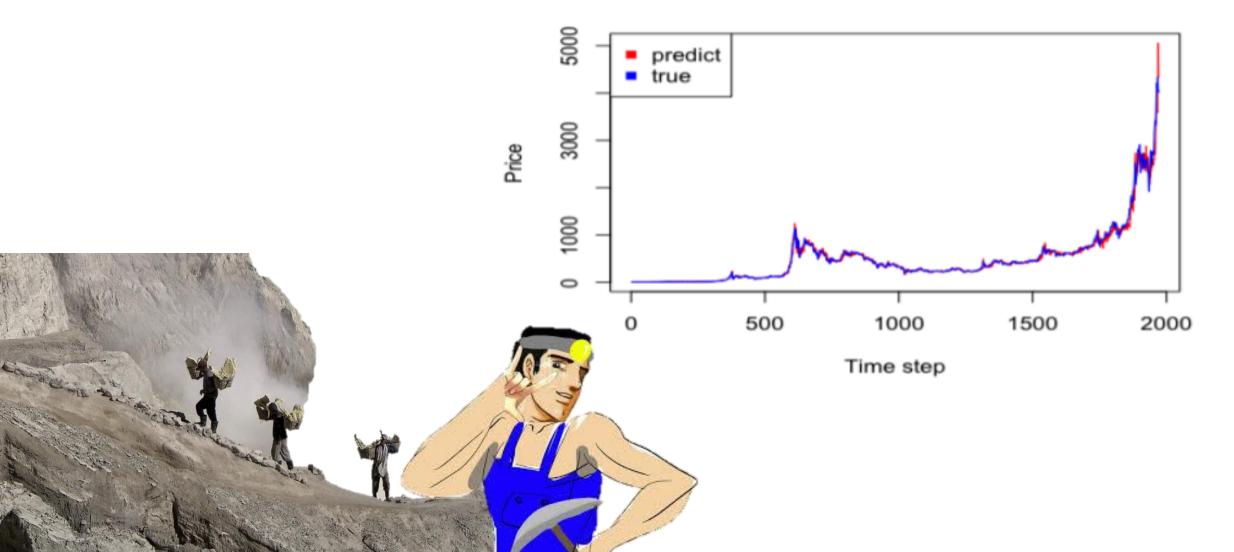
# Predicting Bitcoin Prices by using Rolling Window LSTM model

統研所碩一 張泳樺

# 研究背景

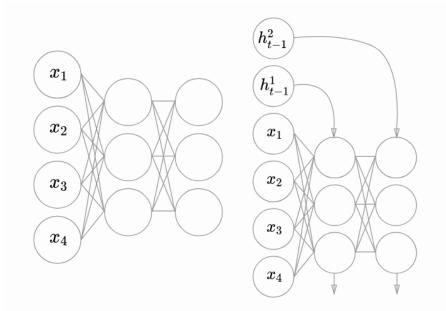
## USD 5 to **USD 4000**



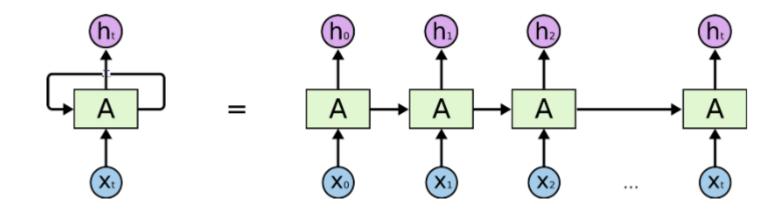
# 分析方法-RNN

## 處理時間序列資料

- ➤ long-term dependencies
- ▶ gradients vanishing
- >LSTM



The overall structure of RNNs is very similar to that of feedforward networks.



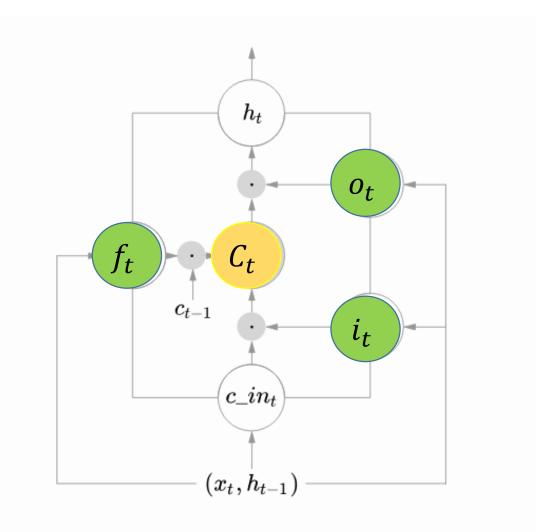
## 分析方法-LSTM

### Gate

- 1) input gate
- 2) forget gate
- 3) output gate

## Memory ceil:

$$c_{-}in_t = tanh(W_{xc}x_t + W_{hc}h_{t-1} + b_{c_{-}in})$$



## Rolling window

將某個點的取值擴大到包含這個點的一段區間,用區間來進行判斷,此區間即窗口。

如:有100個觀測,用10個觀測

作為一個rolling window。

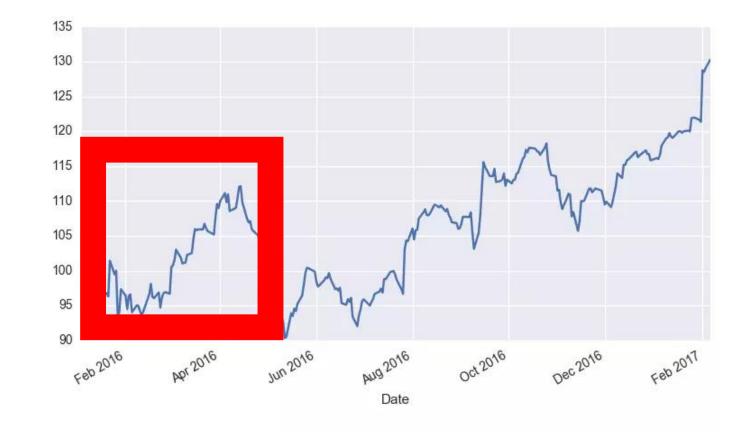
$$\widehat{y^{11}} = F(x^1, x^2 \dots x^{10})$$

$$\widehat{y^{12}}$$
=F( $x^2$ ,  $x^3$ ....  $x^{11}$ )

.

.

•



# 文獻回顧

## **Bootstrap & fixed-size rolling window**

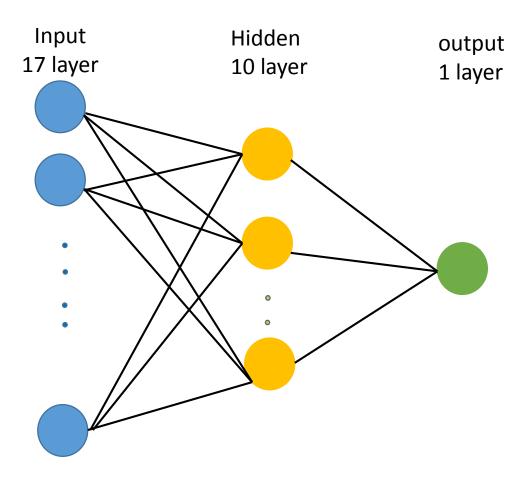
➤ Difficulty in building a test set for model selection

## Rolling window with regression model & Bayesian neural network

➤ Cannot reflect sequential characteristic in the model

## Model setting-(many-to-one)

T = t



**Table 2: common hyperparameter setting** 

Hyperparameter	
output dimension of hidden layer	10
activation function of hidden layer	hyperbolic tangent
loss function	sum of the squares
optimizer	Adam
learning rate	0.01

# 參數介紹

N: # of sample data

Hyperparameter tuning

P: sequence of time

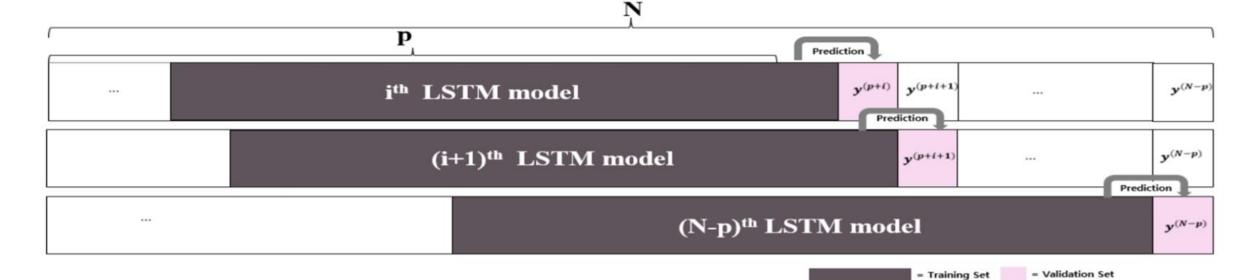
q: the  $i^{th}LSTM's$  training length(time-step)

N-p: models that predict N-P time series, and these model don't share weights

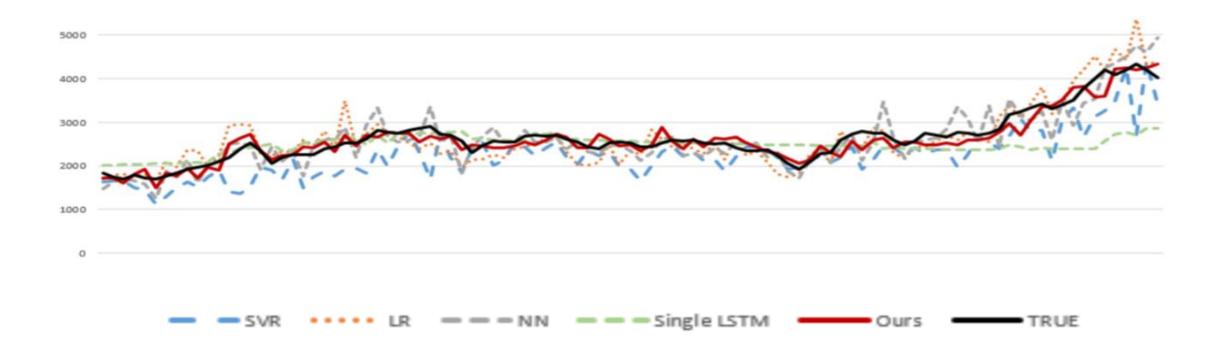
## Model

$$\hat{y}^{j+i+q} = F_i(x^{i+j-1} \dots x^{i+j+q-1}; \theta^i)$$

$$\theta^{i*} = \underset{\theta}{\operatorname{argmin}} \sum_{j=1}^{p-q} (\hat{y}^{j+i} - y^{j+i})^2$$



## conclusion





## validation data

Common Maching learning:

➤ Dividing into train VS test

Rolling window LSTM model:

➤ Building a model structure

# 資料介紹

- 1. In particular, we thought that the degree of change of the blockchain variable had a significant effect on the target variable
- 2. The bitcoin price is the weighted sum of the daily open price, closing price, lowest price, and highest price of the bitstamp exchange

# 資料介紹

#### time:

as daily data from September 2011 to August 2017

## Input:

#### **Global Economic Measure:**

1 prices of crude oil, 2 SSE, 3 gold, 4 VIX, 5 FTSE100

#### **Global Currency**:

**6** USD/CNY, **7** USD/JPY, and **8** USD/CHF

#### **Blockchain Information:**

9block size, 10median confirmation time, 11hash rate, 12 miner's revenue, 13cost per transaction, 14 confirmed transactions per day, 15 the number of transactions, excluding popular addresses

#### Trading Information:

16 The trading volume of the bitstamp exchange \ 17 the historical value of the target variable

## Hyperparameter tuning

(sequence length, training length):
(50, 25), (100, 50), (150,100), (200,150), (500,250), (1000,500), (1000,750), (1500,500), (1500,750), (1500,1000)

# 資料整理