

## 12.lstm-seq2seq-vae

September 29, 2021

```
[1]: import sys
import warnings

if not sys.warnoptions:
    warnings.simplefilter('ignore')
```

```
[2]: import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
from datetime import datetime
from datetime import timedelta
from tqdm import tqdm
sns.set()
tf.compat.v1.random.set_random_seed(1234)
```

```
[3]: df = pd.read_csv('../dataset/G00G-year.csv')
df.head()
```

```
[3]:
```

	Date	Open	High	Low	Close	Adj Close	\
0	2016-11-02	778.200012	781.650024	763.450012	768.700012	768.700012	
1	2016-11-03	767.250000	769.950012	759.030029	762.130005	762.130005	
2	2016-11-04	750.659973	770.359985	750.560974	762.020020	762.020020	
3	2016-11-07	774.500000	785.190002	772.549988	782.520020	782.520020	
4	2016-11-08	783.400024	795.632996	780.190002	790.510010	790.510010	

	Volume
0	1872400
1	1943200
2	2134800
3	1585100
4	1350800

```
[4]: minmax = MinMaxScaler().fit(df.iloc[:, 4:5].astype('float32')) # Close index
df_log = minmax.transform(df.iloc[:, 4:5].astype('float32')) # Close index
```

```
df_log = pd.DataFrame(df_log)
df_log.head()
```

```
[4]:      0
0  0.112708
1  0.090008
2  0.089628
3  0.160459
4  0.188066
```

## 0.1 Split train and test

I will cut the dataset to train and test datasets,

1. Train dataset derived from starting timestamp until last 30 days
2. Test dataset derived from last 30 days until end of the dataset

So we will let the model do forecasting based on last 30 days, and we will going to repeat the experiment for 10 times. You can increase it locally if you want, and tuning parameters will help you by a lot.

```
[5]: test_size = 30
simulation_size = 10

df_train = df_log.iloc[:-test_size]
df_test = df_log.iloc[-test_size:]
df.shape, df_train.shape, df_test.shape
```

```
[5]: ((252, 7), (222, 1), (30, 1))
```

```
[6]: class Model:
    def __init__(
        self,
        learning_rate,
        num_layers,
        size,
        size_layer,
        output_size,
        forget_bias = 0.1,
        lambda_coeff = 0.5
    ):
        def lstm_cell(size_layer):
            return tf.nn.rnn_cell.LSTMCell(size_layer, state_is_tuple = False)

        rnn_cells = tf.nn.rnn_cell.MultiRNNCell(
            [lstm_cell(size_layer) for _ in range(num_layers)],
            state_is_tuple = False,
        )
```

```

self.X = tf.placeholder(tf.float32, (None, None, size))
self.Y = tf.placeholder(tf.float32, (None, output_size))
drop = tf.contrib.rnn.DropoutWrapper(
    rnn_cells, output_keep_prob = forget_bias
)
self.hidden_layer = tf.placeholder(
    tf.float32, (None, num_layers * 2 * size_layer)
)
_, last_state = tf.nn.dynamic_rnn(
    drop, self.X, initial_state = self.hidden_layer, dtype = tf.float32
)

self.z_mean = tf.layers.dense(last_state, size)
self.z_log_sigma = tf.layers.dense(last_state, size)

epsilon = tf.random_normal(tf.shape(self.z_log_sigma))
self.z_vector = self.z_mean + tf.exp(self.z_log_sigma)

with tf.variable_scope('decoder', reuse = False):
    rnn_cells_dec = tf.nn.rnn_cell.MultiRNNCell(
        [lstm_cell(size_layer) for _ in range(num_layers)],
        state_is_tuple = False
    )
    drop_dec = tf.contrib.rnn.DropoutWrapper(
        rnn_cells_dec, output_keep_prob = forget_bias
    )
    x = tf.concat([tf.expand_dims(self.z_vector, axis=0), self.X], axis=
    1)

    self.outputs, self.last_state = tf.nn.dynamic_rnn(
        drop_dec, self.X, initial_state = last_state, dtype = tf.float32
    )

self.logits = tf.layers.dense(self.outputs[-1], output_size)
self.lambda_coeff = lambda_coeff

self.kl_loss = -0.5 * tf.reduce_sum(1.0 + 2 * self.z_log_sigma - self.
z_mean ** 2 -
                                tf.exp(2 * self.z_log_sigma), 1)
self.kl_loss = tf.scalar_mul(self.lambda_coeff, self.kl_loss)
self.cost = tf.reduce_mean(tf.square(self.Y - self.logits) + self.
kl_loss)
self.optimizer = tf.train.AdamOptimizer(learning_rate).minimize(
    self.cost
)

def calculate_accuracy(real, predict):
    real = np.array(real) + 1

```

```

predict = np.array(predict) + 1
percentage = 1 - np.sqrt(np.mean(np.square((real - predict) / real)))
return percentage * 100

def anchor(signal, weight):
    buffer = []
    last = signal[0]
    for i in signal:
        smoothed_val = last * weight + (1 - weight) * i
        buffer.append(smoothed_val)
        last = smoothed_val
    return buffer

```

```

[7]: num_layers = 1
size_layer = 128
timestamp = 5
epoch = 300
dropout_rate = 0.8
future_day = test_size
learning_rate = 0.01

```

```

[8]: def forecast():
    tf.reset_default_graph()
    modelnn = Model(
        learning_rate, num_layers, df_log.shape[1], size_layer, df_log.
↪shape[1], dropout_rate
    )
    sess = tf.InteractiveSession()
    sess.run(tf.global_variables_initializer())
    date_ori = pd.to_datetime(df.iloc[:, 0]).tolist()

    pbar = tqdm(range(epoch), desc = 'train loop')
    for i in pbar:
        init_value = np.zeros((1, num_layers * 2 * size_layer))
        total_loss, total_acc = [], []
        for k in range(0, df_train.shape[0] - 1, timestamp):
            index = min(k + timestamp, df_train.shape[0] - 1)
            batch_x = np.expand_dims(
                df_train.iloc[k : index, :].values, axis = 0
            )
            batch_x = np.random.binomial(1, 0.5, batch_x.shape) * batch_x
            batch_y = df_train.iloc[k + 1 : index + 1, :].values
            logits, last_state, _, loss = sess.run(
                [modelnn.logits, modelnn.last_state, modelnn.optimizer, modelnn.
↪cost],
                feed_dict = {
                    modelnn.X: batch_x,

```

```

        modelnn.Y: batch_y,
        modelnn.hidden_layer: init_value,
    },
)
init_value = last_state
total_loss.append(loss)
total_acc.append(calculate_accuracy(batch_y[:, 0], logits[:, 0]))
pbar.set_postfix(cost = np.mean(total_loss), acc = np.mean(total_acc))

future_day = test_size

output_predict = np.zeros((df_train.shape[0] + future_day, df_train.
↪shape[1]))
output_predict[0] = df_train.iloc[0]
upper_b = (df_train.shape[0] // timestamp) * timestamp
init_value = np.zeros((1, num_layers * 2 * size_layer))

for k in range(0, (df_train.shape[0] // timestamp) * timestamp, timestamp):
    out_logits, last_state = sess.run(
        [modelnn.logits, modelnn.last_state],
        feed_dict = {
            modelnn.X: np.expand_dims(
                df_train.iloc[k : k + timestamp], axis = 0
            ),
            modelnn.hidden_layer: init_value,
        },
    )
    init_value = last_state
    output_predict[k + 1 : k + timestamp + 1] = out_logits

if upper_b != df_train.shape[0]:
    out_logits, last_state = sess.run(
        [modelnn.logits, modelnn.last_state],
        feed_dict = {
            modelnn.X: np.expand_dims(df_train.iloc[upper_b:], axis = 0),
            modelnn.hidden_layer: init_value,
        },
    )
    output_predict[upper_b + 1 : df_train.shape[0] + 1] = out_logits
    future_day -= 1
    date_ori.append(date_ori[-1] + timedelta(days = 1))

init_value = last_state

for i in range(future_day):
    o = output_predict[-future_day - timestamp + i:-future_day + i]
    out_logits, last_state = sess.run(

```

```

        [modelnn.logits, modelnn.last_state],
        feed_dict = {
            modelnn.X: np.expand_dims(o, axis = 0),
            modelnn.hidden_layer: init_value,
        },
    )
    init_value = last_state
    output_predict[-future_day + i] = out_logits[-1]
    date_ori.append(date_ori[-1] + timedelta(days = 1))

    output_predict = minmax.inverse_transform(output_predict)
    deep_future = anchor(output_predict[:, 0], 0.3)

    return deep_future[-test_size:]

```

```

[9]: results = []
    for i in range(simulation_size):
        print('simulation %d'%(i + 1))
        results.append(forecast())

```

WARNING: Logging before flag parsing goes to stderr.

W0816 15:26:45.502804 139658996016960 deprecation.py:323] From <ipython-input-6-d907d7a4dee6>:13: LSTMCell.\_\_init\_\_ (from tensorflow.python.ops.rnn\_cell\_impl) is deprecated and will be removed in a future version.

Instructions for updating:

This class is equivalent as tf.keras.layers.LSTMCell, and will be replaced by that in Tensorflow 2.0.

W0816 15:26:45.505823 139658996016960 rnn\_cell\_impl.py:893] <tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f04dbc873c8>: Using a concatenated state is slower and will soon be deprecated. Use state\_is\_tuple=True.

W0816 15:26:45.507445 139658996016960 deprecation.py:323] From <ipython-input-6-d907d7a4dee6>:17: MultiRNNCell.\_\_init\_\_ (from tensorflow.python.ops.rnn\_cell\_impl) is deprecated and will be removed in a future version.

Instructions for updating:

This class is equivalent as tf.keras.layers.StackedRNNCells, and will be replaced by that in Tensorflow 2.0.

simulation 1

W0816 15:26:45.829126 139658996016960 lazy\_loader.py:50] The TensorFlow contrib module will not be included in TensorFlow 2.0. For more information, please see:

- \* <https://github.com/tensorflow/community/blob/master/rfcs/20180907-contrib-sunset.md>
- \* <https://github.com/tensorflow/addons>

\* <https://github.com/tensorflow/io> (for I/O related ops)  
If you depend on functionality not listed there, please file an issue.

W0816 15:26:45.832581 139658996016960 deprecation.py:323] From <ipython-input-6-d907d7a4dee6>:28: dynamic\_rnn (from tensorflow.python.ops.rnn) is deprecated and will be removed in a future version.

Instructions for updating:

Please use ``keras.layers.RNN(cell)``, which is equivalent to this API

W0816 15:26:46.024316 139658996016960 deprecation.py:506] From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/init\_ops.py:1251: calling VarianceScaling.\_\_init\_\_ (from tensorflow.python.ops.init\_ops) with dtype is deprecated and will be removed in a future version.

Instructions for updating:

Call initializer instance with the dtype argument instead of passing it to the constructor

W0816 15:26:46.031064 139658996016960 deprecation.py:506] From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/rnn\_cell\_impl.py:961: calling Zeros.\_\_init\_\_ (from tensorflow.python.ops.init\_ops) with dtype is deprecated and will be removed in a future version.

Instructions for updating:

Call initializer instance with the dtype argument instead of passing it to the constructor

W0816 15:26:46.507349 139658996016960 deprecation.py:323] From <ipython-input-6-d907d7a4dee6>:31: dense (from tensorflow.python.layers.core) is deprecated and will be removed in a future version.

Instructions for updating:

Use `keras.layers.dense` instead.

W0816 15:26:46.696353 139658996016960 rnn\_cell\_impl.py:893] <tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f04778236d8>: Using a concatenated state is slower and will soon be deprecated. Use `state_is_tuple=True`.

W0816 15:26:46.879564 139658996016960 deprecation.py:323] From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/math\_grad.py:1205: add\_dispatch\_support.<locals>.wrapper (from tensorflow.python.ops.array\_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use `tf.where` in 2.0, which has the same broadcast rule as `np.where`  
train loop: 100%| | 300/300 [01:47<00:00, 2.80it/s, acc=97, cost=0.00235]

W0816 15:28:35.363878 139658996016960 rnn\_cell\_impl.py:893] <tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f0455d7deb8>: Using a concatenated state is slower and will soon be deprecated. Use `state_is_tuple=True`.

W0816 15:28:35.471002 139658996016960 rnn\_cell\_impl.py:893] <tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f044b3bf198>: Using a concatenated state is slower and will soon be deprecated. Use `state_is_tuple=True`.

simulation 2

train loop: 100%| | 300/300 [01:46<00:00, 2.82it/s, acc=96.9,  
cost=0.00305]

W0816 15:30:22.970038 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f044b349f60>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.

W0816 15:30:23.075726 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f03f03e0e10>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.

simulation 3

train loop: 100%| | 300/300 [01:47<00:00, 2.77it/s, acc=95.1,  
cost=0.00633]

W0816 15:32:11.926008 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f03f043bfd0>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.

W0816 15:32:12.031505 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f03a4bb2a58>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.

simulation 4

train loop: 100%| | 300/300 [01:47<00:00, 2.86it/s, acc=95.9,  
cost=0.00422]

W0816 15:34:00.252120 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f03f0194f60>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.

simulation 5

W0816 15:34:00.478516 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f044c079320>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.

train loop: 100%| | 300/300 [01:48<00:00, 2.75it/s, acc=96.3,  
cost=0.00351]

W0816 15:35:49.588577 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f03965d9940>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.

W0816 15:35:49.693055 139658996016960 rnn\_cell\_impl.py:893]

<tensorflow.python.ops.rnn\_cell\_impl.LSTMCell object at 0x7f0393f722e8>: Using a  
concatenated state is slower and will soon be deprecated. Use  
state\_is\_tuple=True.



simulation 6

```
train loop: 100%|      | 300/300 [01:47<00:00, 2.79it/s, acc=96.2,
cost=0.00384]
W0816 15:37:38.517486 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f0393f489e8>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
W0816 15:37:38.625684 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f039189c940>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
```

simulation 7

```
train loop: 100%|      | 300/300 [01:47<00:00, 2.79it/s, acc=95.6,
cost=0.00472]
W0816 15:39:27.256033 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f0391929fd0>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
W0816 15:39:27.363451 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f038f1f4d68>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
```

simulation 8

```
train loop: 100%|      | 300/300 [01:47<00:00, 2.78it/s, acc=96.1,
cost=0.00394]
W0816 15:41:15.619689 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f038f286940>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
W0816 15:41:15.724680 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f038cb1fda0>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
```

simulation 9

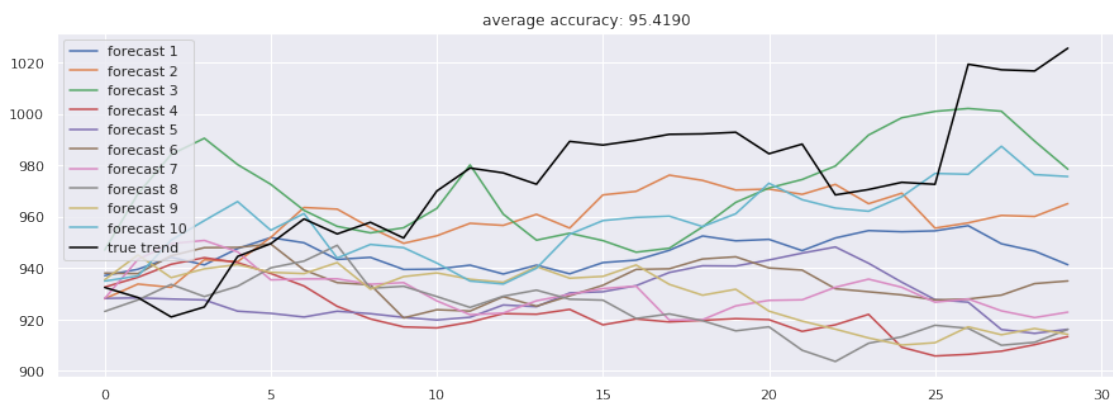
```
train loop: 100%|      | 300/300 [01:47<00:00, 2.82it/s, acc=97.3,
cost=0.00223]
W0816 15:43:04.145420 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f038d74f630>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
W0816 15:43:04.251741 139658996016960 rnn_cell_impl.py:893]
<tensorflow.python.ops.rnn_cell_impl.LSTMCell object at 0x7f0388c03d68>: Using a
concatenated state is slower and will soon be deprecated. Use
state_is_tuple=True.
```

simulation 10

train loop: 100% | 300/300 [01:45<00:00, 2.82it/s, acc=96.6,  
cost=0.00292]

```
[10]: accuracies = [calculate_accuracy(df['Close'].iloc[-test_size:].values, r) for r in results]

plt.figure(figsize = (15, 5))
for no, r in enumerate(results):
    plt.plot(r, label = 'forecast %d'%(no + 1))
plt.plot(df['Close'].iloc[-test_size:].values, label = 'true trend', c = 'black')
plt.legend()
plt.title('average accuracy: %.4f'%(np.mean(accuracies)))
plt.show()
```



[ ]: