STUDYPE

**Predicting Financial Time Series using Deep Learning** 

Module 2. Recurrent Neural Network

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Note. This content mainly refers the summer session of KAIST organized by Jiyong Park(2018)



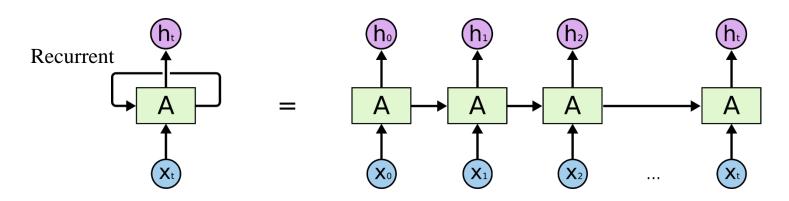
# **Recurrent Neural Network**



### **Review: What is RNN?**



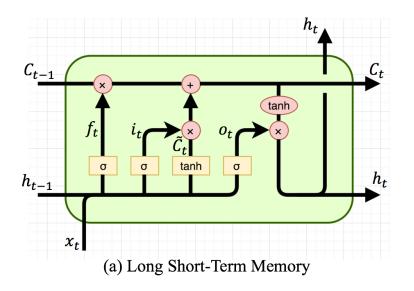
- Unlike CNN, RNN can use past information to learn the present task.
  - ➤ Example: Natural Language Processing (NLP)
    - "The clouds are in the ( )."
    - "I grew up in France ..... I speak fluent ( )."
  - ➤ Vanishing gradient problem
    - As that gap grows, RNN becomes unable to learn to connect the information. (the past information would be vanishing or exploding)

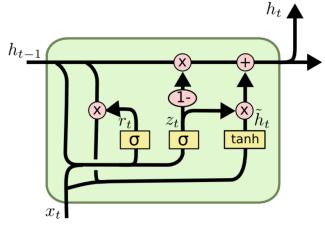


### **Review: What is RNN?**



• Long Short-Term Memory (LSTM) and Gated Recurrent Unit (GRU) allows RNN to learn how much past information would pass to the next.





(b) Gated Recurrent Unit

### **RNN** in Keras



 We can stack RNN, LSTM or GRU by passing a list of layer using Keras Sequential API

• Although RNN and GRU yield two states (output and hidden), LSTM yields three states (output, hidden, and cell)

```
from tensorflow.keras.layers import LSTM, GRU
encoder_outputs, h, c = LSTM(LATENT_DIM, name='encoder_lstm')(x)
encoder_outputs, h = GRU(LATENT_DIM, name='encoder_gru')(x)
```





# Hands-on-Labs Recurrent Neural Network







## **Recurrent Neural Network**

Lab4\_Recurrent Neural Network.ipynb

https://colab.research.google.com/drive/1RH5tnTrTPRZkWfNj7PAuPaM\_E55coxwp



# **RNN Input Shape**



- Regular Sequence Data Shape
  - (N, T, F)
  - N: Number of sequences
  - T: Number of time step for each sequence
  - F: Number of feature for each sample
- LSTM(32, input\_shape = (T, F))
  - Latent Dimension: 32
  - Input Shape: T, F (Number of Sequences will be inferred during training)

# **RNN Model for Image Classification**



- Let's "pretend" the image is a sequence of vectors
- Height as Time / Width as Features

#### **Features**

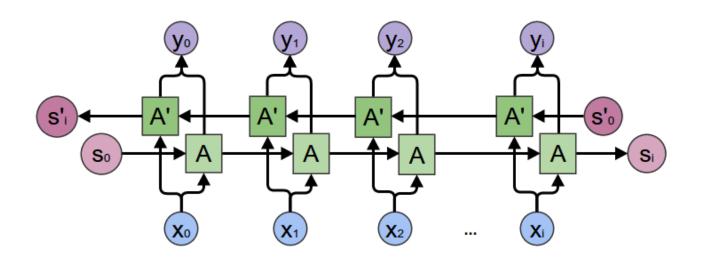
Time



### **Bidirectional RNN**



- Bidirectional Recurrent Neural Networks (BRNN) connects two hidden layers of opposite directions to the same output
- BRNNs were introduced to increase the amount of input information available to the network



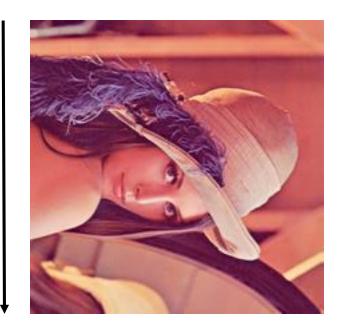
### **Bidirectional RNN**



- Bidirectional RNN can improve performance
- Apply Bidirectional RNN, Top -> Bottom and Bottom -> Top
- Also, we can rotate images so that we go in 4 directions!

Features Features

Time



Time



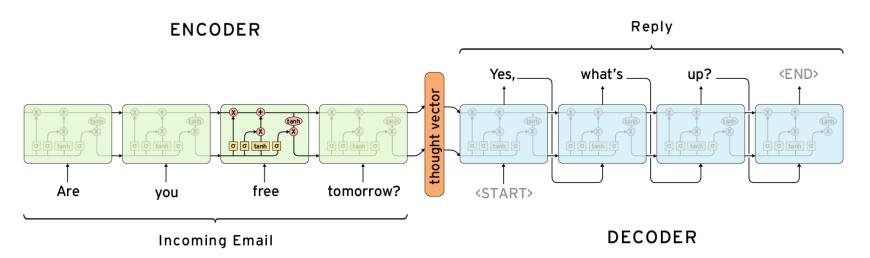
# Sequence-to-Sequence Model



## **Sequence-to-Sequence Model**



- Our understanding of contexts
  - Answer on Question / Translation etc
  - ightharpoonup (Listening/Reading)  $\rightarrow$  (Understanding)  $\rightarrow$  (Speaking/Writing)
- Understanding of sequence-to-sequence models
  - $\triangleright$  (Encoding)  $\rightarrow$  (Thought Vector)  $\rightarrow$  (Decoding)



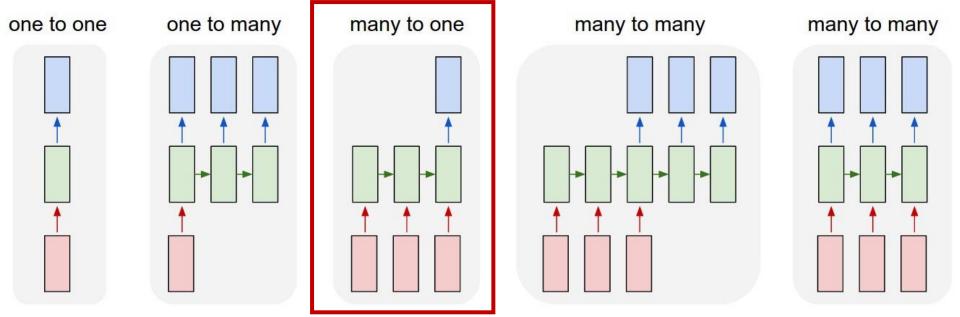
# **Seq2Seq Model for Time Series Prediction**



• For time series prediction, we can apply seq2seq model

• Input: x1, x2, ... xN

• Output: xN+1



# Thank you ©

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### **References**

• Jiyong Park (2018), KAIST Summer Session, Retrieved from <a href="https://sites.google.com/view/kaist-mis-session2018/overview?authuser=0">https://sites.google.com/view/kaist-mis-session2018/overview?authuser=0</a>