

# 机器学习与量化交易实战

---

## 第一讲

# 机器学习与量化交易实战



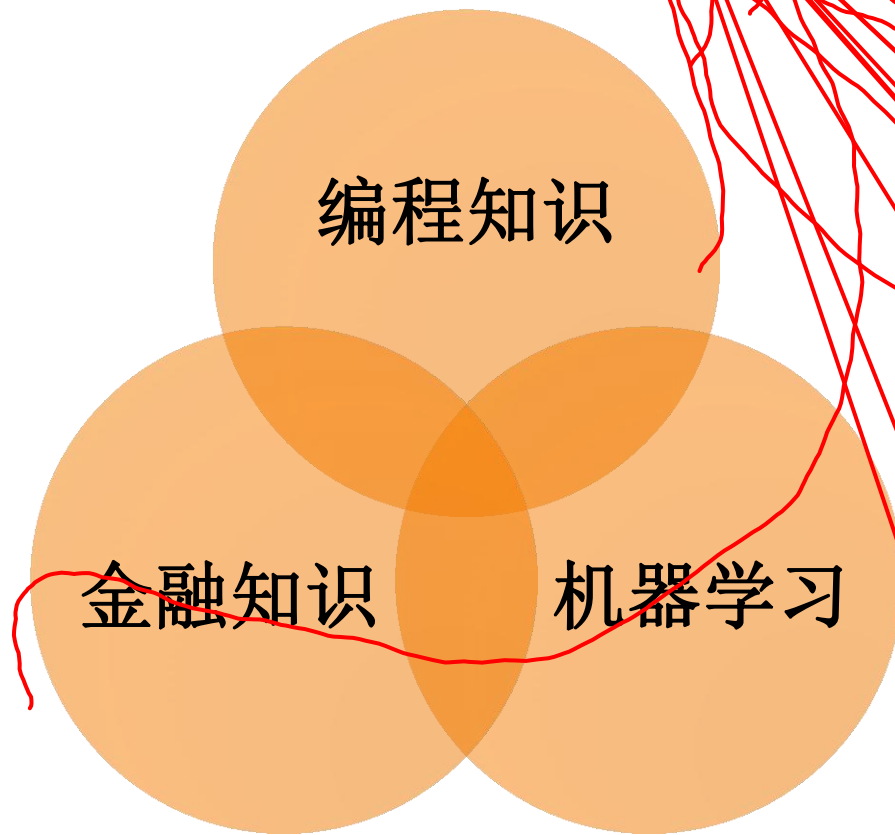
1. 为什么要做算法交易
2. 交易系统的开发与设计
3. 时间序列分析
4. 策略建模及其优化方法
5. 策略评价与回测
6. 风险管理
7. 交易策略的实现
8. 交易策略的执行

重点

This is a team work, but you need to know the big picture and see which role suits you.

# The Big Picture And How The Course Is Organized

---



Office Hours:

8-10 pm 周二 / 周四

Weekly Readings

Homework

Extra Projects For Motivated Students.



# 算法交易综述

---

## Algorithmic Trading Without Bullshit

# 何谓算法交易

---

Algorithmic Trading

利用自动化平台，执行预先设置的一系列规则完成交易行为。

# 算法交易：优势

---

1. 历史数据评估
2. 执行高效
3. 无主观情绪输入
4. 可度量评价
5. 交易频率

# 算法交易：劣势

---

1. 成本

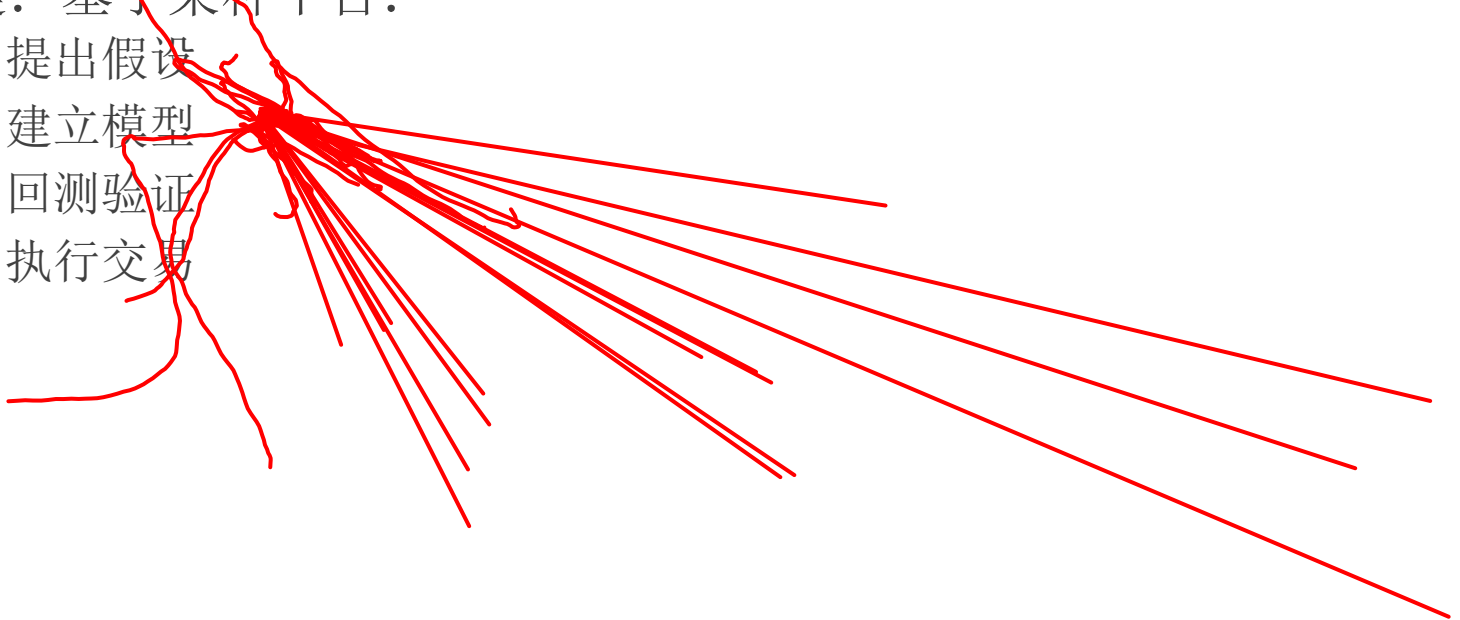
2. 技巧

A red handwritten mark consisting of a horizontal line with an arrow pointing right, and a diagonal line extending downwards and to the right from the end of the horizontal line.

# 算法交易流程

---

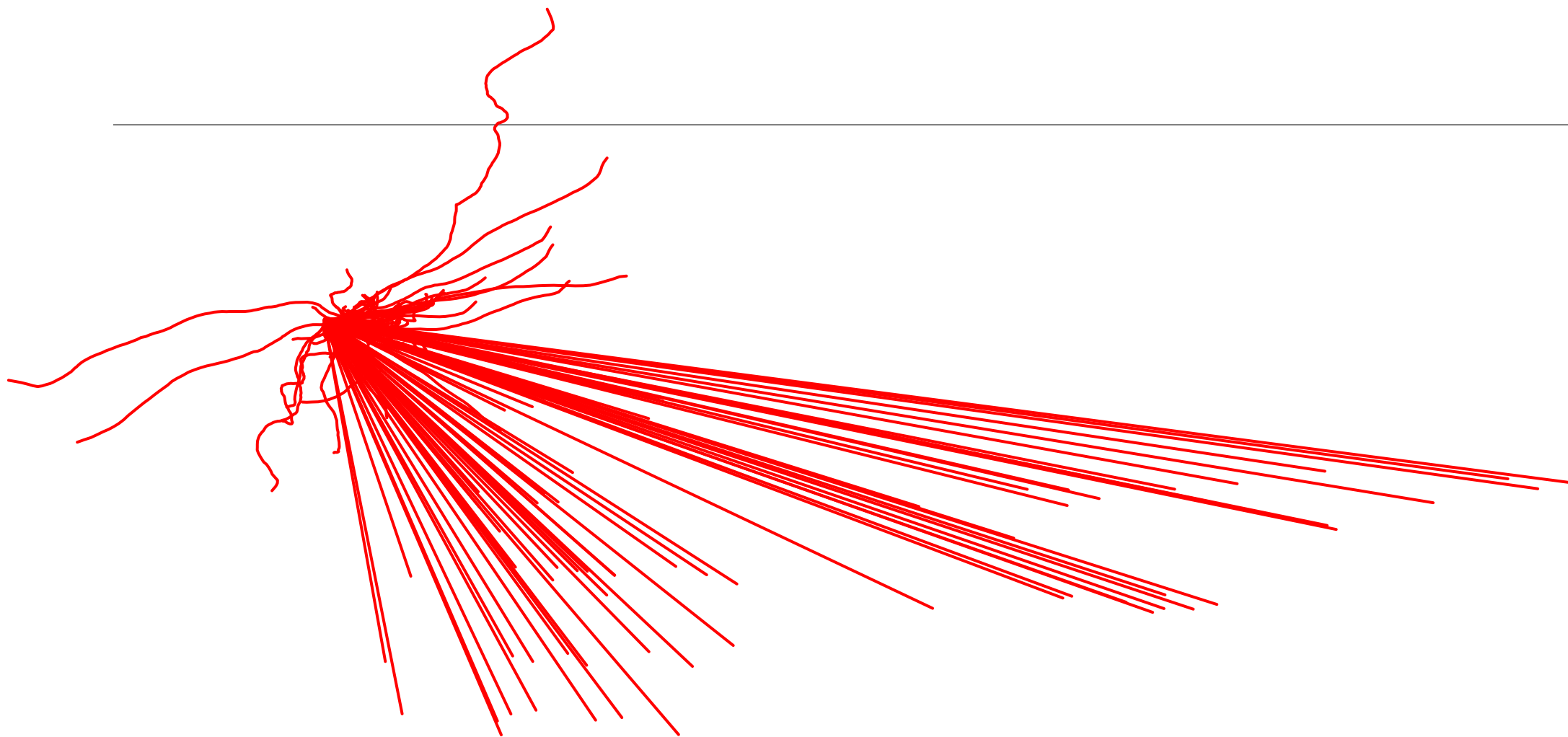
大前提：基于某种平台：

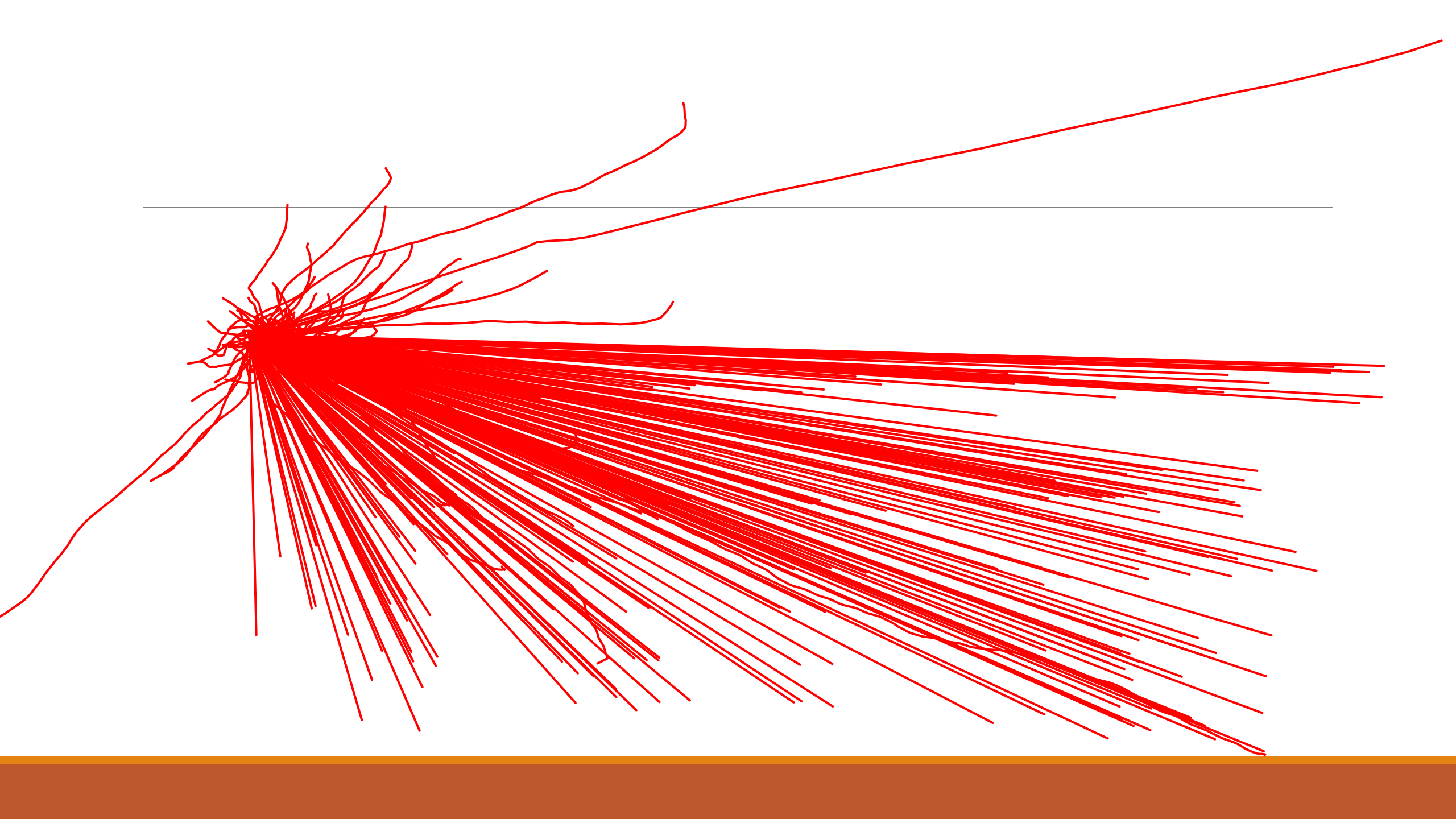
1. 提出假设
  2. 建立模型
  3. 回测验证
  4. 执行交易
- 

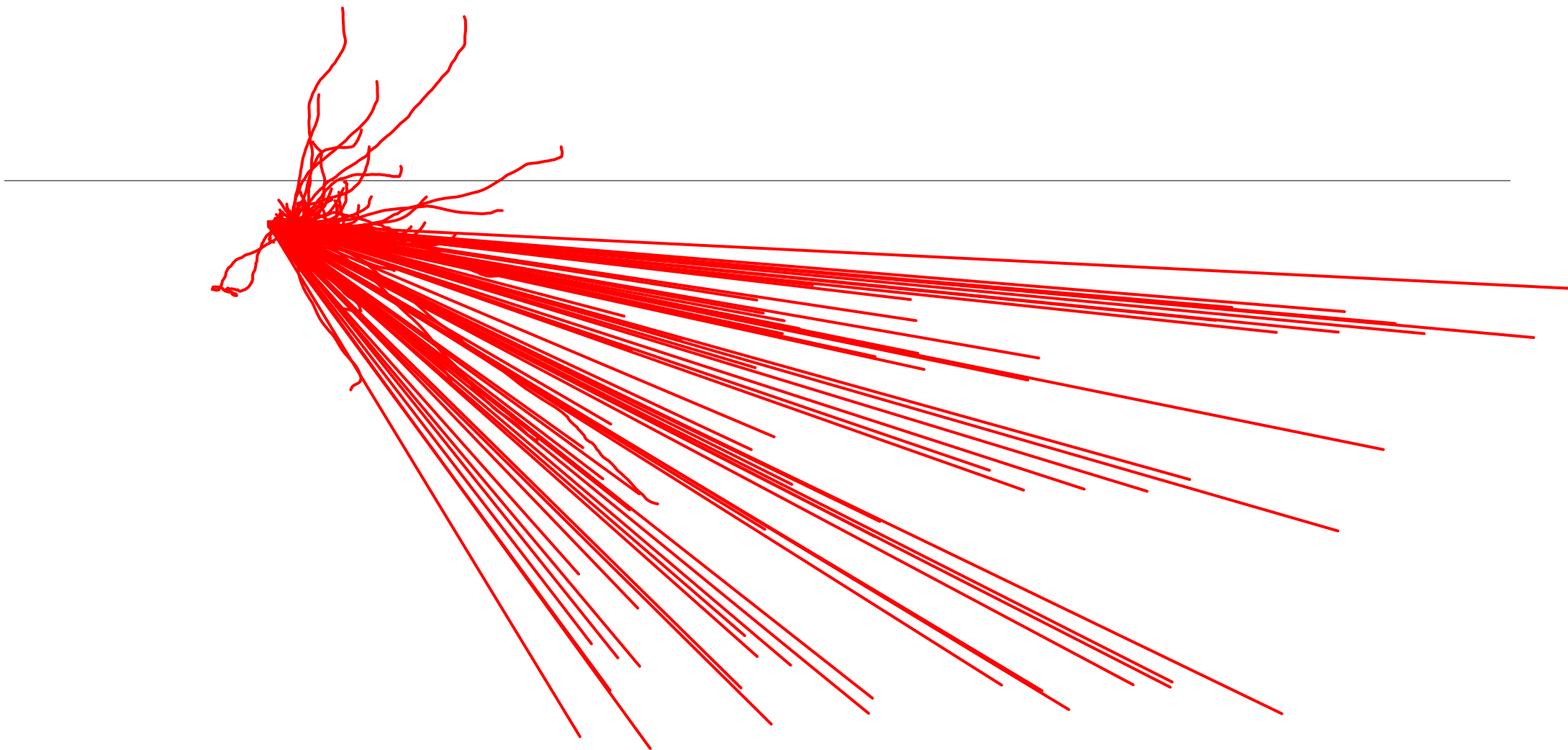
# 交易策略的来源

---

1. 市场微观结构研究 (for HFT mostly)
  2. 基金结构套利(fund structure arbitrage)
  3. 机器学习 / 人工智能
- 

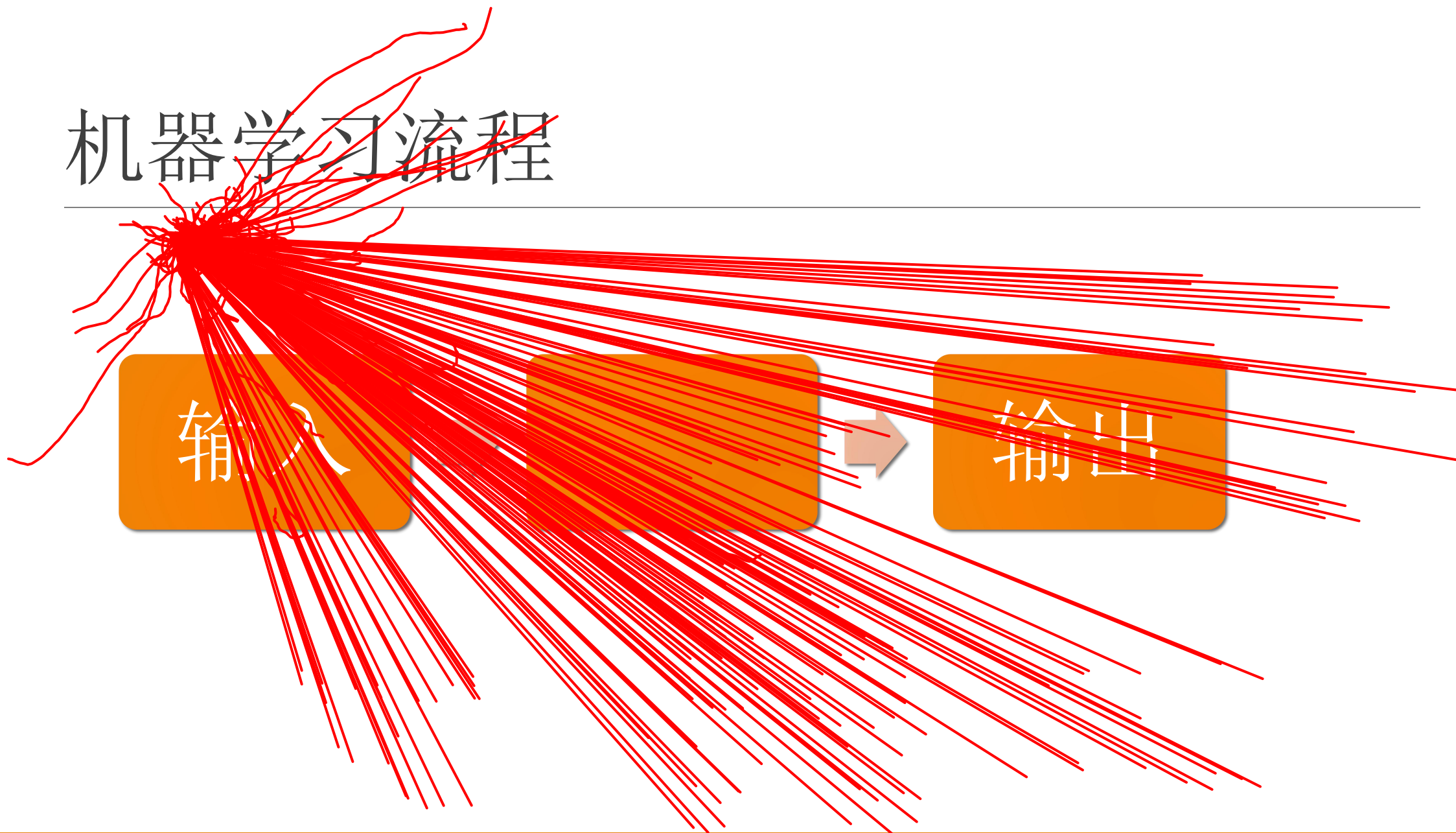




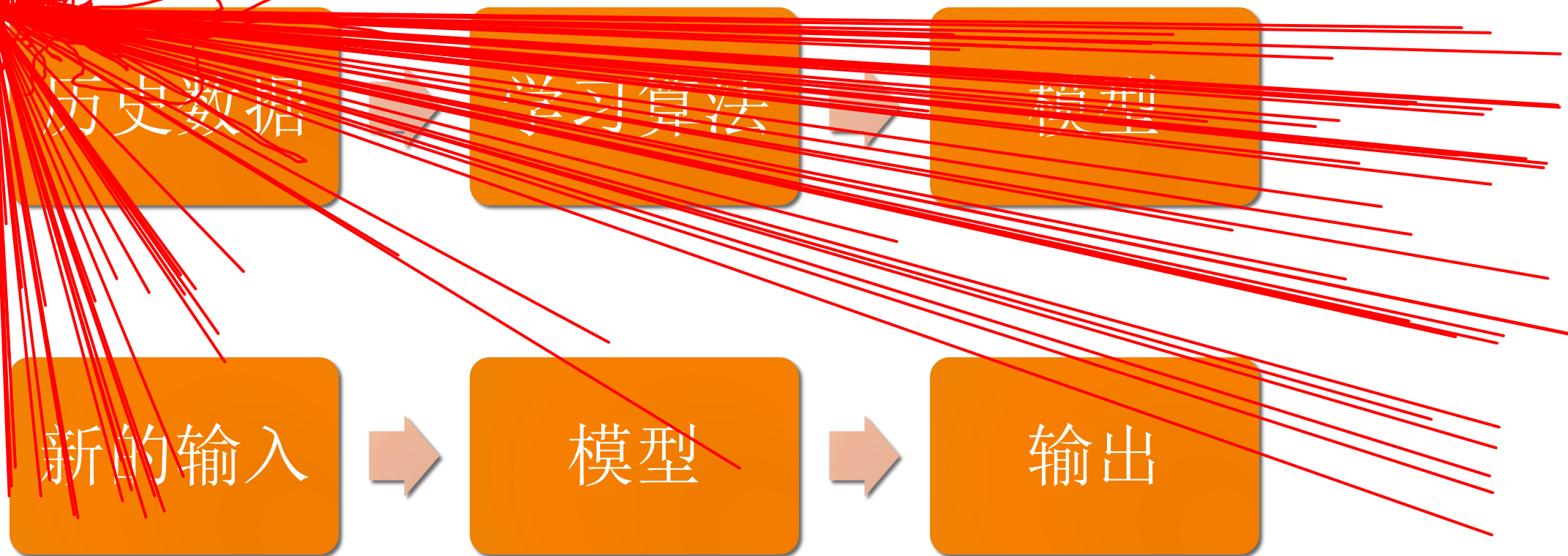


# 机器学习流程

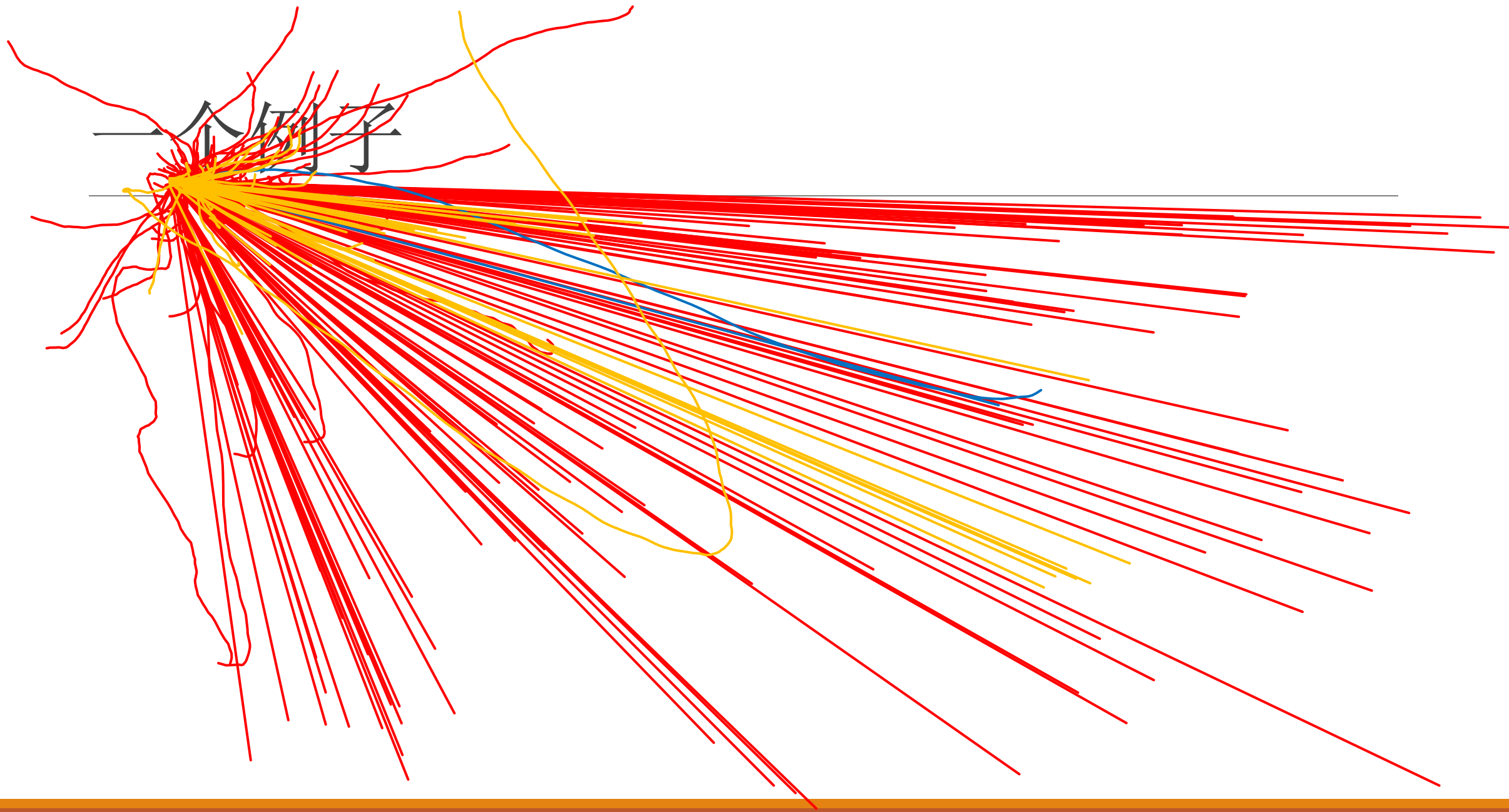
---



# 机器学习流程



一个例子



---

# Four paradigms of machine learning

---

Connectionism

Symbolism

Frequentists

Bayesian



# A (super) Brief History of A.I

---

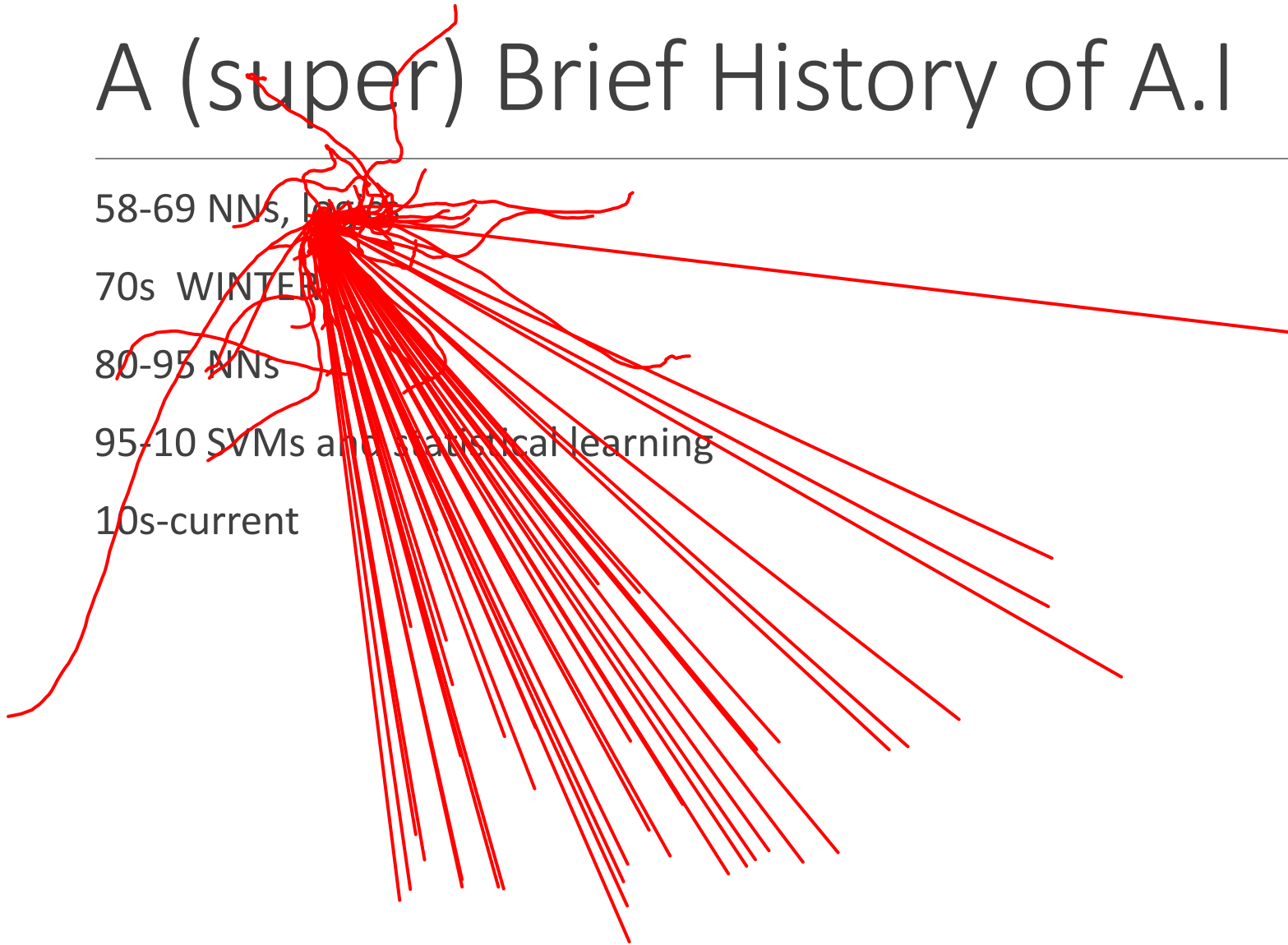
58-69 NNs, logic

70s WINTER

80-95 NNs

95-10 SVMs and statistical learning

10s-current



# Machine Learning in a nutshell

---

Data

Model & Objective Function

Optimization



# Machine Learning & Trading

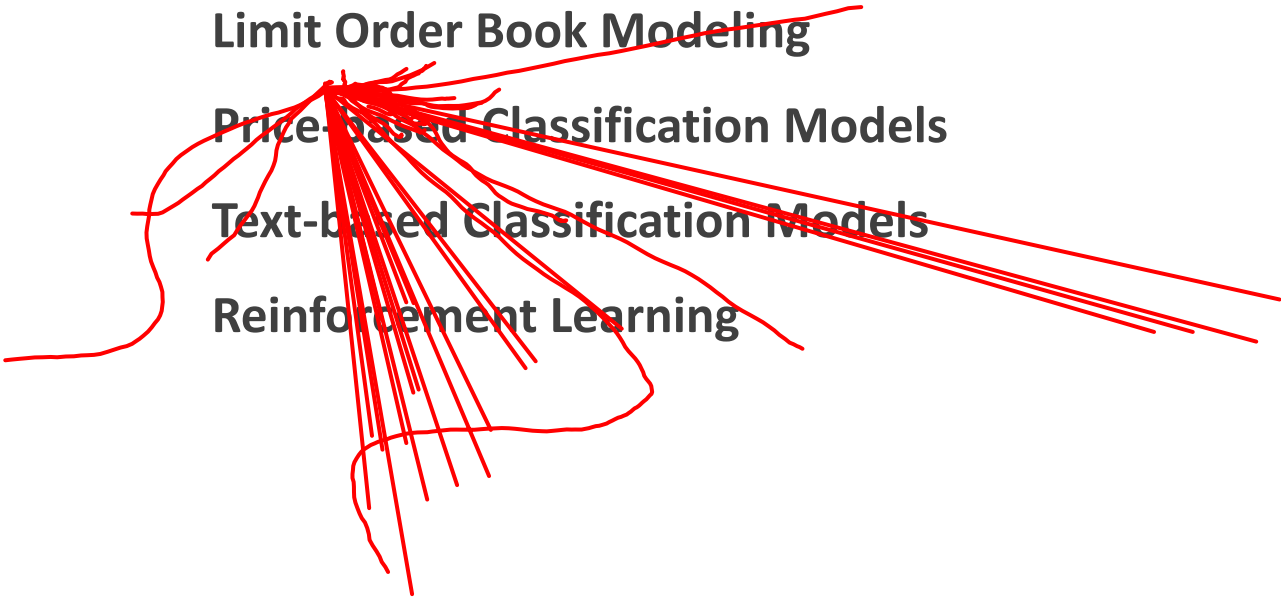
---

**Limit Order Book Modeling**

**Price-based Classification Models**

**Text-based Classification Models**

**Reinforcement Learning**



# 4 Key Factors that makes magic happens

---

Good Model and Efficient Training Algorithms

Hardware( GPU/CPU)

(high quality) DATA

Platform(keras/tensorflow/sklearn)

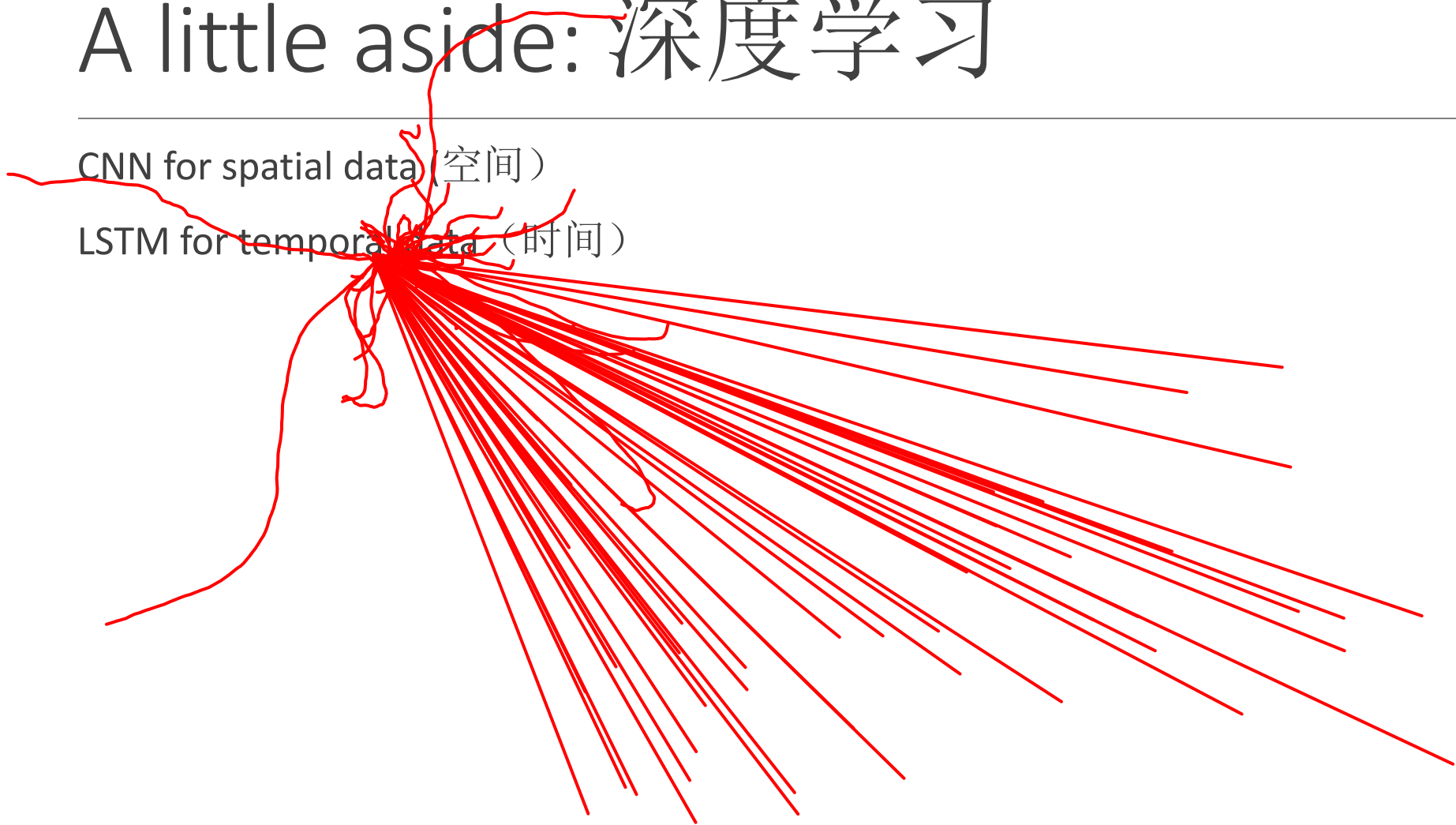


# A little aside: 深度学习

---

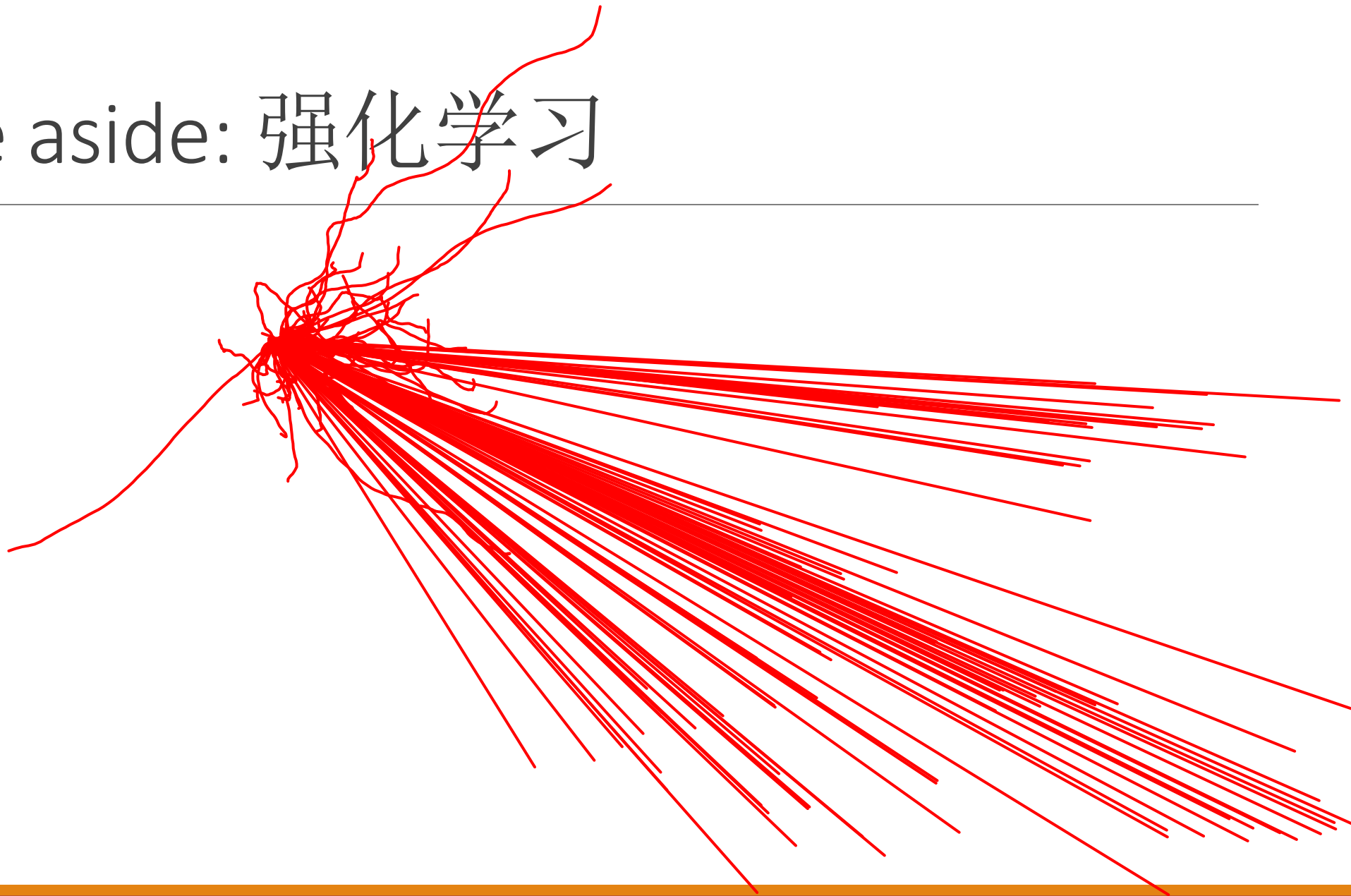
CNN for spatial data (空间)

LSTM for temporal data (时间)

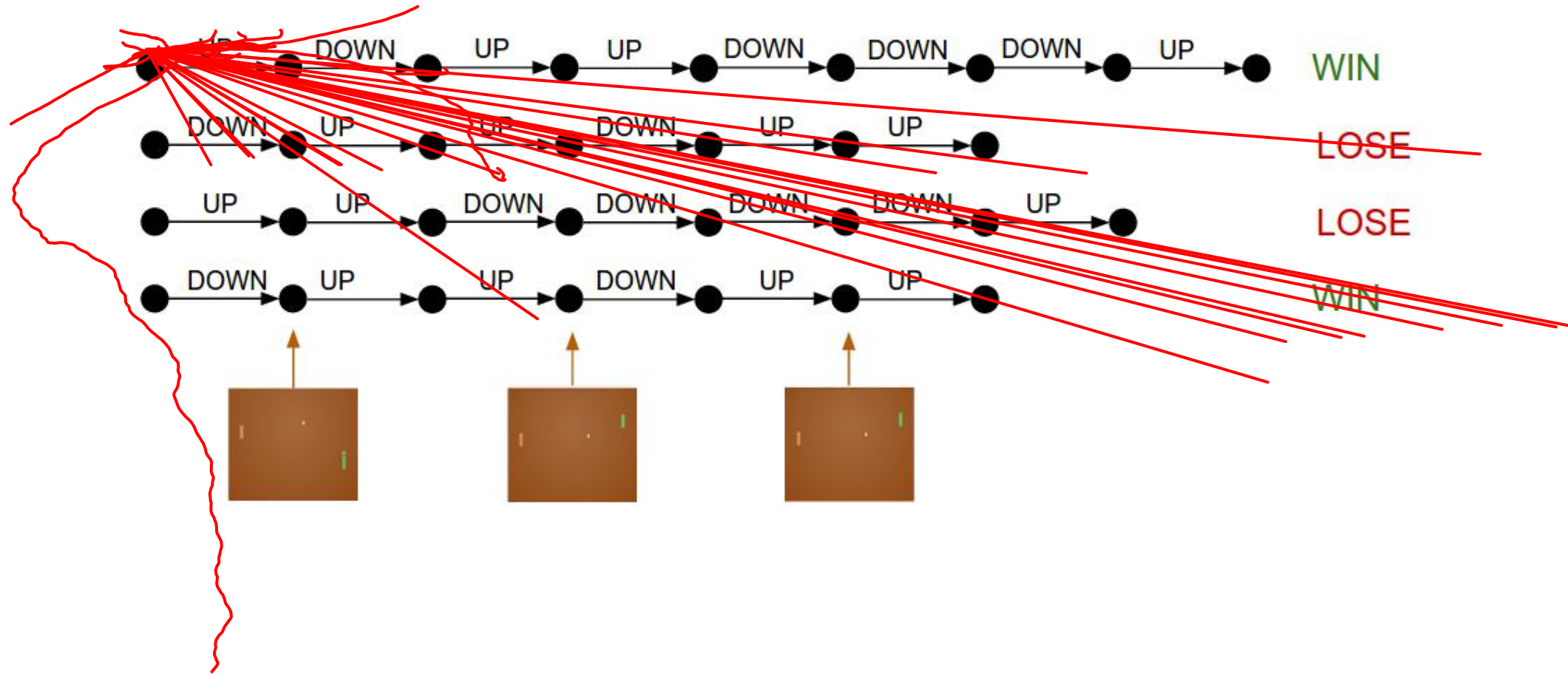


# A little aside: 强化学习

---



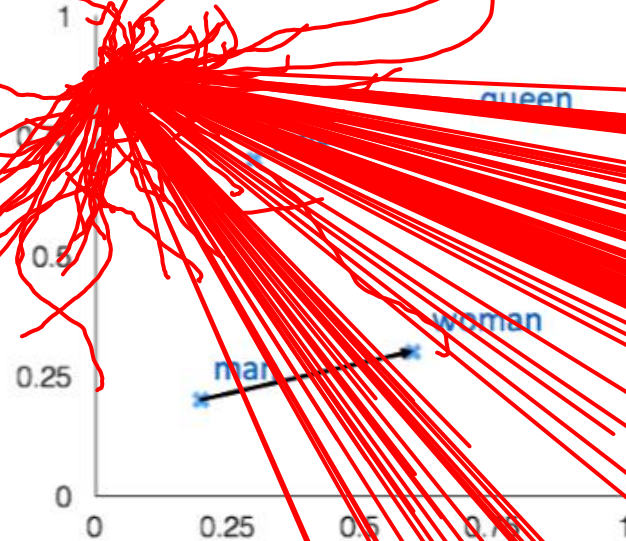
# A little aside: 强化学习



# A little aside: 自然语言处理

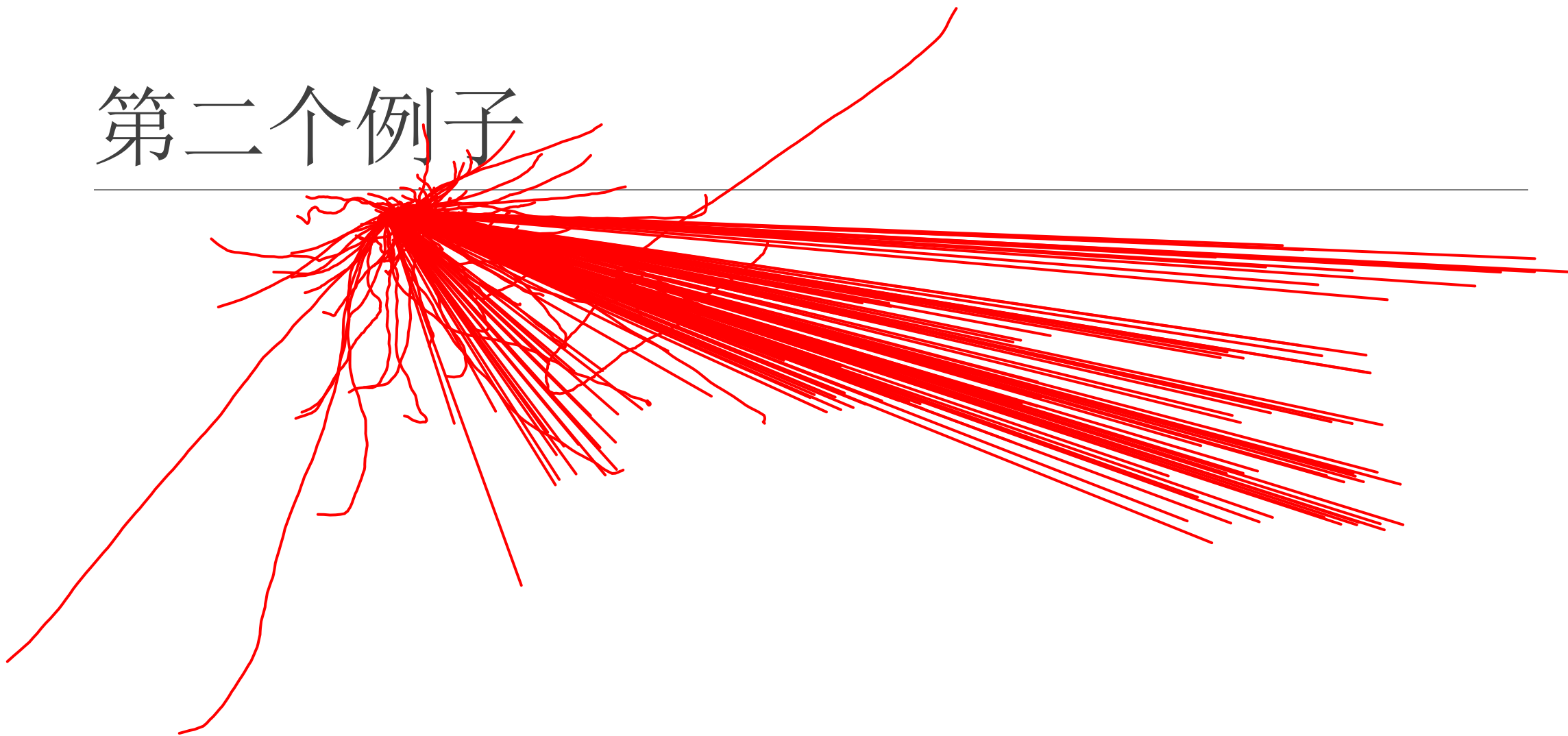
man:woman :: king:?

+ king	[ 0.30 0.70 ]
- man	[ 0.20 0.20 ]
+ woman	[ 0.60 0.30 ]
<hr/>	
queen	[ 0.70 0.80 ]



## 第二个例子

---

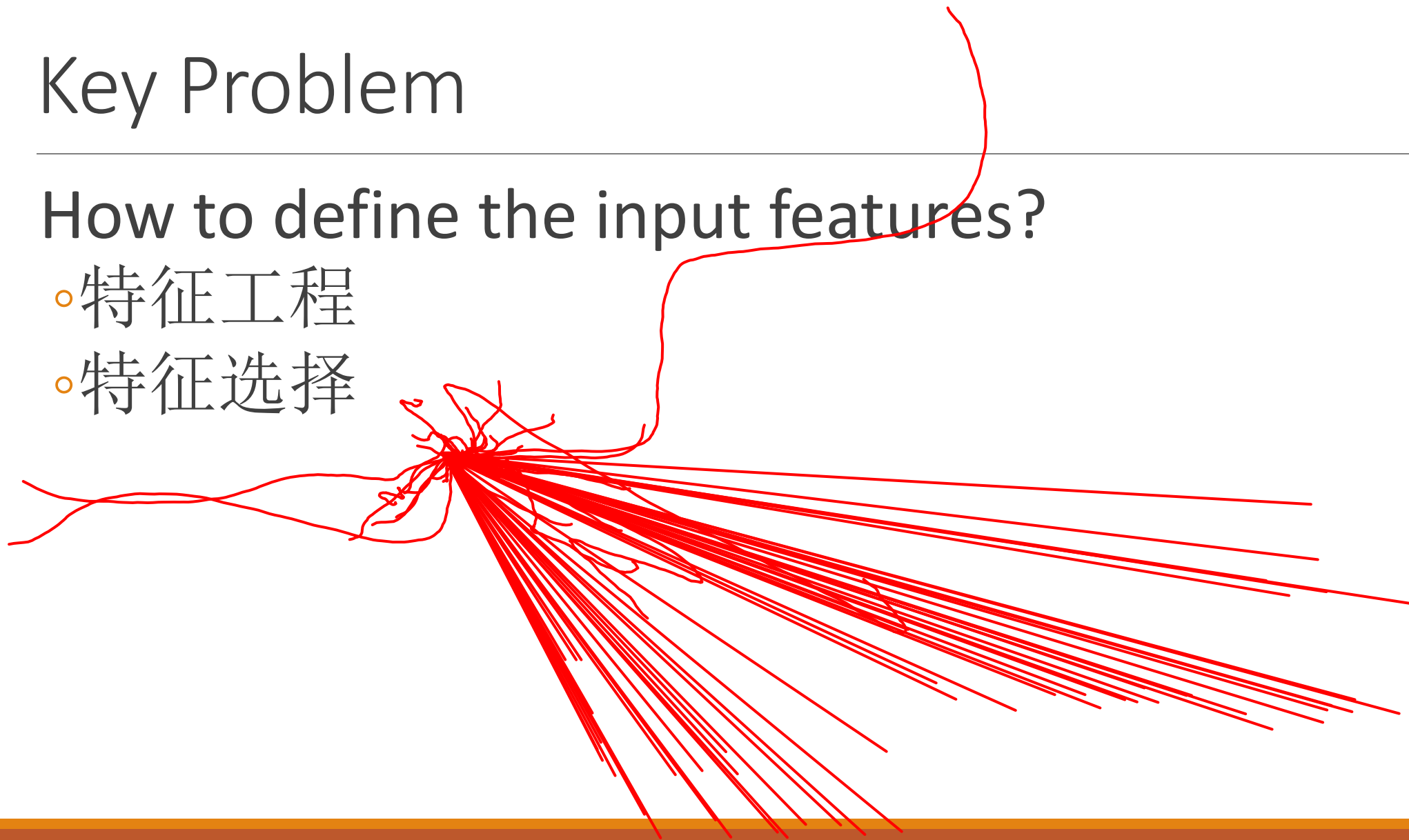


# Key Problem

---

How to define the input features?

- 特征工程
- 特征选择



# 交易策略的评估

1. 策略基本假设
2. Sharp Ratio
3. 杠杆
4. 频率
5. 风险
6. W/L
7. 模型复杂度
8. 最大亏损(Maxium drawdown)
9. Benchmarking

# 回测

---

何谓回测？

将交易策略在历史数据中进行合理验证的过程。

科学的回测十分重要（大部分人死在回测上）

# 回测的意义

---

1. 策略筛选

2. 策略优化

3. 策略验证

# 错误的回测方法

---

很多情况下，回测结果不错，实盘交易不尽如人意。  
造成的偏差原因主要有：

1. 乐观主义偏差。(special look back region)
2. 时间旅行。
  1. 程序Bug
  2. Train/Val/Test set
3. 幸存者误差

# 工具和语言

---

Python

- Sklearn
- Pandas
- And more...

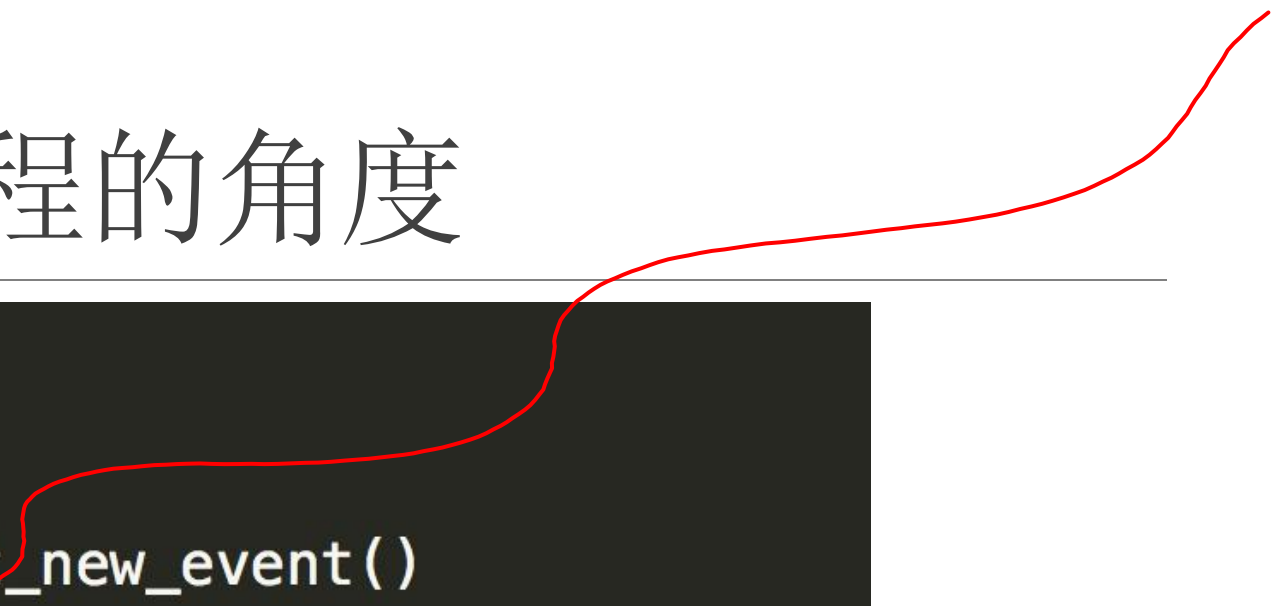

# 量化交易：从工程的角度

```
#event driven

while True:
    new_event = get_new_event()

    if new_event.something == "whatever"
        do_something()
    if new_event.something == "all right"
        do_something_else()

    tick(50) #wait 50 milliseconds
```



# 量化交易：从工程的角度

---

Event

Event Queue

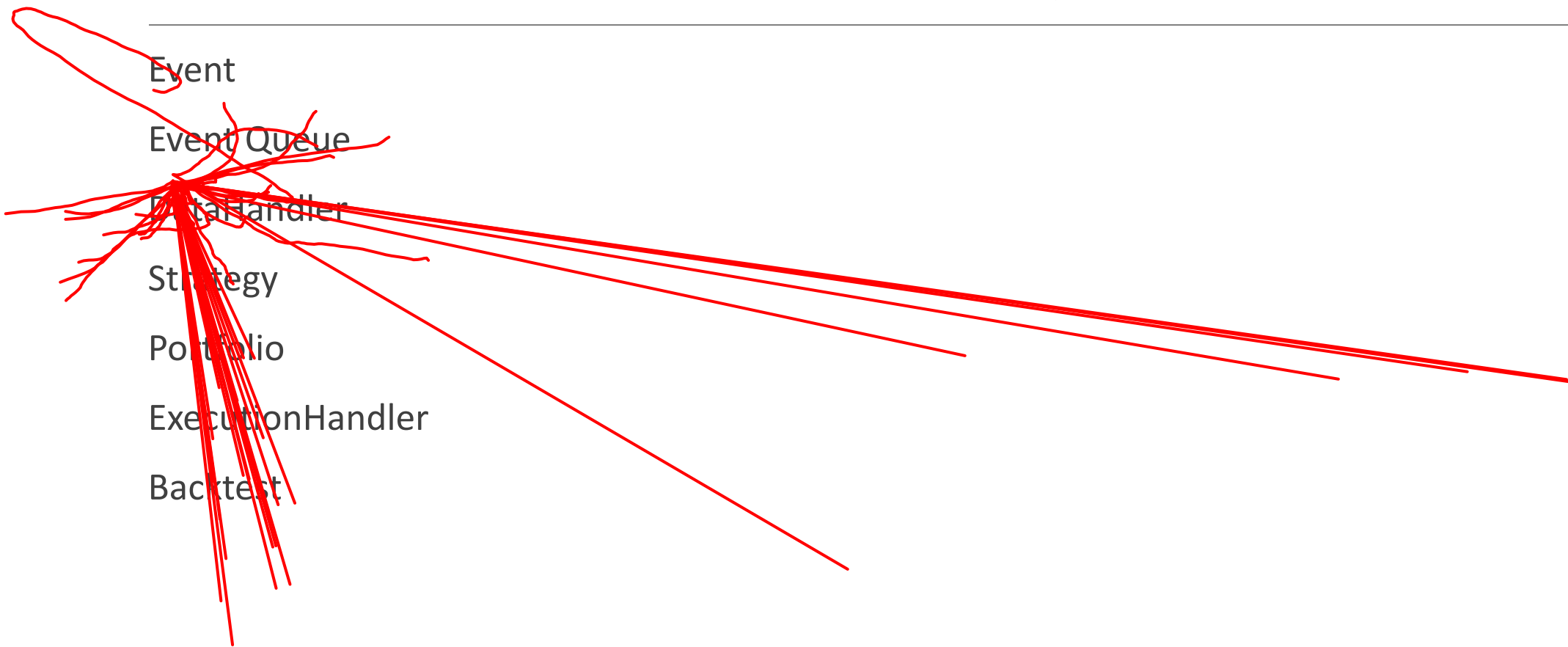
EventHandler

Strategy

Portfolio

ExecutionHandler

Backtest



# Next Time

---

一些金融基本知识

以及Python..

