

## Equivalence between maximizing log-likelihood of label and minimizing criterion function with negative log likelihood

Aiming to minimize the log likelihood function given by:

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

for m ranging from digit 0 to 9

Let  $p(t_m|x; w)$  be a categorical probability distribution :

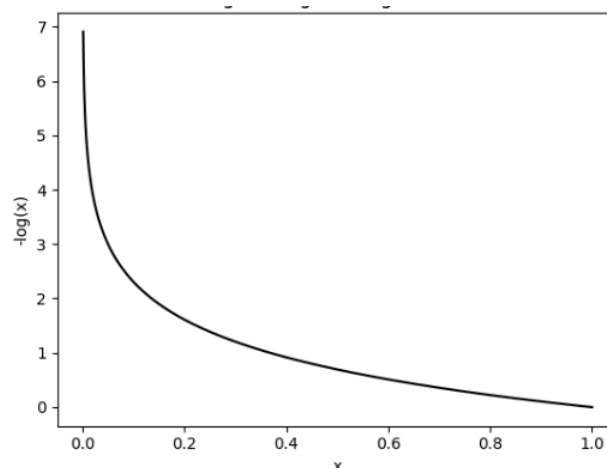
$$p(t_m|x; w) = y_m(x; w)^{t_m}$$

$$t_m \in \{0, 1\} \text{ and } \sum_m t_m = 1$$

Let  $y_m(x; w)$  be softmax probabilities from the output nodes i.e.

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

The curve for negative log-likelihood looks as follows:



which means for it to be minimum the value of the likelihood must be close to 1.

The negative log likelihood can be reduced to :

$$J(w) = -\sum_n \sum_{m=0}^9 t_m^{(n)} \log(y_m^{(n)}(x^{(n)}; w))$$

$t_m$  is one-hot encoded vector thus for all wrong class labels the terms reduces to 0. If we denote the correct label to be 'c' for every feature sample we get:

$$J(w) = -\sum_n 1 * \log(y_c^{(n)}(x^{(n)}; w)) = -\log(\prod_n (y_c^{(n)}(x^{(n)}; w)))$$

To minimize the negative log-likelihood function, we need to maximize  $(\prod_n (y_c^{(n)}(x^{(n)}; w)))$  to make it as close to 1 as possible.

Thus we can prove that a neural network to maximize the log likelihood of the correct label is one that has softmax output nodes and minimizes the criterion function of the negative log probability of training data set.

### Equivalence between maximizing posterior likelihood of training data and minimizing cost function with L2 regularization

Let's say that the output  $y$  is dependent on input feature  $x$  using the following equation:

$$y_i = wx_i + \epsilon$$

where  $w$  is the set of weights and  $\epsilon$  is Gaussian noise  $\mathcal{N}(0, \sigma^2)$

$w$  also has the Gaussian prior  $\mathcal{N}(0, \alpha^{-1})$

Using Bayes Rule:

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

Ignoring  $P(D)$  we can write:

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

The posterior probability we aim to maximize over  $w$  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 and ignoring any constants:

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

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

This proves that maximizing the posterior likelihood of observing the training data is the same as minimizing the cost function with L2 regularization with  $\alpha$  being the regularization constant.

The more we maximize the log of posterior likelihood of training data  $\log P(w|D)$ , the more cost function with L2 regularization  $\sum_{i=1}^N (y_i - wx_i)^2 + \alpha w^2$  is minimized.

### Importing the Library functions

```
In [91]: import numpy as np
import tensorflow as tf
from keras.datasets import mnist
from numpy import loadtxt
from keras.models import Sequential
from keras.layers import Dense
(X_train, y_train), (X_test, y_test) = mnist.load_data()
```

```
In [92]: import matplotlib.pyplot as plt
```

```
In [93]: X_train.shape
```

```
Out[93]: (60000, 28, 28)
```

## Reshaping the data

```
In [94]: X_train=X_train.reshape(60000,784)
X_test=X_test.reshape(10000,784)
y_train.shape
```

```
Out[94]: (60000,)
```

## Function to extraxt 1000 training and testing data with each of the 10 classes having 100 samples of data

```
In [95]: def get_data(x_data,y_data):
    data_list=[]
    label_list=[]
    count=[0,0,0,0,0,0,0,0,0,0]
    flag=[0,0,0,0,0,0,0,0,0,0]
    for i in range(len(x_data)):
        c=0
        if(count[y_data[i]]<100):
            count[y_data[i]]+=1
            data_list.append(x_data[i])
            label_list.append(y_data[i])
        else:
            continue
        for j in range(len(flag)):
            if(flag[j]==1):
                c+=1
        if(c==10):
            break
    return np.array(data_list),np.array(label_list)
```

```
In [96]: X_train,y_train=get_data(X_train,y_train)
X_test, y_test=get_data(X_test, y_test)
```

```
In [97]: X_train.shape
```

```
Out[97]: (1000, 784)
```

```
In [98]: y_train.shape
```

```
Out[98]: (1000,)
```

## Performing One Hot encoding on the training and testing label data

```
In [99]: y_z=np.zeros((1000,10))
         for i in range(len(y_z)):
             y_z[i][y_train[i]]=1
         y_train=y_z

         y_z=np.zeros((1000,10))
         for i in range(len(y_z)):
             y_z[i][y_test[i]]=1
         y_test=y_z
```

```
In [100]: y_train.shape
```

```
Out[100]: (1000, 10)
```

## Normalising the data

```
In [101]: #Normalizing the data
         from sklearn.preprocessing import StandardScaler
         sc = StandardScaler()
         X_train = sc.fit_transform(X_train)
         X_test=sc.fit_transform(X_test)
```

## Defining the Models with the given parameters in the question

### Model with One Layer

```
In [102]: def model_nn(X,Y,L2):
            model = Sequential()
            if(L2==True):
                model.add(Dense(X, input_dim=784, activation='sigmoid', kernel_regularizer=tf.keras.regularizers.l2(5)))
            else:
                model.add(Dense(X, input_dim=784, activation='sigmoid'))
            model.add(Dense(Y, activation='softmax'))
            model.summary()
            adam=tf.keras.optimizers.Adam(learning_rate=0.1)
            model.compile(loss='categorical_crossentropy', optimizer=adam, metrics=['accuracy'])

            return model
```

## Model with Two Layers

```
In [14]: def model_nn_2(X,Y,L2):
            model = Sequential()
            if(L2==True):
                model.add(Dense(X, input_dim=784, activation='sigmoid', kernel_regularizer=tf.keras.regularizers.l2(5)))
                model.add(Dense(X, input_dim=30, activation='sigmoid', kernel_regularizer=tf.keras.regularizers.l2(5)))
            else:
                model.add(Dense(X, input_dim=784, activation='sigmoid'))
                model.add(Dense(X, input_dim=30, activation='sigmoid'))

            model.add(Dense(Y, activation='softmax'))
            model.summary()
            adam=tf.keras.optimizers.Adam(learning_rate=0.1)
            model.compile(loss='categorical_crossentropy', optimizer=adam, metrics=['accuracy'])

            return model
```

## Model with Three Layers

```
In [15]: def model_nn_3(X,Y,L2):
          model = Sequential()
          if(L2==True):
              model.add(Dense(X, input_dim=784, activation='sigmoid',kernel_regularizer=tf.keras.regularizers.l2(5)))
              model.add(Dense(X, input_dim=30, activation='sigmoid',kernel_regularizer=tf.keras.regularizers.l2(5)))
              model.add(Dense(X, input_dim=30, activation='sigmoid',kernel_regularizer=tf.keras.regularizers.l2(5)))
          else:
              model.add(Dense(X, input_dim=784, activation='sigmoid'))
              model.add(Dense(X, input_dim=30, activation='sigmoid'))
              model.add(Dense(X, input_dim=30, activation='sigmoid'))

          model.add(Dense(Y, activation='softmax'))
          model.summary()
          adam=tf.keras.optimizers.Adam(learning_rate=0.1)
          model.compile(loss='categorical_crossentropy', optimizer=adam,metrics=[ 'accuracy' ])

          return model
```

### Function for Model Accuracy

```
In [16]: def model_accuracy():

          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()
```

### Function for Model Loss

```
In [17]: def model_loss():
          import matplotlib.pyplot as plt
          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()
```

### Function for Zero-one error

```

In [18]: def error_model(model):

    X_train_err=[]
    X_train_acc=[]
    X_test_err=[]
    X_test_acc=[]

    for i in range(30):
        model.fit(X_train,y_train,epochs=1,batch_size=10)
        pred_train=model.predict_classes(X_train)
        pred_test=model.predict_classes(X_test)

        hits_train=0
        for i in range(1000):
            if(y_train[i][pred_train[i]]==1):
                hits_train+=1

        X_train_acc.append(hits_train/1000)
        X_train_err.append(1-(hits_train/1000))

        hits_test=0
        for i in range(1000):
            if(y_test[i][pred_test[i]]==1):
                hits_test+=1

        X_test_acc.append(hits_test/1000)
        X_test_err.append(1-(hits_test/1000))

    plt.plot(X_train_err)
    plt.plot(X_test_err)
    plt.title('Model Error')
    plt.ylabel('Error')
    plt.xlabel('Epoch')
    plt.legend(['Train', 'Test'], loc='upper left')
    plt.show()

    plt.plot(X_train_acc)
    plt.plot(X_test_acc)
    plt.title('Model Accuracy')
    plt.ylabel('Accuracy')
    plt.xlabel('Epoch')
    plt.legend(['Train', 'Test'], loc='upper left')
    plt.show()

```

## Function for Learning Speed

```
In [19]: def LearningRate(model,parameters):
    i=0
    W_diff_list=[]
    #model=model_nn(30,10)
    while(i<30):
        W_old=[]
        W_new=[]
        W_diff=[]

        old=model.get_weights()
        for j in range(len(old)):
            W_old.append(old[j])

        model.fit(X_train,y_train,validation_data=(X_test,y_test),e
pochs=1,batch_size=10)

        new=model.get_weights()
        for j in range(len(new)):
            W_new.append(new[j])

        for j in range(len(W_new)):
            W_diff.append(np.absolute((W_old[j]- W_new[j])/W_new[j]
))

        W_diff_avg=0
        for j in range(len(W_diff)):
            W_diff_avg+=(np.sum(W_diff[j]))
        W_diff_avg/=(parameters)
        W_diff_list.append(W_diff_avg)
        i+=1
    return W_diff_list
```

## Learning Speed Plot function

```
In [20]: def LearningRate_plot(W_diff_list):
    plt.plot(W_diff_list)
    plt.title('Learning Rate of Hidden Layers')
    plt.ylabel('Learning rate')
    plt.xlabel('Epoch')
    plt.show()
```

## One Layer Model without regularization

```
In [112]: model=model_nn(30,10,False)
history = model.fit(X_train,y_train,validation_data=(X_test,y_test)
,epochs=30, batch_size=10)

Model: "sequential_28"
```



```
model = Sequential_20
```

Layer (type)	Output Shape	Param #
dense_75 (Dense)	(None, 30)	23550
dense_76 (Dense)	(None, 10)	310

```
Total params: 23,860
```

```
Trainable params: 23,860
```

```
Non-trainable params: 0
```

```
Train on 1000 samples, validate on 1000 samples
```

```
Epoch 1/30
```

```
1000/1000 [=====] - 0s 315us/step - loss: 1.3813 - accuracy: 0.5820 - val_loss: 1.1861 - val_accuracy: 0.6080
```

```
Epoch 2/30
```

```
1000/1000 [=====] - 0s 242us/step - loss: 0.8243 - accuracy: 0.7420 - val_loss: 1.0511 - val_accuracy: 0.7020
```

```
Epoch 3/30
```

```
1000/1000 [=====] - 0s 244us/step - loss: 0.7777 - accuracy: 0.7810 - val_loss: 1.0581 - val_accuracy: 0.6830
```

```
Epoch 4/30
```

```
1000/1000 [=====] - 0s 248us/step - loss: 0.6566 - accuracy: 0.8030 - val_loss: 1.1093 - val_accuracy: 0.6890
```

```
Epoch 5/30
```

```
1000/1000 [=====] - 0s 280us/step - loss: 0.7526 - accuracy: 0.7960 - val_loss: 1.0422 - val_accuracy: 0.7220
```

```
Epoch 6/30
```

```
1000/1000 [=====] - 0s 270us/step - loss: 0.6560 - accuracy: 0.8060 - val_loss: 1.2998 - val_accuracy: 0.6580
```

```
Epoch 7/30
```

```
1000/1000 [=====] - 0s 257us/step - loss: 0.5523 - accuracy: 0.8380 - val_loss: 1.3144 - val_accuracy: 0.7020
```

```
Epoch 8/30
```

```
1000/1000 [=====] - 0s 310us/step - loss: 0.5658 - accuracy: 0.8280 - val_loss: 1.1843 - val_accuracy: 0.6970
```

```
Epoch 9/30
```

```
1000/1000 [=====] - 0s 238us/step - loss: 0.5778 - accuracy: 0.8310 - val_loss: 1.6351 - val_accuracy: 0.6230
```

```
Epoch 10/30
```

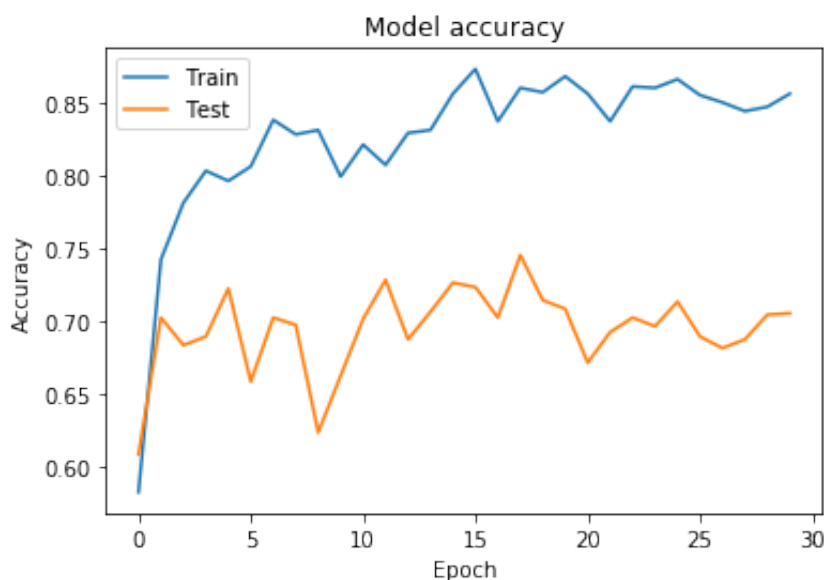
```
1000/1000 [=====] - 0s 276us/step - loss: 0.6377 - accuracy: 0.7990 - val_loss: 1.5187 - val_accuracy: 0.6620
```

```
Epoch 11/30
```

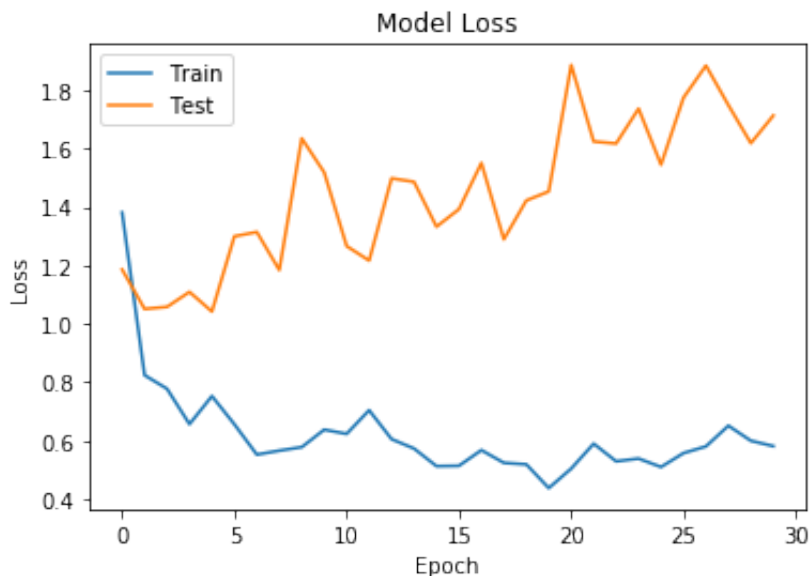
```
1000/1000 [=====] - 0s 283us/step - loss:
0.6232 - accuracy: 0.8210 - val_loss: 1.2663 - val_accuracy: 0.701
0
Epoch 12/30
1000/1000 [=====] - 0s 259us/step - loss:
0.7043 - accuracy: 0.8070 - val_loss: 1.2167 - val_accuracy: 0.728
0
Epoch 13/30
1000/1000 [=====] - 0s 264us/step - loss:
0.6053 - accuracy: 0.8290 - val_loss: 1.4983 - val_accuracy: 0.687
0
Epoch 14/30
1000/1000 [=====] - 0s 252us/step - loss:
0.5730 - accuracy: 0.8310 - val_loss: 1.4864 - val_accuracy: 0.706
0
Epoch 15/30
1000/1000 [=====] - 0s 321us/step - loss:
0.5127 - accuracy: 0.8560 - val_loss: 1.3327 - val_accuracy: 0.726
0
Epoch 16/30
1000/1000 [=====] - 0s 295us/step - loss:
0.5138 - accuracy: 0.8730 - val_loss: 1.3927 - val_accuracy: 0.723
0
Epoch 17/30
1000/1000 [=====] - 0s 310us/step - loss:
0.5675 - accuracy: 0.8370 - val_loss: 1.5509 - val_accuracy: 0.702
0
Epoch 18/30
1000/1000 [=====] - 0s 297us/step - loss:
0.5240 - accuracy: 0.8600 - val_loss: 1.2896 - val_accuracy: 0.745
0
Epoch 19/30
1000/1000 [=====] - 0s 355us/step - loss:
0.5188 - accuracy: 0.8570 - val_loss: 1.4224 - val_accuracy: 0.714
0
Epoch 20/30
1000/1000 [=====] - 0s 437us/step - loss:
0.4376 - accuracy: 0.8680 - val_loss: 1.4534 - val_accuracy: 0.708
0
Epoch 21/30
1000/1000 [=====] - 0s 355us/step - loss:
0.5043 - accuracy: 0.8560 - val_loss: 1.8864 - val_accuracy: 0.671
0
Epoch 22/30
1000/1000 [=====] - 0s 345us/step - loss:
0.5897 - accuracy: 0.8370 - val_loss: 1.6244 - val_accuracy: 0.692
0
Epoch 23/30
1000/1000 [=====] - 0s 308us/step - loss:
0.5293 - accuracy: 0.8610 - val_loss: 1.6176 - val_accuracy: 0.702
0
Epoch 24/30
1000/1000 [=====] - 0s 326us/step - loss:
```

```
0.5387 - accuracy: 0.8600 - val_loss: 1.7369 - val_accuracy: 0.696
0
Epoch 25/30
1000/1000 [=====] - 0s 287us/step - loss:
0.5098 - accuracy: 0.8660 - val_loss: 1.5451 - val_accuracy: 0.713
0
Epoch 26/30
1000/1000 [=====] - 0s 294us/step - loss:
0.5573 - accuracy: 0.8550 - val_loss: 1.7745 - val_accuracy: 0.689
0
Epoch 27/30
1000/1000 [=====] - 0s 263us/step - loss:
0.5803 - accuracy: 0.8500 - val_loss: 1.8844 - val_accuracy: 0.681
0
Epoch 28/30
1000/1000 [=====] - 0s 321us/step - loss:
0.6512 - accuracy: 0.8440 - val_loss: 1.7489 - val_accuracy: 0.687
0
Epoch 29/30
1000/1000 [=====] - 0s 270us/step - loss:
0.5997 - accuracy: 0.8470 - val_loss: 1.6187 - val_accuracy: 0.704
0
Epoch 30/30
1000/1000 [=====] - 0s 266us/step - loss:
0.5813 - accuracy: 0.8560 - val_loss: 1.7132 - val_accuracy: 0.705
0
```

```
In [113]: model_accuracy()
```



```
In [114]: model_loss()
```



```
In [115]: model=model_nn(30,10,False)
          error_model(model)
```

Model: "sequential\_29"

Layer (type)	Output Shape	Param #
dense_77 (Dense)	(None, 30)	23550
dense_78 (Dense)	(None, 10)	310

Total params: 23,860

Trainable params: 23,860

Non-trainable params: 0

Epoch 1/1

1000/1000 [=====] - 0s 277us/step - loss: 1.2557 - accuracy: 0.5900

Epoch 1/1

1000/1000 [=====] - 0s 228us/step - loss: 0.8311 - accuracy: 0.7400

Epoch 1/1

1000/1000 [=====] - 0s 232us/step - loss: 0.8485 - accuracy: 0.7470

Epoch 1/1

1000/1000 [=====] - 0s 184us/step - loss: 0.8604 - accuracy: 0.7410

Epoch 1/1

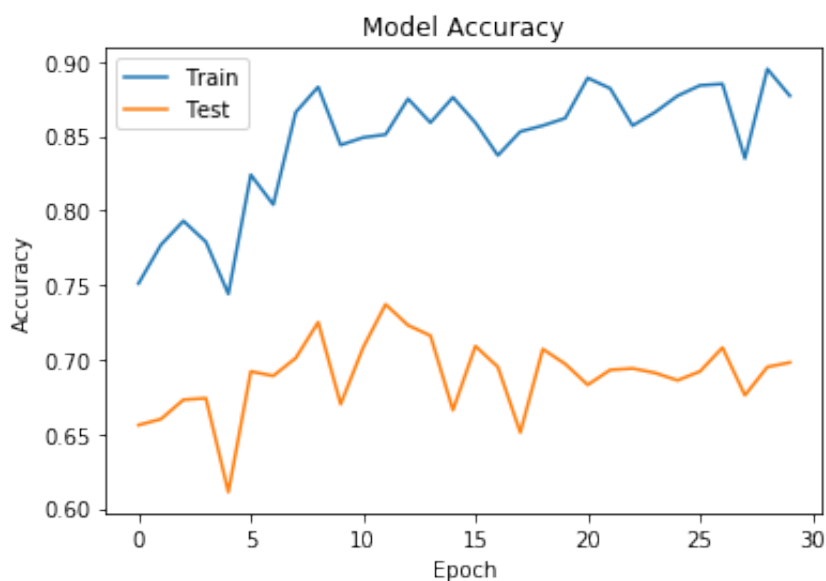
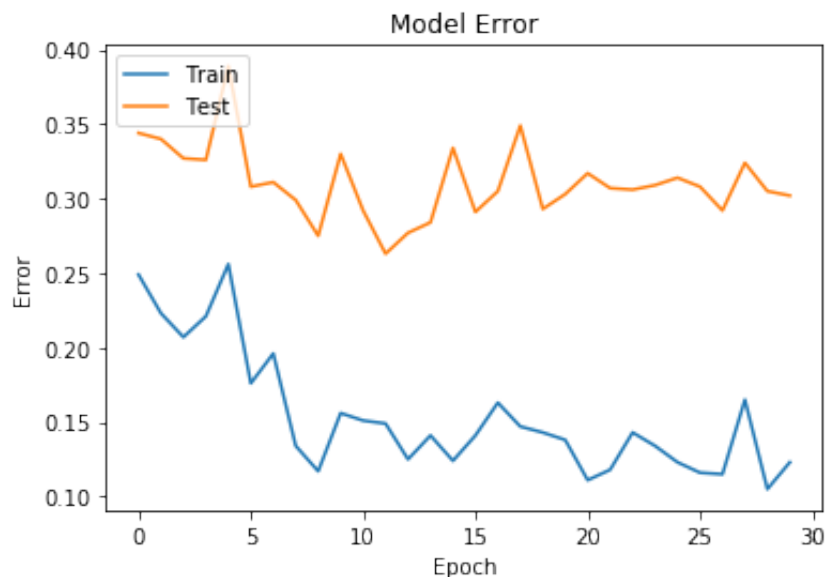
1000/1000 [=====] - 0s 152us/step - loss: 0.8221 - accuracy: 0.7730

Epoch 1/1

1000/1000 [=====] - 0s 149us/step - loss: 0.7245 - accuracy: 0.7930

```
Epoch 1/1
1000/1000 [=====] - 0s 152us/step - loss:
0.6639 - accuracy: 0.8000
Epoch 1/1
1000/1000 [=====] - 0s 181us/step - loss:
0.6385 - accuracy: 0.8310
Epoch 1/1
1000/1000 [=====] - 0s 171us/step - loss:
0.6803 - accuracy: 0.8090
Epoch 1/1
1000/1000 [=====] - 0s 181us/step - loss:
0.5553 - accuracy: 0.8350
Epoch 1/1
1000/1000 [=====] - 0s 175us/step - loss:
0.5202 - accuracy: 0.8410
Epoch 1/1
1000/1000 [=====] - 0s 177us/step - loss:
0.6903 - accuracy: 0.8150
Epoch 1/1
1000/1000 [=====] - 0s 145us/step - loss:
0.5876 - accuracy: 0.8540
Epoch 1/1
1000/1000 [=====] - 0s 146us/step - loss:
0.6494 - accuracy: 0.8300
Epoch 1/1
1000/1000 [=====] - 0s 147us/step - loss:
0.5848 - accuracy: 0.8570
Epoch 1/1
1000/1000 [=====] - 0s 147us/step - loss:
0.5124 - accuracy: 0.8560
Epoch 1/1
1000/1000 [=====] - 0s 147us/step - loss:
0.5490 - accuracy: 0.8400
Epoch 1/1
1000/1000 [=====] - 0s 147us/step - loss:
0.5866 - accuracy: 0.8290
Epoch 1/1
1000/1000 [=====] - 0s 152us/step - loss:
0.6858 - accuracy: 0.8280
Epoch 1/1
1000/1000 [=====] - 0s 166us/step - loss:
0.7074 - accuracy: 0.8320
Epoch 1/1
1000/1000 [=====] - 0s 195us/step - loss:
0.6447 - accuracy: 0.8380
Epoch 1/1
1000/1000 [=====] - 0s 196us/step - loss:
0.5707 - accuracy: 0.8480
Epoch 1/1
1000/1000 [=====] - 0s 183us/step - loss:
0.4940 - accuracy: 0.8660
Epoch 1/1
1000/1000 [=====] - 0s 174us/step - loss:
```

```
0.5199 - accuracy: 0.8690
Epoch 1/1
1000/1000 [=====] - 0s 154us/step - loss:
0.5080 - accuracy: 0.8600
Epoch 1/1
1000/1000 [=====] - 0s 148us/step - loss:
0.5122 - accuracy: 0.8660
Epoch 1/1
1000/1000 [=====] - 0s 146us/step - loss:
0.5599 - accuracy: 0.8480
Epoch 1/1
1000/1000 [=====] - 0s 153us/step - loss:
0.5447 - accuracy: 0.8690
Epoch 1/1
1000/1000 [=====] - 0s 151us/step - loss:
0.5849 - accuracy: 0.8500
Epoch 1/1
1000/1000 [=====] - 0s 147us/step - loss:
0.5972 - accuracy: 0.8410
```



```
In [116]: model=model_nn(30,10,False)
          W=LearningRate(model,(23550+310))
```

Model: "sequential\_30"

Layer (type)	Output Shape	Param #
dense_79 (Dense)	(None, 30)	23550
dense_80 (Dense)	(None, 10)	310

Total params: 23,860

Trainable params: 23,860

Non-trainable params: 0

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 391us/step - loss: 1.3404 - accuracy: 0.5840 - val\_loss: 1.1035 - val\_accuracy: 0.6510

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 366us/step - loss: 0.8214 - accuracy: 0.7520 - val\_loss: 1.0649 - val\_accuracy: 0.6710

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 340us/step - loss: 0.8646 - accuracy: 0.7170 - val\_loss: 1.5765 - val\_accuracy: 0.6110

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 323us/step - loss: 0.8428 - accuracy: 0.7640 - val\_loss: 0.9702 - val\_accuracy: 0.7160

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 299us/step - loss: 0.7332 - accuracy: 0.7820 - val\_loss: 1.0934 - val\_accuracy: 0.7170

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 254us/step - loss: 0.7188 - accuracy: 0.7840 - val\_loss: 1.1495 - val\_accuracy: 0.7130

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 302us/step - loss: 0.7578 - accuracy: 0.7890 - val\_loss: 1.2277 - val\_accuracy: 0.7070

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 241us/step - loss:

```
0.7382 - accuracy: 0.7950 - val_loss: 1.3397 - val_accuracy: 0.688
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 341us/step - loss:
0.6214 - accuracy: 0.8250 - val_loss: 1.2497 - val_accuracy: 0.708
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 292us/step - loss:
0.6024 - accuracy: 0.8160 - val_loss: 1.1809 - val_accuracy: 0.738
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 269us/step - loss:
0.5961 - accuracy: 0.8400 - val_loss: 1.3578 - val_accuracy: 0.692
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 232us/step - loss:
0.6621 - accuracy: 0.8120 - val_loss: 1.3480 - val_accuracy: 0.698
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 268us/step - loss:
0.5636 - accuracy: 0.8380 - val_loss: 1.2664 - val_accuracy: 0.718
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 323us/step - loss:
0.5421 - accuracy: 0.8490 - val_loss: 1.3030 - val_accuracy: 0.707
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 236us/step - loss:
0.4924 - accuracy: 0.8600 - val_loss: 1.3042 - val_accuracy: 0.728
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 256us/step - loss:
0.4973 - accuracy: 0.8540 - val_loss: 1.6256 - val_accuracy: 0.699
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 305us/step - loss:
0.5775 - accuracy: 0.8520 - val_loss: 1.3143 - val_accuracy: 0.750
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 291us/step - loss:
0.5433 - accuracy: 0.8460 - val_loss: 1.3032 - val_accuracy: 0.733
0
Train on 1000 samples, validate on 1000 samples
```



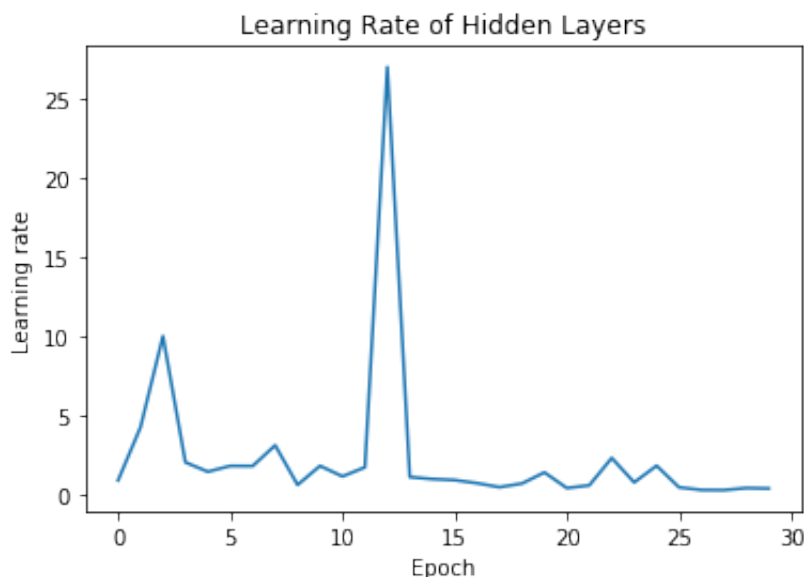
```
Epoch 1/1
1000/1000 [=====] - 0s 262us/step - loss:
0.5624 - accuracy: 0.8460 - val_loss: 1.4631 - val_accuracy: 0.705
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 335us/step - loss:
0.5327 - accuracy: 0.8400 - val_loss: 1.5068 - val_accuracy: 0.705
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 268us/step - loss:
0.5627 - accuracy: 0.8460 - val_loss: 1.4558 - val_accuracy: 0.726
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 253us/step - loss:
0.6012 - accuracy: 0.8470 - val_loss: 1.5806 - val_accuracy: 0.735
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 262us/step - loss:
0.5935 - accuracy: 0.8570 - val_loss: 1.6910 - val_accuracy: 0.704
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 248us/step - loss:
0.6661 - accuracy: 0.8270 - val_loss: 1.4802 - val_accuracy: 0.706
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 253us/step - loss:
0.6006 - accuracy: 0.8370 - val_loss: 1.4876 - val_accuracy: 0.721
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 254us/step - loss:
0.6010 - accuracy: 0.8570 - val_loss: 1.5184 - val_accuracy: 0.715
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 256us/step - loss:
0.5882 - accuracy: 0.8500 - val_loss: 1.6943 - val_accuracy: 0.682
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 313us/step - loss:
0.5309 - accuracy: 0.8700 - val_loss: 1.7444 - val_accuracy: 0.668
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 302us/step - loss:
0.5464 - accuracy: 0.8560 - val_loss: 1.8721 - val_accuracy: 0.671
```

```

0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 283us/step - loss:
0.5182 - accuracy: 0.8620 - val_loss: 1.5086 - val_accuracy: 0.712
0

```

In [117]: LearningRate\_plot(W)



## One Layer Model with regularization

```

In [118]: model=model_nn(30,10,True)
history = model.fit(X_train,y_train,validation_data=(X_test,y_test)
,epochs=30, batch_size=10)

```

Model: "sequential\_31"

Layer (type)	Output Shape	Param #
dense_81 (Dense)	(None, 30)	23550
dense_82 (Dense)	(None, 10)	310

Total params: 23,860

Trainable params: 23,860

Non-trainable params: 0

Train on 1000 samples, validate on 1000 samples

Epoch 1/30

```

1000/1000 [=====] - 0s 332us/step - loss:
30.7283 - accuracy: 0.1450 - val_loss: 3.5953 - val_accuracy: 0.20
70

```

Epoch 2/30

```

1000/1000 [=====] - 0s 255us/step - loss:

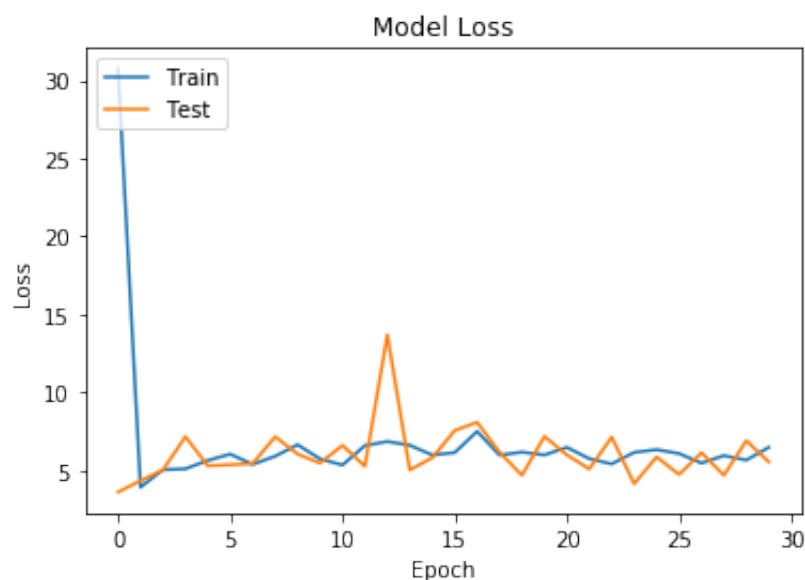
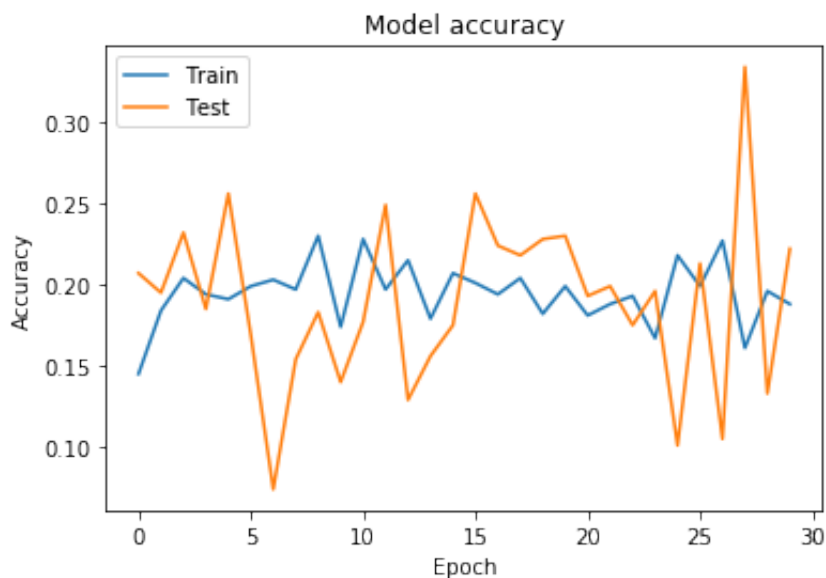
```

```
3.9097 - accuracy: 0.1840 - val_loss: 4.3267 - val_accuracy: 0.195
0
Epoch 3/30
1000/1000 [=====] - 0s 254us/step - loss:
5.0275 - accuracy: 0.2040 - val_loss: 5.0154 - val_accuracy: 0.232
0
Epoch 4/30
1000/1000 [=====] - 0s 258us/step - loss:
5.0857 - accuracy: 0.1940 - val_loss: 7.1511 - val_accuracy: 0.185
0
Epoch 5/30
1000/1000 [=====] - 0s 254us/step - loss:
5.6280 - accuracy: 0.1910 - val_loss: 5.2823 - val_accuracy: 0.256
0
Epoch 6/30
1000/1000 [=====] - 0s 263us/step - loss:
6.0281 - accuracy: 0.1990 - val_loss: 5.3458 - val_accuracy: 0.169
0
Epoch 7/30
1000/1000 [=====] - 0s 262us/step - loss:
5.3717 - accuracy: 0.2030 - val_loss: 5.4259 - val_accuracy: 0.074
0
Epoch 8/30
1000/1000 [=====] - 0s 259us/step - loss:
5.8973 - accuracy: 0.1970 - val_loss: 7.1485 - val_accuracy: 0.154
0
Epoch 9/30
1000/1000 [=====] - 0s 257us/step - loss:
6.6315 - accuracy: 0.2300 - val_loss: 6.0361 - val_accuracy: 0.183
0
Epoch 10/30
1000/1000 [=====] - 0s 277us/step - loss:
5.7259 - accuracy: 0.1740 - val_loss: 5.4541 - val_accuracy: 0.140
0
Epoch 11/30
1000/1000 [=====] - 0s 281us/step - loss:
5.3292 - accuracy: 0.2280 - val_loss: 6.5923 - val_accuracy: 0.177
0
Epoch 12/30
1000/1000 [=====] - 0s 304us/step - loss:
6.5765 - accuracy: 0.1970 - val_loss: 5.2524 - val_accuracy: 0.249
0
Epoch 13/30
1000/1000 [=====] - 0s 325us/step - loss:
6.8387 - accuracy: 0.2150 - val_loss: 13.6791 - val_accuracy: 0.12
90
Epoch 14/30
1000/1000 [=====] - 0s 325us/step - loss:
6.6008 - accuracy: 0.1790 - val_loss: 5.0134 - val_accuracy: 0.156
0
Epoch 15/30
1000/1000 [=====] - 0s 299us/step - loss:
5.9707 - accuracy: 0.2070 - val_loss: 5.7958 - val_accuracy: 0.175
```

```
0
Epoch 16/30
1000/1000 [=====] - 0s 255us/step - loss:
6.1286 - accuracy: 0.2010 - val_loss: 7.5381 - val_accuracy: 0.256
0
Epoch 17/30
1000/1000 [=====] - 0s 269us/step - loss:
7.4863 - accuracy: 0.1940 - val_loss: 8.0740 - val_accuracy: 0.224
0
Epoch 18/30
1000/1000 [=====] - 0s 268us/step - loss:
5.9644 - accuracy: 0.2040 - val_loss: 6.1484 - val_accuracy: 0.218
0
Epoch 19/30
1000/1000 [=====] - 0s 257us/step - loss:
6.1578 - accuracy: 0.1820 - val_loss: 4.6749 - val_accuracy: 0.228
0
Epoch 20/30
1000/1000 [=====] - 0s 256us/step - loss:
5.9762 - accuracy: 0.1990 - val_loss: 7.1665 - val_accuracy: 0.230
0
Epoch 21/30
1000/1000 [=====] - 0s 262us/step - loss:
6.4633 - accuracy: 0.1810 - val_loss: 5.9801 - val_accuracy: 0.193
0
Epoch 22/30
1000/1000 [=====] - 0s 260us/step - loss:
5.7364 - accuracy: 0.1880 - val_loss: 5.0777 - val_accuracy: 0.199
0
Epoch 23/30
1000/1000 [=====] - 0s 260us/step - loss:
5.4011 - accuracy: 0.1930 - val_loss: 7.1246 - val_accuracy: 0.175
0
Epoch 24/30
1000/1000 [=====] - 0s 256us/step - loss:
6.1415 - accuracy: 0.1670 - val_loss: 4.1313 - val_accuracy: 0.196
0
Epoch 25/30
1000/1000 [=====] - 0s 260us/step - loss:
6.3176 - accuracy: 0.2180 - val_loss: 5.8490 - val_accuracy: 0.101
0
Epoch 26/30
1000/1000 [=====] - 0s 260us/step - loss:
6.0663 - accuracy: 0.1990 - val_loss: 4.7250 - val_accuracy: 0.213
0
Epoch 27/30
1000/1000 [=====] - 0s 263us/step - loss:
5.4542 - accuracy: 0.2270 - val_loss: 6.1186 - val_accuracy: 0.105
0
Epoch 28/30
1000/1000 [=====] - 0s 357us/step - loss:
5.9321 - accuracy: 0.1610 - val_loss: 4.6736 - val_accuracy: 0.334
0
```

```
Epoch 29/30
1000/1000 [=====] - 0s 295us/step - loss:
5.6481 - accuracy: 0.1960 - val_loss: 6.8911 - val_accuracy: 0.133
0
Epoch 30/30
1000/1000 [=====] - 0s 311us/step - loss:
6.4662 - accuracy: 0.1880 - val_loss: 5.5241 - val_accuracy: 0.222
0
```

```
In [119]: model_accuracy()
model_loss()
```



```
In [120]: model=model_nn(30,10,True)
error_model(model)
```

Model: "sequential\_32"

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

dense_83 (Dense)	(None, 30)	23550
------------------	------------	-------

dense_84 (Dense)	(None, 10)	310
------------------	------------	-----

=====

Total params: 23,860

Trainable params: 23,860

Non-trainable params: 0

Epoch 1/1

1000/1000 [=====] - 0s 228us/step - loss: 30.6951 - accuracy: 0.1230

Epoch 1/1

1000/1000 [=====] - 0s 254us/step - loss: 3.7179 - accuracy: 0.1740

Epoch 1/1

1000/1000 [=====] - 0s 237us/step - loss: 4.6770 - accuracy: 0.1780

Epoch 1/1

1000/1000 [=====] - 0s 204us/step - loss: 4.8421 - accuracy: 0.2070

Epoch 1/1

1000/1000 [=====] - 0s 191us/step - loss: 5.0461 - accuracy: 0.1980

Epoch 1/1

1000/1000 [=====] - 0s 165us/step - loss: 5.1083 - accuracy: 0.1840

Epoch 1/1

1000/1000 [=====] - 0s 153us/step - loss: 5.7680 - accuracy: 0.2050

Epoch 1/1

1000/1000 [=====] - 0s 153us/step - loss: 5.5682 - accuracy: 0.2080

Epoch 1/1

1000/1000 [=====] - 0s 160us/step - loss: 5.9194 - accuracy: 0.2230

Epoch 1/1

1000/1000 [=====] - 0s 163us/step - loss: 6.0916 - accuracy: 0.1960

Epoch 1/1

1000/1000 [=====] - 0s 175us/step - loss: 5.4155 - accuracy: 0.2070

Epoch 1/1

1000/1000 [=====] - 0s 211us/step - loss: 5.9289 - accuracy: 0.1810

Epoch 1/1

1000/1000 [=====] - 0s 220us/step - loss: 5.4300 - accuracy: 0.1860

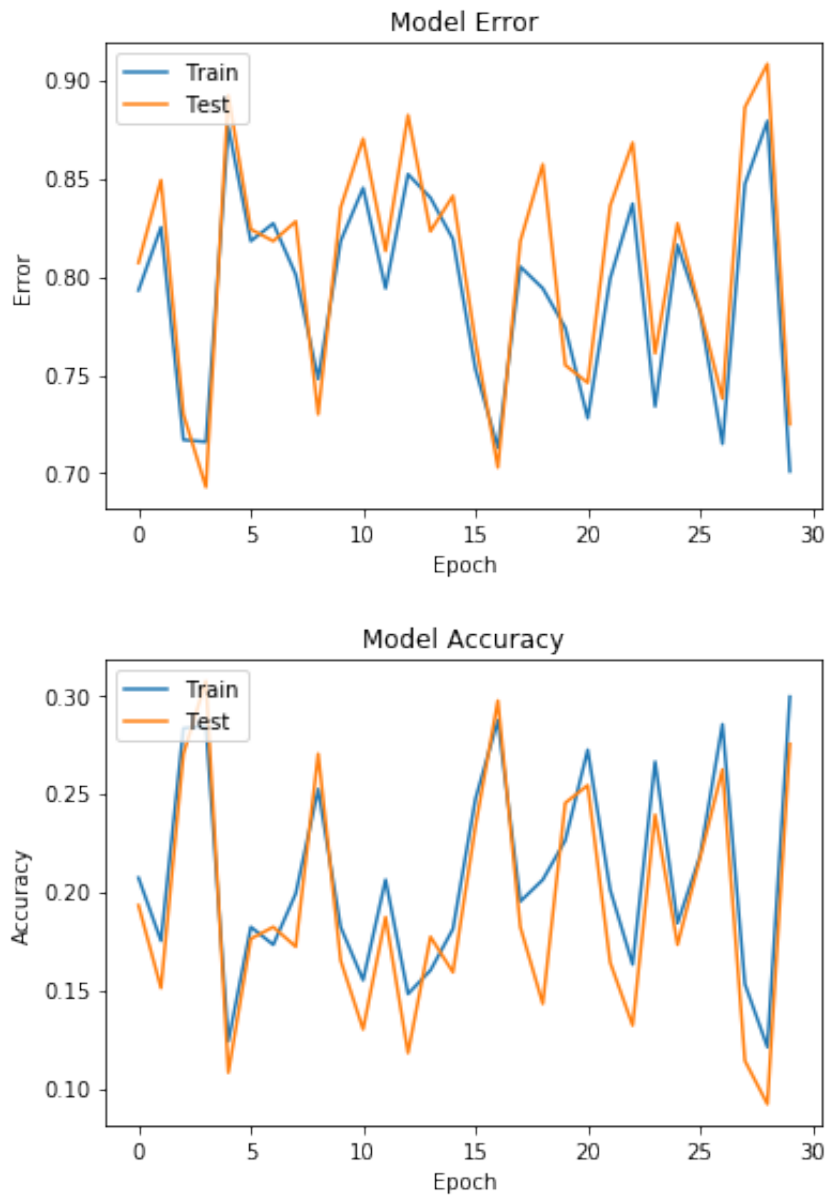
Epoch 1/1

1000/1000 [=====] - 0s 203us/step - loss: 7.2165 - accuracy: 0.2050

Epoch 1/1

1000/1000 [=====] - 0s 159us/step - loss: 5.4273 - accuracy: 0.1610

```
Epoch 1/1
1000/1000 [=====] - 0s 159us/step - loss:
5.7662 - accuracy: 0.2180
Epoch 1/1
1000/1000 [=====] - 0s 170us/step - loss:
6.1361 - accuracy: 0.2080
Epoch 1/1
1000/1000 [=====] - 0s 163us/step - loss:
6.8780 - accuracy: 0.1960
Epoch 1/1
1000/1000 [=====] - 0s 163us/step - loss:
6.7338 - accuracy: 0.2210
Epoch 1/1
1000/1000 [=====] - 0s 157us/step - loss:
6.1083 - accuracy: 0.2150
Epoch 1/1
1000/1000 [=====] - 0s 160us/step - loss:
6.6663 - accuracy: 0.1930
Epoch 1/1
1000/1000 [=====] - 0s 157us/step - loss:
6.0698 - accuracy: 0.1900
Epoch 1/1
1000/1000 [=====] - 0s 169us/step - loss:
6.8879 - accuracy: 0.2230
Epoch 1/1
1000/1000 [=====] - 0s 214us/step - loss:
5.9167 - accuracy: 0.1850
Epoch 1/1
1000/1000 [=====] - 0s 205us/step - loss:
5.6246 - accuracy: 0.1920
Epoch 1/1
1000/1000 [=====] - 0s 189us/step - loss:
4.8924 - accuracy: 0.1750
Epoch 1/1
1000/1000 [=====] - 0s 162us/step - loss:
5.4305 - accuracy: 0.2020
Epoch 1/1
1000/1000 [=====] - 0s 160us/step - loss:
5.8140 - accuracy: 0.1830
Epoch 1/1
1000/1000 [=====] - 0s 168us/step - loss:
5.2853 - accuracy: 0.1620
Epoch 1/1
1000/1000 [=====] - 0s 154us/step - loss:
4.9126 - accuracy: 0.1930
```



```
In [121]: model=model_nn(30,10,True)
          W=LearningRate(model,(23550+310))
```

Model: "sequential\_33"

Layer (type)	Output Shape	Param #
dense_85 (Dense)	(None, 30)	23550
dense_86 (Dense)	(None, 10)	310

Total params: 23,860

Trainable params: 23,860

Non-trainable params: 0

Train on 1000 samples, validate on 1000 samples

Epoch 1/1

1000/1000 [=====] - 0s 373us/step - loss: 30.8374 - accuracy: 0.1340 - val\_loss: 3.0594 - val\_accuracy: 0.12

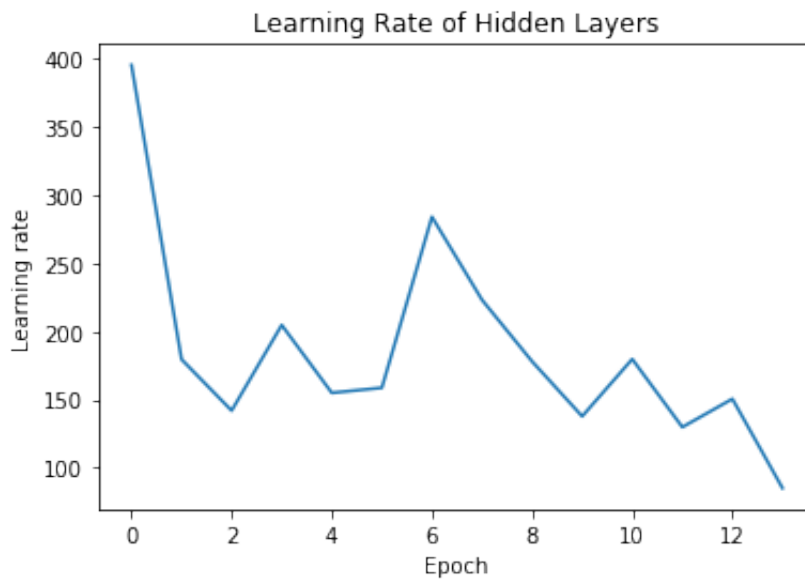


```
50
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 268us/step - loss:
3.9004 - accuracy: 0.1730 - val_loss: 4.5355 - val_accuracy: 0.175
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 365us/step - loss:
4.7082 - accuracy: 0.1800 - val_loss: 6.2007 - val_accuracy: 0.204
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 282us/step - loss:
5.2843 - accuracy: 0.2160 - val_loss: 5.0703 - val_accuracy: 0.267
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 283us/step - loss:
5.6730 - accuracy: 0.2260 - val_loss: 5.5772 - val_accuracy: 0.166
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 305us/step - loss:
5.7531 - accuracy: 0.2020 - val_loss: 6.2607 - val_accuracy: 0.248
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 398us/step - loss:
6.0863 - accuracy: 0.2320 - val_loss: 7.2565 - val_accuracy: 0.167
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 362us/step - loss:
5.4375 - accuracy: 0.2040 - val_loss: 5.1689 - val_accuracy: 0.184
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 261us/step - loss:
7.0406 - accuracy: 0.2190 - val_loss: 6.3355 - val_accuracy: 0.306
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 264us/step - loss:
6.1361 - accuracy: 0.2030 - val_loss: 5.4977 - val_accuracy: 0.269
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 282us/step - loss:
5.8373 - accuracy: 0.2080 - val_loss: 5.6226 - val_accuracy: 0.166
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
```

```
1000/1000 [=====] - 0s 259us/step - loss:
6.1951 - accuracy: 0.2010 - val_loss: 4.2848 - val_accuracy: 0.168
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 328us/step - loss:
5.3097 - accuracy: 0.1770 - val_loss: 5.3263 - val_accuracy: 0.157
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 338us/step - loss:
6.0497 - accuracy: 0.1910 - val_loss: 6.0244 - val_accuracy: 0.286
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 298us/step - loss:
6.6310 - accuracy: 0.2090 - val_loss: 6.4096 - val_accuracy: 0.194
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 266us/step - loss:
6.2579 - accuracy: 0.2080 - val_loss: 6.5253 - val_accuracy: 0.156
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 274us/step - loss:
6.2270 - accuracy: 0.1940 - val_loss: 6.8317 - val_accuracy: 0.239
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 269us/step - loss:
6.9005 - accuracy: 0.2080 - val_loss: 6.7235 - val_accuracy: 0.272
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 261us/step - loss:
6.8350 - accuracy: 0.1780 - val_loss: 7.0064 - val_accuracy: 0.191
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 326us/step - loss:
6.3891 - accuracy: 0.1950 - val_loss: 5.7205 - val_accuracy: 0.146
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 345us/step - loss:
6.6304 - accuracy: 0.1880 - val_loss: 5.2031 - val_accuracy: 0.167
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 355us/step - loss:
5.5980 - accuracy: 0.1990 - val_loss: 4.8382 - val_accuracy: 0.143
0
```

```
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 262us/step - loss:
5.9083 - accuracy: 0.2100 - val_loss: 6.1855 - val_accuracy: 0.283
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 263us/step - loss:
6.4476 - accuracy: 0.2320 - val_loss: 9.4106 - val_accuracy: 0.153
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 270us/step - loss:
6.3843 - accuracy: 0.2030 - val_loss: 7.6869 - val_accuracy: 0.186
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 252us/step - loss:
6.0698 - accuracy: 0.2010 - val_loss: 4.6945 - val_accuracy: 0.168
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 345us/step - loss:
6.2490 - accuracy: 0.2180 - val_loss: 7.9760 - val_accuracy: 0.258
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 346us/step - loss:
6.2856 - accuracy: 0.1960 - val_loss: 5.8173 - val_accuracy: 0.180
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 354us/step - loss:
6.7535 - accuracy: 0.1920 - val_loss: 7.1048 - val_accuracy: 0.165
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 267us/step - loss:
5.9238 - accuracy: 0.2110 - val_loss: 5.5343 - val_accuracy: 0.172
0
```

In [122]: LearningRate\_plot(W)



## Two Layer Model without regularization

```
In [27]: model=model_nn_2(30,10,False)
history = model.fit(X_train,y_train,validation_data=(X_test,y_test)
,epochs=30, batch_size=10)
```

Model: "sequential\_4"

Layer (type)	Output Shape	Param #
dense_7 (Dense)	(None, 30)	23550
dense_8 (Dense)	(None, 30)	930
dense_9 (Dense)	(None, 10)	310

Total params: 24,790

Trainable params: 24,790

Non-trainable params: 0

Train on 1000 samples, validate on 1000 samples

Epoch 1/30

1000/1000 [=====] - 0s 388us/step - loss: 1.5217 - accuracy: 0.4720 - val\_loss: 1.3268 - val\_accuracy: 0.5520

Epoch 2/30

1000/1000 [=====] - 0s 243us/step - loss: 1.0205 - accuracy: 0.6630 - val\_loss: 1.2290 - val\_accuracy: 0.5690

Epoch 3/30

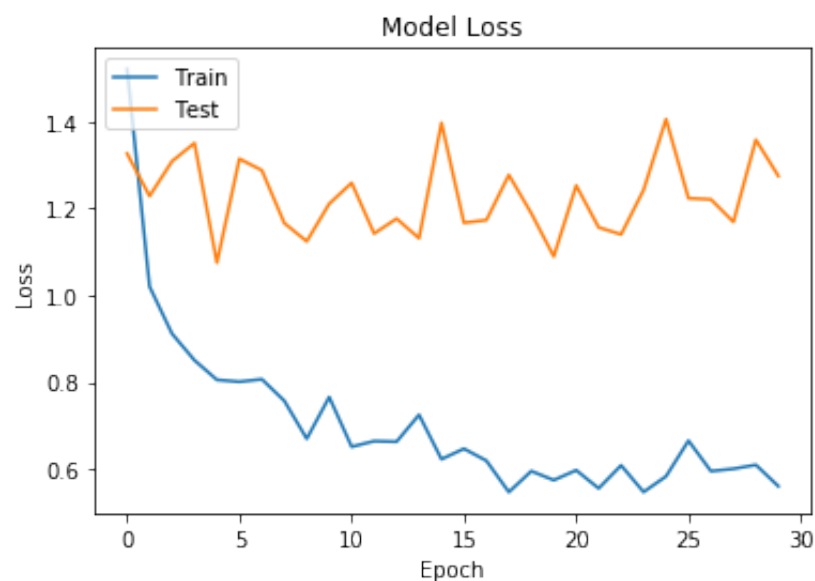
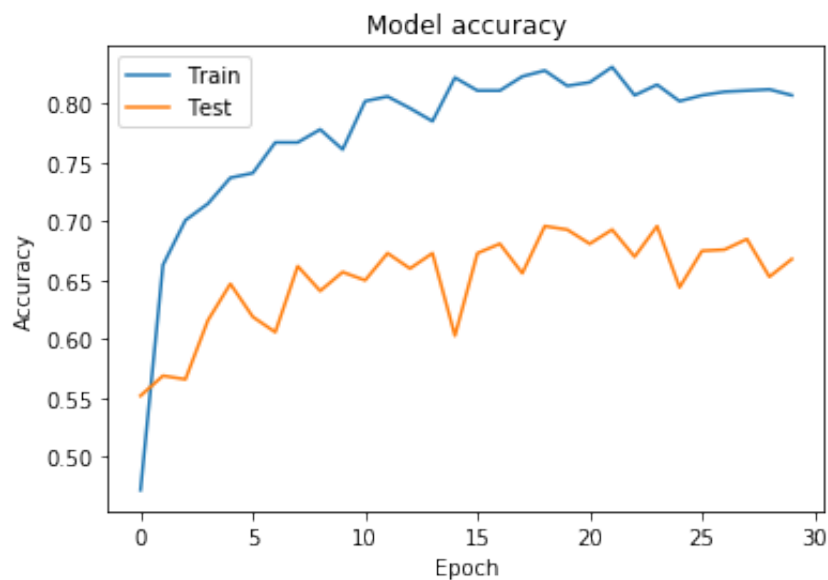
1000/1000 [=====] - 0s 271us/step - loss: 0.9120 - accuracy: 0.7010 - val\_loss: 1.3091 - val\_accuracy: 0.566

```
0
Epoch 4/30
1000/1000 [=====] - 0s 242us/step - loss:
0.8509 - accuracy: 0.7150 - val_loss: 1.3510 - val_accuracy: 0.616
0
Epoch 5/30
1000/1000 [=====] - 0s 319us/step - loss:
0.8057 - accuracy: 0.7370 - val_loss: 1.0755 - val_accuracy: 0.647
0
Epoch 6/30
1000/1000 [=====] - 0s 320us/step - loss:
0.8014 - accuracy: 0.7410 - val_loss: 1.3148 - val_accuracy: 0.619
0
Epoch 7/30
1000/1000 [=====] - 0s 324us/step - loss:
0.8071 - accuracy: 0.7670 - val_loss: 1.2884 - val_accuracy: 0.606
0
Epoch 8/30
1000/1000 [=====] - 0s 260us/step - loss:
0.7575 - accuracy: 0.7670 - val_loss: 1.1667 - val_accuracy: 0.662
0
Epoch 9/30
1000/1000 [=====] - 0s 310us/step - loss:
0.6700 - accuracy: 0.7780 - val_loss: 1.1247 - val_accuracy: 0.641
0
Epoch 10/30
1000/1000 [=====] - 0s 265us/step - loss:
0.7660 - accuracy: 0.7610 - val_loss: 1.2113 - val_accuracy: 0.657
0
Epoch 11/30
1000/1000 [=====] - 0s 258us/step - loss:
0.6516 - accuracy: 0.8020 - val_loss: 1.2593 - val_accuracy: 0.650
0
Epoch 12/30
1000/1000 [=====] - 0s 365us/step - loss:
0.6646 - accuracy: 0.8060 - val_loss: 1.1427 - val_accuracy: 0.673
0
Epoch 13/30
1000/1000 [=====] - 0s 244us/step - loss:
0.6633 - accuracy: 0.7960 - val_loss: 1.1770 - val_accuracy: 0.660
0
Epoch 14/30
1000/1000 [=====] - 0s 250us/step - loss:
0.7252 - accuracy: 0.7850 - val_loss: 1.1318 - val_accuracy: 0.673
0
Epoch 15/30
1000/1000 [=====] - 0s 305us/step - loss:
0.6230 - accuracy: 0.8220 - val_loss: 1.3976 - val_accuracy: 0.603
0
Epoch 16/30
1000/1000 [=====] - 0s 260us/step - loss:
0.6471 - accuracy: 0.8110 - val_loss: 1.1675 - val_accuracy: 0.673
0
```

```
Epoch 17/30
1000/1000 [=====] - 0s 252us/step - loss:
0.6195 - accuracy: 0.8110 - val_loss: 1.1737 - val_accuracy: 0.681
0
Epoch 18/30
1000/1000 [=====] - 0s 256us/step - loss:
0.5472 - accuracy: 0.8230 - val_loss: 1.2776 - val_accuracy: 0.656
0
Epoch 19/30
1000/1000 [=====] - 0s 233us/step - loss:
0.5955 - accuracy: 0.8280 - val_loss: 1.1894 - val_accuracy: 0.696
0
Epoch 20/30
1000/1000 [=====] - 0s 237us/step - loss:
0.5747 - accuracy: 0.8150 - val_loss: 1.0899 - val_accuracy: 0.693
0
Epoch 21/30
1000/1000 [=====] - 0s 250us/step - loss:
0.5975 - accuracy: 0.8180 - val_loss: 1.2533 - val_accuracy: 0.681
0
Epoch 22/30
1000/1000 [=====] - 0s 277us/step - loss:
0.5558 - accuracy: 0.8310 - val_loss: 1.1568 - val_accuracy: 0.693
0
Epoch 23/30
1000/1000 [=====] - 0s 239us/step - loss:
0.6087 - accuracy: 0.8070 - val_loss: 1.1407 - val_accuracy: 0.670
0
Epoch 24/30
1000/1000 [=====] - 0s 236us/step - loss:
0.5474 - accuracy: 0.8160 - val_loss: 1.2430 - val_accuracy: 0.696
0
Epoch 25/30
1000/1000 [=====] - 0s 251us/step - loss:
0.5834 - accuracy: 0.8020 - val_loss: 1.4061 - val_accuracy: 0.644
0
Epoch 26/30
1000/1000 [=====] - 0s 244us/step - loss:
0.6656 - accuracy: 0.8070 - val_loss: 1.2240 - val_accuracy: 0.675
0
Epoch 27/30
1000/1000 [=====] - 0s 305us/step - loss:
0.5954 - accuracy: 0.8100 - val_loss: 1.2214 - val_accuracy: 0.676
0
Epoch 28/30
1000/1000 [=====] - 0s 283us/step - loss:
0.6008 - accuracy: 0.8110 - val_loss: 1.1694 - val_accuracy: 0.685
0
Epoch 29/30
1000/1000 [=====] - 0s 302us/step - loss:
0.6098 - accuracy: 0.8120 - val_loss: 1.3587 - val_accuracy: 0.653
0
Epoch 30/30
```

```
1000/1000 [=====] - 0s 294us/step - loss:
0.5608 - accuracy: 0.8070 - val_loss: 1.2754 - val_accuracy: 0.668
0
```

```
In [28]: model_accuracy()
model_loss()
```



```
In [29]: model=model_nn_2(30,10,False)
error_model(model)
```

Model: "sequential\_5"

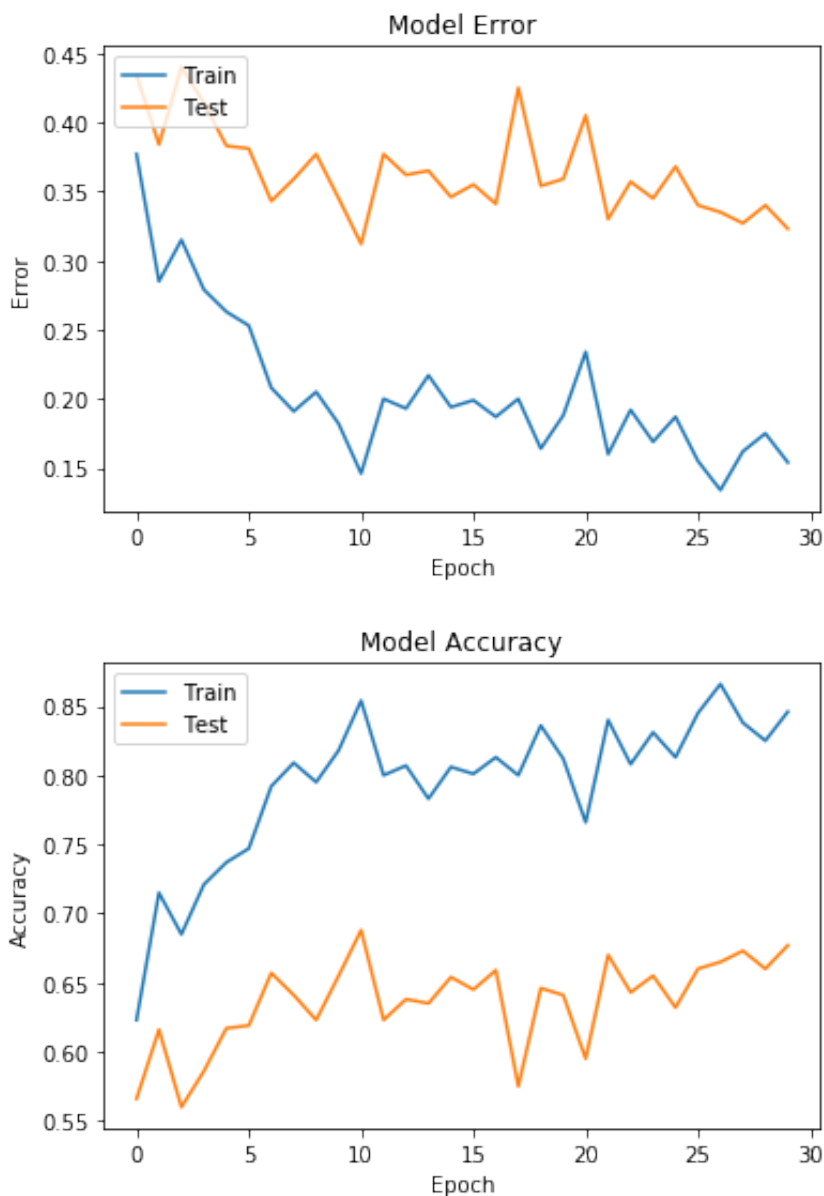
Layer (type)	Output Shape	Param #
dense_10 (Dense)	(None, 30)	23550
dense_11 (Dense)	(None, 30)	930
dense_12 (Dense)	(None, 10)	310

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

Epoch 1/1
1000/1000 [=====] - 0s 243us/step - loss:
1.5514 - accuracy: 0.4320
Epoch 1/1
1000/1000 [=====] - 0s 184us/step - loss:
1.0612 - accuracy: 0.6320
Epoch 1/1
1000/1000 [=====] - 0s 150us/step - loss:
1.0390 - accuracy: 0.6740
Epoch 1/1
1000/1000 [=====] - 0s 165us/step - loss:
1.0138 - accuracy: 0.6930
Epoch 1/1
1000/1000 [=====] - 0s 159us/step - loss:
0.9350 - accuracy: 0.6990
Epoch 1/1
1000/1000 [=====] - 0s 163us/step - loss:
0.8352 - accuracy: 0.7240
Epoch 1/1
1000/1000 [=====] - 0s 170us/step - loss:
0.8036 - accuracy: 0.7480
Epoch 1/1
1000/1000 [=====] - 0s 152us/step - loss:
0.7505 - accuracy: 0.7770
Epoch 1/1
1000/1000 [=====] - 0s 150us/step - loss:
0.6336 - accuracy: 0.7950
Epoch 1/1
1000/1000 [=====] - 0s 148us/step - loss:
0.6915 - accuracy: 0.7760
Epoch 1/1
1000/1000 [=====] - 0s 149us/step - loss:
0.6812 - accuracy: 0.7790
Epoch 1/1
1000/1000 [=====] - 0s 149us/step - loss:
0.6426 - accuracy: 0.7960
Epoch 1/1
1000/1000 [=====] - 0s 152us/step - loss:
0.5851 - accuracy: 0.7970
Epoch 1/1
1000/1000 [=====] - 0s 213us/step - loss:
0.6521 - accuracy: 0.7880
Epoch 1/1
1000/1000 [=====] - 0s 189us/step - loss:
0.7033 - accuracy: 0.7740
Epoch 1/1
1000/1000 [=====] - 0s 205us/step - loss:
0.6875 - accuracy: 0.7820
```



```
Epoch 1/1
1000/1000 [=====] - 0s 158us/step - loss:
0.7111 - accuracy: 0.7840
Epoch 1/1
1000/1000 [=====] - 0s 148us/step - loss:
0.6399 - accuracy: 0.7840
Epoch 1/1
1000/1000 [=====] - 0s 161us/step - loss:
0.6203 - accuracy: 0.8030
Epoch 1/1
1000/1000 [=====] - 0s 155us/step - loss:
0.6716 - accuracy: 0.7930
Epoch 1/1
1000/1000 [=====] - 0s 164us/step - loss:
0.6271 - accuracy: 0.7950
Epoch 1/1
1000/1000 [=====] - 0s 149us/step - loss:
0.6147 - accuracy: 0.8120
Epoch 1/1
1000/1000 [=====] - 0s 153us/step - loss:
0.6186 - accuracy: 0.8000
Epoch 1/1
1000/1000 [=====] - 0s 177us/step - loss:
0.5723 - accuracy: 0.8250
Epoch 1/1
1000/1000 [=====] - 0s 151us/step - loss:
0.5857 - accuracy: 0.8070
Epoch 1/1
1000/1000 [=====] - 0s 154us/step - loss:
0.6289 - accuracy: 0.7900
Epoch 1/1
1000/1000 [=====] - 0s 161us/step - loss:
0.5474 - accuracy: 0.8240
Epoch 1/1
1000/1000 [=====] - 0s 151us/step - loss:
0.5329 - accuracy: 0.8390
Epoch 1/1
1000/1000 [=====] - 0s 152us/step - loss:
0.5649 - accuracy: 0.8150
Epoch 1/1
1000/1000 [=====] - 0s 152us/step - loss:
0.5775 - accuracy: 0.8120
```



```
In [30]: model=model_nn_2(30,10,False)
W=LearningRate(model,(24790))
LearningRate_plot(W)
```

Model: "sequential\_6"

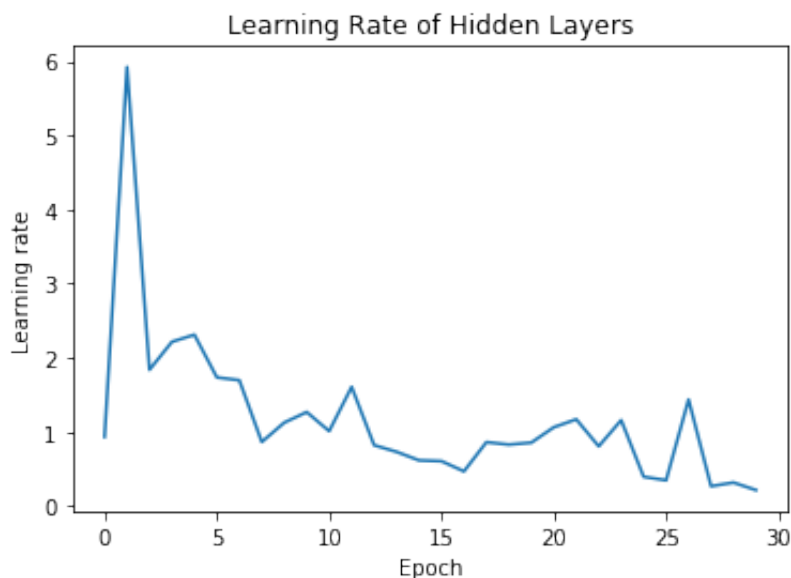
Layer (type)	Output Shape	Param #
dense_13 (Dense)	(None, 30)	23550
dense_14 (Dense)	(None, 30)	930
dense_15 (Dense)	(None, 10)	310
Total params: 24,790		
Trainable params: 24,790		
Non-trainable params: 0		

Train on 1000 samples, validate on 1000 samples

```
Epoch 1/1
1000/1000 [=====] - 0s 381us/step - loss:
1.6348 - accuracy: 0.4370 - val_loss: 1.1676 - val_accuracy: 0.625
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 327us/step - loss:
0.9744 - accuracy: 0.6670 - val_loss: 1.2375 - val_accuracy: 0.603
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 303us/step - loss:
0.8864 - accuracy: 0.7070 - val_loss: 1.2527 - val_accuracy: 0.624
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 289us/step - loss:
0.7607 - accuracy: 0.7580 - val_loss: 1.0989 - val_accuracy: 0.656
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 277us/step - loss:
0.7365 - accuracy: 0.7710 - val_loss: 1.2092 - val_accuracy: 0.631
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 242us/step - loss:
0.6508 - accuracy: 0.7950 - val_loss: 1.2522 - val_accuracy: 0.635
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 241us/step - loss:
0.7625 - accuracy: 0.7400 - val_loss: 1.2239 - val_accuracy: 0.673
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 334us/step - loss:
0.6754 - accuracy: 0.7880 - val_loss: 1.3125 - val_accuracy: 0.619
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 284us/step - loss:
0.6287 - accuracy: 0.8060 - val_loss: 1.1229 - val_accuracy: 0.692
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 263us/step - loss:
0.6979 - accuracy: 0.7790 - val_loss: 1.2111 - val_accuracy: 0.647
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 337us/step - loss:
0.7371 - accuracy: 0.7760 - val_loss: 1.2419 - val_accuracy: 0.673
```

```
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 326us/step - loss:
0.7865 - accuracy: 0.7430 - val_loss: 1.1829 - val_accuracy: 0.650
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 305us/step - loss:
0.6862 - accuracy: 0.7780 - val_loss: 1.2147 - val_accuracy: 0.645
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 320us/step - loss:
0.6424 - accuracy: 0.7980 - val_loss: 1.1490 - val_accuracy: 0.656
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 320us/step - loss:
0.6348 - accuracy: 0.7910 - val_loss: 1.0250 - val_accuracy: 0.706
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 254us/step - loss:
0.5291 - accuracy: 0.8270 - val_loss: 1.0832 - val_accuracy: 0.651
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 253us/step - loss:
0.5403 - accuracy: 0.8090 - val_loss: 1.1473 - val_accuracy: 0.678
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 280us/step - loss:
0.5685 - accuracy: 0.8230 - val_loss: 1.3053 - val_accuracy: 0.644
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 252us/step - loss:
0.6827 - accuracy: 0.7700 - val_loss: 1.1737 - val_accuracy: 0.644
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 247us/step - loss:
0.7070 - accuracy: 0.7790 - val_loss: 1.2195 - val_accuracy: 0.649
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 324us/step - loss:
0.6382 - accuracy: 0.7750 - val_loss: 1.1927 - val_accuracy: 0.654
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
```

```
1000/1000 [=====] - 0s 327us/step - loss:
0.6127 - accuracy: 0.8000 - val_loss: 1.1720 - val_accuracy: 0.640
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 329us/step - loss:
0.6526 - accuracy: 0.7930 - val_loss: 1.3098 - val_accuracy: 0.660
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 285us/step - loss:
0.6720 - accuracy: 0.7870 - val_loss: 1.1865 - val_accuracy: 0.653
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 253us/step - loss:
0.6819 - accuracy: 0.7890 - val_loss: 1.0842 - val_accuracy: 0.688
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 272us/step - loss:
0.5799 - accuracy: 0.8160 - val_loss: 1.4624 - val_accuracy: 0.646
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 265us/step - loss:
0.5994 - accuracy: 0.8110 - val_loss: 1.2177 - val_accuracy: 0.685
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 257us/step - loss:
0.5275 - accuracy: 0.8210 - val_loss: 1.2296 - val_accuracy: 0.660
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 267us/step - loss:
0.5231 - accuracy: 0.8310 - val_loss: 1.1760 - val_accuracy: 0.685
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 331us/step - loss:
0.5255 - accuracy: 0.8300 - val_loss: 1.1911 - val_accuracy: 0.675
0
```



## Two Layer Model with regularization

```
In [31]: model=model_nn_2(30,10,True)
history = model.fit(X_train,y_train,validation_data=(X_test,y_test)
,epochs=30, batch_size=10)
```

Model: "sequential\_7"

Layer (type)	Output Shape	Param #
dense_16 (Dense)	(None, 30)	23550
dense_17 (Dense)	(None, 30)	930
dense_18 (Dense)	(None, 10)	310

Total params: 24,790

Trainable params: 24,790

Non-trainable params: 0

Train on 1000 samples, validate on 1000 samples

Epoch 1/30

1000/1000 [=====] - 0s 411us/step - loss: 35.3744 - accuracy: 0.0950 - val\_loss: 2.3468 - val\_accuracy: 0.1000

Epoch 2/30

1000/1000 [=====] - 0s 299us/step - loss: 2.3355 - accuracy: 0.0860 - val\_loss: 2.3122 - val\_accuracy: 0.1000

Epoch 3/30

1000/1000 [=====] - 0s 269us/step - loss: 2.3358 - accuracy: 0.0980 - val\_loss: 2.3313 - val\_accuracy: 0.1000

Epoch 4/30

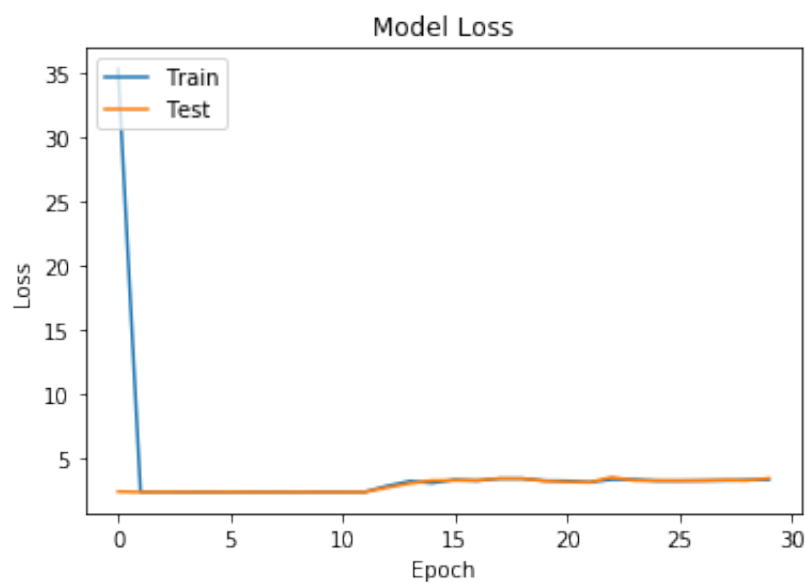
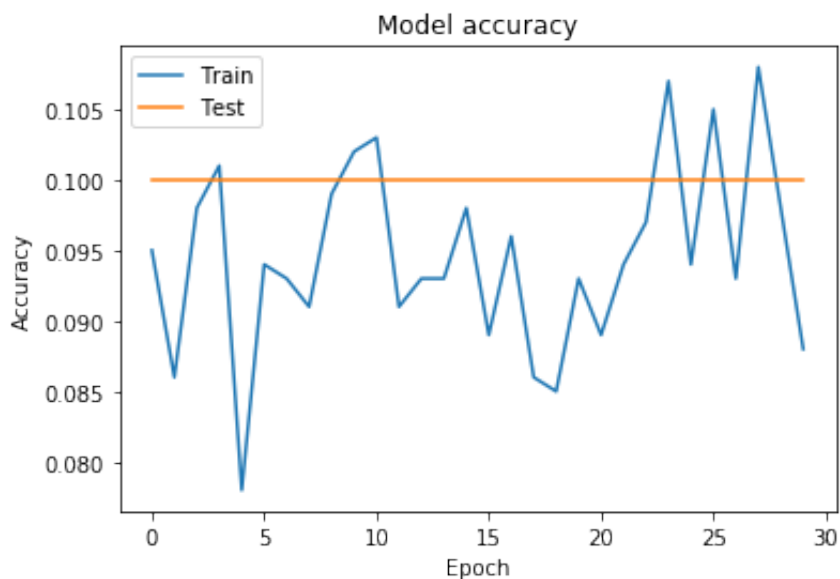
```
1000/1000 [=====] - 0s 261us/step - loss:
2.3237 - accuracy: 0.1010 - val_loss: 2.3196 - val_accuracy: 0.100
0
Epoch 5/30
1000/1000 [=====] - 0s 268us/step - loss:
2.3307 - accuracy: 0.0780 - val_loss: 2.3124 - val_accuracy: 0.100
0
Epoch 6/30
1000/1000 [=====] - 0s 261us/step - loss:
2.3272 - accuracy: 0.0940 - val_loss: 2.3306 - val_accuracy: 0.100
0
Epoch 7/30
1000/1000 [=====] - 0s 264us/step - loss:
2.3334 - accuracy: 0.0930 - val_loss: 2.3133 - val_accuracy: 0.100
0
Epoch 8/30
1000/1000 [=====] - 0s 268us/step - loss:
2.3308 - accuracy: 0.0910 - val_loss: 2.3102 - val_accuracy: 0.100
0
Epoch 9/30
1000/1000 [=====] - 0s 269us/step - loss:
2.3253 - accuracy: 0.0990 - val_loss: 2.3266 - val_accuracy: 0.100
0
Epoch 10/30
1000/1000 [=====] - 0s 269us/step - loss:
2.3341 - accuracy: 0.1020 - val_loss: 2.3122 - val_accuracy: 0.100
0
Epoch 11/30
1000/1000 [=====] - 0s 265us/step - loss:
2.3326 - accuracy: 0.1030 - val_loss: 2.3123 - val_accuracy: 0.100
0
Epoch 12/30
1000/1000 [=====] - 0s 263us/step - loss:
2.3252 - accuracy: 0.0910 - val_loss: 2.3213 - val_accuracy: 0.100
0
Epoch 13/30
1000/1000 [=====] - 0s 268us/step - loss:
2.7882 - accuracy: 0.0930 - val_loss: 2.6617 - val_accuracy: 0.100
0
Epoch 14/30
1000/1000 [=====] - 0s 327us/step - loss:
3.1588 - accuracy: 0.0930 - val_loss: 2.9804 - val_accuracy: 0.100
0
Epoch 15/30
1000/1000 [=====] - 0s 350us/step - loss:
3.0242 - accuracy: 0.0980 - val_loss: 3.2150 - val_accuracy: 0.100
0
Epoch 16/30
1000/1000 [=====] - 0s 327us/step - loss:
3.2754 - accuracy: 0.0890 - val_loss: 3.2276 - val_accuracy: 0.100
0
Epoch 17/30
1000/1000 [=====] - 0s 276us/step - loss:
```

```
3.2059 - accuracy: 0.0960 - val_loss: 3.2298 - val_accuracy: 0.100
0
Epoch 18/30
1000/1000 [=====] - 0s 287us/step - loss:
3.3675 - accuracy: 0.0860 - val_loss: 3.3688 - val_accuracy: 0.100
0
Epoch 19/30
1000/1000 [=====] - 0s 279us/step - loss:
3.3742 - accuracy: 0.0850 - val_loss: 3.3581 - val_accuracy: 0.100
0
Epoch 20/30
1000/1000 [=====] - 0s 270us/step - loss:
3.2005 - accuracy: 0.0930 - val_loss: 3.1862 - val_accuracy: 0.100
0
Epoch 21/30
1000/1000 [=====] - 0s 268us/step - loss:
3.1839 - accuracy: 0.0890 - val_loss: 3.1087 - val_accuracy: 0.100
0
Epoch 22/30
1000/1000 [=====] - 0s 265us/step - loss:
3.0820 - accuracy: 0.0940 - val_loss: 3.0726 - val_accuracy: 0.100
0
Epoch 23/30
1000/1000 [=====] - 0s 267us/step - loss:
3.2921 - accuracy: 0.0970 - val_loss: 3.4771 - val_accuracy: 0.100
0
Epoch 24/30
1000/1000 [=====] - 0s 264us/step - loss:
3.2909 - accuracy: 0.1070 - val_loss: 3.2132 - val_accuracy: 0.100
0
Epoch 25/30
1000/1000 [=====] - 0s 268us/step - loss:
3.2000 - accuracy: 0.0940 - val_loss: 3.2102 - val_accuracy: 0.100
0
Epoch 26/30
1000/1000 [=====] - 0s 274us/step - loss:
3.1983 - accuracy: 0.1050 - val_loss: 3.2025 - val_accuracy: 0.100
0
Epoch 27/30
1000/1000 [=====] - 0s 264us/step - loss:
3.2142 - accuracy: 0.0930 - val_loss: 3.2074 - val_accuracy: 0.100
0
Epoch 28/30
1000/1000 [=====] - 0s 264us/step - loss:
3.2232 - accuracy: 0.1080 - val_loss: 3.2593 - val_accuracy: 0.100
0
Epoch 29/30
1000/1000 [=====] - 0s 266us/step - loss:
3.2619 - accuracy: 0.0980 - val_loss: 3.2137 - val_accuracy: 0.100
0
Epoch 30/30
1000/1000 [=====] - 0s 266us/step - loss:
3.2872 - accuracy: 0.0880 - val_loss: 3.3858 - val_accuracy: 0.100
```



0

```
In [32]: model_accuracy()
model_loss()
```



```
In [33]: model=model_nn_2(30,10,True)
error_model(model)
```

Model: "sequential\_8"

Layer (type)	Output Shape	Param #
dense_19 (Dense)	(None, 30)	23550
dense_20 (Dense)	(None, 30)	930
dense_21 (Dense)	(None, 10)	310
Total params: 24,790		

Trainable params: 24,790

Non-trainable params: 0

---

Epoch 1/1  
1000/1000 [=====] - 0s 248us/step - loss:  
35.3365 - accuracy: 0.0980

Epoch 1/1  
1000/1000 [=====] - 0s 186us/step - loss:  
2.3469 - accuracy: 0.1020

Epoch 1/1  
1000/1000 [=====] - 0s 232us/step - loss:  
2.3366 - accuracy: 0.1010

Epoch 1/1  
1000/1000 [=====] - 0s 172us/step - loss:  
2.3296 - accuracy: 0.0950

Epoch 1/1  
1000/1000 [=====] - 0s 152us/step - loss:  
2.3343 - accuracy: 0.0840

Epoch 1/1  
1000/1000 [=====] - 0s 181us/step - loss:  
2.3261 - accuracy: 0.0920

Epoch 1/1  
1000/1000 [=====] - 0s 179us/step - loss:  
2.3406 - accuracy: 0.0870

Epoch 1/1  
1000/1000 [=====] - 0s 198us/step - loss:  
2.3287 - accuracy: 0.0860

Epoch 1/1  
1000/1000 [=====] - 0s 207us/step - loss:  
2.3456 - accuracy: 0.0830

Epoch 1/1  
1000/1000 [=====] - 0s 189us/step - loss:  
2.3290 - accuracy: 0.1050

Epoch 1/1  
1000/1000 [=====] - 0s 200us/step - loss:  
2.3343 - accuracy: 0.0920

Epoch 1/1  
1000/1000 [=====] - 0s 192us/step - loss:  
2.3244 - accuracy: 0.0920

Epoch 1/1  
1000/1000 [=====] - 0s 157us/step - loss:  
2.7683 - accuracy: 0.0930

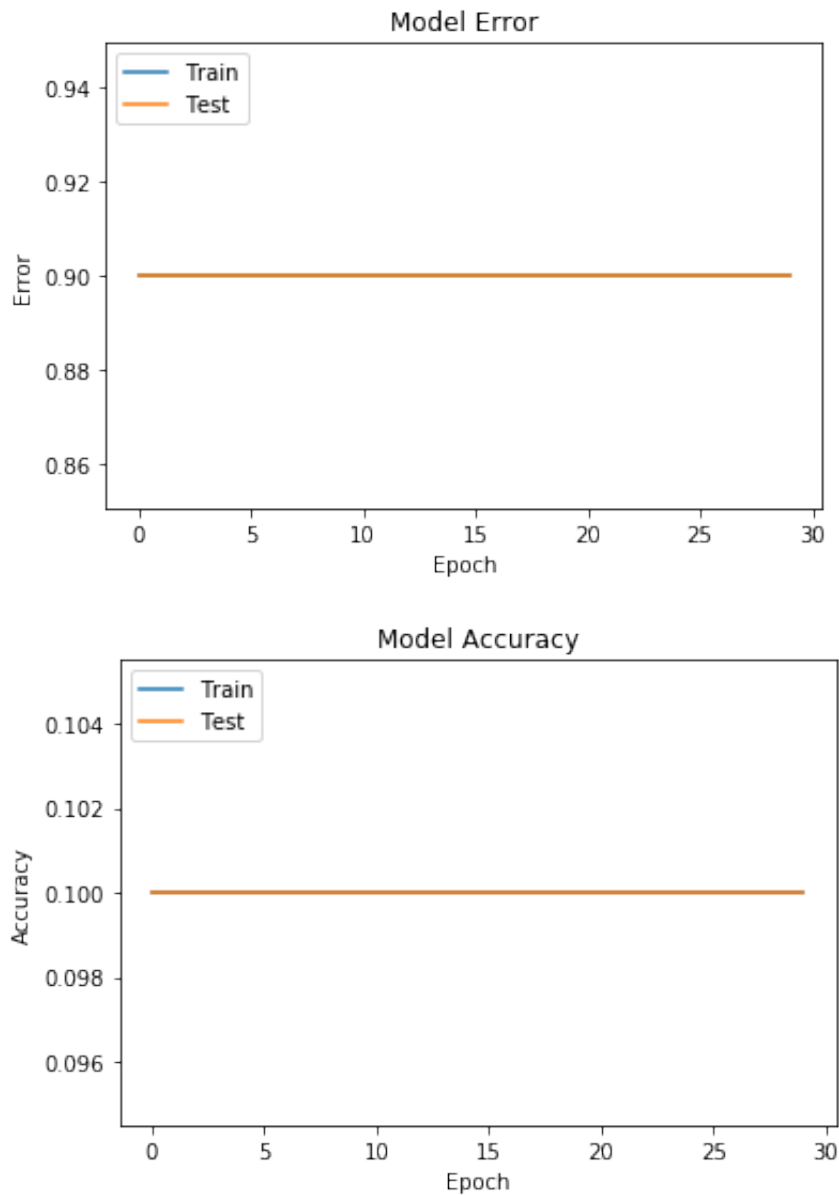
Epoch 1/1  
1000/1000 [=====] - 0s 161us/step - loss:  
3.1566 - accuracy: 0.0960

Epoch 1/1  
1000/1000 [=====] - 0s 165us/step - loss:  
3.0189 - accuracy: 0.0850

Epoch 1/1  
1000/1000 [=====] - 0s 165us/step - loss:  
3.2681 - accuracy: 0.1070

Epoch 1/1  
1000/1000 [=====] - 0s 158us/step - loss:

```
3.1728 - accuracy: 0.1010
Epoch 1/1
1000/1000 [=====] - 0s 157us/step - loss:
3.3784 - accuracy: 0.0960
Epoch 1/1
1000/1000 [=====] - 0s 161us/step - loss:
3.3574 - accuracy: 0.0900
Epoch 1/1
1000/1000 [=====] - 0s 157us/step - loss:
3.1954 - accuracy: 0.1080
Epoch 1/1
1000/1000 [=====] - 0s 160us/step - loss:
3.2192 - accuracy: 0.0930
Epoch 1/1
1000/1000 [=====] - 0s 160us/step - loss:
3.0684 - accuracy: 0.0920
Epoch 1/1
1000/1000 [=====] - 0s 163us/step - loss:
3.2873 - accuracy: 0.1030
Epoch 1/1
1000/1000 [=====] - 0s 164us/step - loss:
3.2926 - accuracy: 0.0880
Epoch 1/1
1000/1000 [=====] - 0s 161us/step - loss:
3.1975 - accuracy: 0.0780
Epoch 1/1
1000/1000 [=====] - 0s 162us/step - loss:
3.2314 - accuracy: 0.1020
Epoch 1/1
1000/1000 [=====] - 0s 161us/step - loss:
3.1762 - accuracy: 0.0820
Epoch 1/1
1000/1000 [=====] - 0s 159us/step - loss:
3.2260 - accuracy: 0.0810
Epoch 1/1
1000/1000 [=====] - 0s 161us/step - loss:
3.2875 - accuracy: 0.0910
Epoch 1/1
1000/1000 [=====] - 0s 160us/step - loss:
3.2493 - accuracy: 0.0870
```



```
In [34]: model=model_nn_2(30,10,True)
W=LearningRate(model,(24790))
LearningRate_plot(W)
```

Model: "sequential\_9"

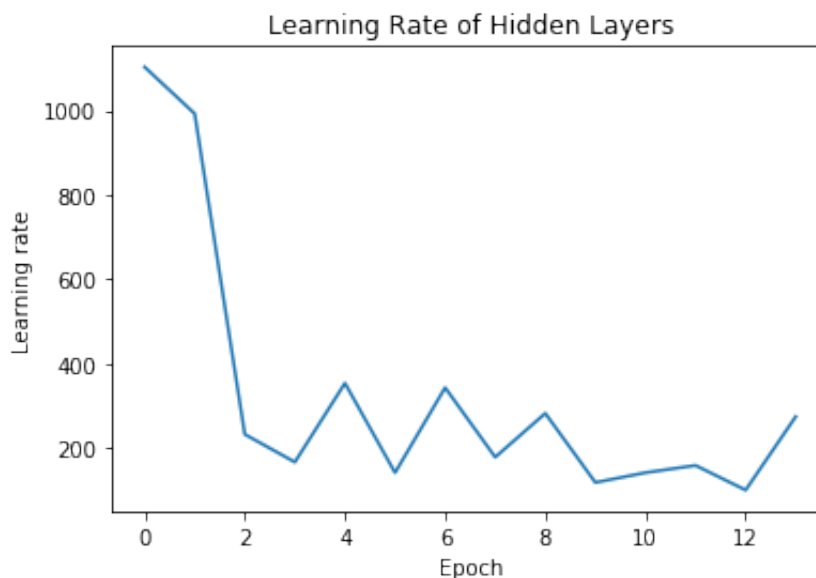
Layer (type)	Output Shape	Param #
dense_22 (Dense)	(None, 30)	23550
dense_23 (Dense)	(None, 30)	930
dense_24 (Dense)	(None, 10)	310
Total params: 24,790		
Trainable params: 24,790		
Non-trainable params: 0		

Train on 1000 samples, validate on 1000 samples

```
Epoch 1/1
1000/1000 [=====] - 0s 393us/step - loss:
35.5635 - accuracy: 0.0870 - val_loss: 2.3364 - val_accuracy: 0.10
00
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 277us/step - loss:
2.3427 - accuracy: 0.0890 - val_loss: 2.3202 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 272us/step - loss:
2.3239 - accuracy: 0.1080 - val_loss: 2.3292 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 350us/step - loss:
2.3285 - accuracy: 0.1080 - val_loss: 2.3199 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 325us/step - loss:
2.3268 - accuracy: 0.1000 - val_loss: 2.3165 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 353us/step - loss:
2.3264 - accuracy: 0.0870 - val_loss: 2.3430 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 332us/step - loss:
2.3328 - accuracy: 0.0960 - val_loss: 2.3126 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 399us/step - loss:
2.3246 - accuracy: 0.0850 - val_loss: 2.3177 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 292us/step - loss:
2.3294 - accuracy: 0.0900 - val_loss: 2.3187 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 276us/step - loss:
2.3297 - accuracy: 0.0970 - val_loss: 2.3310 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 289us/step - loss:
2.3320 - accuracy: 0.0970 - val_loss: 2.3204 - val_accuracy: 0.100
```

```
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 269us/step - loss:
2.3353 - accuracy: 0.0830 - val_loss: 2.3098 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 263us/step - loss:
2.7779 - accuracy: 0.0900 - val_loss: 2.6920 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 264us/step - loss:
3.1551 - accuracy: 0.0970 - val_loss: 3.0345 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 263us/step - loss:
3.0333 - accuracy: 0.0960 - val_loss: 3.3473 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 266us/step - loss:
3.2913 - accuracy: 0.0820 - val_loss: 3.1382 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 270us/step - loss:
3.2017 - accuracy: 0.1020 - val_loss: 3.1455 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 258us/step - loss:
3.3607 - accuracy: 0.0950 - val_loss: 3.3978 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 267us/step - loss:
3.3583 - accuracy: 0.1110 - val_loss: 3.2230 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 270us/step - loss:
3.1981 - accuracy: 0.0970 - val_loss: 3.2989 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 270us/step - loss:
3.1952 - accuracy: 0.0860 - val_loss: 3.0777 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
```

```
1000/1000 [=====] - 0s 270us/step - loss:
3.0801 - accuracy: 0.0890 - val_loss: 2.9939 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 275us/step - loss:
3.2927 - accuracy: 0.0920 - val_loss: 3.4838 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 270us/step - loss:
3.2493 - accuracy: 0.0970 - val_loss: 3.2052 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 265us/step - loss:
3.2396 - accuracy: 0.0860 - val_loss: 3.2858 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 264us/step - loss:
3.2099 - accuracy: 0.0890 - val_loss: 3.1612 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 276us/step - loss:
3.1604 - accuracy: 0.0990 - val_loss: 3.1999 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 338us/step - loss:
3.2390 - accuracy: 0.1040 - val_loss: 3.2264 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 327us/step - loss:
3.2582 - accuracy: 0.1030 - val_loss: 3.1761 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 306us/step - loss:
3.2963 - accuracy: 0.0850 - val_loss: 3.5017 - val_accuracy: 0.100
0
```



### Three Layer Model without regularization

```
In [35]: model=model_nn_3(30,10,False)
history = model.fit(X_train,y_train,validation_data=(X_test,y_test)
,epochs=30, batch_size=10)
model_accuracy()
model_loss()
```

Model: "sequential\_10"

Layer (type)	Output Shape	Param #
dense_25 (Dense)	(None, 30)	23550
dense_26 (Dense)	(None, 30)	930
dense_27 (Dense)	(None, 30)	930
dense_28 (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

1000/1000 [=====] - 0s 377us/step - loss: 2.3781 - accuracy: 0.0850 - val\_loss: 2.3173 - val\_accuracy: 0.1000

Epoch 2/30

1000/1000 [=====] - 0s 268us/step - loss: 2.1821 - accuracy: 0.1650 - val\_loss: 2.0271 - val\_accuracy: 0.2580

Epoch 3/30



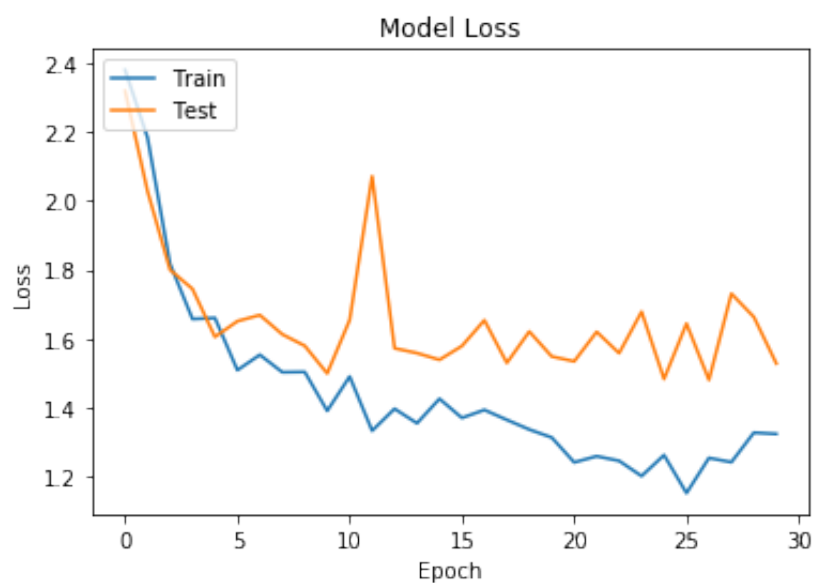
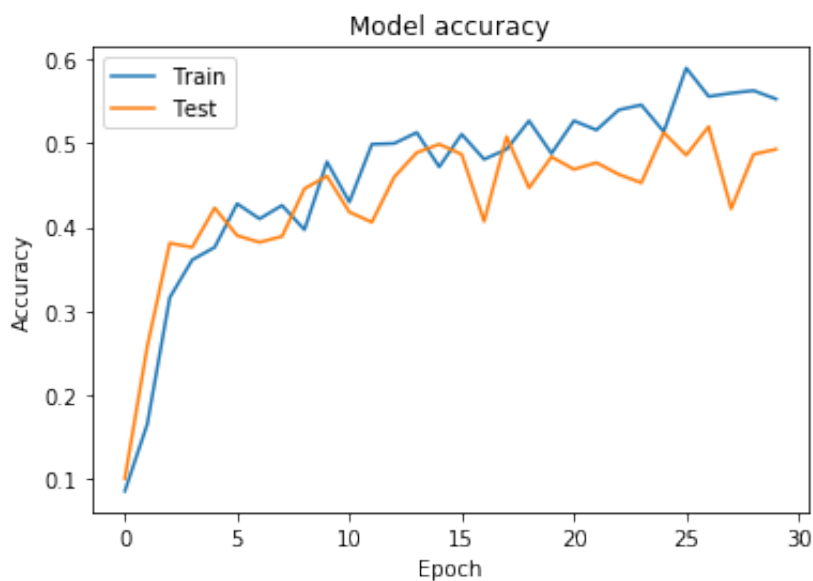
```
1000/1000 [=====] - 0s 342us/step - loss:
1.8174 - accuracy: 0.3160 - val_loss: 1.8001 - val_accuracy: 0.381
0
Epoch 4/30
1000/1000 [=====] - 0s 369us/step - loss:
1.6577 - accuracy: 0.3610 - val_loss: 1.7447 - val_accuracy: 0.376
0
Epoch 5/30
1000/1000 [=====] - 0s 273us/step - loss:
1.6607 - accuracy: 0.3760 - val_loss: 1.6059 - val_accuracy: 0.423
0
Epoch 6/30
1000/1000 [=====] - 0s 267us/step - loss:
1.5097 - accuracy: 0.4280 - val_loss: 1.6514 - val_accuracy: 0.390
0
Epoch 7/30
1000/1000 [=====] - 0s 286us/step - loss:
1.5541 - accuracy: 0.4100 - val_loss: 1.6691 - val_accuracy: 0.382
0
Epoch 8/30
1000/1000 [=====] - 0s 270us/step - loss:
1.5038 - accuracy: 0.4260 - val_loss: 1.6133 - val_accuracy: 0.389
0
Epoch 9/30
1000/1000 [=====] - 0s 265us/step - loss:
1.5046 - accuracy: 0.3970 - val_loss: 1.5802 - val_accuracy: 0.446
0
Epoch 10/30
1000/1000 [=====] - 0s 265us/step - loss:
1.3919 - accuracy: 0.4780 - val_loss: 1.5002 - val_accuracy: 0.461
0
Epoch 11/30
1000/1000 [=====] - 0s 265us/step - loss:
1.4913 - accuracy: 0.4300 - val_loss: 1.6549 - val_accuracy: 0.418
0
Epoch 12/30
1000/1000 [=====] - 0s 306us/step - loss:
1.3343 - accuracy: 0.4990 - val_loss: 2.0707 - val_accuracy: 0.406
0
Epoch 13/30
1000/1000 [=====] - 0s 319us/step - loss:
1.3979 - accuracy: 0.5000 - val_loss: 1.5730 - val_accuracy: 0.460
0
Epoch 14/30
1000/1000 [=====] - 0s 274us/step - loss:
1.3561 - accuracy: 0.5130 - val_loss: 1.5590 - val_accuracy: 0.489
0
Epoch 15/30
1000/1000 [=====] - 0s 279us/step - loss:
1.4269 - accuracy: 0.4720 - val_loss: 1.5397 - val_accuracy: 0.499
0
Epoch 16/30
1000/1000 [=====] - 0s 275us/step - loss:
```

```
1.3716 - accuracy: 0.5110 - val_loss: 1.5796 - val_accuracy: 0.487
0
Epoch 17/30
1000/1000 [=====] - 0s 276us/step - loss:
1.3947 - accuracy: 0.4810 - val_loss: 1.6541 - val_accuracy: 0.407
0
Epoch 18/30
1000/1000 [=====] - 0s 317us/step - loss:
1.3659 - accuracy: 0.4930 - val_loss: 1.5306 - val_accuracy: 0.508
0
Epoch 19/30
1000/1000 [=====] - 0s 257us/step - loss:
1.3380 - accuracy: 0.5270 - val_loss: 1.6215 - val_accuracy: 0.447
0
Epoch 20/30
1000/1000 [=====] - 0s 278us/step - loss:
1.3148 - accuracy: 0.4880 - val_loss: 1.5492 - val_accuracy: 0.484
0
Epoch 21/30
1000/1000 [=====] - 0s 273us/step - loss:
1.2430 - accuracy: 0.5270 - val_loss: 1.5352 - val_accuracy: 0.469
0
Epoch 22/30
1000/1000 [=====] - 0s 314us/step - loss:
1.2606 - accuracy: 0.5160 - val_loss: 1.6205 - val_accuracy: 0.477
0
Epoch 23/30
1000/1000 [=====] - 0s 325us/step - loss:
1.2473 - accuracy: 0.5400 - val_loss: 1.5589 - val_accuracy: 0.463
0
Epoch 24/30
1000/1000 [=====] - 0s 332us/step - loss:
1.2033 - accuracy: 0.5460 - val_loss: 1.6787 - val_accuracy: 0.453
0
Epoch 25/30
1000/1000 [=====] - 0s 333us/step - loss:
1.2640 - accuracy: 0.5140 - val_loss: 1.4841 - val_accuracy: 0.513
0
Epoch 26/30
1000/1000 [=====] - 0s 318us/step - loss:
1.1540 - accuracy: 0.5900 - val_loss: 1.6444 - val_accuracy: 0.486
0
Epoch 27/30
1000/1000 [=====] - 0s 319us/step - loss:
1.2556 - accuracy: 0.5560 - val_loss: 1.4814 - val_accuracy: 0.520
0
Epoch 28/30
1000/1000 [=====] - 0s 347us/step - loss:
1.2435 - accuracy: 0.5600 - val_loss: 1.7310 - val_accuracy: 0.422
0
Epoch 29/30
1000/1000 [=====] - 0s 311us/step - loss:
1.3289 - accuracy: 0.5630 - val_loss: 1.6636 - val_accuracy: 0.487
```

0

Epoch 30/30

1000/1000 [=====] - 0s 289us/step - loss:  
 1.3253 - accuracy: 0.5530 - val\_loss: 1.5290 - val\_accuracy: 0.493  
 0



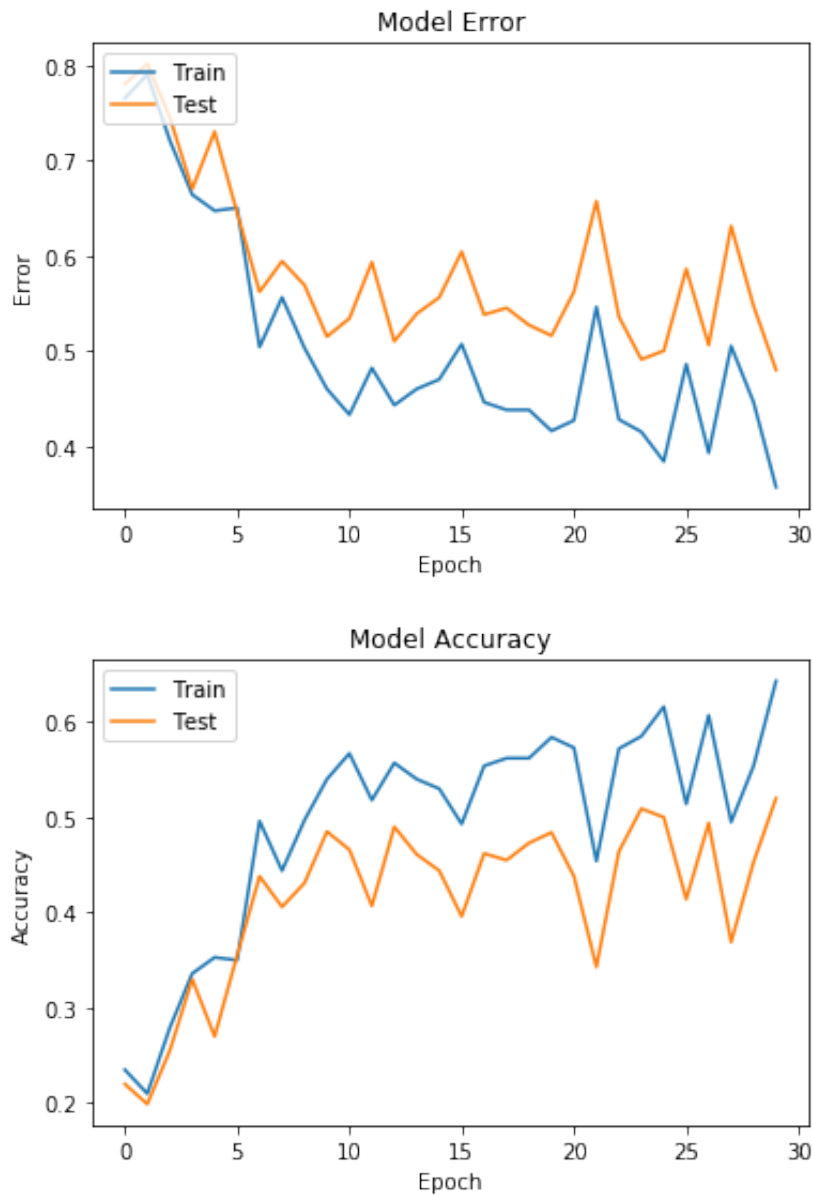
```
In [38]: model=model_nn_3(30,10,False)
         error_model(model)
```

Model: "sequential\_13"

Layer (type)	Output Shape	Param #
dense_37 (Dense)	(None, 30)	23550
dense_38 (Dense)	(None, 30)	930
dense_39 (Dense)	(None, 30)	930

```
dense_40 (Dense)                (None, 10)                310
=====
Total params: 25,720
Trainable params: 25,720
Non-trainable params: 0
=====
Epoch 1/1
1000/1000 [=====] - 0s 260us/step - loss:
2.3267 - accuracy: 0.1190
Epoch 1/1
1000/1000 [=====] - 0s 178us/step - loss:
2.0401 - accuracy: 0.2010
Epoch 1/1
1000/1000 [=====] - 0s 202us/step - loss:
1.9195 - accuracy: 0.2560
Epoch 1/1
1000/1000 [=====] - 0s 171us/step - loss:
1.7588 - accuracy: 0.3090
Epoch 1/1
1000/1000 [=====] - 0s 165us/step - loss:
1.6189 - accuracy: 0.3660
Epoch 1/1
1000/1000 [=====] - 0s 159us/step - loss:
1.6184 - accuracy: 0.4210
Epoch 1/1
1000/1000 [=====] - 0s 164us/step - loss:
1.5744 - accuracy: 0.4580
Epoch 1/1
1000/1000 [=====] - 0s 189us/step - loss:
1.4426 - accuracy: 0.4990
Epoch 1/1
1000/1000 [=====] - 0s 163us/step - loss:
1.5098 - accuracy: 0.4870
Epoch 1/1
1000/1000 [=====] - 0s 172us/step - loss:
1.3939 - accuracy: 0.5090
Epoch 1/1
1000/1000 [=====] - 0s 174us/step - loss:
1.3892 - accuracy: 0.5220
Epoch 1/1
1000/1000 [=====] - 0s 168us/step - loss:
1.3021 - accuracy: 0.5450
Epoch 1/1
1000/1000 [=====] - 0s 162us/step - loss:
1.4805 - accuracy: 0.5030
Epoch 1/1
1000/1000 [=====] - 0s 172us/step - loss:
1.3706 - accuracy: 0.5330
Epoch 1/1
1000/1000 [=====] - 0s 172us/step - loss:
1.3775 - accuracy: 0.5100
Epoch 1/1
1000/1000 [=====] - 0s 163us/step - loss:
```

```
1.3309 - accuracy: 0.5120
Epoch 1/1
1000/1000 [=====] - 0s 199us/step - loss:
1.3773 - accuracy: 0.5220
Epoch 1/1
1000/1000 [=====] - 0s 206us/step - loss:
1.4196 - accuracy: 0.4980
Epoch 1/1
1000/1000 [=====] - 0s 193us/step - loss:
1.3176 - accuracy: 0.5350
Epoch 1/1
1000/1000 [=====] - 0s 230us/step - loss:
1.3127 - accuracy: 0.5500
Epoch 1/1
1000/1000 [=====] - 0s 177us/step - loss:
1.4921 - accuracy: 0.4940
Epoch 1/1
1000/1000 [=====] - 0s 161us/step - loss:
1.3964 - accuracy: 0.5150
Epoch 1/1
1000/1000 [=====] - 0s 175us/step - loss:
1.4125 - accuracy: 0.5380
Epoch 1/1
1000/1000 [=====] - 0s 168us/step - loss:
1.3543 - accuracy: 0.5490
Epoch 1/1
1000/1000 [=====] - 0s 166us/step - loss:
1.4085 - accuracy: 0.5190
Epoch 1/1
1000/1000 [=====] - 0s 164us/step - loss:
1.2611 - accuracy: 0.5640
Epoch 1/1
1000/1000 [=====] - 0s 163us/step - loss:
1.2508 - accuracy: 0.5810
Epoch 1/1
1000/1000 [=====] - 0s 172us/step - loss:
1.3344 - accuracy: 0.5560
Epoch 1/1
1000/1000 [=====] - 0s 226us/step - loss:
1.2962 - accuracy: 0.5710
Epoch 1/1
1000/1000 [=====] - 0s 166us/step - loss:
1.2448 - accuracy: 0.6010
```



```
In [37]: model=model_nn_3(30,10,False)
W=LearningRate(model,(24790))
LearningRate_plot(W)
```

Model: "sequential\_12"

Layer (type)	Output Shape	Param #
dense_33 (Dense)	(None, 30)	23550
dense_34 (Dense)	(None, 30)	930
dense_35 (Dense)	(None, 30)	930
dense_36 (Dense)	(None, 10)	310

=====  
Total params: 25,720

Trainable params: 25,720

Non-trainable params: 0

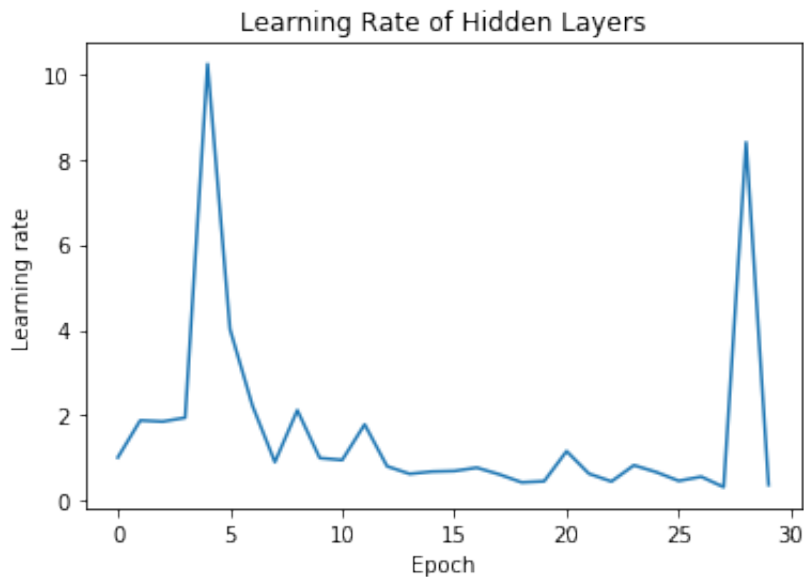
---

```
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 396us/step - loss:
2.2507 - accuracy: 0.1540 - val_loss: 2.1032 - val_accuracy: 0.177
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 288us/step - loss:
1.8492 - accuracy: 0.2680 - val_loss: 1.8162 - val_accuracy: 0.256
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 287us/step - loss:
1.5780 - accuracy: 0.4040 - val_loss: 1.5697 - val_accuracy: 0.418
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 366us/step - loss:
1.4318 - accuracy: 0.4690 - val_loss: 1.5633 - val_accuracy: 0.364
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 347us/step - loss:
1.4519 - accuracy: 0.4600 - val_loss: 1.6301 - val_accuracy: 0.441
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 287us/step - loss:
1.3433 - accuracy: 0.5350 - val_loss: 1.6292 - val_accuracy: 0.460
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 265us/step - loss:
1.3035 - accuracy: 0.5390 - val_loss: 1.4576 - val_accuracy: 0.507
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 289us/step - loss:
1.3012 - accuracy: 0.5530 - val_loss: 1.4668 - val_accuracy: 0.510
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 323us/step - loss:
1.2242 - accuracy: 0.5740 - val_loss: 1.2963 - val_accuracy: 0.579
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 350us/step - loss:
1.2400 - accuracy: 0.5910 - val_loss: 1.5393 - val_accuracy: 0.463
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
```

```
1000/1000 [=====] - 0s 349us/step - loss:
1.1613 - accuracy: 0.6240 - val_loss: 1.6911 - val_accuracy: 0.462
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 298us/step - loss:
1.2474 - accuracy: 0.6090 - val_loss: 1.5336 - val_accuracy: 0.509
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 270us/step - loss:
1.1576 - accuracy: 0.6300 - val_loss: 1.3663 - val_accuracy: 0.580
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 302us/step - loss:
1.1351 - accuracy: 0.6610 - val_loss: 1.4352 - val_accuracy: 0.552
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 317us/step - loss:
1.1092 - accuracy: 0.6660 - val_loss: 1.4057 - val_accuracy: 0.573
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 340us/step - loss:
1.0824 - accuracy: 0.6660 - val_loss: 1.3471 - val_accuracy: 0.577
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 327us/step - loss:
1.1116 - accuracy: 0.6600 - val_loss: 1.3435 - val_accuracy: 0.571
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 268us/step - loss:
1.0706 - accuracy: 0.6620 - val_loss: 1.4118 - val_accuracy: 0.562
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 288us/step - loss:
1.0189 - accuracy: 0.6920 - val_loss: 1.2614 - val_accuracy: 0.586
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 270us/step - loss:
0.9120 - accuracy: 0.7350 - val_loss: 1.2826 - val_accuracy: 0.605
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 273us/step - loss:
0.9882 - accuracy: 0.7120 - val_loss: 1.3823 - val_accuracy: 0.608
0
```



```
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 294us/step - loss:
0.9810 - accuracy: 0.7150 - val_loss: 1.3732 - val_accuracy: 0.589
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 350us/step - loss:
0.9436 - accuracy: 0.7140 - val_loss: 1.3441 - val_accuracy: 0.604
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 353us/step - loss:
0.9516 - accuracy: 0.7220 - val_loss: 1.4234 - val_accuracy: 0.582
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 336us/step - loss:
0.9703 - accuracy: 0.6870 - val_loss: 1.3739 - val_accuracy: 0.565
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 284us/step - loss:
0.9880 - accuracy: 0.7140 - val_loss: 1.3223 - val_accuracy: 0.592
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 290us/step - loss:
0.9670 - accuracy: 0.6910 - val_loss: 1.3103 - val_accuracy: 0.604
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 356us/step - loss:
0.9717 - accuracy: 0.6910 - val_loss: 1.3959 - val_accuracy: 0.582
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 336us/step - loss:
1.0063 - accuracy: 0.6910 - val_loss: 1.4337 - val_accuracy: 0.556
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 339us/step - loss:
1.0367 - accuracy: 0.6600 - val_loss: 1.3723 - val_accuracy: 0.556
0
```



### Three Layer Model with regularization

```
In [39]: model=model_nn_3(30,10,True)
history = model.fit(X_train,y_train,validation_data=(X_test,y_test)
,epochs=30, batch_size=10)
model_accuracy()
model_loss()
```

Model: "sequential\_14"

Layer (type)	Output Shape	Param #
dense_41 (Dense)	(None, 30)	23550
dense_42 (Dense)	(None, 30)	930
dense_43 (Dense)	(None, 30)	930
dense_44 (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

1000/1000 [=====] - 1s 518us/step - loss: 40.9483 - accuracy: 0.0810 - val\_loss: 2.3330 - val\_accuracy: 0.1000

Epoch 2/30

1000/1000 [=====] - 0s 314us/step - loss: 2.3452 - accuracy: 0.0920 - val\_loss: 2.3288 - val\_accuracy: 0.1000

Epoch 3/30

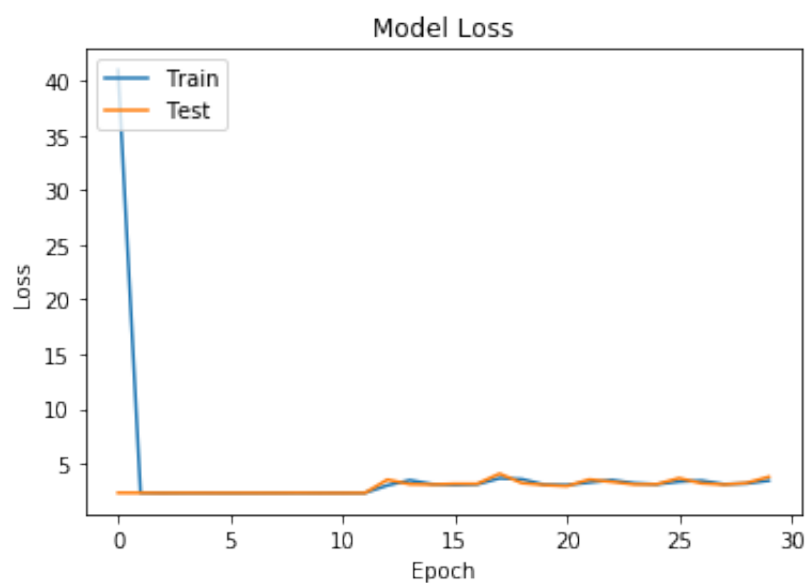
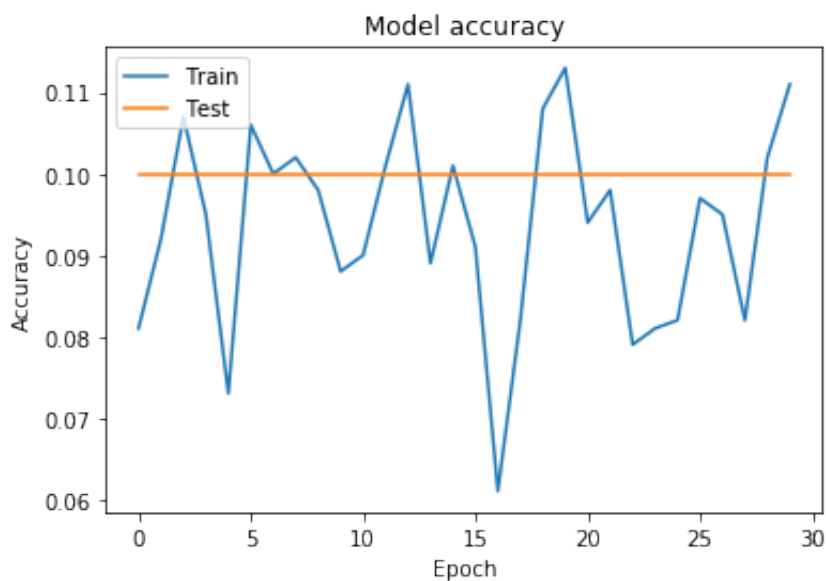
```
1000/1000 [=====] - 0s 258us/step - loss:
2.3267 - accuracy: 0.1070 - val_loss: 2.3122 - val_accuracy: 0.100
0
Epoch 4/30
1000/1000 [=====] - 0s 264us/step - loss:
2.3298 - accuracy: 0.0950 - val_loss: 2.3229 - val_accuracy: 0.100
0
Epoch 5/30
1000/1000 [=====] - 0s 261us/step - loss:
2.3334 - accuracy: 0.0730 - val_loss: 2.3163 - val_accuracy: 0.100
0
Epoch 6/30
1000/1000 [=====] - 0s 265us/step - loss:
2.3234 - accuracy: 0.1060 - val_loss: 2.3178 - val_accuracy: 0.100
0
Epoch 7/30
1000/1000 [=====] - 0s 263us/step - loss:
2.3276 - accuracy: 0.1000 - val_loss: 2.3195 - val_accuracy: 0.100
0
Epoch 8/30
1000/1000 [=====] - 0s 268us/step - loss:
2.3293 - accuracy: 0.1020 - val_loss: 2.3136 - val_accuracy: 0.100
0
Epoch 9/30
1000/1000 [=====] - 0s 265us/step - loss:
2.3299 - accuracy: 0.0980 - val_loss: 2.3266 - val_accuracy: 0.100
0
Epoch 10/30
1000/1000 [=====] - 0s 264us/step - loss:
2.3303 - accuracy: 0.0880 - val_loss: 2.3116 - val_accuracy: 0.100
0
Epoch 11/30
1000/1000 [=====] - 0s 270us/step - loss:
2.3318 - accuracy: 0.0900 - val_loss: 2.3106 - val_accuracy: 0.100
0
Epoch 12/30
1000/1000 [=====] - 0s 266us/step - loss:
2.3285 - accuracy: 0.1010 - val_loss: 2.3227 - val_accuracy: 0.100
0
Epoch 13/30
1000/1000 [=====] - 0s 271us/step - loss:
3.0105 - accuracy: 0.1110 - val_loss: 3.5358 - val_accuracy: 0.100
0
Epoch 14/30
1000/1000 [=====] - 0s 267us/step - loss:
3.4854 - accuracy: 0.0890 - val_loss: 3.1242 - val_accuracy: 0.100
0
Epoch 15/30
1000/1000 [=====] - 0s 280us/step - loss:
3.1300 - accuracy: 0.1010 - val_loss: 3.0899 - val_accuracy: 0.100
0
Epoch 16/30
1000/1000 [=====] - 0s 259us/step - loss:
```

```
3.0634 - accuracy: 0.0910 - val_loss: 3.1579 - val_accuracy: 0.100
0
Epoch 17/30
1000/1000 [=====] - 0s 258us/step - loss:
3.1053 - accuracy: 0.0610 - val_loss: 3.1554 - val_accuracy: 0.100
0
Epoch 18/30
1000/1000 [=====] - 0s 262us/step - loss:
3.6642 - accuracy: 0.0820 - val_loss: 4.0776 - val_accuracy: 0.100
0
Epoch 19/30
1000/1000 [=====] - 0s 267us/step - loss:
3.5802 - accuracy: 0.1080 - val_loss: 3.2243 - val_accuracy: 0.100
0
Epoch 20/30
1000/1000 [=====] - 0s 261us/step - loss:
3.0764 - accuracy: 0.1130 - val_loss: 3.0521 - val_accuracy: 0.100
0
Epoch 21/30
1000/1000 [=====] - 0s 273us/step - loss:
3.0462 - accuracy: 0.0940 - val_loss: 2.9504 - val_accuracy: 0.100
0
Epoch 22/30
1000/1000 [=====] - 0s 265us/step - loss:
3.2782 - accuracy: 0.0980 - val_loss: 3.5513 - val_accuracy: 0.100
0
Epoch 23/30
1000/1000 [=====] - 0s 268us/step - loss:
3.4983 - accuracy: 0.0790 - val_loss: 3.3394 - val_accuracy: 0.100
0
Epoch 24/30
1000/1000 [=====] - 0s 265us/step - loss:
3.2159 - accuracy: 0.0810 - val_loss: 3.0996 - val_accuracy: 0.100
0
Epoch 25/30
1000/1000 [=====] - 0s 263us/step - loss:
3.1089 - accuracy: 0.0820 - val_loss: 3.1162 - val_accuracy: 0.100
0
Epoch 26/30
1000/1000 [=====] - 0s 265us/step - loss:
3.3532 - accuracy: 0.0970 - val_loss: 3.6727 - val_accuracy: 0.100
0
Epoch 27/30
1000/1000 [=====] - 0s 264us/step - loss:
3.4477 - accuracy: 0.0950 - val_loss: 3.2031 - val_accuracy: 0.100
0
Epoch 28/30
1000/1000 [=====] - 0s 268us/step - loss:
3.1250 - accuracy: 0.0820 - val_loss: 3.0840 - val_accuracy: 0.100
0
Epoch 29/30
1000/1000 [=====] - 0s 266us/step - loss:
3.1779 - accuracy: 0.1020 - val_loss: 3.2363 - val_accuracy: 0.100
```

0

Epoch 30/30

1000/1000 [=====] - 0s 262us/step - loss:  
 3.4499 - accuracy: 0.1110 - val\_loss: 3.7903 - val\_accuracy: 0.100  
 0



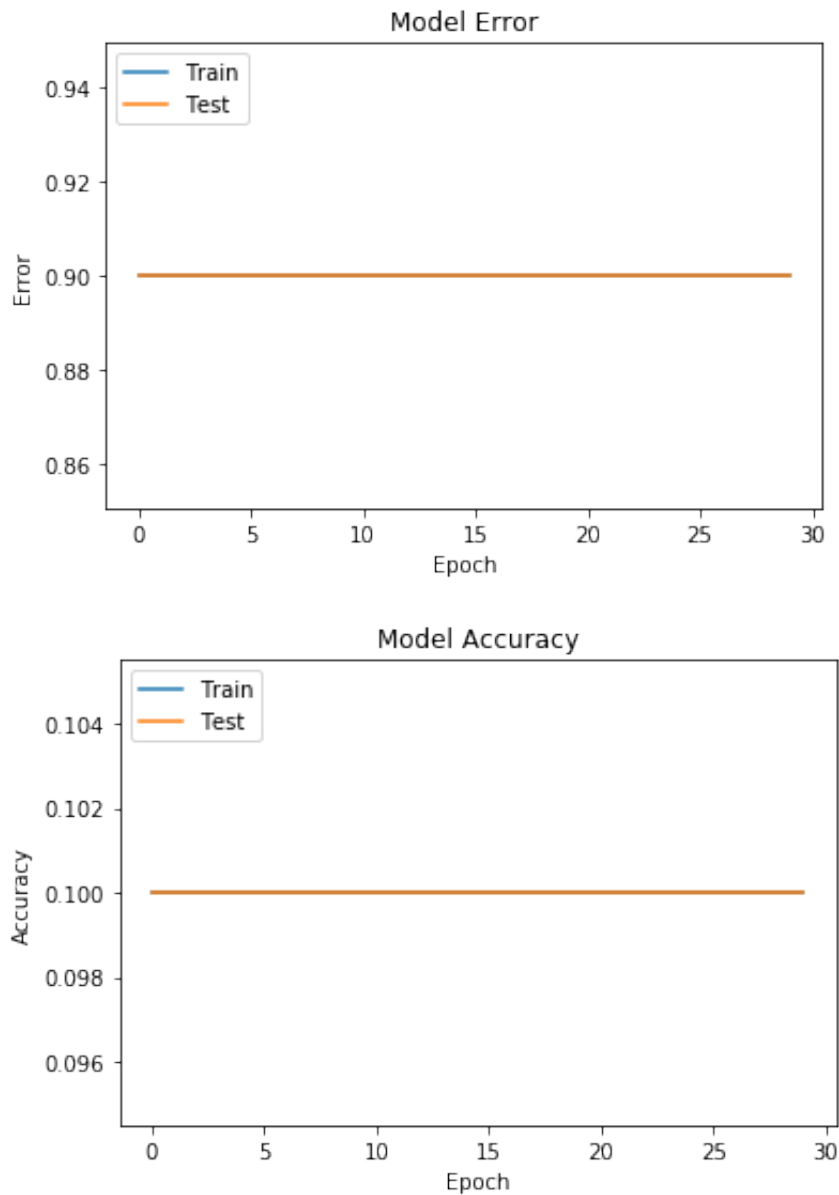
```
In [40]: model=model_nn_3(30,10,True)
         error_model(model)
```

Model: "sequential\_15"

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

```
dense_48 (Dense)                (None, 10)                310
=====
Total params: 25,720
Trainable params: 25,720
Non-trainable params: 0
=====
Epoch 1/1
1000/1000 [=====] - 0s 269us/step - loss:
40.8970 - accuracy: 0.1110
Epoch 1/1
1000/1000 [=====] - 0s 188us/step - loss:
2.3433 - accuracy: 0.0940
Epoch 1/1
1000/1000 [=====] - 0s 179us/step - loss:
2.3361 - accuracy: 0.0980
Epoch 1/1
1000/1000 [=====] - 0s 195us/step - loss:
2.3323 - accuracy: 0.0940
Epoch 1/1
1000/1000 [=====] - 0s 180us/step - loss:
2.3304 - accuracy: 0.0880
Epoch 1/1
1000/1000 [=====] - 0s 236us/step - loss:
2.3315 - accuracy: 0.0860
Epoch 1/1
1000/1000 [=====] - 0s 219us/step - loss:
2.3293 - accuracy: 0.0970
Epoch 1/1
1000/1000 [=====] - 0s 221us/step - loss:
2.3380 - accuracy: 0.0920
Epoch 1/1
1000/1000 [=====] - 0s 174us/step - loss:
2.3266 - accuracy: 0.1140
Epoch 1/1
1000/1000 [=====] - 0s 175us/step - loss:
2.3339 - accuracy: 0.0890
Epoch 1/1
1000/1000 [=====] - 0s 187us/step - loss:
2.3286 - accuracy: 0.0880
Epoch 1/1
1000/1000 [=====] - 0s 181us/step - loss:
2.3272 - accuracy: 0.0900
Epoch 1/1
1000/1000 [=====] - 0s 173us/step - loss:
3.0129 - accuracy: 0.1030
Epoch 1/1
1000/1000 [=====] - 0s 174us/step - loss:
3.5045 - accuracy: 0.0920
Epoch 1/1
1000/1000 [=====] - 0s 173us/step - loss:
3.1295 - accuracy: 0.0890
Epoch 1/1
1000/1000 [=====] - 0s 175us/step - loss:
```

```
3.0997 - accuracy: 0.0870
Epoch 1/1
1000/1000 [=====] - 0s 176us/step - loss:
3.0516 - accuracy: 0.1090
Epoch 1/1
1000/1000 [=====] - 0s 176us/step - loss:
3.6236 - accuracy: 0.1060
Epoch 1/1
1000/1000 [=====] - 0s 180us/step - loss:
3.6269 - accuracy: 0.0950
Epoch 1/1
1000/1000 [=====] - 0s 180us/step - loss:
3.1198 - accuracy: 0.0830
Epoch 1/1
1000/1000 [=====] - 0s 169us/step - loss:
3.0338 - accuracy: 0.1030
Epoch 1/1
1000/1000 [=====] - 0s 172us/step - loss:
3.1860 - accuracy: 0.1000
Epoch 1/1
1000/1000 [=====] - 0s 173us/step - loss:
3.5229 - accuracy: 0.0850
Epoch 1/1
1000/1000 [=====] - 0s 172us/step - loss:
3.2443 - accuracy: 0.0880
Epoch 1/1
1000/1000 [=====] - 0s 177us/step - loss:
3.1260 - accuracy: 0.0910
Epoch 1/1
1000/1000 [=====] - 0s 182us/step - loss:
3.2801 - accuracy: 0.0950
Epoch 1/1
1000/1000 [=====] - 0s 173us/step - loss:
3.3932 - accuracy: 0.0940
Epoch 1/1
1000/1000 [=====] - 0s 176us/step - loss:
3.2184 - accuracy: 0.0990
Epoch 1/1
1000/1000 [=====] - 0s 172us/step - loss:
3.2048 - accuracy: 0.0890
Epoch 1/1
1000/1000 [=====] - 0s 177us/step - loss:
3.3562 - accuracy: 0.0890
```



```
In [41]: model=model_nn_3(30,10,True)
W=LearningRate(model,(24790))
LearningRate_plot(W)
```

Model: "sequential\_16"

Layer (type)	Output Shape	Param #
dense_49 (Dense)	(None, 30)	23550
dense_50 (Dense)	(None, 30)	930
dense_51 (Dense)	(None, 30)	930
dense_52 (Dense)	(None, 10)	310

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

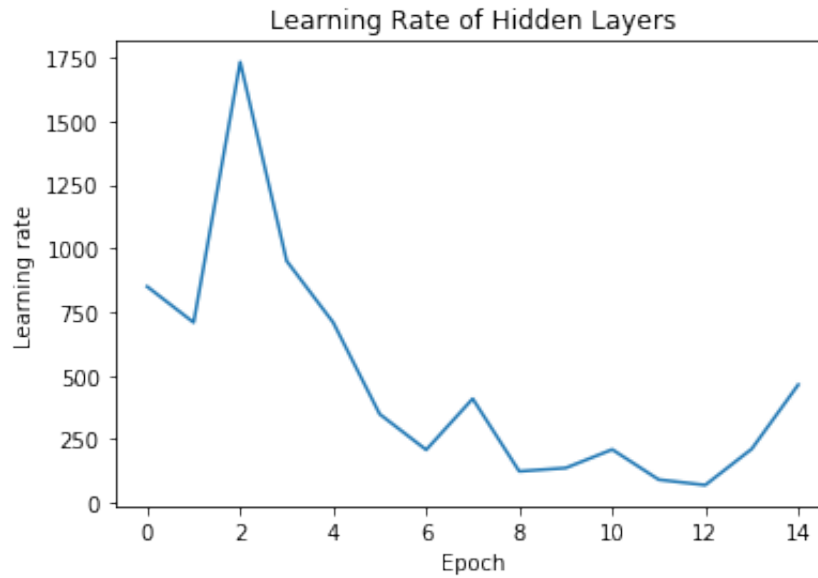


---

```
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 405us/step - loss:
40.6369 - accuracy: 0.1070 - val_loss: 2.3445 - val_accuracy: 0.10
00
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 277us/step - loss:
2.3492 - accuracy: 0.0930 - val_loss: 2.3382 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 281us/step - loss:
2.3296 - accuracy: 0.1010 - val_loss: 2.3533 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 368us/step - loss:
2.3319 - accuracy: 0.1080 - val_loss: 2.3329 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 359us/step - loss:
2.3296 - accuracy: 0.0970 - val_loss: 2.3300 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 318us/step - loss:
2.3252 - accuracy: 0.0990 - val_loss: 2.3208 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 389us/step - loss:
2.3365 - accuracy: 0.0890 - val_loss: 2.3343 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 300us/step - loss:
2.3299 - accuracy: 0.0800 - val_loss: 2.3182 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 304us/step - loss:
2.3208 - accuracy: 0.1120 - val_loss: 2.3241 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 299us/step - loss:
2.3284 - accuracy: 0.0830 - val_loss: 2.3153 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
```

```
1000/1000 [=====] - 0s 320us/step - loss:
2.3231 - accuracy: 0.1040 - val_loss: 2.3265 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 278us/step - loss:
2.3284 - accuracy: 0.0970 - val_loss: 2.3248 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 287us/step - loss:
3.0088 - accuracy: 0.0900 - val_loss: 3.4482 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 293us/step - loss:
3.4916 - accuracy: 0.0960 - val_loss: 3.2517 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 277us/step - loss:
3.1169 - accuracy: 0.0880 - val_loss: 3.0378 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 275us/step - loss:
3.0654 - accuracy: 0.1100 - val_loss: 3.0522 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 279us/step - loss:
3.0898 - accuracy: 0.0820 - val_loss: 3.2365 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 277us/step - loss:
3.6430 - accuracy: 0.0900 - val_loss: 4.1971 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 283us/step - loss:
3.5887 - accuracy: 0.0900 - val_loss: 3.2497 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 284us/step - loss:
3.1262 - accuracy: 0.0840 - val_loss: 3.1051 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 281us/step - loss:
3.0207 - accuracy: 0.1000 - val_loss: 3.0554 - val_accuracy: 0.100
0
```

```
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 280us/step - loss:
3.2636 - accuracy: 0.0740 - val_loss: 3.4238 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 284us/step - loss:
3.5086 - accuracy: 0.1060 - val_loss: 3.3532 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 273us/step - loss:
3.2125 - accuracy: 0.0890 - val_loss: 3.1706 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 274us/step - loss:
3.0988 - accuracy: 0.1130 - val_loss: 3.0557 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 279us/step - loss:
3.3321 - accuracy: 0.0960 - val_loss: 3.5043 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 277us/step - loss:
3.4340 - accuracy: 0.0850 - val_loss: 3.1911 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 271us/step - loss:
3.1545 - accuracy: 0.0810 - val_loss: 3.1151 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 288us/step - loss:
3.1840 - accuracy: 0.0920 - val_loss: 3.2157 - val_accuracy: 0.100
0
Train on 1000 samples, validate on 1000 samples
Epoch 1/1
1000/1000 [=====] - 0s 284us/step - loss:
3.4083 - accuracy: 0.0930 - val_loss: 3.8243 - val_accuracy: 0.100
0
```



## Defining CNN Model

```
In [126]: from keras.models import Sequential, Model
from keras.layers import Conv2D, MaxPooling2D, AveragePooling2D, Lambda, Input, Dense, Dropout, Flatten
from keras.layers.advanced_activations import LeakyReLU, ThresholdedReLU
from keras.layers.normalization import BatchNormalization
from keras.preprocessing.image import ImageDataGenerator
from keras.optimizers import SGD, Adam, Nadam, Adamax, TFOptimizer
```

```
In [127]: def cnn():  
  
    # Create Model  
    model = Sequential()  
    model.add(Conv2D(32, kernel_size=(3, 3), activation='linear', input_shape=(28, 28, 1), padding='same'))  
    model.add(LeakyReLU(alpha=0.1))  
    model.add(MaxPooling2D((2, 2), padding='same'))  
    model.add(Dropout(0.2))  
    model.add(Conv2D(64, (3, 3), activation='linear', padding='same'))  
    model.add(LeakyReLU(alpha=0.1))  
    model.add(MaxPooling2D(pool_size=(2, 2), padding='same'))  
    model.add(Dropout(0.2))  
    model.add(Conv2D(128, (3, 3), activation='linear', padding='same'))  
    model.add(LeakyReLU(alpha=0.1))  
    model.add(Dropout(0.2))  
    model.add(MaxPooling2D(pool_size=(2, 2), padding='same'))  
    model.add(Flatten())  
    model.add(Dense(256, activation='linear'))  
    model.add(LeakyReLU(alpha=0.1))  
    model.add(Dense(10, activation='softmax'))  
    model.compile(loss='categorical_crossentropy', optimizer=Adam(learning_rate=0.1), metrics=['accuracy'])  
    model.summary()  
  
    return model
```

```
In [128]: model=cnn()
```

Model: "sequential\_35"

Layer (type)	Output Shape	Param #
conv2d_22 (Conv2D)	(None, 28, 28, 32)	320
leaky_re_lu_29 (LeakyReLU)	(None, 28, 28, 32)	0
max_pooling2d_22 (MaxPooling)	(None, 14, 14, 32)	0
dropout_22 (Dropout)	(None, 14, 14, 32)	0
conv2d_23 (Conv2D)	(None, 14, 14, 64)	18496
leaky_re_lu_30 (LeakyReLU)	(None, 14, 14, 64)	0
max_pooling2d_23 (MaxPooling)	(None, 7, 7, 64)	0
dropout_23 (Dropout)	(None, 7, 7, 64)	0
conv2d_24 (Conv2D)	(None, 7, 7, 128)	73856
leaky_re_lu_31 (LeakyReLU)	(None, 7, 7, 128)	0
dropout_24 (Dropout)	(None, 7, 7, 128)	0
max_pooling2d_24 (MaxPooling)	(None, 4, 4, 128)	0
flatten_8 (Flatten)	(None, 2048)	0
dense_89 (Dense)	(None, 256)	524544
leaky_re_lu_32 (LeakyReLU)	(None, 256)	0
dense_90 (Dense)	(None, 10)	2570
Total params: 619,786		
Trainable params: 619,786		
Non-trainable params: 0		

In [129]: X\_train.shape,X\_test.shape

Out[129]: ((1000, 784), (1000, 784))

In [130]: y\_train.shape

Out[130]: (1000, 10)

In [131]: X\_train=X\_train.reshape(1000,28,28,1)  
X\_test=X\_test.reshape(1000,28,28,1)

```
In [132]: X_train.shape
```

```
Out[132]: (1000, 28, 28, 1)
```

```
In [133]: gen = ImageDataGenerator(width_shift_range=3,height_shift_range=3,rotation_range=3)
batches = gen.flow(X_train, y_train, batch_size=10)
val_batches = gen.flow(X_test, y_test, batch_size=10)

history=model.fit_generator(batches, steps_per_epoch=1000//10, epochs=30,validation_data=val_batches, validation_steps=1000//10)
```

Epoch 1/30

100/100 [=====] - 4s 39ms/step - loss: 16014.3135 - accuracy: 0.1460 - val\_loss: 1971.1371 - val\_accuracy: 0.2670

Epoch 2/30

100/100 [=====] - 3s 34ms/step - loss: 1030.2441 - accuracy: 0.2510 - val\_loss: 67.2700 - val\_accuracy: 0.2990

Epoch 3/30

100/100 [=====] - 3s 35ms/step - loss: 153.7517 - accuracy: 0.3330 - val\_loss: 191.4404 - val\_accuracy: 0.3500

Epoch 4/30

100/100 [=====] - 4s 37ms/step - loss: 1136.5329 - accuracy: 0.2720 - val\_loss: 2472.2087 - val\_accuracy: 0.1500

Epoch 5/30

100/100 [=====] - 4s 36ms/step - loss: 1191.7899 - accuracy: 0.2600 - val\_loss: 5188.4268 - val\_accuracy: 0.1290

Epoch 6/30

100/100 [=====] - 3s 35ms/step - loss: 8073340.3646 - accuracy: 0.1250 - val\_loss: 18144960.0000 - val\_accuracy: 0.1670

Epoch 7/30

100/100 [=====] - 3s 34ms/step - loss: 4579967.4138 - accuracy: 0.2210 - val\_loss: 593103.2500 - val\_accuracy: 0.2340

Epoch 8/30

100/100 [=====] - 4s 37ms/step - loss: 585645.8569 - accuracy: 0.3320 - val\_loss: 212050.1719 - val\_accuracy: 0.3680

Epoch 9/30

100/100 [=====] - 4s 35ms/step - loss: 329384.7674 - accuracy: 0.4060 - val\_loss: 529407.1250 - val\_accuracy: 0.4020

Epoch 10/30

100/100 [=====] - 3s 35ms/step - loss: 196399.3884 - accuracy: 0.4860 - val\_loss: 333899.0938 - val\_accuracy: 0.2550

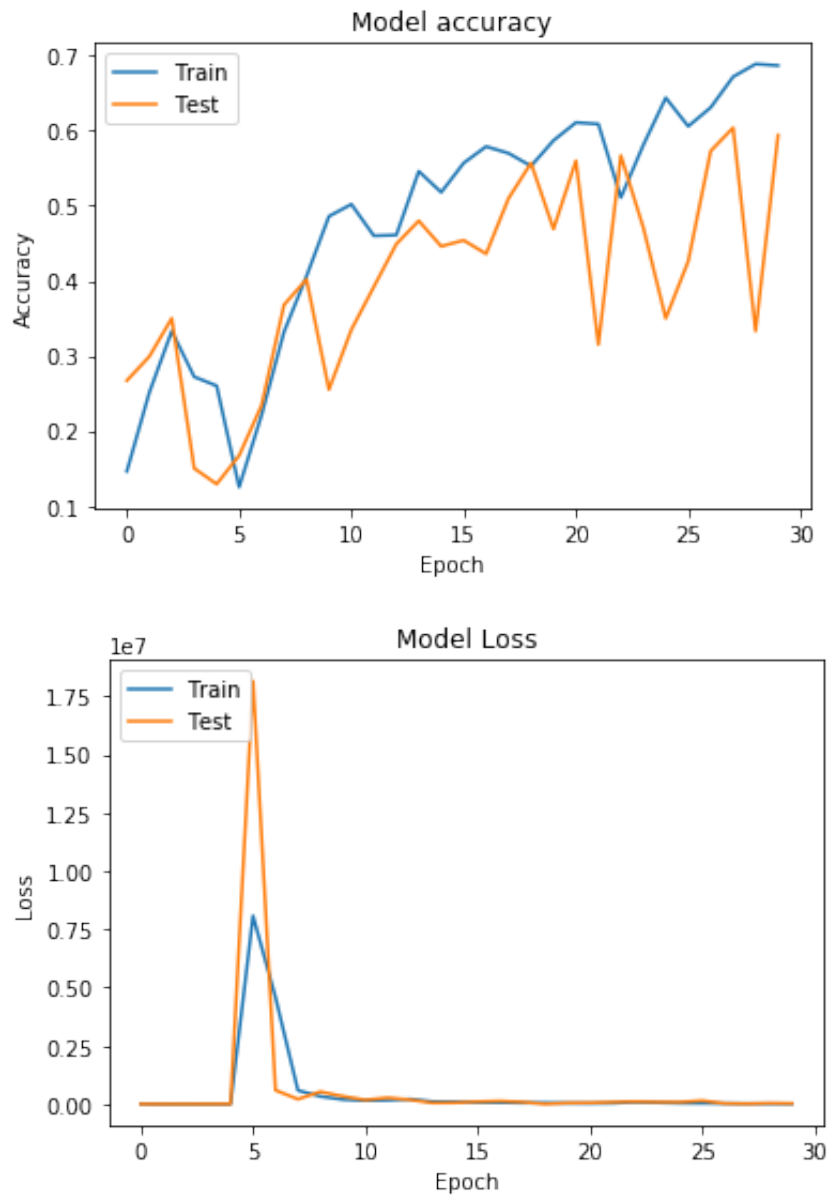
Epoch 11/30

```
100/100 [=====] - 3s 34ms/step - loss: 15
1382.6084 - accuracy: 0.5020 - val_loss: 178355.2500 - val_accurac
y: 0.3350
Epoch 12/30
100/100 [=====] - 4s 35ms/step - loss: 16
7572.5747 - accuracy: 0.4600 - val_loss: 271296.0938 - val_accurac
y: 0.3920
Epoch 13/30
100/100 [=====] - 4s 37ms/step - loss: 21
2337.0158 - accuracy: 0.4610 - val_loss: 183798.0000 - val_accurac
y: 0.4490
Epoch 14/30
100/100 [=====] - 4s 37ms/step - loss: 10
6882.1790 - accuracy: 0.5460 - val_loss: 49503.0000 - val_accuracy
: 0.4800
Epoch 15/30
100/100 [=====] - 4s 35ms/step - loss: 94
829.3142 - accuracy: 0.5180 - val_loss: 56790.2734 - val_accuracy:
0.4460
Epoch 16/30
100/100 [=====] - 3s 35ms/step - loss: 69
711.0031 - accuracy: 0.5570 - val_loss: 105118.8281 - val_accuracy
: 0.4540
Epoch 17/30
100/100 [=====] - 4s 37ms/step - loss: 64
046.1290 - accuracy: 0.5790 - val_loss: 129631.1250 - val_accuracy
: 0.4360
Epoch 18/30
100/100 [=====] - 4s 35ms/step - loss: 52
272.6501 - accuracy: 0.5700 - val_loss: 87196.1406 - val_accuracy:
0.5100
Epoch 19/30
100/100 [=====] - 4s 39ms/step - loss: 56
672.8345 - accuracy: 0.5530 - val_loss: 4202.5562 - val_accuracy:
0.5570
Epoch 20/30
100/100 [=====] - 4s 36ms/step - loss: 42
749.8611 - accuracy: 0.5870 - val_loss: 48134.4492 - val_accuracy:
0.4690
Epoch 21/30
100/100 [=====] - 4s 36ms/step - loss: 35
196.1669 - accuracy: 0.6110 - val_loss: 56403.7734 - val_accuracy:
0.5600
Epoch 22/30
100/100 [=====] - 4s 40ms/step - loss: 40
193.2020 - accuracy: 0.6090 - val_loss: 83542.1875 - val_accuracy:
0.3150
Epoch 23/30
100/100 [=====] - 4s 38ms/step - loss: 83
601.5778 - accuracy: 0.5110 - val_loss: 108153.8281 - val_accuracy
: 0.5670
Epoch 24/30
100/100 [=====] - 3s 34ms/step - loss: 60
```



```
569.3805 - accuracy: 0.5810 - val_loss: 87683.2500 - val_accuracy:
0.4710
Epoch 25/30
100/100 [=====] - 4s 35ms/step - loss: 40
339.9521 - accuracy: 0.6440 - val_loss: 85272.8750 - val_accuracy:
0.3500
Epoch 26/30
100/100 [=====] - 3s 34ms/step - loss: 37
744.4988 - accuracy: 0.6060 - val_loss: 146556.0469 - val_accuracy
: 0.4260
Epoch 27/30
100/100 [=====] - 4s 38ms/step - loss: 29
169.2627 - accuracy: 0.6310 - val_loss: 28230.4375 - val_accuracy:
0.5730
Epoch 28/30
100/100 [=====] - 4s 36ms/step - loss: 22
701.6729 - accuracy: 0.6720 - val_loss: 16576.8320 - val_accuracy:
0.6040
Epoch 29/30
100/100 [=====] - 4s 36ms/step - loss: 21
549.0789 - accuracy: 0.6890 - val_loss: 37421.1172 - val_accuracy:
0.3330
Epoch 30/30
100/100 [=====] - 4s 35ms/step - loss: 17
167.5762 - accuracy: 0.6870 - val_loss: 28015.0312 - val_accuracy:
0.5940
```

```
In [134]: model_accuracy()
          model_loss()
```



## References:

<https://www.kaggle.com/imrandude/fashion-mnist-cnn-imagedatagenerator>  
(<https://www.kaggle.com/imrandude/fashion-mnist-cnn-imagedatagenerator>)