

Double Pendulum Trajectory Forecasting using RNN

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```
In [5]: 1 #What are we working with?
        2 import sys
        3 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]: 1 #Import Libraries
        2 import tensorflow as tf
        3 import pandas as pd
        4 import numpy as np
        5 import os
        6 import matplotlib
        7 import matplotlib.pyplot as plt
        8 import random
        9 %matplotlib inline
       10 import tensorflow as tf
       11 import shutil
       12 import tensorflow.contrib.learn as tflearn
       13 import tensorflow.contrib.layers as tflayers
       14 from tensorflow.contrib.learn.python.learn import learn_runner
       15 import tensorflow.contrib.metrics as metrics
       16 import tensorflow.contrib.rnn as rnn
       17
```

```
In [4]: 1 #TF Version
        2 tf.__version__
```

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

```
In [3]: 1
        2 # for each experiment value of l1,l2,m1,m2 and th1,th2,w1,w2 are s
        3
        4 G = 9.8 # acceleration due to gravity, in m/s^2
        5 L1 = 1.0 # length of pendulum 1 in m
        6 L2 = 1.0 # length of pendulum 2 in m
        7 M1 = 1.0 # mass of pendulum 1 in kg
        8 M2 = 1.0 # mass of pendulum 2 in kg
        9
       10
       11 def derivs(state, t):
       12
       13     dydx = np.zeros_like(state)
       14     dvdx[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'

```

```

69 time_text = ax.text(0.05, 0.9, '', transform=ax.transAxes)
70
71
72 def init():
73     line.set_data([], [])
74     time_text.set_text('')
75     return line, time_text
76
77
78 def animate(i):
79     thisx = [0, x1[i], x2[i]]
80     thisy = [0, y1[i], y2[i]]
81
82     line.set_data(thisx, thisy)
83     time_text.set_text(time_template % (i*dt))
84     return line, time_text
85
86 ani = animation.FuncAnimation(fig, animate, np.arange(1, len(y)), i
87
88

```

```

-----
-----
NameError                                Traceback (most recent call
1 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

```

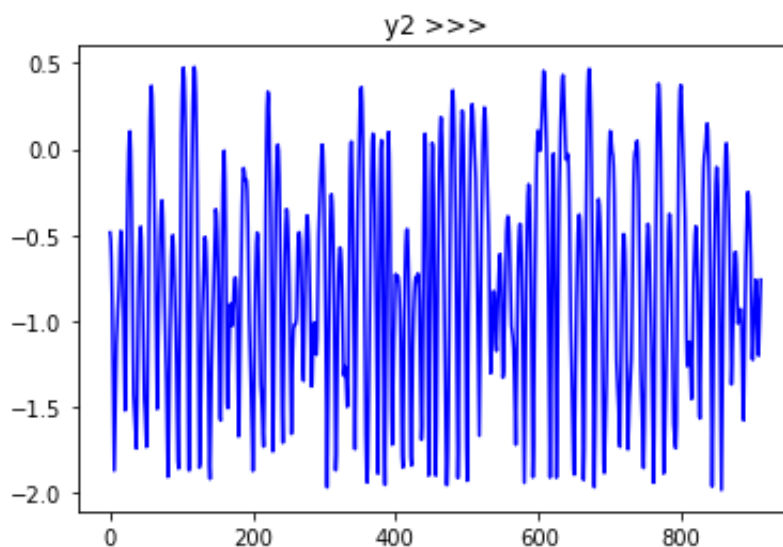
In [7]: 1 random.seed(111)
        2 rng = pd.date_range(start='2000', periods=9, freq='M')
        3 ts = pd.Series(np.random.uniform(-10, 10, size=len(rng)), rng).cumsum()
        4
        5 #ts.head(10)
        6
        7 f = open("data/y2.txt" , "r")
        8 t = open("data/time_slots.txt" , "r")
        9 array1 = []
       10 array2 = []
       11
       12 for line in f.read().split('\n') :
       13     array1.append(line)
       14 for line in t.read().split('\n') :
       15     array2.append(line)
       16

```

```
16 i.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 `numpy.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]: 1 TS = np.array(ts)
2 num_periods = 100
3 f_horizon = 1 #forecast horizon, one period into the future
4
5 x_data = TS[: (len(TS)-(len(TS) % num_periods))]
6 print(x_data.shape)
7 x_batches = x_data.reshape(-1, 100, 1)
8 print (len(x_batches))
9 print (x_batches.shape)
10 #print ("x_batches : ",x_batches)
11 y_data = TS[1:(len(TS)-(len(TS) % num_periods))+f_horizon]
12 print(y_data.shape)
13 y_batches = y_data.reshape(-1, 100, 1)
14
15
16 print ("y_batches : ",y_batches)
17 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]
[ -4.22125552e-01]
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[ -8.01111111e-02]
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[ -1.11111111e-02]
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[ 9.911111
```

Pull out our test data

```
In [9]: 1 def test_data(series,forecast,num_periods):
2         test_x_setup = TS[-(num_periods + forecast):]
3         testX = test_x_setup[:num_periods].reshape(-1, 100, 1)
4         testY = TS[-(num_periods):].reshape(-1, 100, 1)
5         return testX,testY
6
7 X_test, Y_test = test_data(TS,f_horizon,num_periods )
8 print (X_test.shape)
9 #print (X_test)
10 print (Y_test.shape)
11
(1, 100, 1)
(1, 100, 1)
```

```
In [10]: 1 tf.reset_default_graph()    #We didn't have any previous graph object
2
3 num_periods = 100          #number of periods per vector we are using
4 inputs = 1                 #number of vectors submitted
5 hidden = 100              #number of neurons we will recursively work
6 output = 1                #number of output vectors
7
8 X = tf.placeholder(tf.float32, [None, num_periods, inputs])    #create
9 y = tf.placeholder(tf.float32, [None, num_periods, output])
10
11
12 basic_cell = tf.contrib.rnn.BasicRNNCell(num_units=hidden, activation_fn=tf.nn.tanh)
13 rnn_output, states = tf.nn.dynamic_rnn(basic_cell, X, dtype=tf.float32)
14
15 learning_rate = 0.001     #small learning rate so we don't overshoot
16
17 stacked_rnn_output = tf.reshape(rnn_output, [-1, hidden])
18 stacked_outputs = tf.layers.dense(stacked_rnn_output, output)
19 outputs = tf.reshape(stacked_outputs, [-1, num_periods, output])
20
21 loss = tf.reduce_sum(tf.square(outputs - y))    #define the cost function
22 optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate)
23 training_op = optimizer.minimize(loss)          #train the result
24
25 init = tf.global_variables_initializer()        #initialize all variables
```

```
In [11]: 1 with tf.Session() as sess:
2         writer = tf.summary.FileWriter("outputs_y", sess.graph)
3         print(sess.run(init))
4         writer.close()
```

None

In [12]:

```

1 epochs = 2000      #number of iterations or training cycles, includ
2 errors = []
3 iterations = []
4 with tf.Session() as sess:
5     init.run()
6     for ep in range(epochs):
7         sess.run(training_op, feed_dict={X: x_batches, y: y_batches})
8         errors.append(loss.eval(feed_dict={X: x_batches, y: y_batches}))
9         iterations.append(ep)
10        if ep % 100 == 0:
11            mse = loss.eval(feed_dict={X: x_batches, y: y_batches})
12            print(ep, "\tMSE:", mse)
13        y_pred = sess.run(outputs, feed_dict={X: X_test})
14        print(y_pred)
15

```

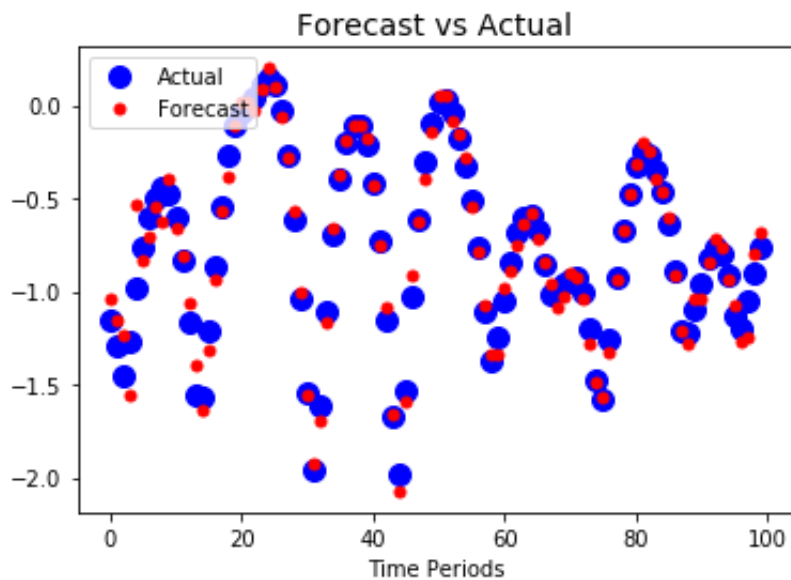
```

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
1900   MSE: 0.61007

```

In [14]:

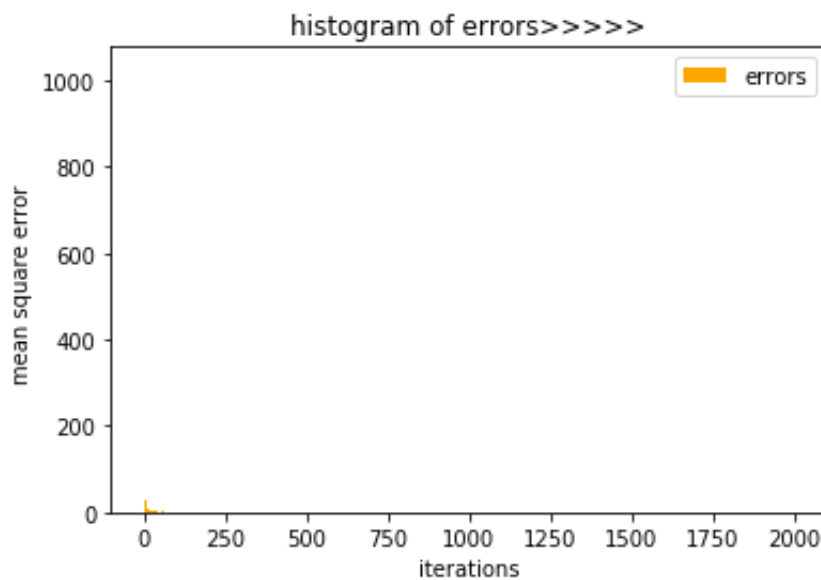
```
1 plt.title("Forecast vs Actual", fontsize=14)
2 plt.plot(pd.Series(np.ravel(Y_test)), "bo", markersize=10, label="Actual")
3 #plt.plot(pd.Series(np.ravel(Y_test)), "w*", markersize=10)
4 plt.plot(pd.Series(np.ravel(y_pred)), "r.", markersize=10, label="Forecast")
5 plt.legend(loc="upper left")
6 plt.xlabel("Time Periods")
7
8 plt.show()
9
10
```



In [15]:

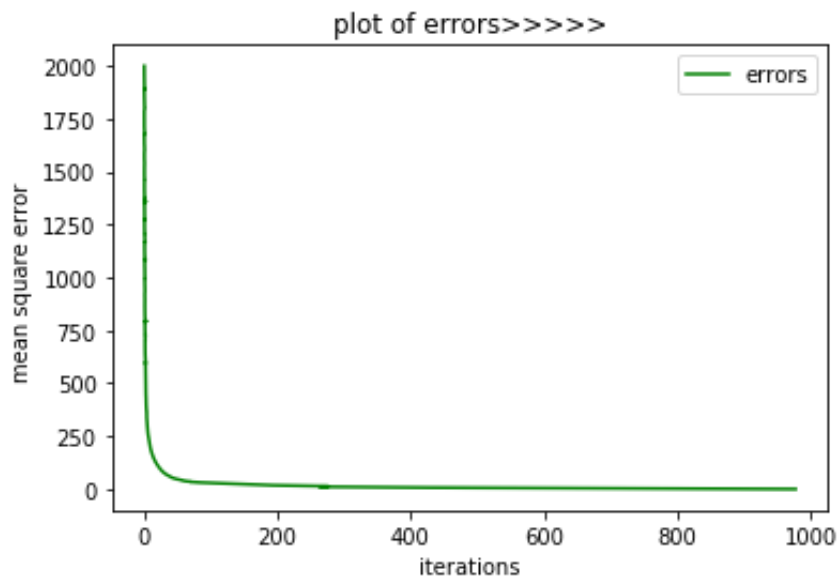
```
1 #!/usr/bin/env python
2 import numpy as np
3 import matplotlib.mlab as mlab
4 import matplotlib.pyplot as plt
5
6 errors=np.array(errors)
7 iterations=np.array(iterations)
8 print(errors.shape)
9 #print(errors)
10
11 plt.hist(errors,iterations,label='errors', facecolor='orange')
12
13 plt.xlabel('iterations')
14 plt.ylabel('mean square error ')
15 plt.title('histogram of errors>>>>')
16 plt.legend()
17 plt.show()
```

(2000,)



```
In [16]: 1 #!/usr/bin/env python
2 import numpy as np
3 import matplotlib.mlab as mlab
4 import matplotlib.pyplot as plt
5
6 errors=np.array(errors)
7 iterations=np.array(iterations)
8 print(errors.shape)
9 #print(errors)
10
11 plt.plot(errors,iterations,label='errors',color='green')
12
13 plt.xlabel('iterations')
14 plt.ylabel('mean square error ')
15 plt.title('plot of errors>>>>')
16 plt.legend()
17 plt.show()
```

(2000,)



```
In [ ]: 1
```