 0.0780
Train on 1000 samples
s: 3.1083 - accuracy: 0.0900
Train on 1000 samples
s: 3.3003 - accuracy: 0.0820
Train on 1000 samples
1000/1000 [============ ] - 0s 204us/sample - los
s: 2.9490 - accuracy: 0.0970
Train on 1000 samples
s: 2.9410 - accuracy: 0.0870
```

```
Train on 1000 samples
s: 2.9352 - accuracy: 0.0870
Train on 1000 samples
s: 3.5555 - accuracy: 0.0890
Train on 1000 samples
s: 3.5269 - accuracy: 0.0780
Train on 1000 samples
1000/1000 [============] - 0s 202us/sample - los
s: 3.1053 - accuracy: 0.1070
Train on 1000 samples
s: 3.0694 - accuracy: 0.0840
Train on 1000 samples
1000/1000 [============ ] - 0s 202us/sample - los
s: 3.3600 - accuracy: 0.0900
Train on 1000 samples
1000/1000 [============ ] - 0s 203us/sample - los
s: 3.4318 - accuracy: 0.0970
Train on 1000 samples
s: 3.2234 - accuracy: 0.0910
Train on 1000 samples
s: 3.1541 - accuracy: 0.0940
Train on 1000 samples
s: 3.4287 - accuracy: 0.0960
Train on 1000 samples
s: 3.3139 - accuracy: 0.0960
Train on 1000 samples
s: 3.1531 - accuracy: 0.0940
Train on 1000 samples
s: 3.2569 - accuracy: 0.1010
Train on 1000 samples
s: 3.4481 - accuracy: 0.0880
```





# Plot for learning rate

In [52]: model=neural\_net\_3hidden(True)
learning\_rate\_plot(25720, model)

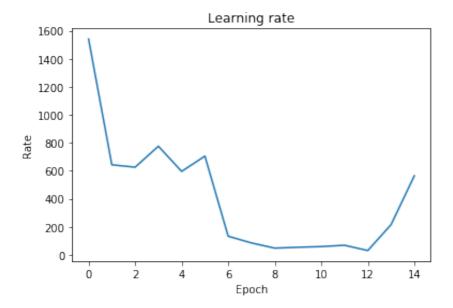
Model: "sequential\_20"

Layer (type)	Output Shape	Param #
dense_56 (Dense)	(None, 30)	23550
dense_57 (Dense)	(None, 30)	930
dense_58 (Dense)	(None, 30)	930
dense_59 (Dense)	(None, 10)	310

Total params: 25,720
Trainable params: 25,720
Non-trainable params: 0

```
Train on 1000 samples
s: 40.6000 - accuracy: 0.0870
Train on 1000 samples
1000/1000 [============ ] - 0s 213us/sample - los
s: 2.3403 - accuracy: 0.0950
Train on 1000 samples
s: 2.3368 - accuracy: 0.09000s - loss: 2.3385 - accuracy: 0.
Train on 1000 samples
s: 2.3359 - accuracy: 0.0980
Train on 1000 samples
1000/1000 [============= ] - 0s 223us/sample - los
s: 2.3272 - accuracy: 0.0920
Train on 1000 samples
s: 2.3344 - accuracy: 0.0980
Train on 1000 samples
s: 2.3256 - accuracy: 0.0910
Train on 1000 samples
s: 2.3293 - accuracy: 0.0920
Train on 1000 samples
s: 2.3290 - accuracy: 0.0880
Train on 1000 samples
s: 2.3378 - accuracy: 0.0920
Train on 1000 samples
s: 2.3372 - accuracy: 0.0870
Train on 1000 samples
s: 2.3270 - accuracy: 0.0900
Train on 1000 samples
1000/1000 [============= ] - 0s 206us/sample - los
s: 3.1312 - accuracy: 0.0720
Train on 1000 samples
s: 3.2935 - accuracy: 0.0900
Train on 1000 samples
s: 2.9718 - accuracy: 0.0920
Train on 1000 samples
1000/1000 [============ ] - 0s 201us/sample - los
s: 2.9480 - accuracy: 0.1140
Train on 1000 samples
```

```
1000/1000 [============ ] - 0s 203us/sample - los
s: 2.9606 - accuracy: 0.0940
Train on 1000 samples
1000/1000 [============ ] - 0s 201us/sample - los
s: 3.5531 - accuracy: 0.1040
Train on 1000 samples
s: 3.5676 - accuracy: 0.0730
Train on 1000 samples
1000/1000 [============= ] - 0s 203us/sample - los
s: 3.1163 - accuracy: 0.0990
Train on 1000 samples
1000/1000 [============ ] - 0s 221us/sample - los
s: 3.0721 - accuracy: 0.0950
Train on 1000 samples
s: 3.2771 - accuracy: 0.0930
Train on 1000 samples
s: 3.4924 - accuracy: 0.0840
Train on 1000 samples
1000/1000 [============= ] - 0s 201us/sample - los
s: 3.2208 - accuracy: 0.0980
Train on 1000 samples
1000/1000 [============ ] - 0s 209us/sample - los
s: 3.1684 - accuracy: 0.0940
Train on 1000 samples
s: 3.3135 - accuracy: 0.0940
Train on 1000 samples
1000/1000 [============ ] - 0s 201us/sample - los
s: 3.4041 - accuracy: 0.0920
Train on 1000 samples
s: 3.2010 - accuracy: 0.0780
Train on 1000 samples
s: 3.2152 - accuracy: 0.0880
Train on 1000 samples
s: 3.4112 - accuracy: 0.0860
```



# Question 2)c) CNN

```
In [53]: from keras.datasets import mnist from keras.models import Sequential from keras.layers import Dense, Dropout, Activation, Flatten from keras.optimizers import Adam from keras.layers.normalization import BatchNormalization from keras.utils import np_utils from keras.layers import Conv2D, MaxPooling2D, ZeroPadding2D, Globa lAveragePooling2D from keras.layers.advanced_activations import LeakyReLU from keras.preprocessing.image import ImageDataGenerator
```

```
In [64]: model = Sequential()
         model.add(Conv2D(32, (3, 3), input_shape=(28,28,1)))
         model.add(LeakyReLU(alpha=0.1))
         model.add(Conv2D(32, (3, 3)))
         model.add(LeakyReLU(alpha=0.1))
         model.add(MaxPooling2D(pool size=(2,2)))
         model.add(Conv2D(64,(3, 3)))
         model.add(LeakyReLU(alpha=0.1))
         model.add(Conv2D(64, (3, 3)))
         model.add(LeakyReLU(alpha=0.1))
         model.add(MaxPooling2D(pool_size=(2,2)))
         model.add(Flatten())
         model.add(Dense(512))
         model.add(LeakyReLU(alpha=0.1))
         model.add(Dropout(0.2))
         model.add(Dense(10))
         model.add(Activation('softmax'))
```

```
In [65]: model.summary()
```

Model: "sequential\_2"

Layer (type)	Output	Shape	Param #
conv2d_5 (Conv2D)	(None,	26, 26, 32)	320
leaky_re_lu_1 (LeakyReLU)	(None,	26, 26, 32)	0
conv2d_6 (Conv2D)	(None,	24, 24, 32)	9248
leaky_re_lu_2 (LeakyReLU)	(None,	24, 24, 32)	0
max_pooling2d_3 (MaxPooling2	(None,	12, 12, 32)	0
conv2d_7 (Conv2D)	(None,	10, 10, 64)	18496
leaky_re_lu_3 (LeakyReLU)	(None,	10, 10, 64)	0
conv2d_8 (Conv2D)	(None,	8, 8, 64)	36928
leaky_re_lu_4 (LeakyReLU)	(None,	8, 8, 64)	0
max_pooling2d_4 (MaxPooling2	(None,	4, 4, 64)	0
flatten_2 (Flatten)	(None,	1024)	0
dense_3 (Dense)	(None,	512)	524800
leaky_re_lu_5 (LeakyReLU)	(None,	512)	0
dropout_2 (Dropout)	(None,	512)	0
dense_4 (Dense)	(None,	10)	5130

Total params: 594,922 Trainable params: 594,922 Non-trainable params: 0

```
In [66]: mini_train=mini_train.reshape((1000,28,28,1))
   mini_test=mini_test.reshape((1000,28,28,1))
```

```
In [67]: mini_train.shape,mini_test.shape
```

```
Out[67]: ((1000, 28, 28, 1), (1000, 28, 28, 1))
```

```
In [68]: train_labels.shape,test_labels.shape
```

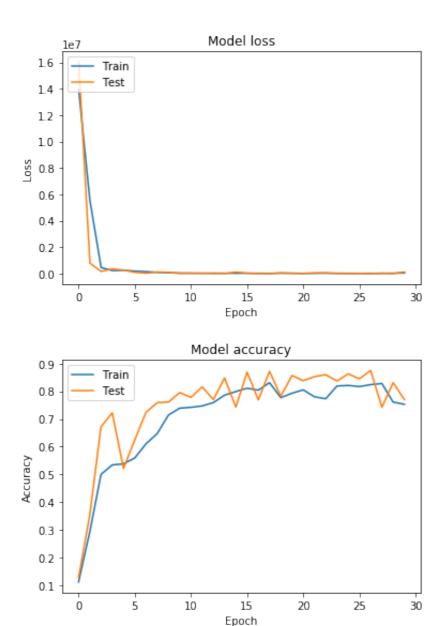
Out[68]: ((1000, 10), (1000, 10))

```
In [69]: model.compile(loss='categorical crossentropy', optimizer=Adam(learn
       ing rate=0.1), metrics=['accuracy'])
In [70]: train gen = ImageDataGenerator(rotation range=3, width shift range=
       3, height shift range=3)
       test gen = ImageDataGenerator()
In [71]: train generator = train gen.flow(mini train, train labels, batch si
       ze=10)
       test generator = test gen.flow(mini test, test labels, batch size=1
       0)
In [72]: history=model.fit_generator(train generator, steps per epoch=1000//
       10, epochs=30, validation data=test generator, validation steps=100
       0//10)
       Epoch 1/30
       100/100 [============= ] - 5s 49ms/step - loss: 13
       943718.7768 - accuracy: 0.1110 - val loss: 16028184.0000 - val acc
       uracy: 0.1260
       Epoch 2/30
       100/100 [=============] - 4s 44ms/step - loss: 55
       29694.5050 - accuracy: 0.2920 - val loss: 788680.6250 - val accura
       cy: 0.3530
       Epoch 3/30
       100/100 [============== ] - 4s 44ms/step - loss: 45
       1844.4041 - accuracy: 0.5000 - val loss: 159742.6250 - val accurac
       y: 0.6720
       Epoch 4/30
       2572.9458 - accuracy: 0.5340 - val loss: 352928.4062 - val accurac
       y: 0.7220
       Epoch 5/30
       100/100 [============= ] - 4s 43ms/step - loss: 24
       8917.9212 - accuracy: 0.5380 - val loss: 259498.5000 - val accurac
       y: 0.5210
       Epoch 6/30
       100/100 [============= ] - 5s 45ms/step - loss: 18
       4655.6699 - accuracy: 0.5590 - val loss: 90099.7891 - val accuracy
       : 0.6240
       Epoch 7/30
       4813.8153 - accuracy: 0.6100 - val_loss: 25548.2871 - val accuracy
       : 0.7240
       Epoch 8/30
       076.4708 - accuracy: 0.6470 - val loss: 118837.2266 - val accuracy
       : 0.7590
       Epoch 9/30
```

```
656.4339 - accuracy: 0.7140 - val loss: 87568.8125 - val accuracy:
0.7610
Epoch 10/30
100/100 [============= ] - 5s 46ms/step - loss: 39
367.4359 - accuracy: 0.7390 - val loss: 13037.2109 - val accuracy:
0.7950
Epoch 11/30
173.6685 - accuracy: 0.7420 - val loss: 27648.9473 - val accuracy:
0.7780
Epoch 12/30
554.4343 - accuracy: 0.7470 - val loss: 21893.1211 - val accuracy:
0.8160
Epoch 13/30
150.1836 - accuracy: 0.7590 - val loss: 5697.6123 - val accuracy:
0.7700
Epoch 14/30
701.2793 - accuracy: 0.7860 - val loss: 11643.3145 - val accuracy:
0.8480
Epoch 15/30
100/100 [============= ] - 5s 47ms/step - loss: 20
547.3794 - accuracy: 0.7990 - val_loss: 110618.9766 - val_accuracy
: 0.7430
Epoch 16/30
918.9665 - accuracy: 0.8110 - val loss: 34346.0859 - val accuracy:
0.8690
Epoch 17/30
114.8533 - accuracy: 0.8040 - val loss: 3444.9390 - val accuracy:
0.7690
Epoch 18/30
100/100 [============= ] - 5s 47ms/step - loss: 15
939.2175 - accuracy: 0.8310 - val loss: 0.0000e+00 - val accuracy:
0.8720
Epoch 19/30
562.0061 - accuracy: 0.7770 - val_loss: 52312.0234 - val_accuracy:
0.7820
Epoch 20/30
577.5042 - accuracy: 0.7930 - val loss: 24177.2129 - val accuracy:
0.8570
Epoch 21/30
100/100 [============] - 5s 46ms/step - loss: 14
180.6516 - accuracy: 0.8050 - val loss: 9576.2246 - val accuracy:
0.8380
Epoch 22/30
100/100 [============= ] - 4s 45ms/step - loss: 21
023.0222 - accuracy: 0.7800 - val loss: 51836.2148 - val accuracy:
```

```
0.8530
Epoch 23/30
084.9780 - accuracy: 0.7730 - val loss: 56501.9062 - val accuracy:
0.8600
Epoch 24/30
363.0631 - accuracy: 0.8190 - val loss: 11489.7939 - val accuracy:
0.8370
Epoch 25/30
100/100 [============ ] - 5s 45ms/step - loss: 14
810.4845 - accuracy: 0.8210 - val loss: 0.0000e+00 - val accuracy:
0.8630
Epoch 26/30
678.5440 - accuracy: 0.8170 - val loss: 15532.1309 - val accuracy:
0.8450
Epoch 27/30
543.7392 - accuracy: 0.8240 - val_loss: 0.0000e+00 - val_accuracy:
0.8750
Epoch 28/30
396.9999 - accuracy: 0.8280 - val loss: 29576.3809 - val accuracy:
0.7420
Epoch 29/30
100/100 [============ ] - 4s 42ms/step - loss: 25
180.8621 - accuracy: 0.7610 - val loss: 0.0000e+00 - val accuracy:
0.8310
Epoch 30/30
357.7607 - accuracy: 0.7530 - val_loss: 120708.9219 - val_accuracy
: 0.7700
```

```
In [73]: model_training_plot()
```



#### References:

https://yashk2810.github.io/Applying-Convolutional-Neural-Network-on-the-MNIST-dataset/ (https://yashk2810.github.io/Applying-Convolutional-Neural-Network-on-the-MNIST-dataset/)

https://rohanvarma.me/Regularization/ (https://rohanvarma.me/Regularization/)

https://ljvmiranda921.github.io/notebook/2017/08/13/softmax-and-the-negative-log-likelihood/ (https://ljvmiranda921.github.io/notebook/2017/08/13/softmax-and-the-negative-log-likelihood/)