

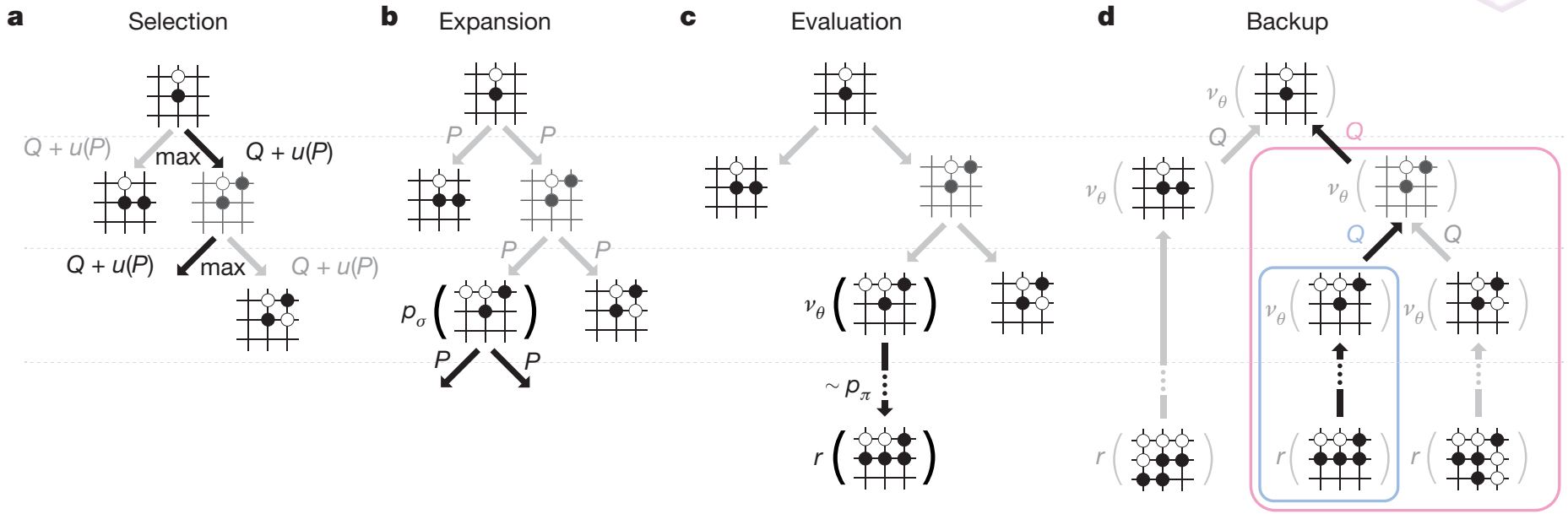
# Lecture 20: Final

# On Artificial Intelligence

[http://cs.nju.edu.cn/yuy/course\\_ai16.ashx](http://cs.nju.edu.cn/yuy/course_ai16.ashx)



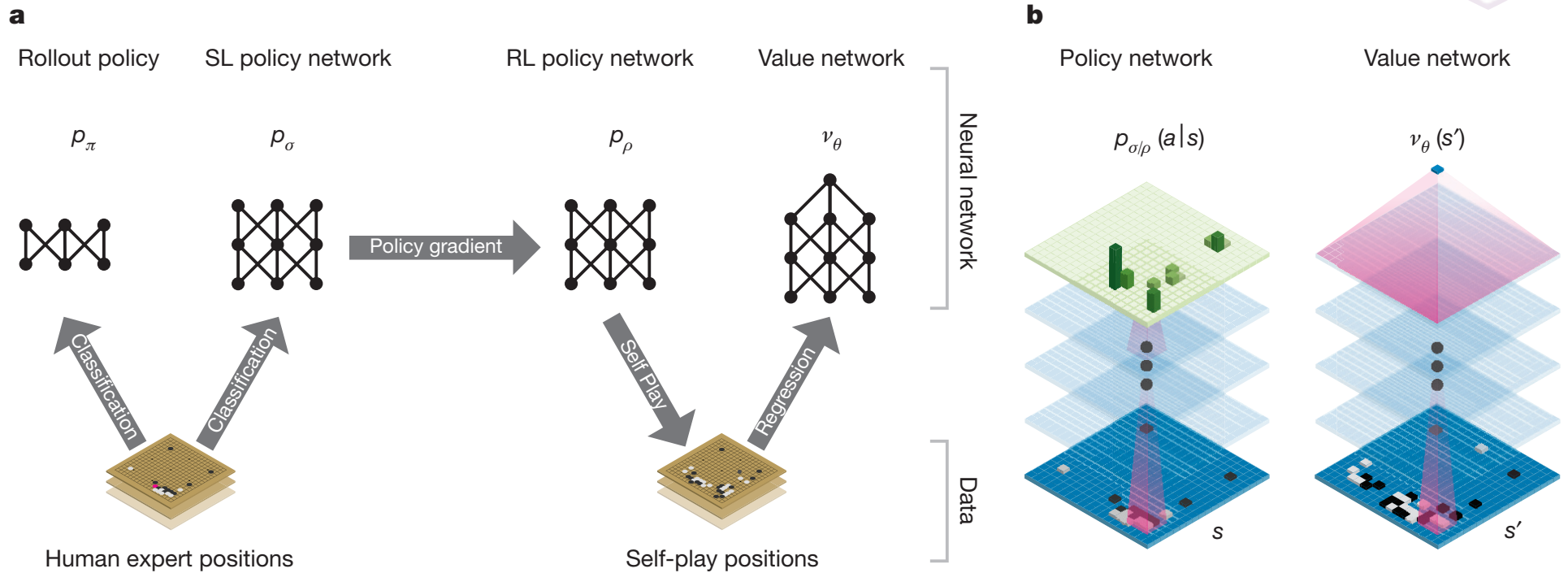
# Review on AlphaGo



**Figure 3 | Monte Carlo tree search in AlphaGo.** **a**, Each simulation traverses the tree by selecting the edge with maximum action value  $Q$ , plus a bonus  $u(P)$  that depends on a stored prior probability  $P$  for that edge. **b**, The leaf node may be expanded; the new node is processed once by the policy network  $p_\sigma$  and the output probabilities are stored as prior probabilities  $P$  for each action. **c**, At the end of a simulation, the leaf node

is evaluated in two ways: using the value network  $v_\theta$ ; and by running a rollout to the end of the game with the fast rollout policy  $p_\pi$ , then computing the winner with function  $r$ . **d**, Action values  $Q$  are updated to track the mean value of all evaluations  $r(\cdot)$  and  $v_\theta(\cdot)$  in the subtree below that action.

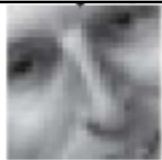
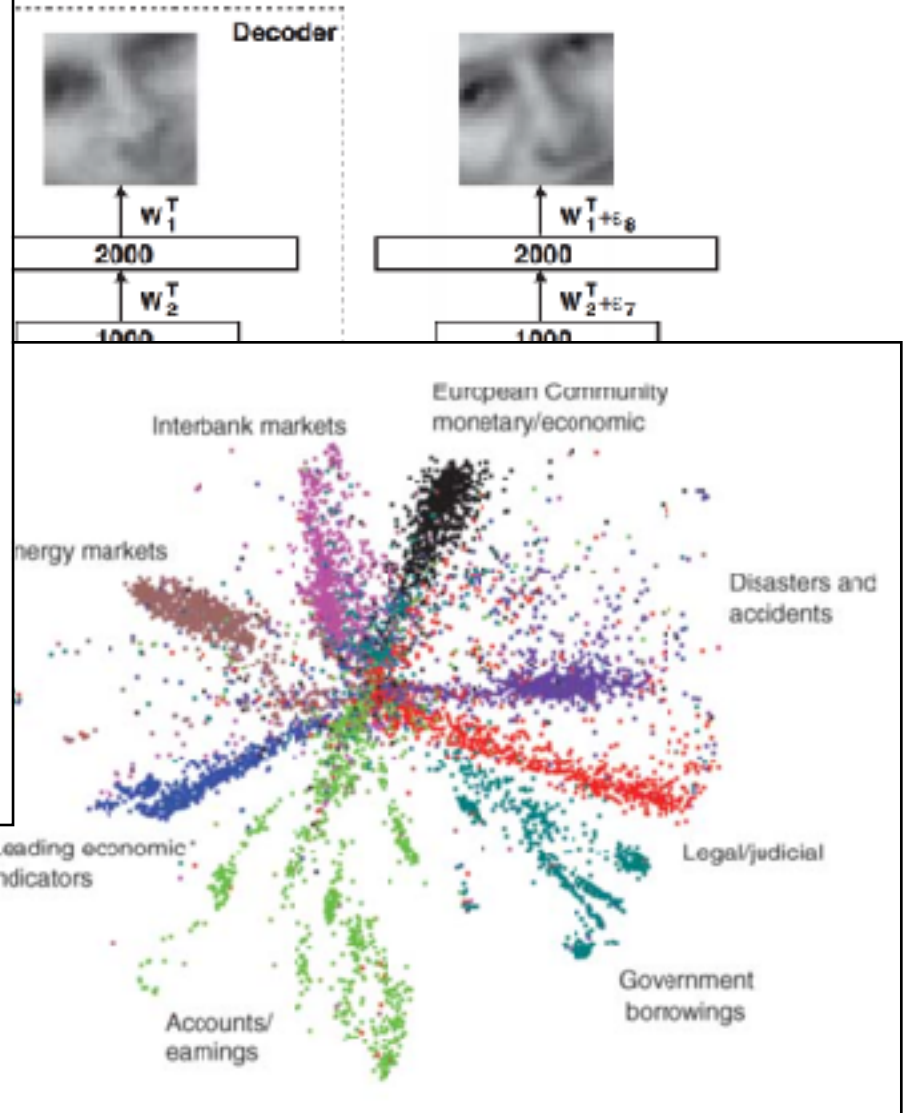
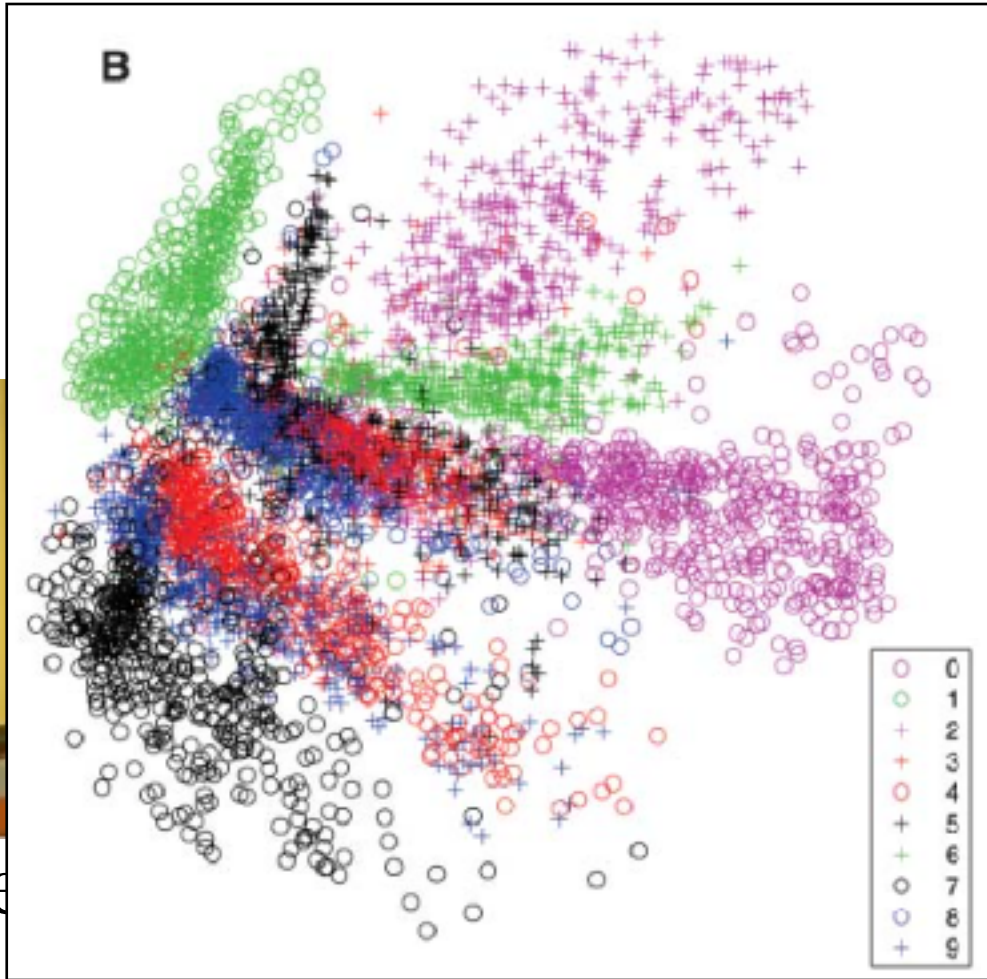
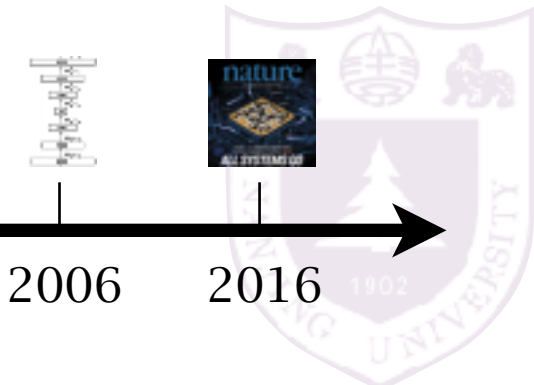
# Review on AlphaGo



**Figure 1 | Neural network training pipeline and architecture.** **a**, A fast rollout policy  $p_\pi$  and supervised learning (SL) policy network  $p_\sigma$  are trained to predict human expert moves in a data set of positions. A reinforcement learning (RL) policy network  $p_\rho$  is initialized to the SL policy network, and is then improved by policy gradient learning to maximize the outcome (that is, winning more games) against previous versions of the policy network. A new data set is generated by playing games of self-play with the RL policy network. Finally, a value network  $v_\theta$  is trained by regression to predict the expected outcome (that is, whether

the current player wins) in positions from the self-play data set. **b**, Schematic representation of the neural network architecture used in AlphaGo. The policy network takes a representation of the board position  $s$  as its input, passes it through many convolutional layers with parameters  $\sigma$  (SL policy network) or  $\rho$  (RL policy network), and outputs a probability distribution  $p_\sigma(a|s)$  or  $p_\rho(a|s)$  over legal moves  $a$ , represented by a probability map over the board. The value network similarly uses many convolutional layers with parameters  $\theta$ , but outputs a scalar value  $v_\theta(s')$  that predicts the expected outcome in position  $s'$ .

# History



Pretraining

RBM End

Ge



**Describes without errors**

**Describes with minor errors**

**Somewhat related to the image**

**Unrelated to the image**



**A person riding a motorcycle on a dirt road.**



**Two dogs play in the grass.**



**A skateboarder does a trick on a ramp.**



**A dog is jumping to catch a frisbee.**



**A group of young people playing a game of frisbee.**



**Two hockey players are fighting over the puck.**



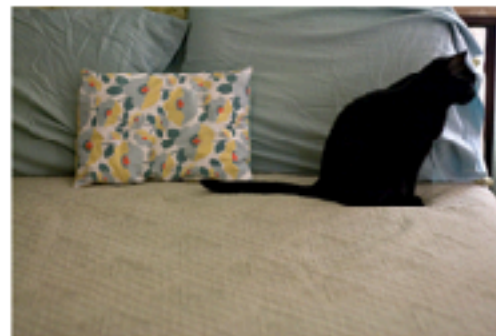
**A little girl in a pink hat is blowing bubbles.**



**A refrigerator filled with lots of food and drinks.**



**A herd of elephants walking across a dry grass field.**



**A close up of a cat laying on a couch.**



**A red motorcycle parked on the side of the road.**



**A yellow school bus parked in a parking lot.**

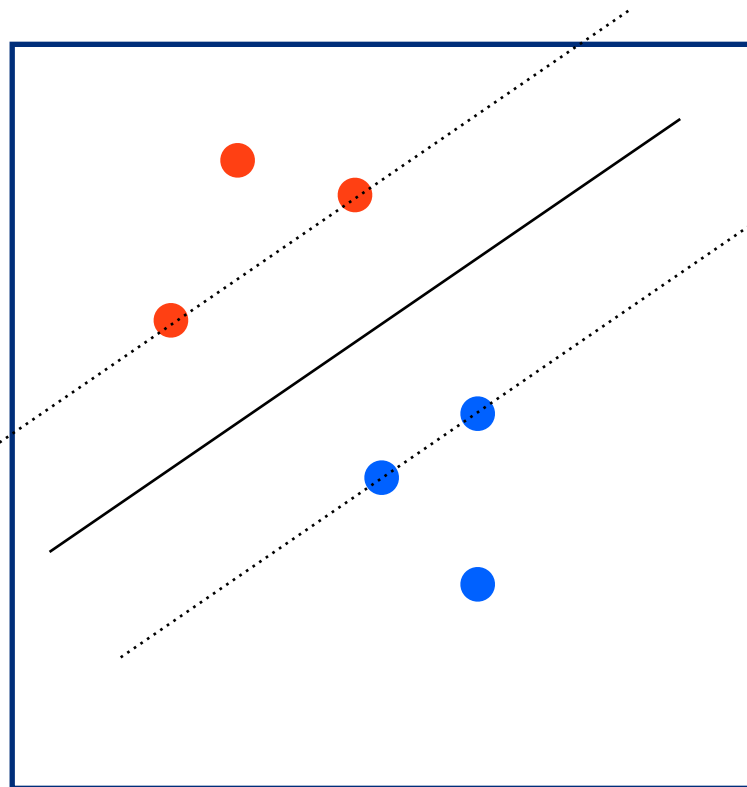
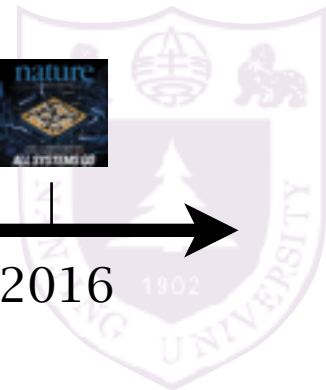
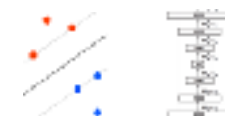


90年代  
中期

1998

2006

2016





### 店铺2级服务概述

好评退款率: 0.0022% 小于 0.0064% (行业均值)      退款自主完成率: 99.90% 大于 95.47% (行业均值)

退款纠纷时长: 0.64天 小于 1.66天 (行业均值)

店铺动态评分: (所属行业: 书籍音像)

商品与服务描述:	4.8分	比同行平均水平 低0.86%	5分: 92.47% (31601人)
商家发货与描述:	4.7分	比同行平均水平 低0.82%	4分: 3.99% (1362人)
商家发货的速度:	4.7分	比同行平均水平 低1.89%	3分: 1.61% (551人)
			2分: 0.58% (198人)
			1分: 1.36% (464人)

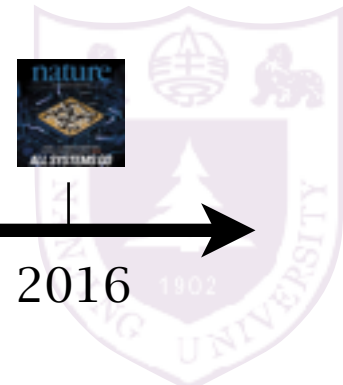
商家承诺: 凡使用支付宝支付付款购买商品, 若存在该商家发货延迟不期, 本店支持退货退款服务并承诺包邮费!

### 精选好书

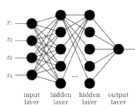
<p>Head First C#</p> <p>¥38.8</p>	<p>人脸识别原理及算法</p> <p>¥65.9</p>	<p>数学之美</p> <p>¥35</p>
<p>智能车辆导航技术</p> <p>¥51.8</p>	<p>视觉机器学习</p> <p>¥42</p>	<p>Offer</p> <p>¥41</p>
<p>统计学习方法</p> <p>¥28.5</p>	<p>程序员的数学</p> <p>¥53.2</p>	<p>深度学习 方法及应用</p> <p>¥35.9</p>



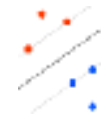




80年代  
初期



90年代  
中期

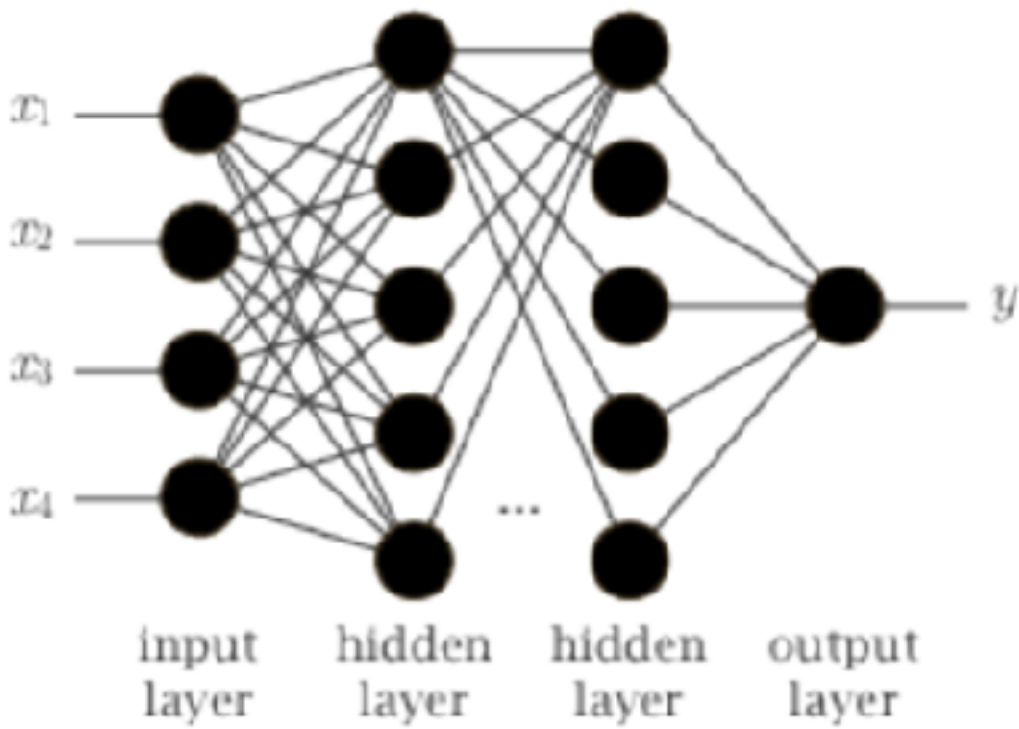


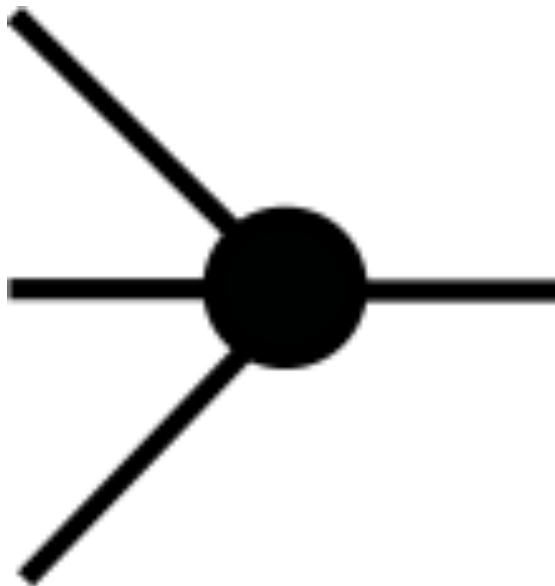
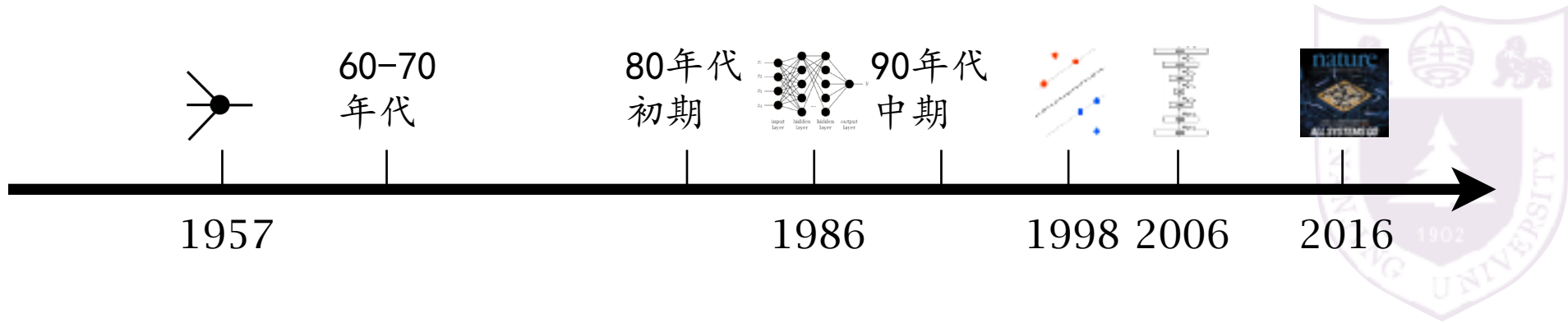
1986

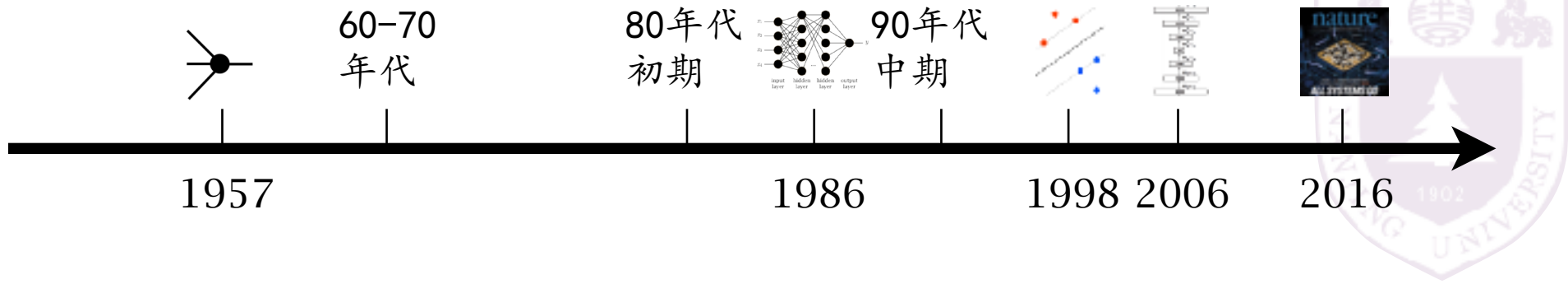
1998

2006

2016



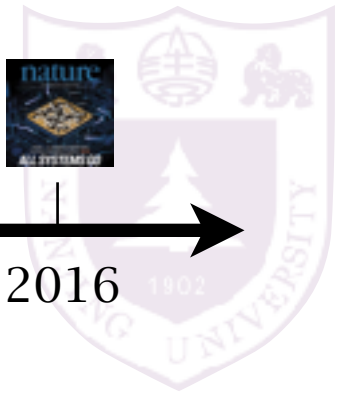
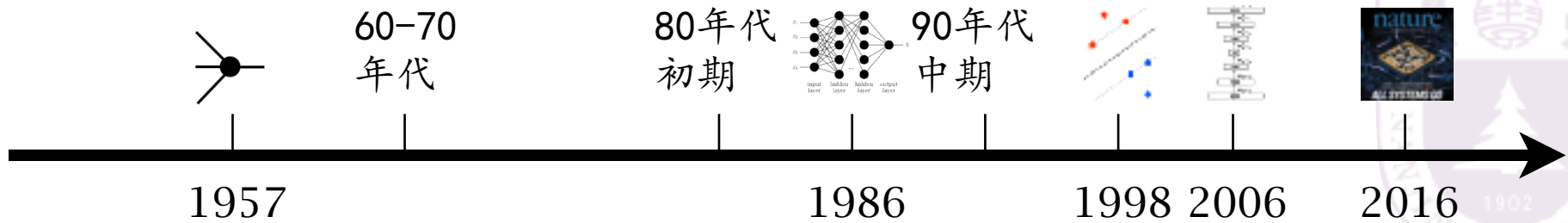




Edward Albert Feigenbaum

## 专家系统





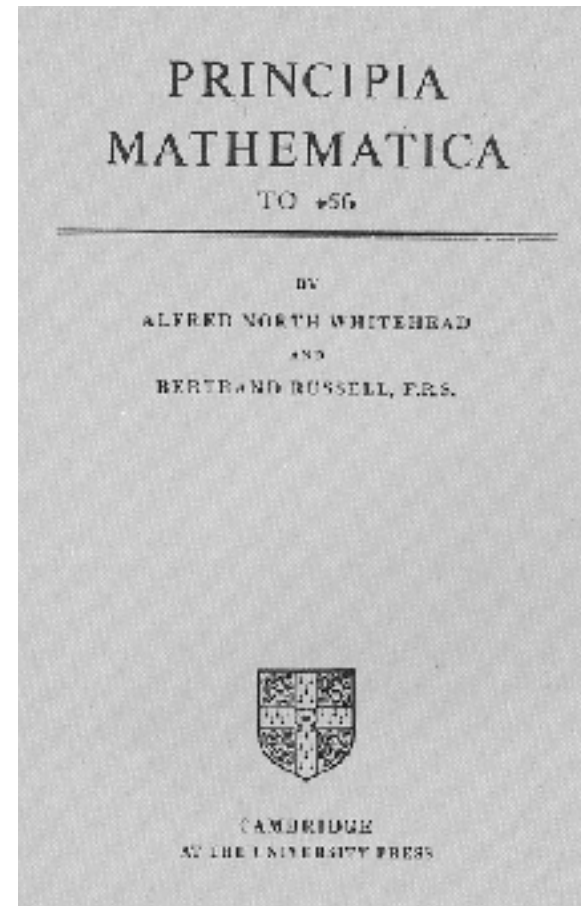
## 逻辑学家



Allen Newell



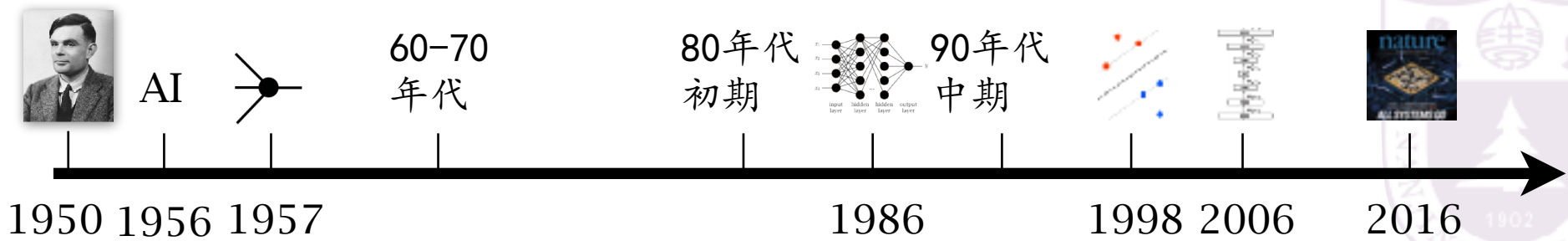
Herbert Simon



# 1956 Dartmouth meeting: “Artificial Intelligence” adopted



- |                      |  |
|----------------------|--|
| <b>John McCarthy</b> | Turing Award (1971)                                  |
| Marvin Minsky        | Turing Award (1969)                                  |
| Claude Shannon       | the father of information theory                     |
| Oliver Selfridge     | father of machine perception                         |
| Herbert A. Simon     | Turing Award (1975), Nobel Prize in Economics (1978) |
| Allen Newell         | Turing Award (1975)                                  |

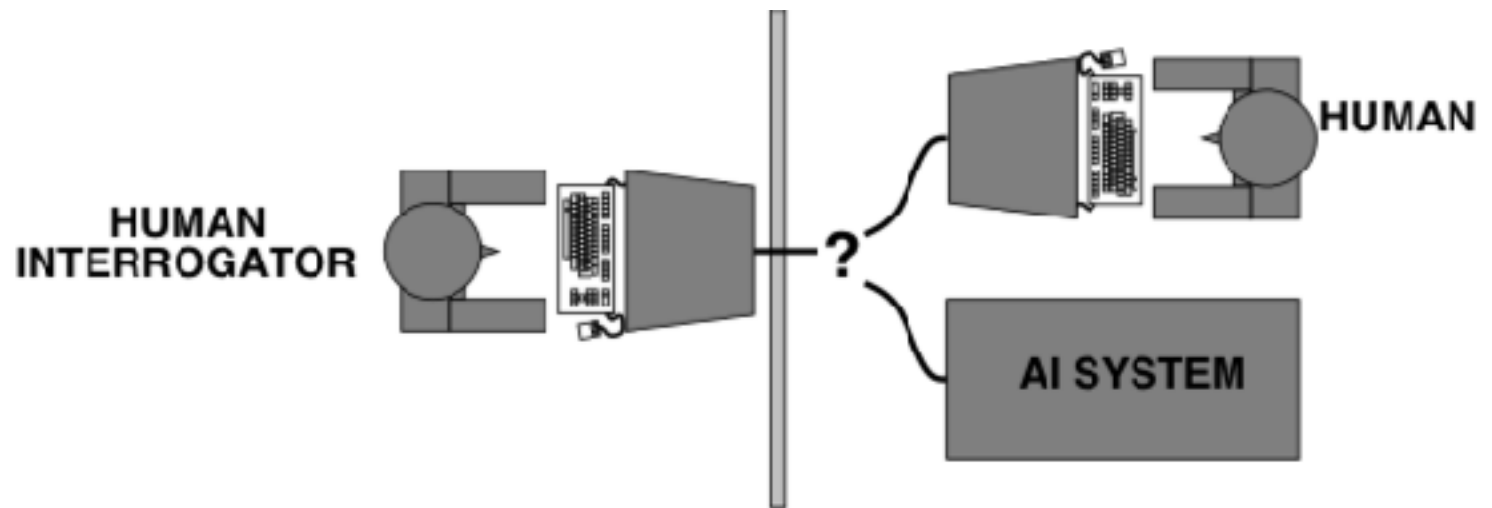


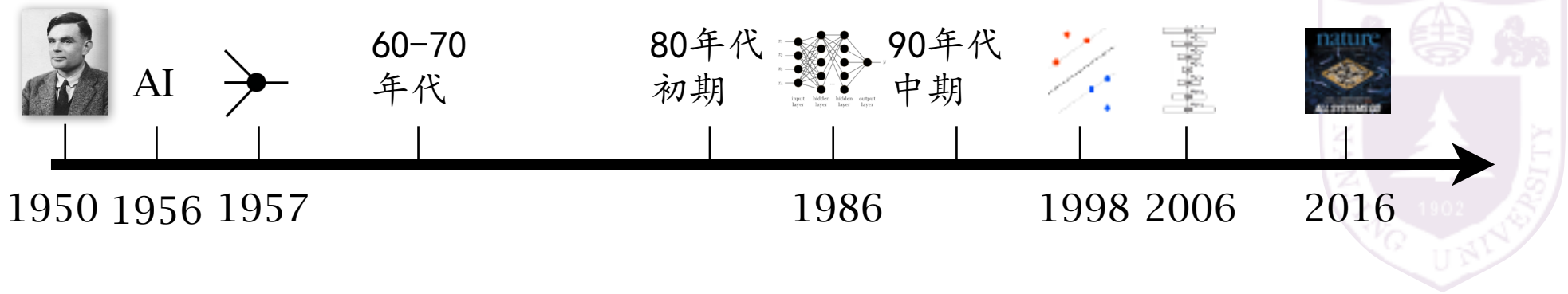
[*Computing machinery and intelligence. Mind* 49: 433-460, 1950.]



Alan Turing  
1912-1954

## Section 1: Imitation game





[*Computing machinery and intelligence. Mind* 49: 433-460, 1950.]



Alan Turing  
1912-1954

## Section 7: Learning machines

“In the process of trying to imitate an adult human mind we are bound to think a good deal about the process which has brought it to the state that it is in. We may notice three components.

- (a) The initial state of the mind, say at birth,
- (b) The education to which it has been subjected,
- (c) Other experience, not to be described as education, to which it has been subjected.”

Instead of trying to produce a programme to simulate the adult mind, why not rather try to produce one which simulates the child's?

# History

学习期



知识期

推理期

AI之冬

AI之冬

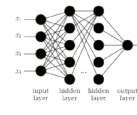


AI



60-70年代

80年代初期



90年代中期



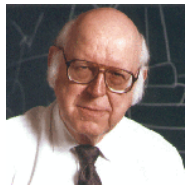
1950 1956 1957

1986

1998 2006

2016

- “解决了神秘的心/身问题，解释了物质构成的系统如何获得心灵的性质。”
- “十年之内，数字计算机将成为国际象棋世界冠军。”
- “二十年内，机器将能完成人能做到的一切工作。”
- “一代之内……创造‘人工智能’的问题将获得实质上的解决。”
- “在三到八年的时间里我们将得到一台具有人类平均智能的机器。”



Allen Newell

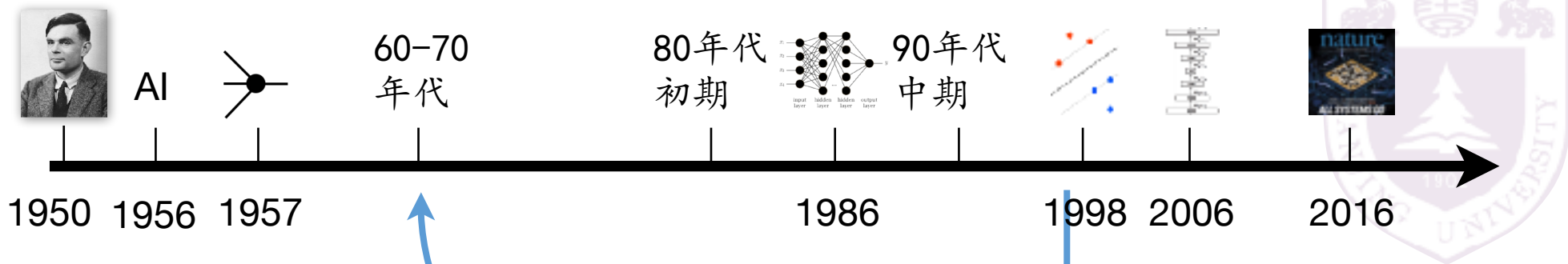


Herbert Simon



Marvin Minsky





1997 深蓝

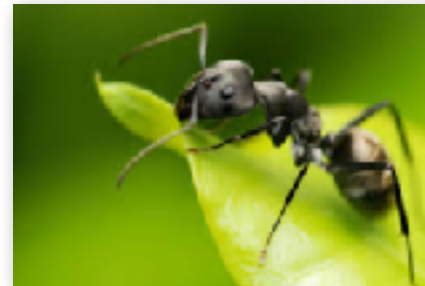


# Potted history of AI



- 1943 McCulloch & Pitts: Boolean circuit model of brain
- 1950 Turing's "Computing Machinery and Intelligence"
- 1952–69 Look, Ma, no hands!
- 1950s Early AI programs, including Samuel's checkers program, Newell & Simon's Logic Theorist, Gelernter's Geometry Engine
- 1956 **Dartmouth meeting: "Artificial Intelligence" adopted**
- 1965 Robinson's complete algorithm for logical reasoning
- 1966–74 AI discovers computational complexity  
Neural network research almost disappears
- 1969–79 Early development of knowledge-based systems
- 1980–88 Expert systems industry booms
- 1988–93 Expert systems industry busts: "AI Winter"
- 1985–95 Neural networks return to popularity
- 1988– Resurgence of probability; general increase in technical depth  
"Nouvelle AI": ALife, GAs, soft computing
- 1995– Agents, agents, everywhere . . .
- 1990– Machine learning quickly develops
- 2003– Human-level AI back on the agenda
- 2006– Machine learning industry booms

# What is intelligence?



# What is intelligence?



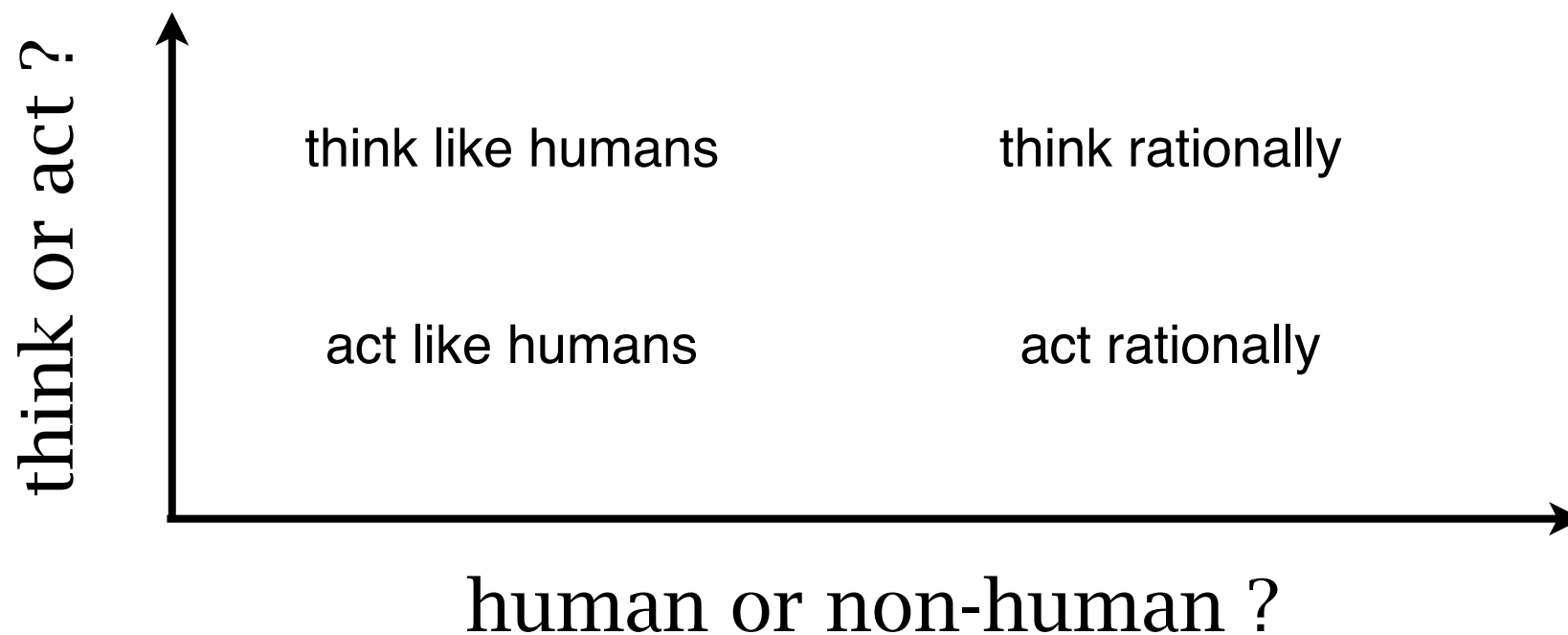
The uncertain about intelligence is a fundamental problem of AI



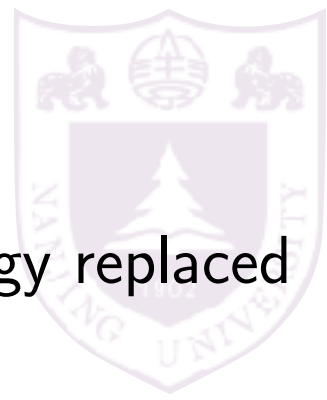
# What is AI?



AI is a system that



# Thinking humanly: Cognitive Science



1960s “cognitive revolution”: information-processing psychology replaced prevailing orthodoxy of behaviorism

Requires scientific theories of internal activities of the brain

- What level of abstraction? “Knowledge” or “circuits”?
- How to validate? Requires
  - 1) Predicting and testing behavior of human subjects (top-down)
  - or 2) Direct identification from neurological data (bottom-up)

Both approaches (roughly, Cognitive Science and Cognitive Neuroscience) are now distinct from AI

Both share with AI the following characteristic:

**the available theories do not explain (or engender) anything resembling human-level general intelligence**

Hence, all three fields share one principal direction!

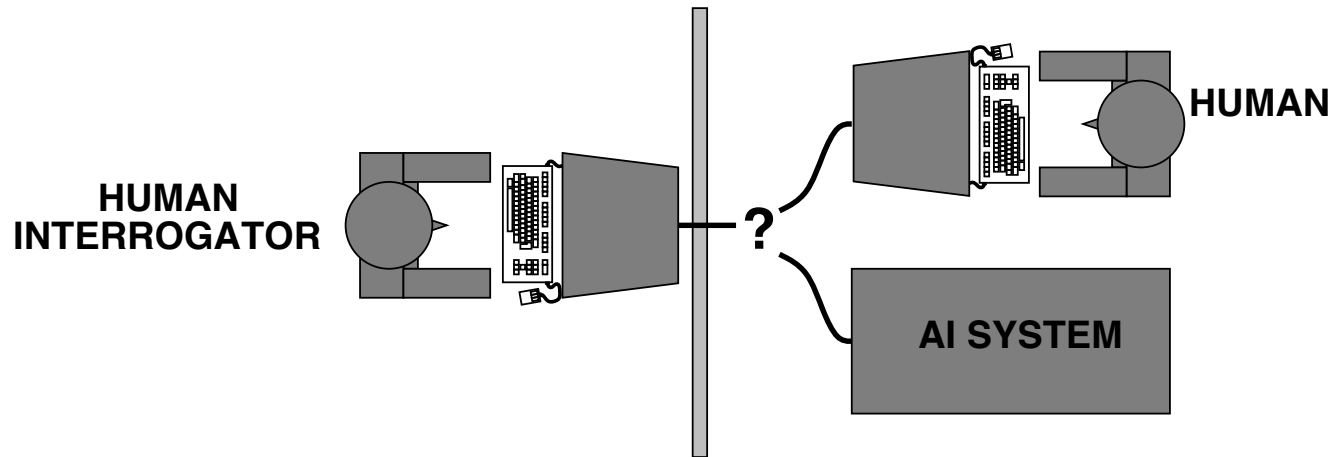


# Acting humanly: The Turing test

Turing (1950) “Computing machinery and intelligence”:

◇ “Can machines think?” → “Can machines behave intelligently?”

◇ Operational test for intelligent behavior: the **Imitation Game**



◇ Predicted that by 2000, a machine might have a 30% chance of fooling a lay person for 5 minutes

◇ Anticipated all major arguments against AI in following 50 years

◇ Suggested major components of AI: knowledge, reasoning, language understanding, learning

Problem: Turing test is not **reproducible**, **constructive**, or amenable to **mathematical analysis**

# Thinking rationally: Laws of Thought



Normative (or prescriptive) rather than descriptive

Aristotle: what are correct arguments/thought processes?

Several Greek schools developed various forms of logic:

**notation** and **rules of derivation** for thoughts;  
may or may not have proceeded to the idea of mechanization

Direct line through mathematics and philosophy to modern AI

Problems:

- 1) Not all intelligent behavior is mediated by logical deliberation
- 2) **What is the purpose of thinking?** What thoughts **should** I have out of all the thoughts (logical or otherwise) that I **could** have?



# Acting rationally



**Rational** behavior: doing the right thing

The right thing: that which is expected to maximize goal achievement, given the available information

Doesn't necessarily involve thinking—e.g., blinking reflex—but thinking should be in the service of rational action

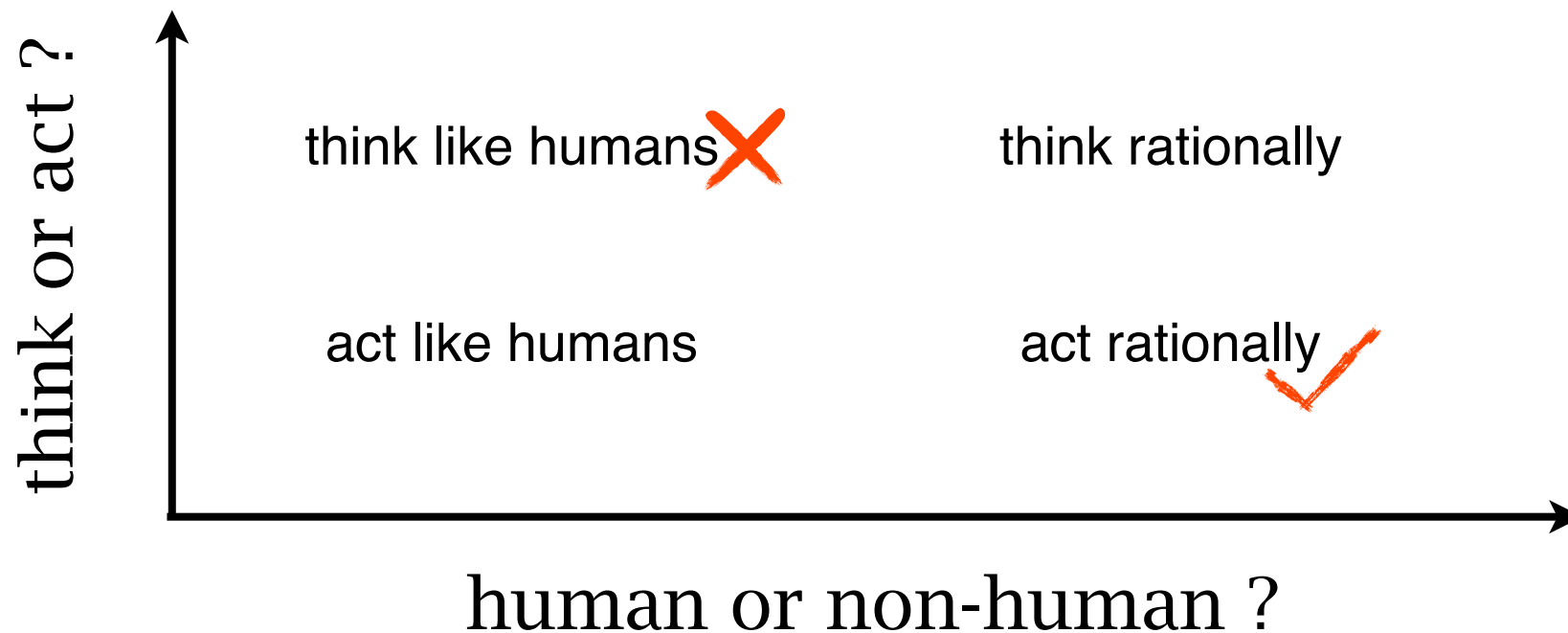
Aristotle (Nicomachean Ethics):

**Every art and every inquiry, and similarly every action and pursuit, is thought to aim at some good**

# What is AI?



AI is a system that





AI IS BLOOMING

HOPE YOU ENJOY

THANK YOU ALL!