

In [1]:



```
# import libraries

import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import SimpleRNN
from tensorflow.keras.layers import Dense

import numpy as np
from numpy import array
import os
import pandas as pd
import matplotlib.pyplot as plt
```

In [2]:



```
# define input sequence

raw_seq = [10, 20, 30, 40, 50, 60, 70, 80, 90]

# choose a number of time steps
seq_len = 3

# choose number of future time steps to predict
n_steps = 2
```

In [3]:



```
def split_sequence(sequence, seq_len, n_steps):
    X, y = list(), list()
    for i in range(len(sequence)):
        # find the end of this pattern
        end_ix = i + seq_len
        # check if we are beyond the sequence
        if end_ix + n_steps > len(sequence):
            break
        # gather input and output parts of the pattern
        seq_x, seq_y = sequence[i:end_ix], sequence[end_ix:end_ix+n_steps]
        X.append(seq_x)
        y.append(seq_y)
    return array(X), array(y)
```

In [4]:



```
# split into sequences

seq_X, seq_Y = split_sequence(raw_seq, seq_len, n_steps)

# summarize the data
for i in range(len(seq_X)):
    print(seq_X[i], seq_Y[i])
```

```
[10 20 30] [40 50]
[20 30 40] [50 60]
[30 40 50] [60 70]
[40 50 60] [70 80]
[50 60 70] [80 90]
```

In [5]:



```
# split the sequence generated into final training X and Y sequence to be fed to RNN model

X = seq_X.reshape((seq_X.shape[0], seq_X.shape[1], 1))
Y = seq_Y

print(X.shape, Y.shape)
```

```
(5, 3, 1) (5, 2)
```

In [8]:



```
#RNN
model = Sequential()
model.add(SimpleRNN(50, activation='relu', input_shape=(seq_len, 1)))
model.add(Dense(n_steps))
model.compile(optimizer='adam', loss='mse')

print(model.summary())
```

```
Model: "sequential_2"
```

Layer (type)	Output Shape	Param #
=====		
simple_rnn_1 (SimpleRNN)	(None, 50)	2600
dense_1 (Dense)	(None, 2)	102
=====		
Total params: 2,702		
Trainable params: 2,702		
Non-trainable params: 0		
None		

For Simple RNN layer:

Number of parameters is given by the formula:

$(\text{num_features} + \text{num_units}) * \text{num_units} + \text{num_units}$

num_units = equals the number of units in the RNN

num_features = equals the number features of your input

Now you have two things happening in your RNN.

First you have the recurrent loop, where the state is fed recurrently into the model to generate the next step. Weights for the recurrent step are:

$\text{recurrent_weights} = \text{num_units} * \text{num_units}$

The secondly you have new input of your sequence at each step.

$\text{input_weights} = \text{num_features} * \text{num_units}$

So now we have the weights, whats missing are the biases - for every unit one bias:

$\text{biases} = \text{num_units} * 1$

So finally we have the formula:

$(\text{num_features} + \text{num_units}) * \text{num_units} + \text{num_units} = (50 + 1) * 50 + 50 = 2600$

For Output layer:

$**\text{output_channel_number} * (\text{input_channel_number} + 1) = 2 * (50 + 1) = 102$

In [9]:



```
# model training

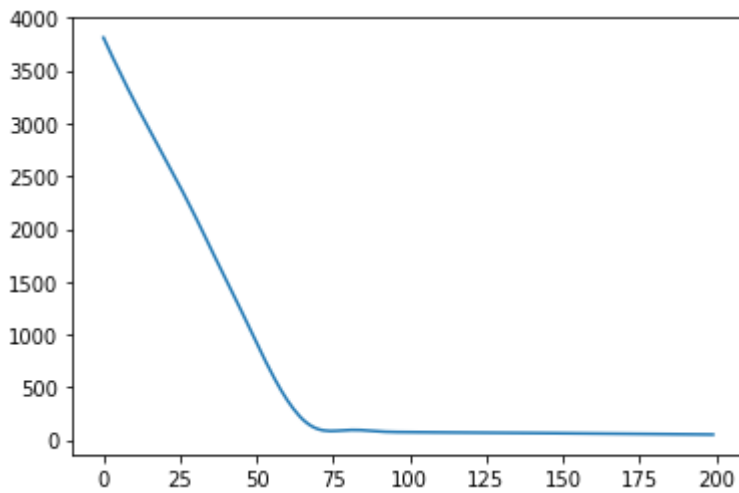
history = model.fit(X, Y, epochs= 200, verbose=0)
```

In [12]:

```
# plot training loss  
plt.plot(history.history["loss"])
```

Out[12]:

```
[<matplotlib.lines.Line2D at 0x7f57b9dcf2e8>]
```



In [11]:

```
# Test Prediction  
  
X_test = array([70, 80, 90])  
X_test = X_test.reshape((1, seq_len, 1))  
Y_test = model.predict(X_test)  
print(Y_test)
```

```
[[114.02077 127.01842]]
```

In [10]:

```
keras.backend.clear_session()
```