Double Pendulum Trajectory Forecasting using RNN

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In [5]:
```
#What are we working with?
import sys
sys.version
```

Out[5]:
```
'3.6.3 |Anaconda custom (64-bit)| (default, Oct  6 2017, 12:04:38) \
n[GCC 4.2.1 Compatible Clang 4.0.1 (tags/RELEASE_401/final)]'
```

In [6]:
```
#Import Libraries
import tensorflow as tf
import pandas as pd
import numpy as np
import os
import matplotlib
import matplotlib.pyplot as plt
import random
%matplotlib inline
import tensorflow as tf
import shutil
import tensorflow.contrib.learn as tflearn
import tensorflow.contrib.layers as tflayers
from tensorflow.contrib.learn.python.learn import learn_runner
import tensorflow.contrib.metrics as metrics
import tensorflow.contrib.rnn as rnn

```

In [4]:
```
#TF Version
tf.__version__
```

Out[4]:
```
'1.3.0'
```

In [3]:
```

# for each experiment value of l1,l2,m1,m2 and th1,th2,w1,w2 are s

G = 9.8  # acceleration due to gravity, in m/s^2
L1 = 1.0  # length of pendulum 1 in m
L2 = 1.0  # length of pendulum 2 in m
M1 = 1.0  # mass of pendulum 1 in kg
M2 = 1.0  # mass of pendulum 2 in kg


def derivs(state, t):

    dydx = np.zeros_like(state)
    dydx[0] = state[1]
```

```
15
16      del_ = state[2] - state[0]
17      den1 = (M1 + M2)*L1 - M2*L1*cos(del_)*cos(del_)
18      dydx[1] = (M2*L1*state[1]*state[1]*sin(del_)*cos(del_) +
19                 M2*G*sin(state[2])*cos(del_) +
20                 M2*L2*state[3]*state[3]*sin(del_) -
21                 (M1 + M2)*G*sin(state[0]))/den1
22
23      dydx[2] = state[3]
24
25      den2 = (L2/L1)*den1
26      dydx[3] = (-M2*L2*state[3]*state[3]*sin(del_)*cos(del_) +
27                 (M1 + M2)*G*sin(state[0])*cos(del_) -
28                 (M1 + M2)*L1*state[1]*state[1]*sin(del_) -
29                 (M1 + M2)*G*sin(state[2]))/den2
30
31      return dydx
32
33  # create a time array from 0..100 sampled at 0.05 second steps
34  dt = 0.1
35  t = np.arange(0.0, 100, dt)
36
37  # th1 and th2 are the initial angles (degrees)
38  # w10 and w20 are the initial angular velocities (degrees per seco
39
40  th1 = 120.0
41  w1 = 0.0
42  th2 = 0.0
43  w2 = 0.0
44
45  # initial state
46  state = np.radians([th1, w1, th2, w2])
47
48  # integrate your ODE using scipy.integrate.
49  y = integrate.odeint(derivs, state, t)
50
51  x1 = L1*sin(y[:, 0])
52  y1 = -L1*cos(y[:, 0])
53
54  #print("x1 : ",x1)
55  #print("y1 : ",y1)
56
57  x2 = L2*sin(y[:, 2]) + x1
58  y2 = -L2*cos(y[:, 2]) + y1
59
60  #print("x2 : ",x2)
61  #print("y2 : ",y2)
62
63  fig = plt.figure()
64  ax = fig.add_subplot(111, autoscale_on=False, xlim=(-2, 2), ylim=(
65  ax.grid()
66
67  line, = ax.plot([], [], 'o-', lw=2)
68  time_template = 'time = %.1fs'
```

```
time_text = ax.text(0.05, 0.9, '', transform=ax.transAxes)


def init():
    line.set_data([], [])
    time_text.set_text('')
    return line, time_text


def animate(i):
    thisx = [0, x1[i], x2[i]]
    thisy = [0, y1[i], y2[i]]

    line.set_data(thisx, thisy)
    time_text.set_text(time_template % (i*dt))
    return line, time_text

ani = animation.FuncAnimation(fig, animate, np.arange(1, len(y)),i
```

```
---------------------------------------------------------------------
-------
NameError                                 Traceback (most recent cal
l last)
<ipython-input-3-20c9db25e880> in <module>()
     47
     48 # integrate your ODE using scipy.integrate.
---> 49 y = integrate.odeint(derivs, state, t)
     50
     51 x1 = L1*sin(y[:, 0])

NameError: name 'integrate' is not defined
```

Generate some data

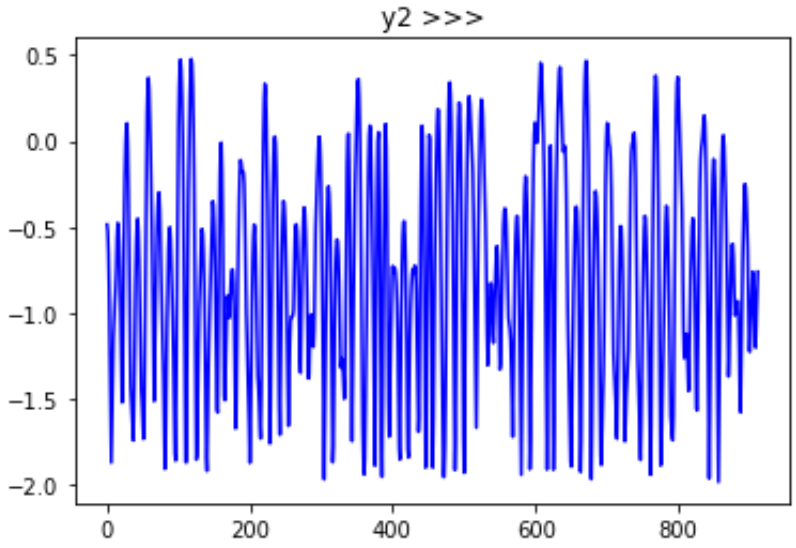
In [7]:

```
random.seed(111)
rng = pd.date_range(start='2000', periods=9, freq='M')
ts = pd.Series(np.random.uniform(-10, 10, size=len(rng)), rng).cum
#ts.head(10)

f = open("data/y2.txt" , "r")
t = open("data/time_slots.txt" , "r")
array1 = []
array2 = []

for line in f.read().split('\n') :
  array1.append(line )
for line in t.read().split('\n') :
  array2.append(line )
```

```
16 r.close()
17 t.close()
18 array1.pop()
19 array2.pop()
20 data = []
21
22 for i in range(len(array1)):
23     data.append([array1[i]])
24 test = np.array(data)
25 mylist = test.astype(np.float)
26
27 print("length of test data : ",len(mylist))
28 print(mylist.shape)
29 mylist.reshape(1,-1)
30 #print(mylist)
31
32 ts = np.delete(mylist,[i for i in range(912,1001)],0)
33
34 print("length of test data : ",len(ts))
35 print(ts.shape)
36 ts.reshape(1,-1)
37 #print(ts)
38
39 plt.plot(ts,c='b')
40 plt.title('y2 >>>')
41 plt.show()
42
43
```

```
length of test data :  1000
(1000, 1)
length of test data :  912
(912, 1)
```

```
/Users/omkarchakradharthawakar/anaconda3/lib/python3.6/site-packages
/ipykernel_launcher.py:32: DeprecationWarning: in the future out of
bounds indices will raise an error instead of being ignored by `nump
y.delete`.
```



Convert data into array that can be broken up into training "batches" that we will feed into our RNN model. Note the shape of the arrays.

In [8]:
```
TS = np.array(ts)
num_periods = 100
f_horizon = 1  #forecast horizon, one period into the future

x_data = TS[:(len(TS)-(len(TS) % num_periods))]
print(x_data.shape)
x_batches = x_data.reshape(-1, 100, 1)
print (len(x_batches))
print (x_batches.shape)
#print ("x_batches : ",x_batches)
y_data = TS[1:(len(TS)-(len(TS) % num_periods))+f_horizon]
print(y_data.shape)
y_batches = y_data.reshape(-1, 100, 1)


print ("y_batches : ",y_batches)
print (y_batches.shape)
```

```
(900, 1)
9
(9, 100, 1)
(900, 1)
y_batches :  [[[ -5.24476449e-01]
  [ -6.48676966e-01]
  [ -8.71019532e-01]
  [ -1.20275951e+00]
  [ -1.62066548e+00]
  [ -1.86911222e+00]
  [ -1.59735687e+00]
  [ -1.31756139e+00]
  [ -1.13473550e+00]
  [ -1.02169831e+00]
  [ -9.32942575e-01]
  [ -8.55187067e-01]
  [ -7.49839803e-01]
  [ -5.67737096e-01]
  [ -4.75580003e-01]
```

Pull out our test data

In [9]:
```
def test_data(series,forecast,num_periods):
    test_x_setup = TS[-(num_periods + forecast):]
    testX = test_x_setup[:num_periods].reshape(-1, 100, 1)
    testY = TS[-(num_periods):].reshape(-1, 100, 1)
    return testX,testY

X_test, Y_test = test_data(TS,f_horizon,num_periods )
print (X_test.shape)
#print (X_test)
print (Y_test.shape)

```

```
(1, 100, 1)
(1, 100, 1)
```

In [10]:
```
tf.reset_default_graph()   #We didn't have any previous graph obje

num_periods = 100      #number of periods per vector we are using
inputs = 1             #number of vectors submitted
hidden = 100           #number of neurons we will recursively work
output = 1             #number of output vectors

X = tf.placeholder(tf.float32, [None, num_periods, inputs])   #cre
y = tf.placeholder(tf.float32, [None, num_periods, output])


basic_cell = tf.contrib.rnn.BasicRNNCell(num_units=hidden, activat
rnn_output, states = tf.nn.dynamic_rnn(basic_cell, X, dtype=tf.flo

learning_rate = 0.001   #small learning rate so we don't overshoot

stacked_rnn_output = tf.reshape(rnn_output, [-1, hidden])
stacked_outputs = tf.layers.dense(stacked_rnn_output, output)
outputs = tf.reshape(stacked_outputs, [-1, num_periods, output])

loss = tf.reduce_sum(tf.square(outputs - y))    #define the cost f
optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate)
training_op = optimizer.minimize(loss)          #train the result

init = tf.global_variables_initializer()           #initialize all
```

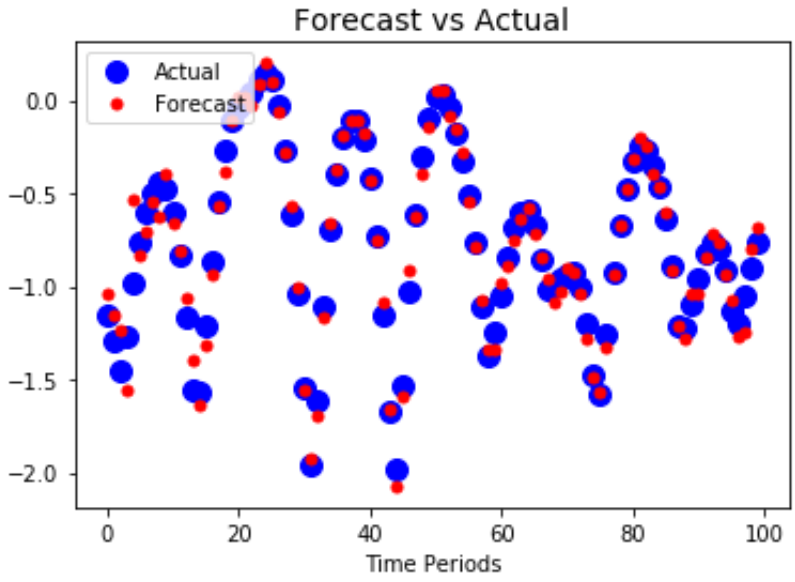
In [11]:
```
with tf.Session() as sess:
    writer = tf.summary.FileWriter("outputs_y", sess.graph)
    print(sess.run(init))
    writer.close()
```

```
None
```

In [12]:
```
epochs = 2000     #number of iterations or training cycles, includ
errors = []
iterations = []
with tf.Session() as sess:
    init.run()
    for ep in range(epochs):
        sess.run(training_op, feed_dict={X: x_batches, y: y_batche
        errors.append(loss.eval(feed_dict={X: x_batches, y: y_batc
        iterations.append(ep)
        if ep % 100 == 0:
            mse = loss.eval(feed_dict={X: x_batches, y: y_batches}
            print(ep, "\tMSE:", mse)
    y_pred = sess.run(outputs, feed_dict={X: X_test})
    print(y_pred)

```

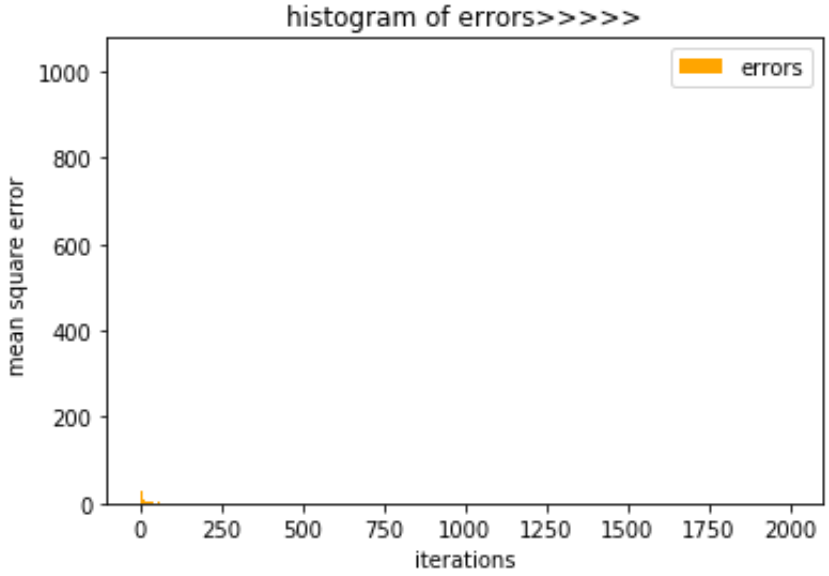
```
0       MSE: 590.14
100     MSE: 16.4718
200     MSE: 8.35422
300     MSE: 5.62936
400     MSE: 4.1909
500     MSE: 3.1443
600     MSE: 2.49152
700     MSE: 2.31495
800     MSE: 1.78809
900     MSE: 1.6103
1000    MSE: 1.82122
1100    MSE: 1.33001
1200    MSE: 1.69053
1300    MSE: 1.08325
1400    MSE: 0.990443
1500    MSE: 0.941035
1600    MSE: 0.875769
1700    MSE: 0.824849
1800    MSE: 0.755943
```

In [14]:
```
plt.title("Forecast vs Actual", fontsize=14)
plt.plot(pd.Series(np.ravel(Y_test)), "bo", markersize=10, label="
#plt.plot(pd.Series(np.ravel(Y_test)), "w*", markersize=10)
plt.plot(pd.Series(np.ravel(y_pred)), "r.", markersize=10, label="
plt.legend(loc="upper left")
plt.xlabel("Time Periods")

plt.show()


```



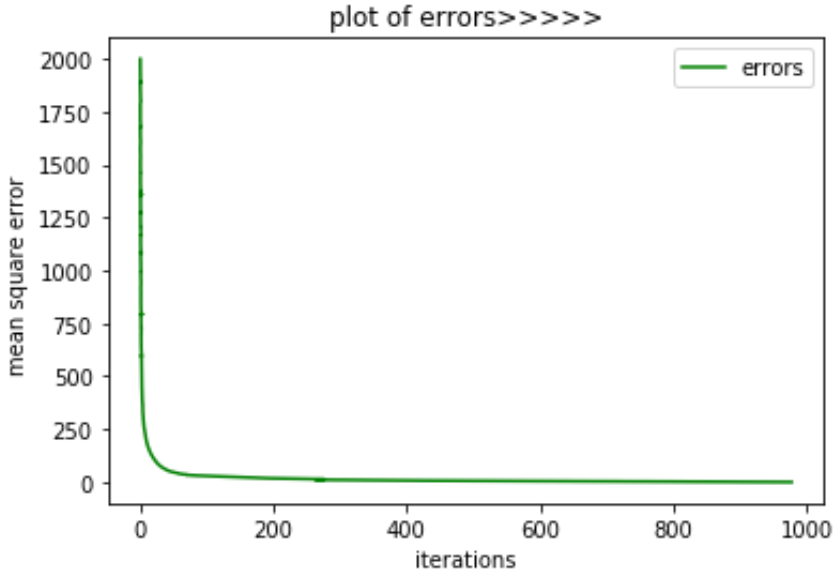
In [15]:
```
#!/usr/bin/env python
import numpy as np
import matplotlib.mlab as mlab
import matplotlib.pyplot as plt

errors=np.array(errors)
iterations=np.array(iterations)
print(errors.shape)
#print(errors)

plt.hist(errors,iterations,label='errors', facecolor='orange')

plt.xlabel('iterations')
plt.ylabel('mean square error ')
plt.title('histogram of errors>>>>>')
plt.legend()
plt.show()
```

```
(2000,)
```



In [16]:
```
#!/usr/bin/env python
import numpy as np
import matplotlib.mlab as mlab
import matplotlib.pyplot as plt

errors=np.array(errors)
iterations=np.array(iterations)
print(errors.shape)
#print(errors)

plt.plot(errors,iterations,label='errors',color='green')

plt.xlabel('iterations')
plt.ylabel('mean square error ')
plt.title('plot of errors>>>>>')
plt.legend()
plt.show()
```

```
(2000,)
```



In [ ]:
```
1
```