

Challenges

# Machine learning on Time Series Analysis

Solutions

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Investing

Gambling

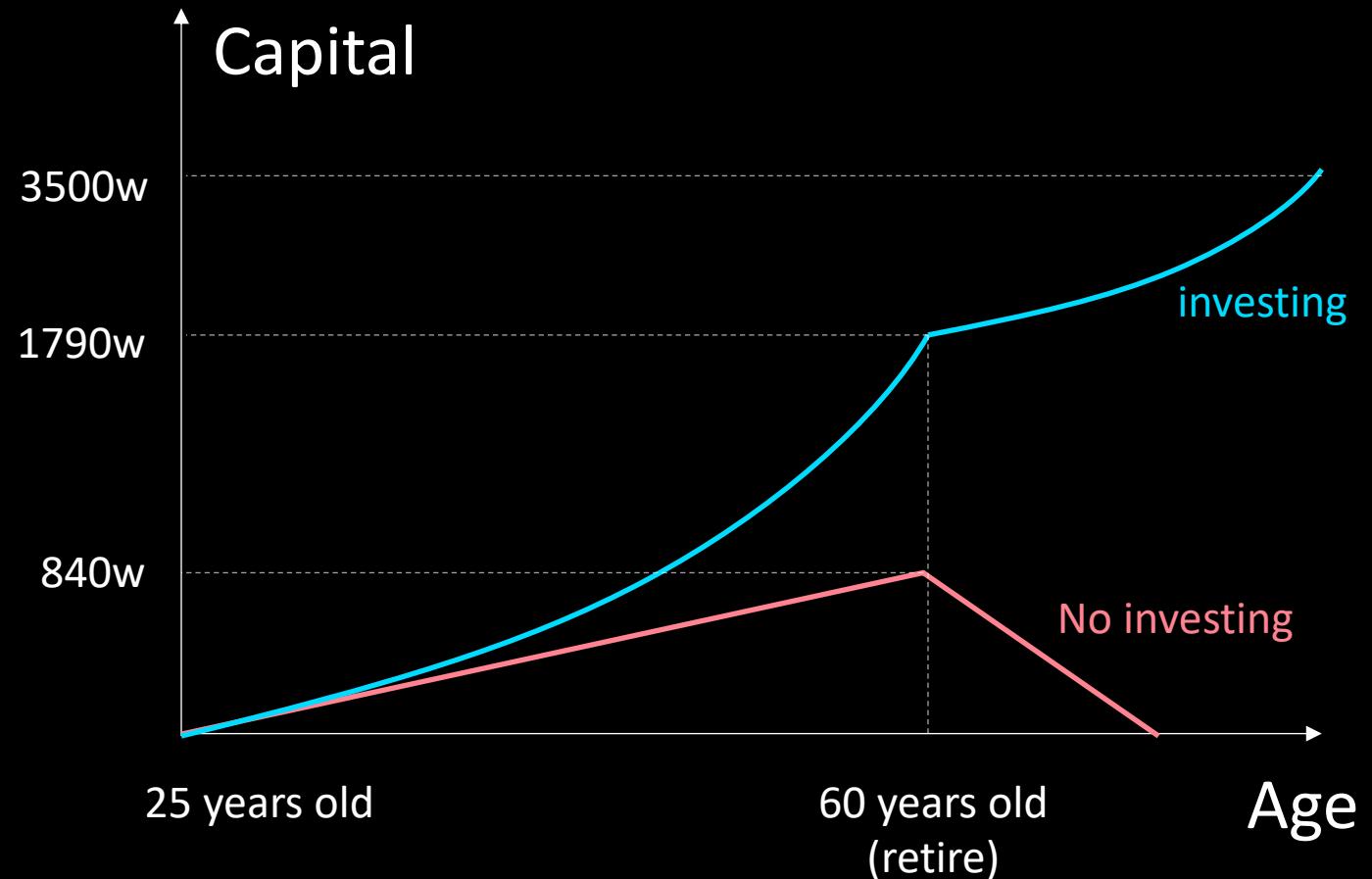


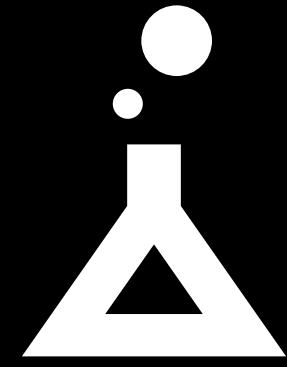


# ■ Personal Life Financial Plan

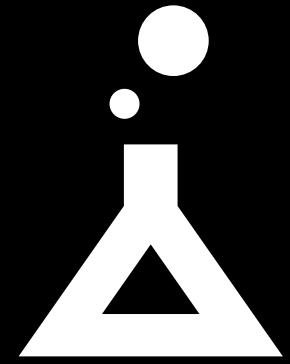
## Capital simulation

Income per month.....5w (NTD)
Spent per month.....3w (NTD)
Investment.....70% (of total capital)
Compound interest....5% (each year)





FinLab

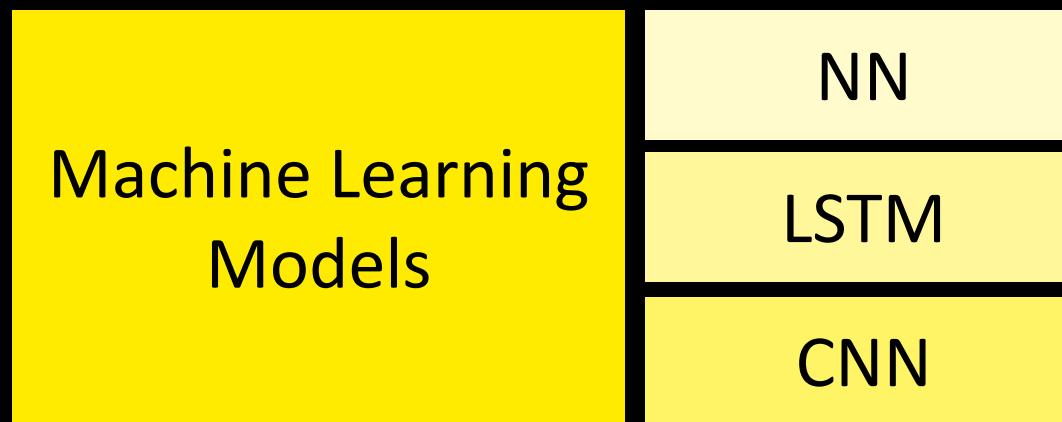


FinLab



<https://www.finlab.tw/mopcon.html>

# ■ Outline

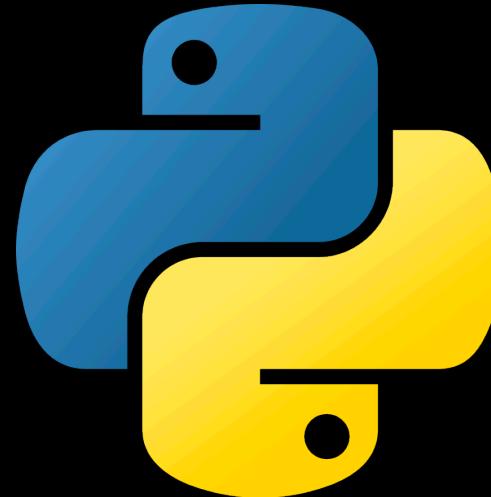


# ■ Trading programming language



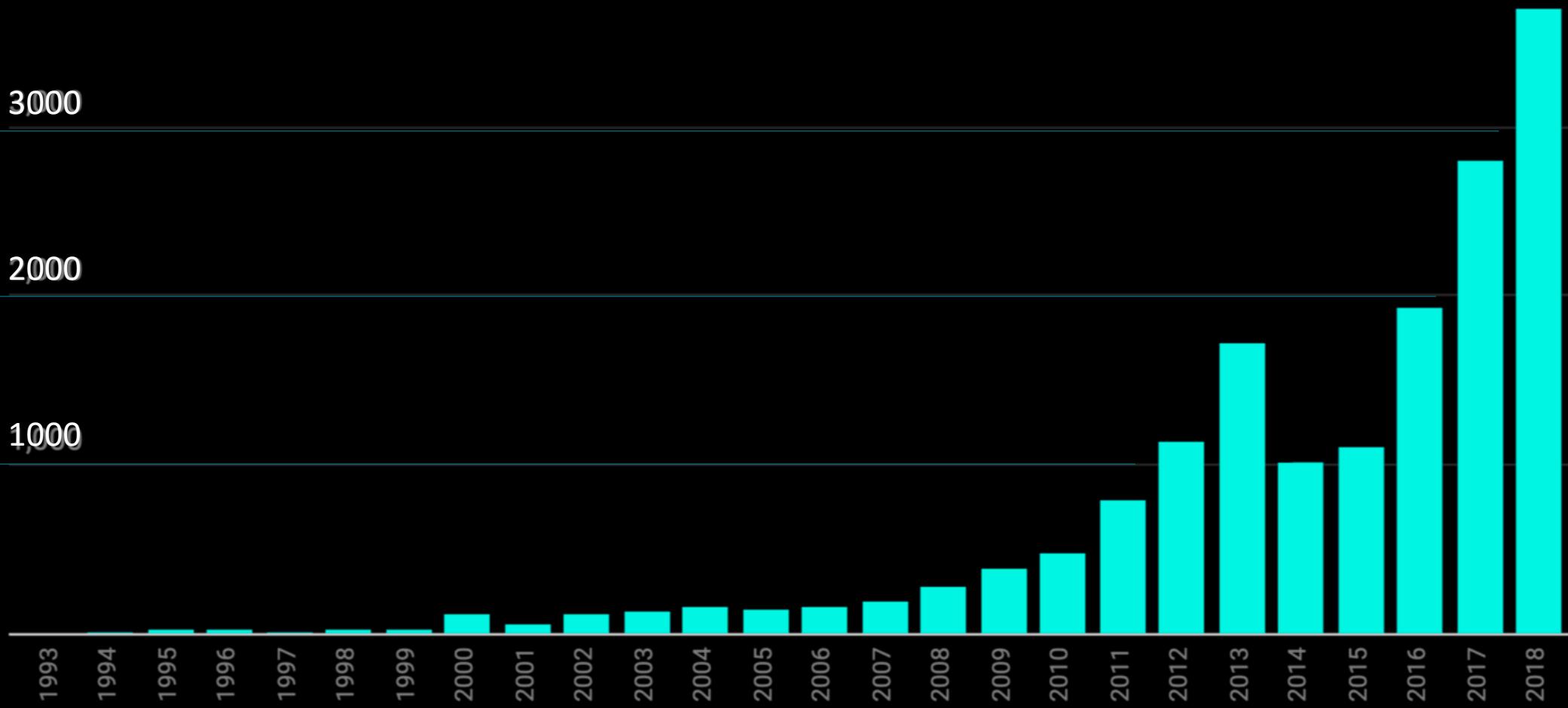
- Easy learning curve for the beginners
- Integrated with language editor in platforms
- Can be extend by external DLL
- Most of the functions are encrypted or the source code is not provided
- Does not support statistic analysis or machine learning toolkit

# ■ Trading programming language



- Friendly statistic toolkit
- Friendly statistic toolkit
- Strong community and widely applied
- Easy to deploy (Flask/Django/...)
- More innovative data science applications

# ■ Artificial Intelligence papers



All of the papers available in the “artificial intelligence” section (arXiv)

## ■ ML algorithms in finance?



# ■ Supervised Machine Learning

Training



Testing





# Financial Data (Features)

# ■ Financial Data Structures

## Fundamental data

Focusing on creating a portrait of a company

- Useful to combine other data types
- Difficult to confirm data release date
- Missing data is often backfilled
- Consider multiple correction

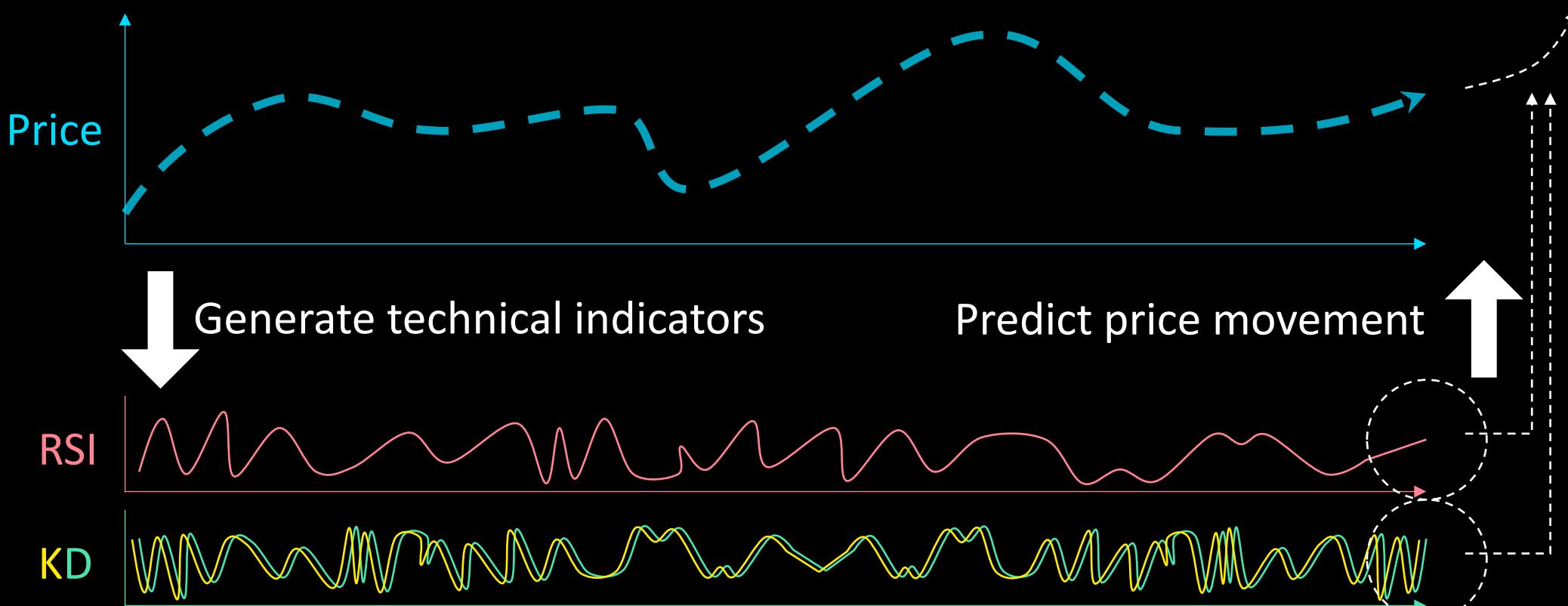
## Trading data

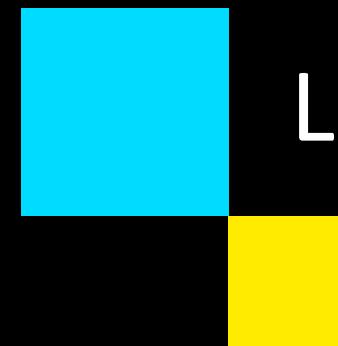
Market participant characteristic footprint  
Trading book, price, broker trading summary...etc

- Data often with timestamp
- Generate extra features (ex: technical indicators)
- Massive amount of data generated in one day
- Some of the data is difficult to obtain

# Creating Technical indicators

Price historical data





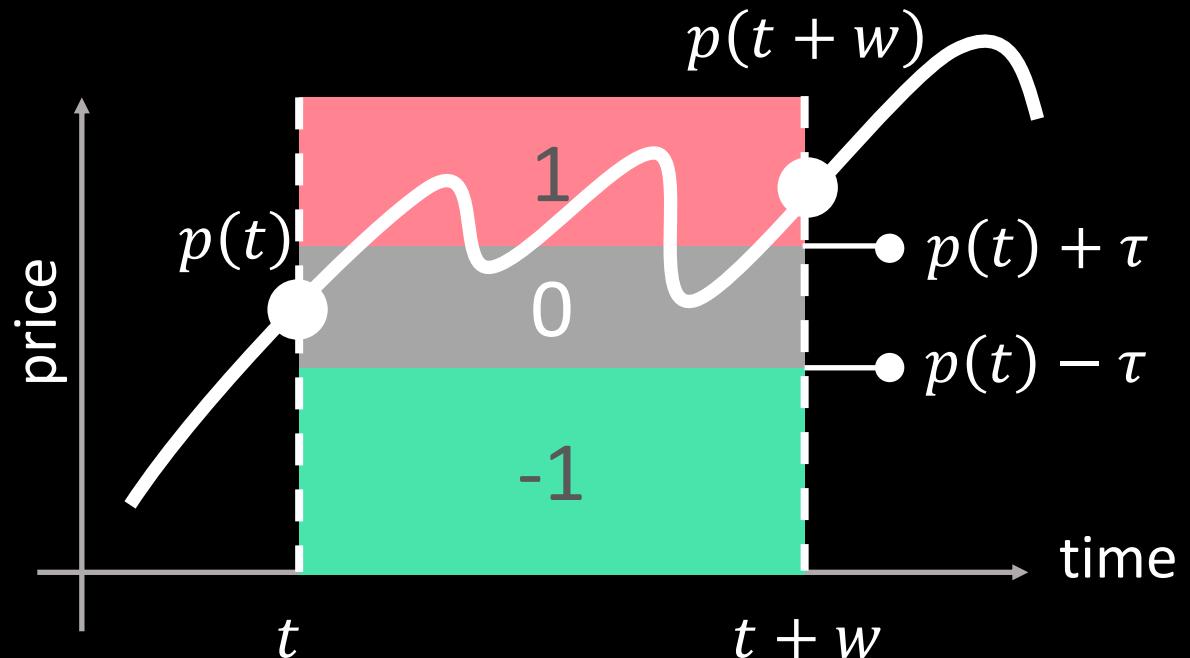
Labeling

# ■ Challenging of Labeling the data

## Fixed time horizon

A popular method in the literature

- $\tau$  is a constant
- Do not have stop-loss limits



# ■ Label Generation Methods

- Triple barrier [Prado 2018]
- Continuous trading signals [Dash 2016]
- Trading Point decision [Chang 2009]

[Prado 2018] Advances in Financial Machine Learning

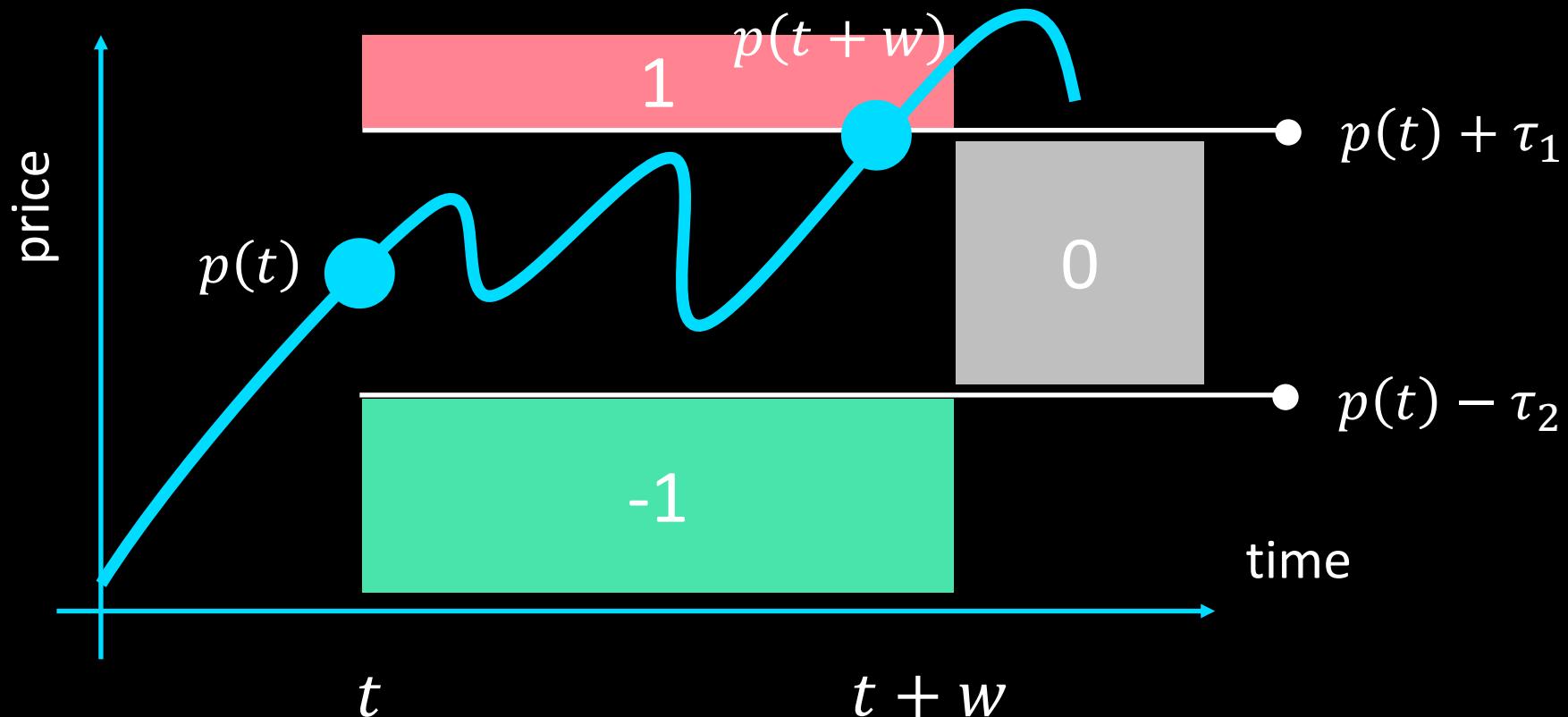
[Tsantekidis 2017] Using Deep Learning to Detect Price Change Indications in Financial Markets

[Dash 2016] A hybrid stock trading framework integrating technical analysis with machine learning techniques

[Chang 2009] Integrating a Piecewise Linear Representation Method and a Neural Network Model for Stock Trading Points Prediction

## ■ Triple barriers [Prado 2018]

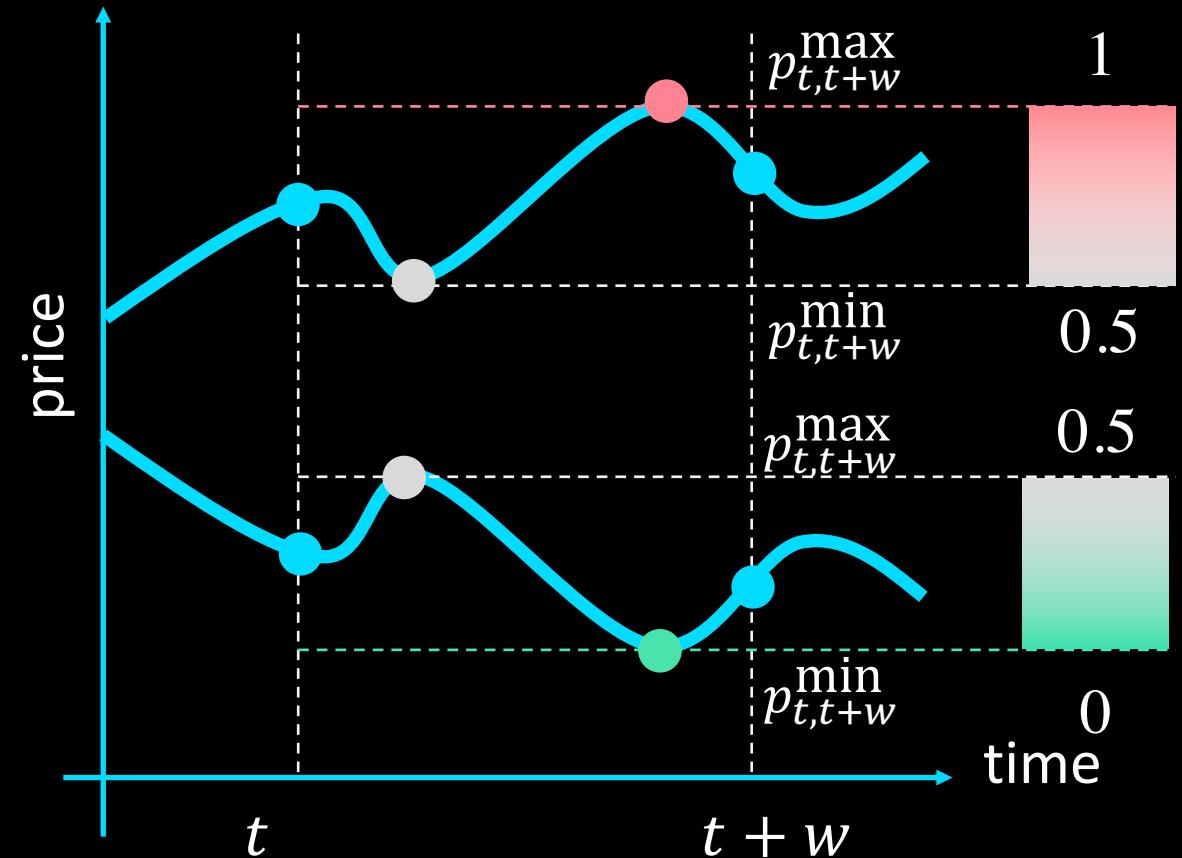
- Horizontal barriers are defined by profit-taking and stop-loss limit
- $\tau_1$  and  $\tau_2$  are dynamic according to estimated volatility



# ■ Continuous trading signals [Dash 2016]

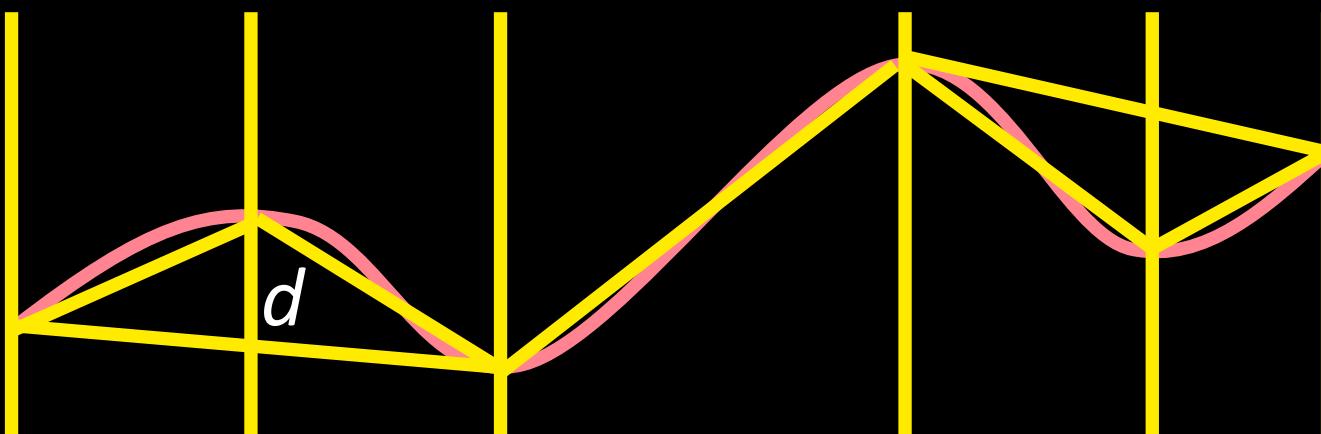
- Using momentum of the stock price
- $y(t)$ 's are continuous
- Provides more detailed information

$$y(t) = \begin{cases} \frac{p_{t+w} - p_{t,t+w}^{\min}}{p_{t,t+w}^{\max} - p_{t,t+w}^{\min}} & \text{if } p_{t+w} > p_t \\ 0.5(1 - \frac{p_{t+w} - p_{t,t+w}^{\min}}{p_{t,t+w}^{\max} - p_{t,t+w}^{\min}}) & \text{else} \end{cases}$$



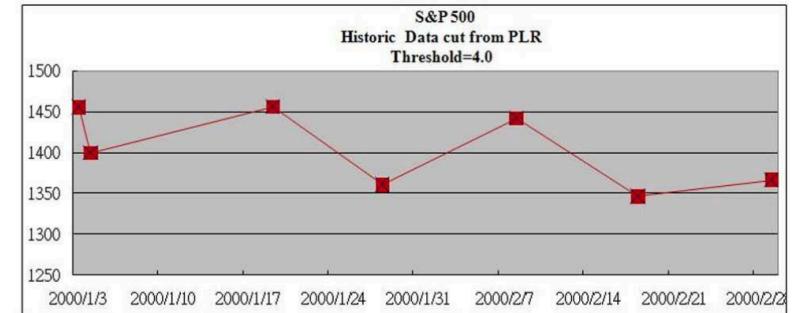
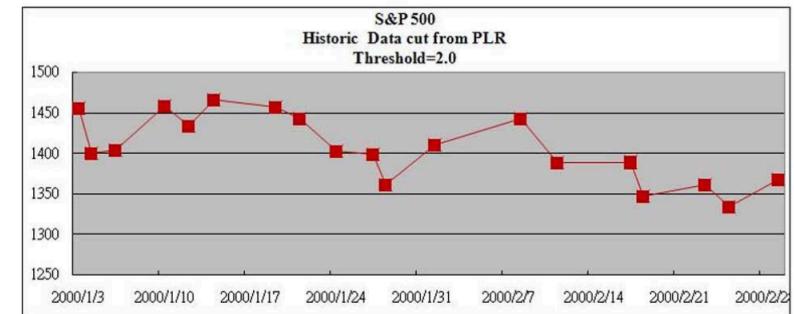
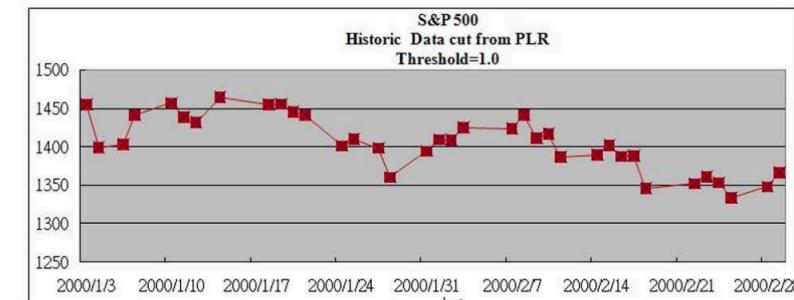
## ■ Trading point decision

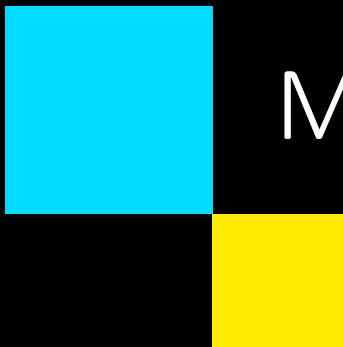
- Find the local minimum and maximum points
- Divide the time series into subsegments
- Threshold value  $d \rightarrow$  length of trend



# ■ Trading point decision

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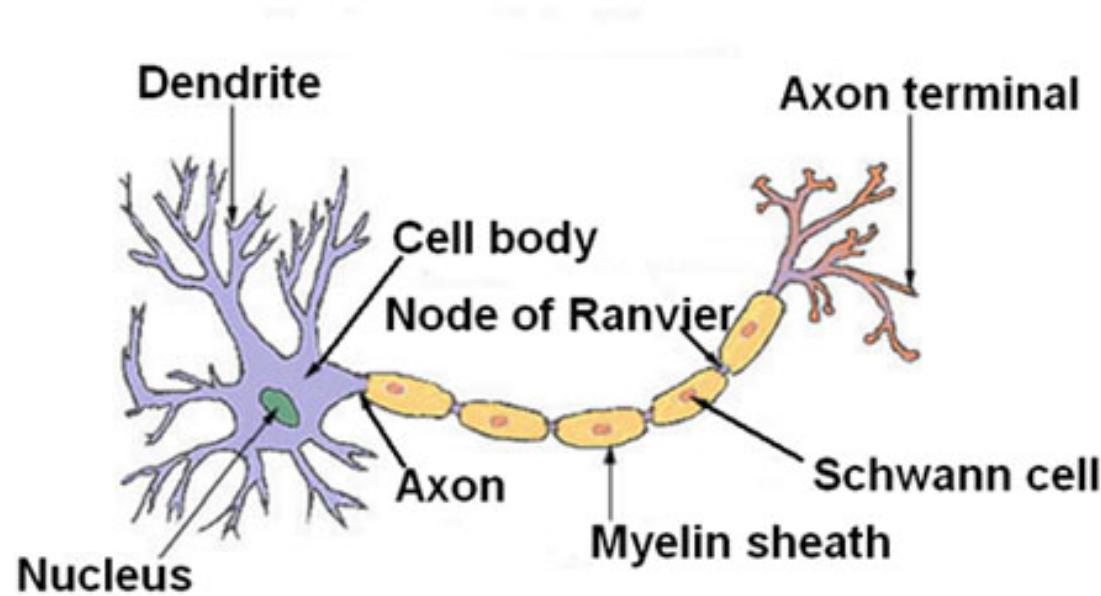


ML Models

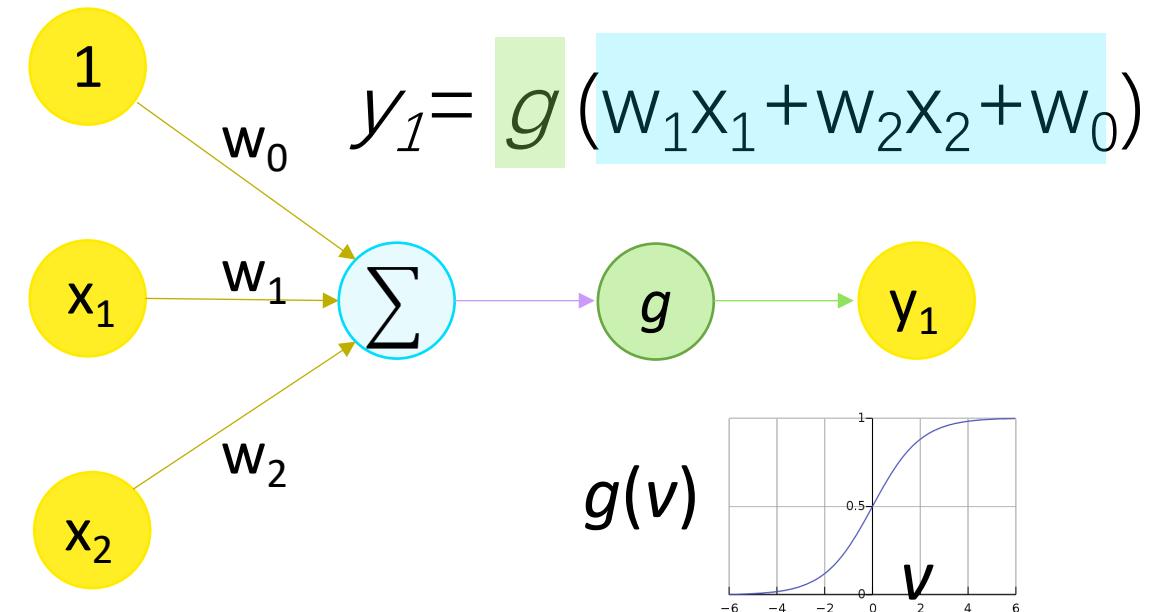
# ■ Neural Network

- Built to model the human brain
- interpret numeric data through a kind of machine perception

Human neuron structure

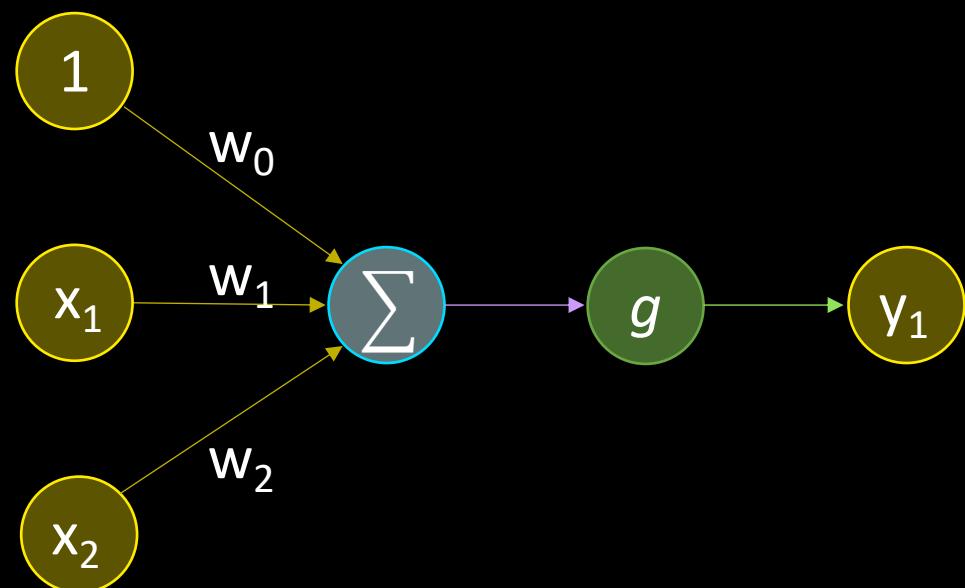


Single neuron model



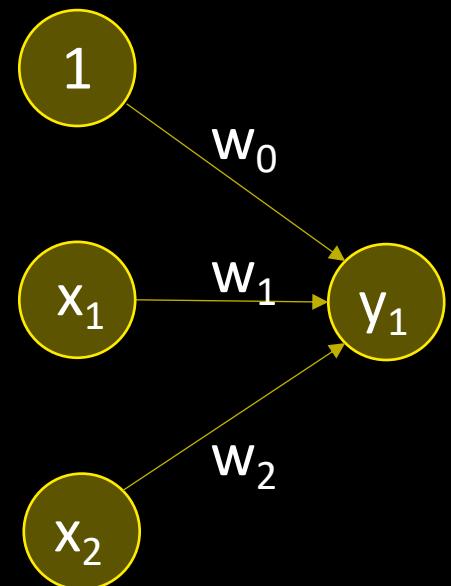
# ■ Neural Network

Single node in neural network



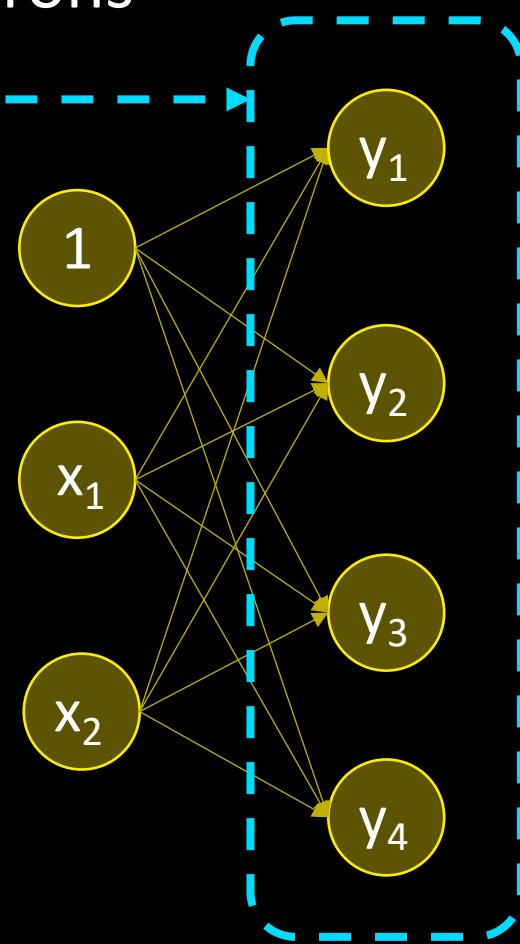
# ■ Neural Network

Simplified expression



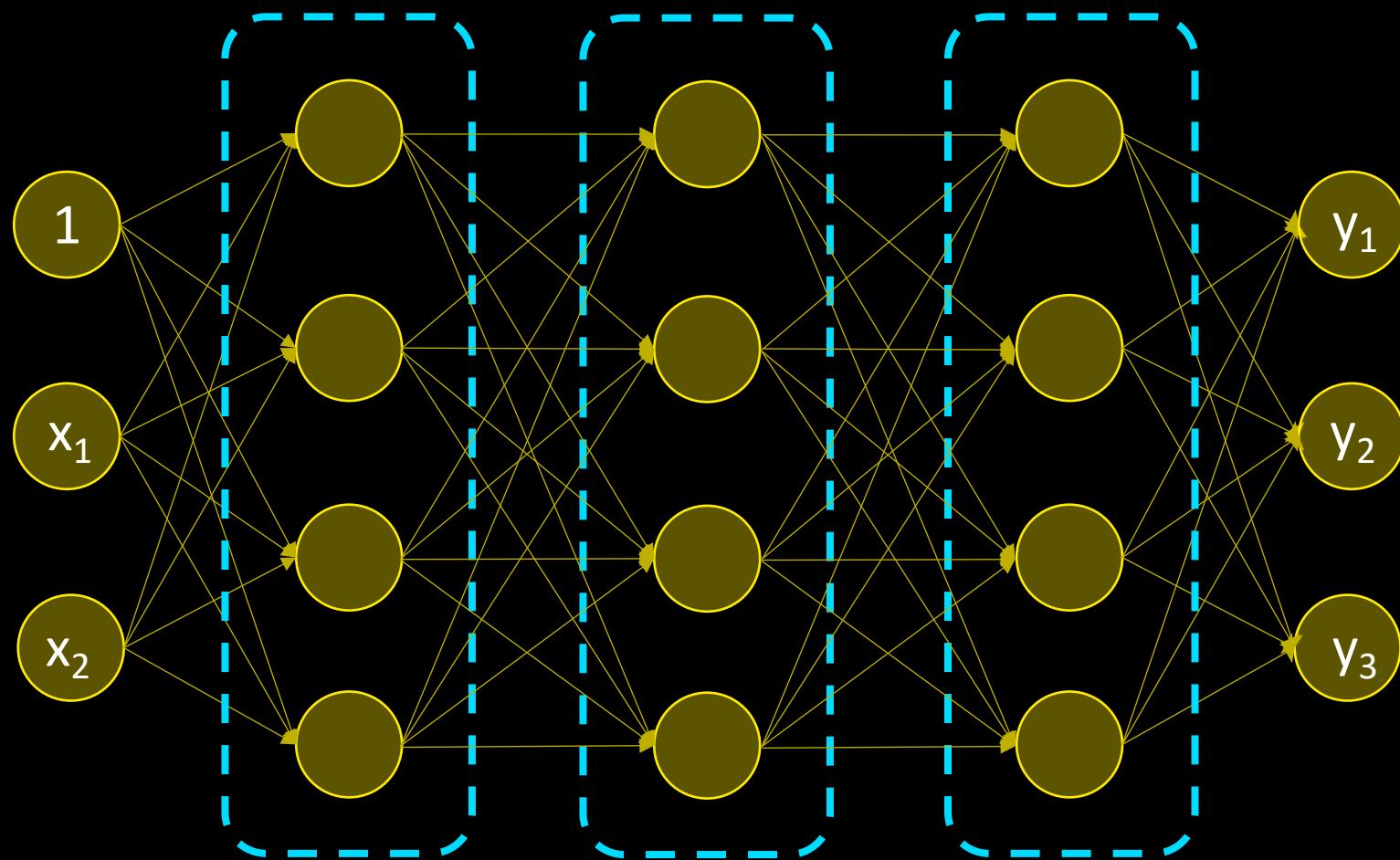
# ■ Neural Network

A layer contain multiple neurons



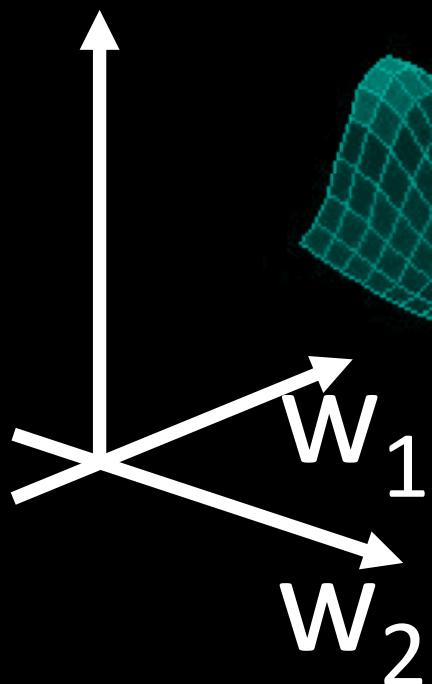
# ■ Deep Neural Network

Multi-layer deep neural network



# ■ Neural Network Optimization

Cost function



# ■ Deep Neural Network Training Result

Asset

Taiwan Capitalization  
Weighted Stock Index

Data split

Train	Validate	Backtest
2006 ~ 2014	2015	2016 ~ 2019-3-1

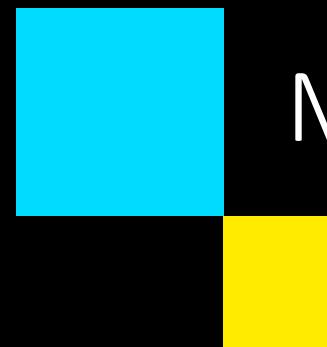
Features

Scaled Technical Indicators

Labels

Fixed time horizon

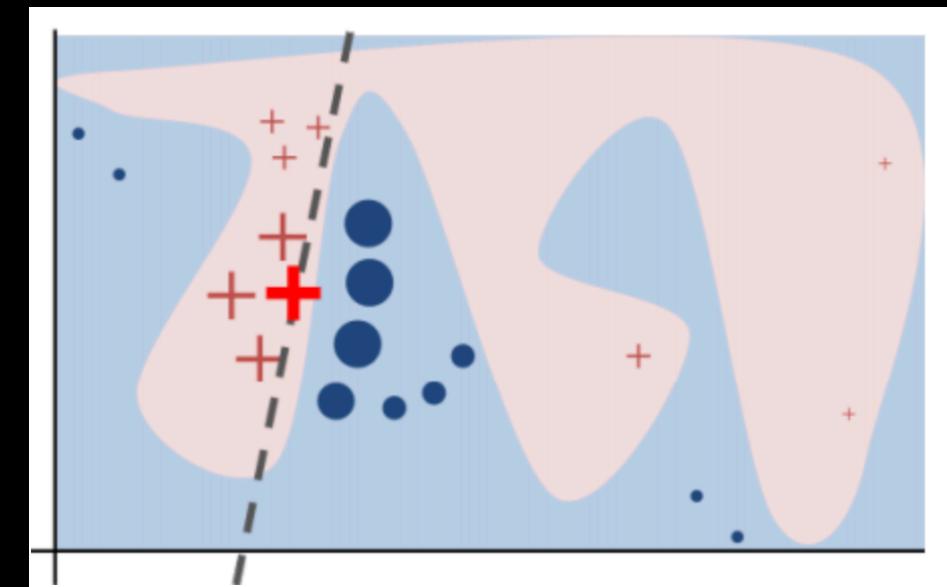
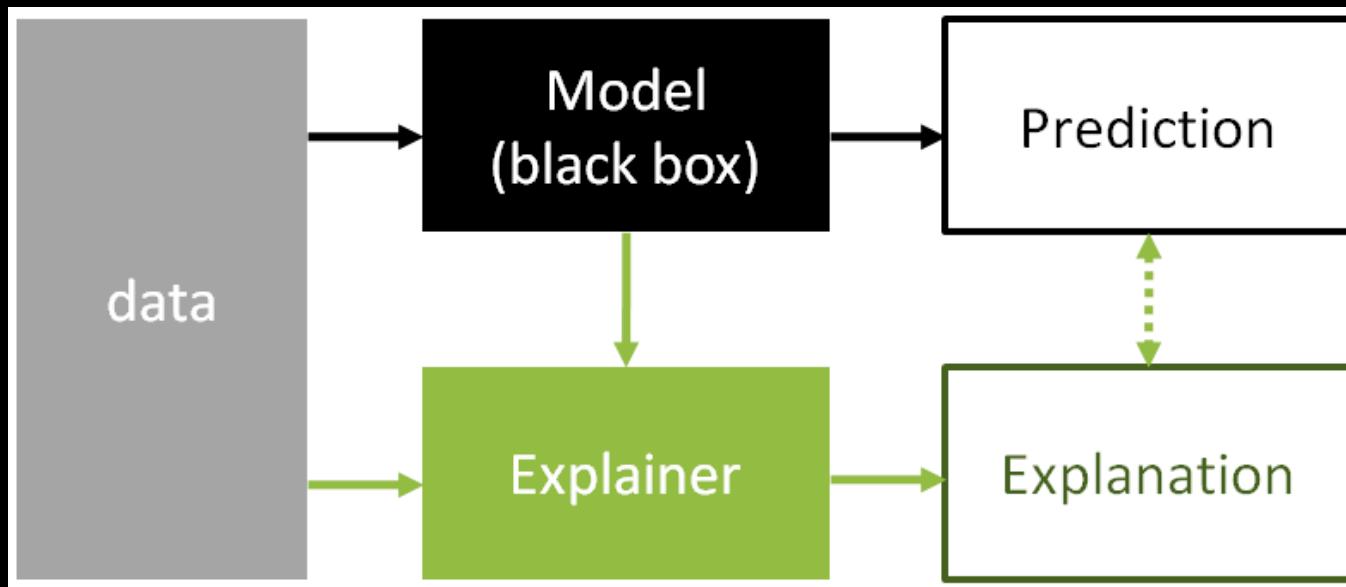




# Model Interpretation

# ■ Deep Neural Network Training Result

- Explain the predictions of your machine learning models
- Approximate the predictions of the underlying black box model
- Focuses on training local surrogate models to explain individual predictions



<https://arxiv.org/pdf/1602.04938.pdf>



## ■ Backtest

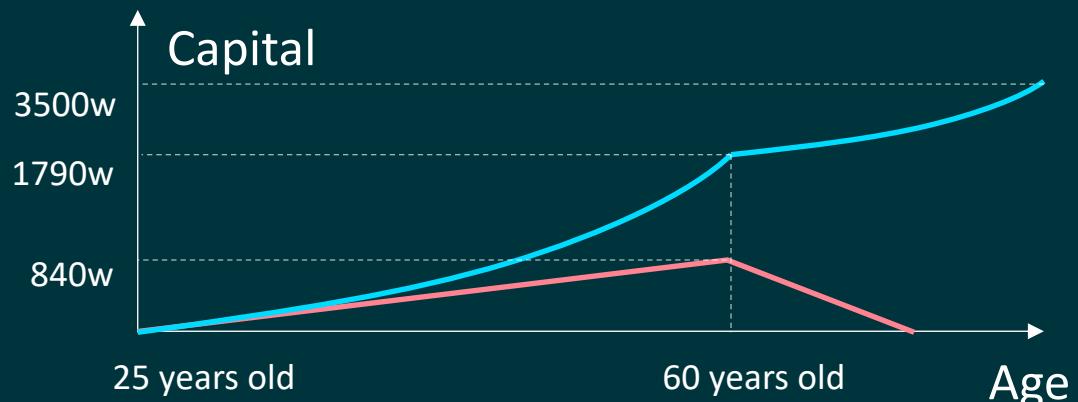
- Survivor bias, lookahead bias, transection cost, outlier, overfitting
- Finding the lottery tickets that won the last game
- Solutions
  - Develop model for entire asset or classes
  - Use Bootstrap aggregating
  - Record every backtest conducted
  - Resist the temptation of reusing a failed strategy

# Conclusion

## Investing

Investing or gambling ?

Life Plan



# Machine Learning

Financial Data

Features

Machine Learning  
Models

NN

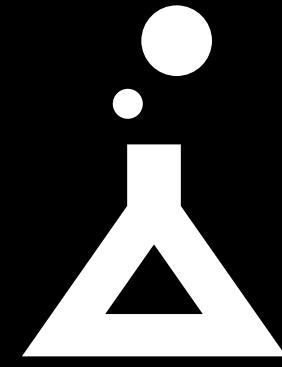
LSTM

CNN

Evaluation

Backtesting

Purged Validation



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