



High Frequency Trading

About SANOU

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Table of Contents

|01

Overview HFT

|02

Architecture Overview

|03

RL in for Agent

HFT Overview!

Some basics

- An exchange is a marketplace where securities, commodities, derivatives and other financial instruments are traded. [1]
- Market Data : The generally adopted stock market data is the sequence at regular intervals of time such as the price open, close, high, low, and volume
- How securities price is defined : **Law of supply and demand**
- A stock exchange is used to raise capital for companies seeking to grow and expand their operations. The first sale of stock by a private company to the public is referred to as an initial public offering (IPO).
- Portfolio : A portfolio is a **collection of financial investments** like stocks, bonds, commodities, cash, and cash equivalents, as well as their fund counterparts.
- **Modern Portfolio Theory(Markowitz's theory)** is a method that can be used by risk-averse investors to construct diversified portfolios that maximize their returns without unacceptable levels of risk.

What is HFT ??

- Trading of equities, options, futures at high speed in large volumes
- Earning money by exploiting the fleeting variation in stock price or demand

HFT accounts

- 70% of all trades in US Markets[1]
- And it continues to grow

HFT involves

- Using computers to place orders based on “predefined” algorithms

Main challenges of HFT

- Scalability
- Latency
 - Execute orders faster than other investors to capture fleeting variations in price and demand in the markets
- Throughput
 - Handle large volume of orders
- Flexibility
 - Adapt to changing risks and trading strategies

Recent Problems in HFT

- Knight Capital bug[2]:
 - Test script executed live trades
 - \$450M loss in 45 minutes
- Nasdaq - Facebook IPO[3]
 - Order confirmations delayed
 - \$62M loss in direct damages
- BATS Failure
 - Software bug in order auctions

[2] <https://www.bugsnag.com/blog/bug-day-460m-loss>

[3]: <https://www.sec.gov/news/press-release/2013-2013-95.htm>

More precisely

HFT tries to :

- Buy low
- Sell High
- In high speed context
- And high scalability context

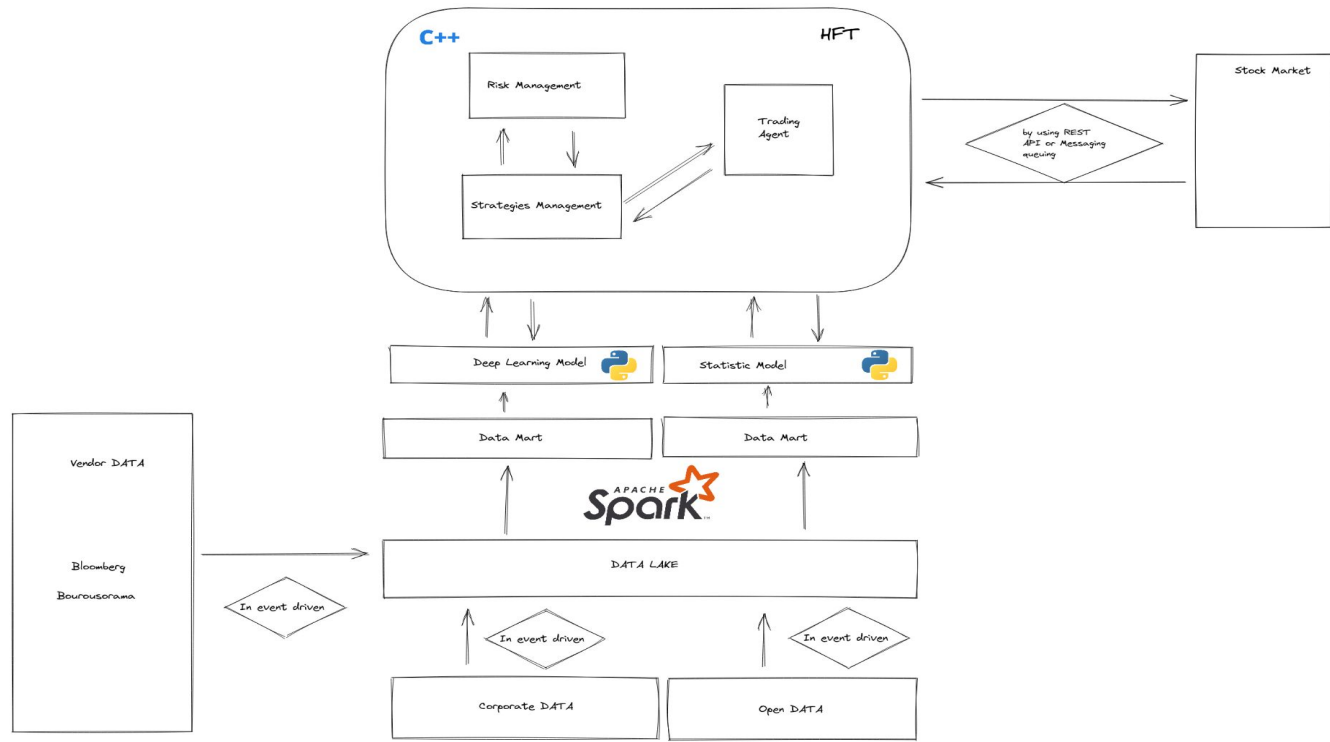
Archi

Overview!

Protocol to interact with exchange

- Generally by rest api
- Also possible by pushing event in topic(event streaming)
- Some stock were created its own protocol on top of http
Like Nasdaq : Nasdaq TotalView-ITCH[3]


Architecture



kafka



RedPanda for Streaming



Take care about Chinese Wall

As we can remark, A whole component for HFT, will be built in C++;

- Memory Management
- Scalability

Heterogeneity of data

The highest-frequency data is a collection of sequential “ticks,” arrivals of the latest quote, trade, price, and volume information. Tick data usually has the following properties:

A timestamp, A financial security identification code, An indicator of what information it carries: bid price, ask price, available bid volume, available ask volume, last trade price, last trade size

Option-specific data, such as implied volatility The market value information, such as the actual numerical value of the price, available volume, or size.

Risk Management Component

A risk management strategy is an important tool for traders. It helps to identify and manage risk, set limits on position sizes, and set up solid trading opportunities. A good risk management strategy can help to protect capital, prevent losses, and maximise profits.

There are many modelling for the risk :

- Value at Risk
- Stochastic modelling
- Deep Risk Model[4]

Strategy Component

- A trading Strategy is a set of rules to apply;
a trading strategy is a fixed plan that is designed to achieve a profitable return by going long or short in markets.
- The aim of this component will be to generate according to the context of market and risk factor of market, strategies to apply in order to achieve a return.

Trading Agent Component

- The trading agent follow **PEAS**[5] paradigm.

PEAS For Performance Measure, Environment, Actuator, Sensor

This Paradigm is more sustainable because it can provide rational agent

This model allows you to keep the strategies and risk component separated

- We can use Reinforcement Learning for the Trading Agent.
 - More difficult to separate a unit of domain (Risk and Strategy)
 - Less use in production, problem a explaining.
 -

Formalize the Trading World in PEAS approach

- PERFORMANCE MEASUREMENT : PnL (Profit and Loss)
- Environment : Stochastic
- Actuator : Be able to buy/ SELL or Hold
- Sensor : tools to get market parameter or knowledge from latent space

Formalize the Trading Agent in (Deep) Reinforcement Learning approach

- “State” are market parameters
- Actions: buy/sell/hold
- Reward : PnL
- Goal : Maximise reward

Some important notions in (D)RL

- Principe According to the **Bellman Equation**, long-term-reward in a **given action is equal to the reward from the current action combined with the expected reward from the future actions** taken at the following time.
- Policy function π : maps state to action, and policy optimization is to find an optimal mapping.
- Value function V : estimates *how good* it is for the agent to be in a given state.
Other version of this function more subntable in trading context estimates how good it is to perform a given action in a given state.

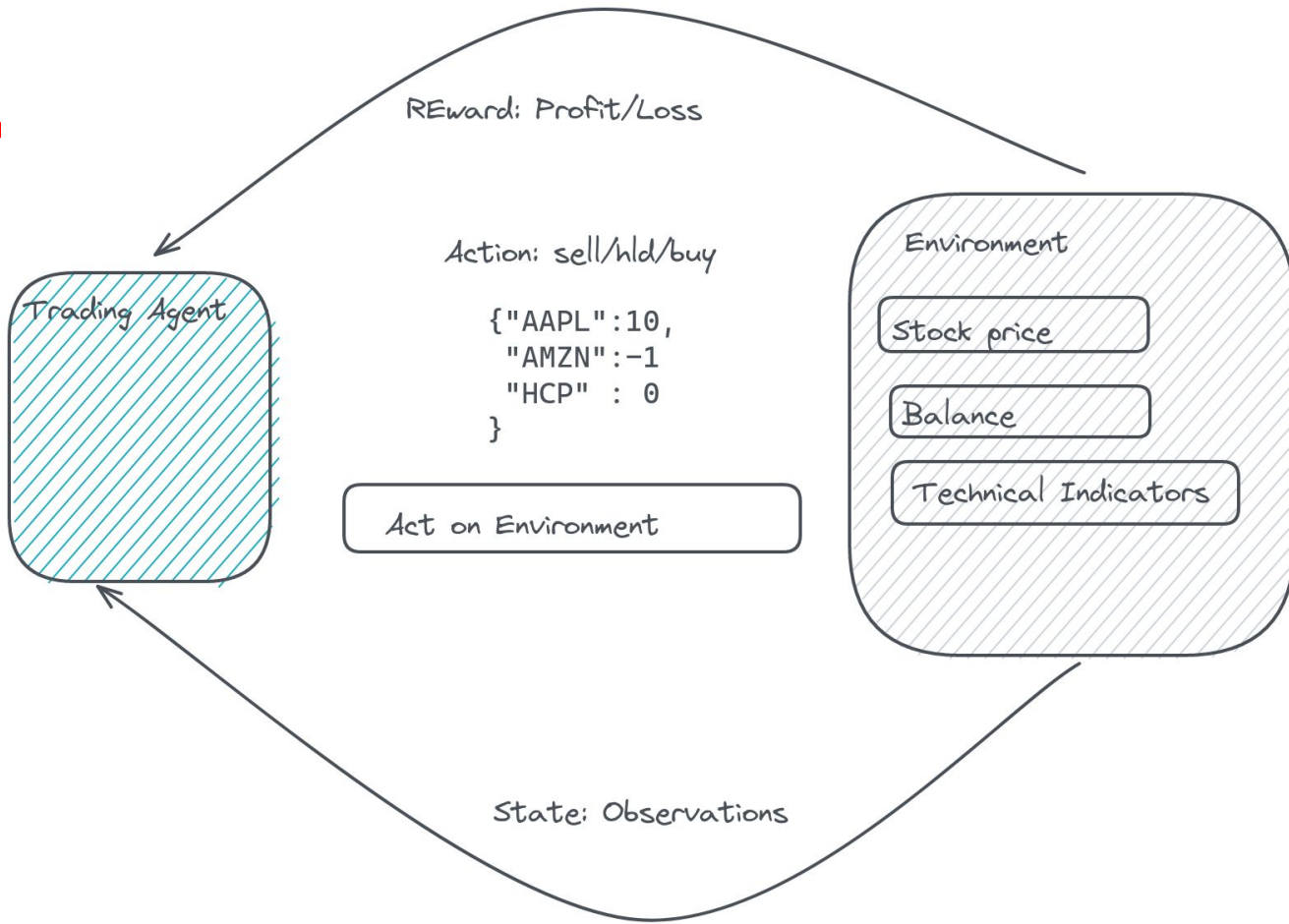
In Trading Context, The Value Functions can be estimated from experience.

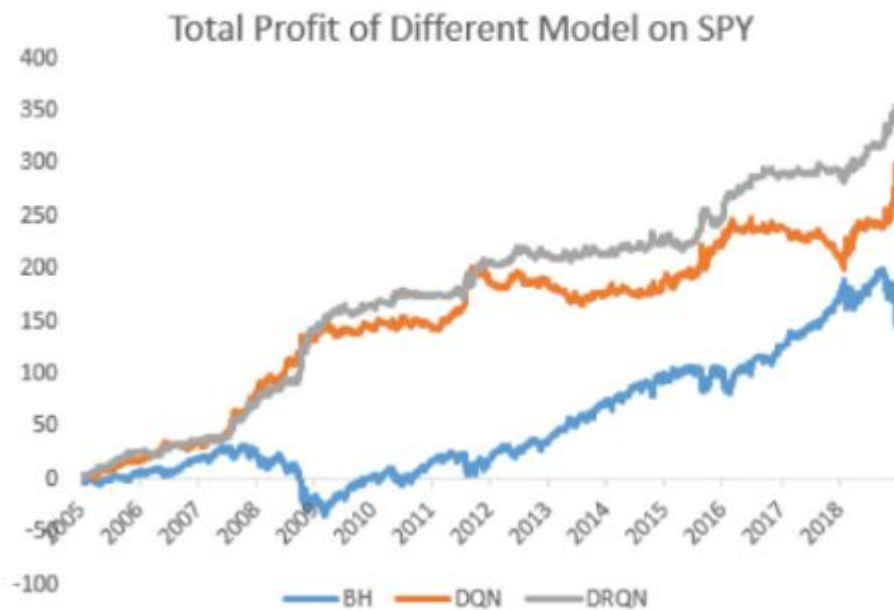
Some important notions in (D)RL

- Episode : one a sequence of states, actions and rewards, which ends with terminal state.

The reward function

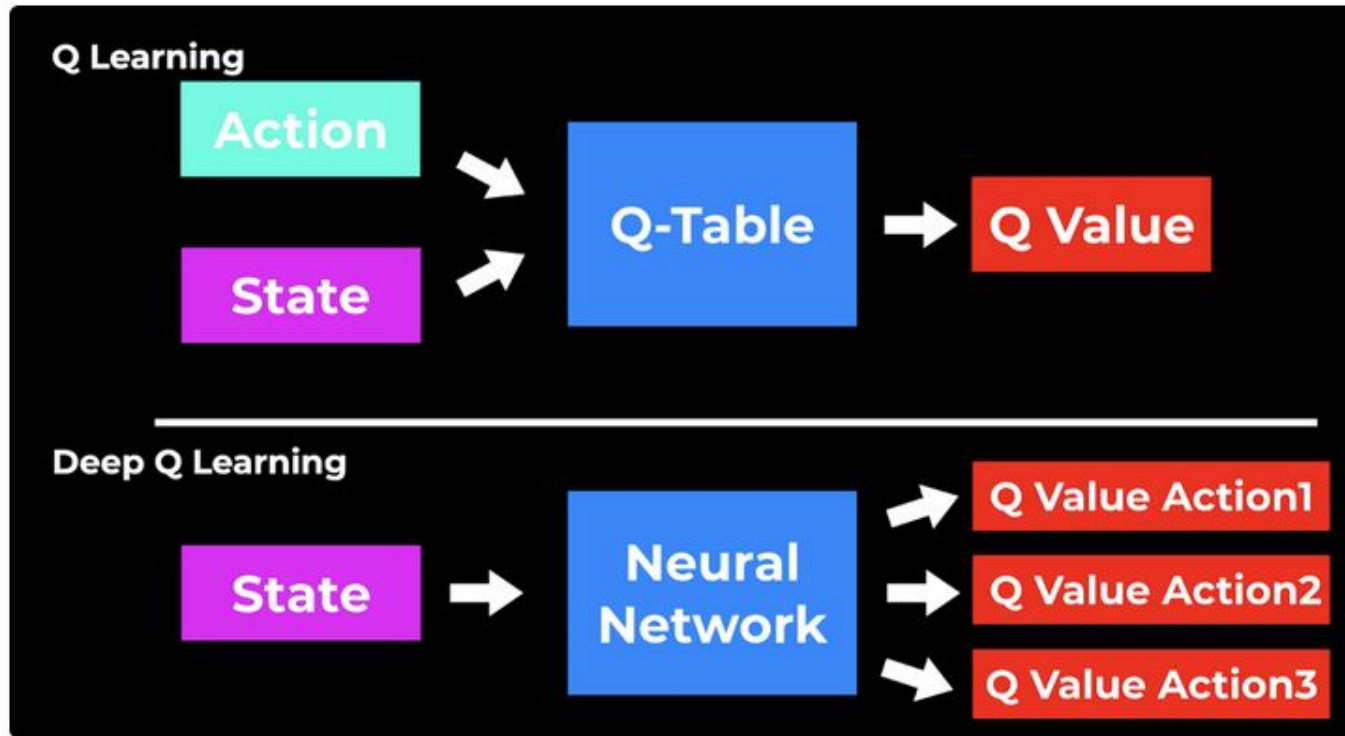
- PnL
- Profit per tick
- Sharpe ratio
- Punishment for long hold times





More deeply in DRL

- Q-Learning[1]
 - It is a value based approach based on the so called Q-Table.
 - The **Q-Table calculates the maximum expected future reward, for each action at each state**. With this information we can then choose the action with the highest reward.
- Deep Q-Learning:
 - Deep Q Learning uses the Q-learning idea and takes it one step further. Instead of using a Q-table, **we use a Neural Network** that takes a state and approximates the Q-values for each action based on that state.
- Deep Recurrent Q-Learning
 - this is a kind of model that adds Recurrent Network layers to original DQN.
 - Indeed, DQN is a powerful model for many problem domains. However, they are limited in the sense that they learn a mapping from a limited number of past states. This is a scenario where Recurrent Network can be reasonably applied. Specifically, to implement Deep Recurrent Q-Network, we may add an LSTM layer to remember states from the past.



Cons RL in HFT

- Difficulties to apply **Modern Portfolio Theory**
- RL needs a lot of realistic data and a lot of computation.

REX

Form my previous experiences :

- Companies avoid to use architectures based on unexplained models.
- Data Quality make difference.
- Don't try to follow the trend on ML World.
- Separate the system in many component as well as possible in order to improve easily each component and to affect dedicated team to each component.
- We don't need to be irrational on risk.
- The best results in trading environment came from Deep Reinforcement Learning.



References

- [1]: Aldridge, I., Krawciw, S., 2017. Real-Time Risk: What Investors Should Know About Fintech, High-Frequency Trading and Flash Crashes. Hoboken: Wiley.
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- [6]: Sutton, Richard S., and Andrew G. Barto. *Reinforcement learning: An introduction*. MIT press, 2018.
- [7]: <https://github.com/yabdellah/Finance>
- [8]: <https://lilianweng.github.io/posts/2018-04-08-policy-gradient/>
- [9]: <https://www.gymnasium.dev/>

Thanks