# 知识图谱不是图谱

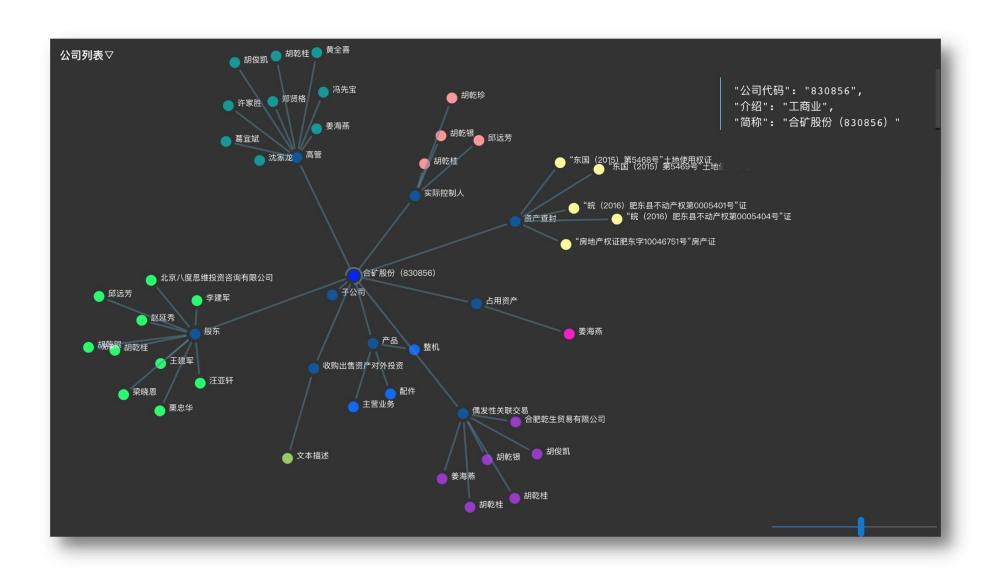
——把智能放在数据而非代码里

鲍捷

2019-03-27✓ 北京文因互联科技有限公司

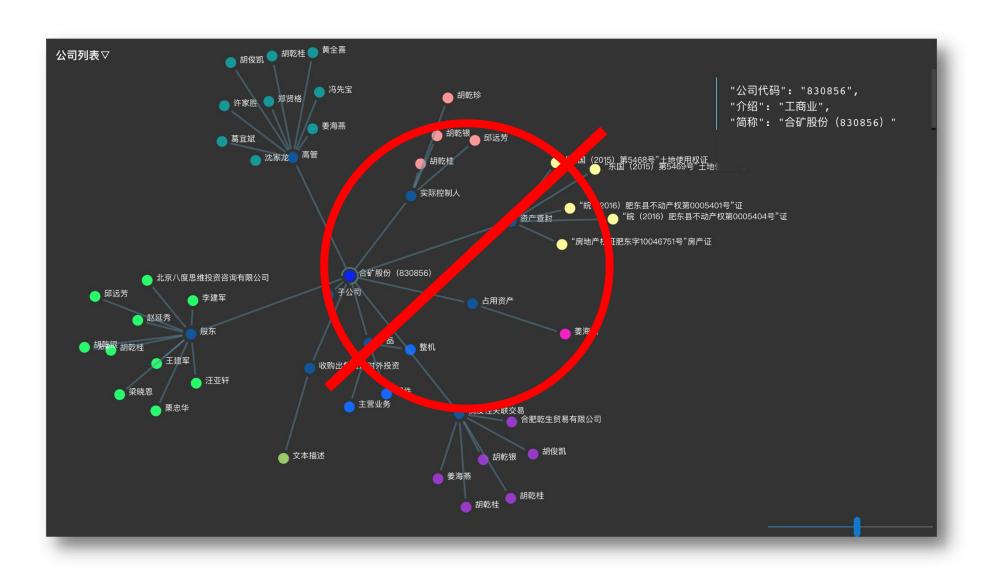
# 图谱





# 知识图谱 ≠图谱







# Semantics Meaning Relation Structure



# Relation gives meaning



# Semantics ≈ Structure



# Knowledge



# Structure that may generate structure

可以产生结构的结构



# Knowledge graph



# Machine readable structure that gives "meaning"

机器可读的定义"意义"的数据结构



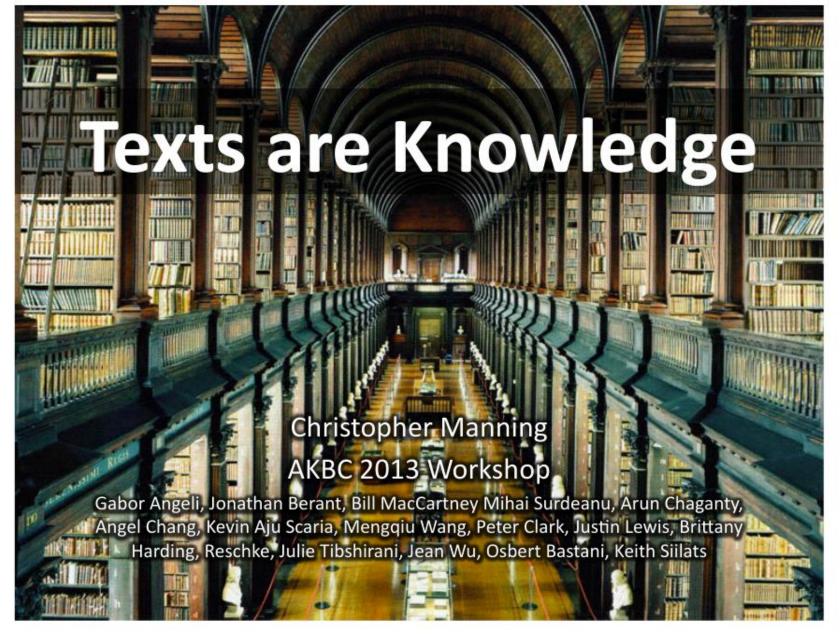
# Knowledge graph

 $\approx$ 

Machine readable "dictionary"

机器可读的"字典"





知识结构怎么来?

# 结构化数据的处理方式



图书馆学:标签

软件工程:一维表、哈希表、树等

数据工程:二维表

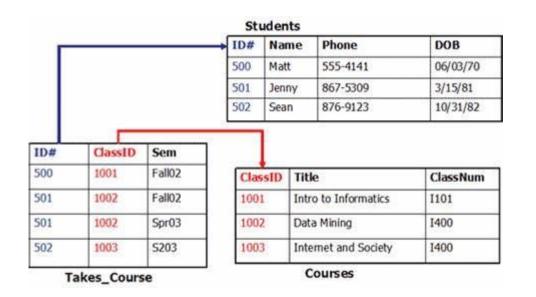
知识工程:剩下的 (那些难的)

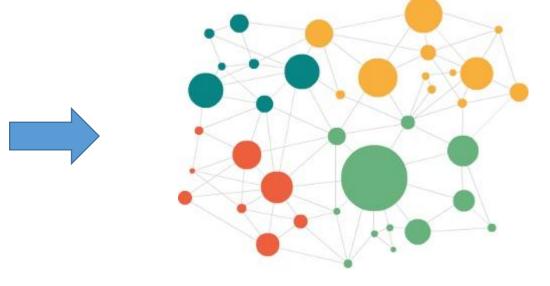


知识工程处理那些能产生数据的数据

# "数据库"和知识图谱有啥不同?







表间关系 列间关系 行间关系 产间关系 全型性约 完整性约 条件触发



知识图谱的真正威力:

把智能从代码里挪到数据里

# 把智能从代码里挪到数据里



DOS=HIGH, UMB

DEVICE=C:\DOS\HIMEM.SYS

DEVICE=C:\DOS\EMM386.EXE RAM

DEVICEHIGH=C:\DOS\ANSI.SYS

FILES=30

SHELL=C:\DOS\COMMAND.COM C:\DOS /E:512 /P

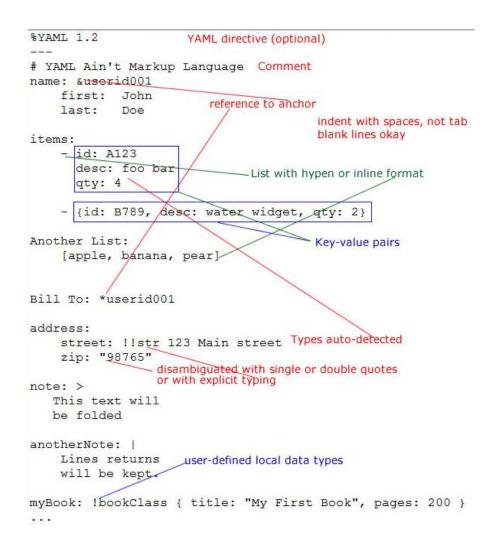
所有的知识表现语言开始都像是个配置文件

# 把智能从代码里挪到数据里



#### 从配置文件到知识语言

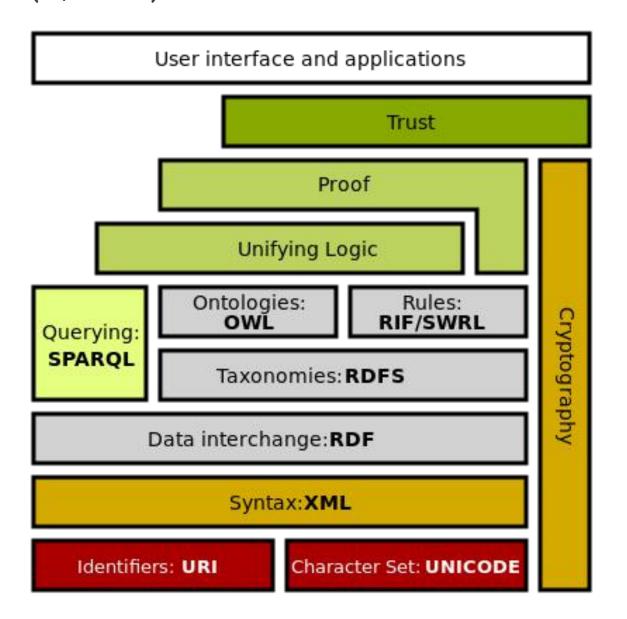
- 配置的配置
- 增强模块化和重用性
- 增强节点命名
- 增强链接
- 规则的需求自然出现



所有的知识表现语言开始都像是个配置文件

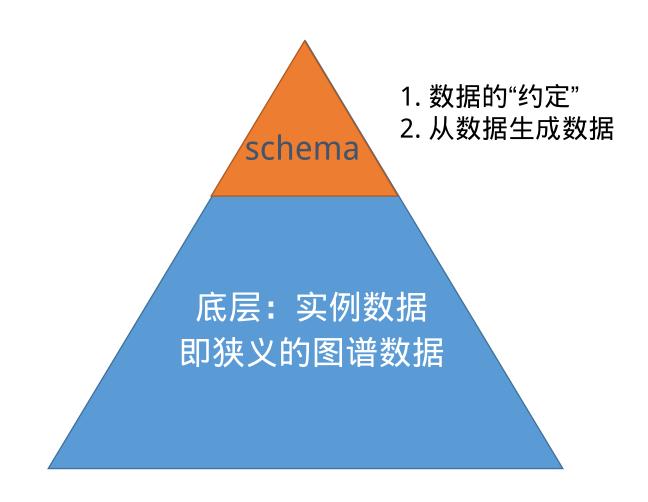
# 知识图谱的(官方)层级





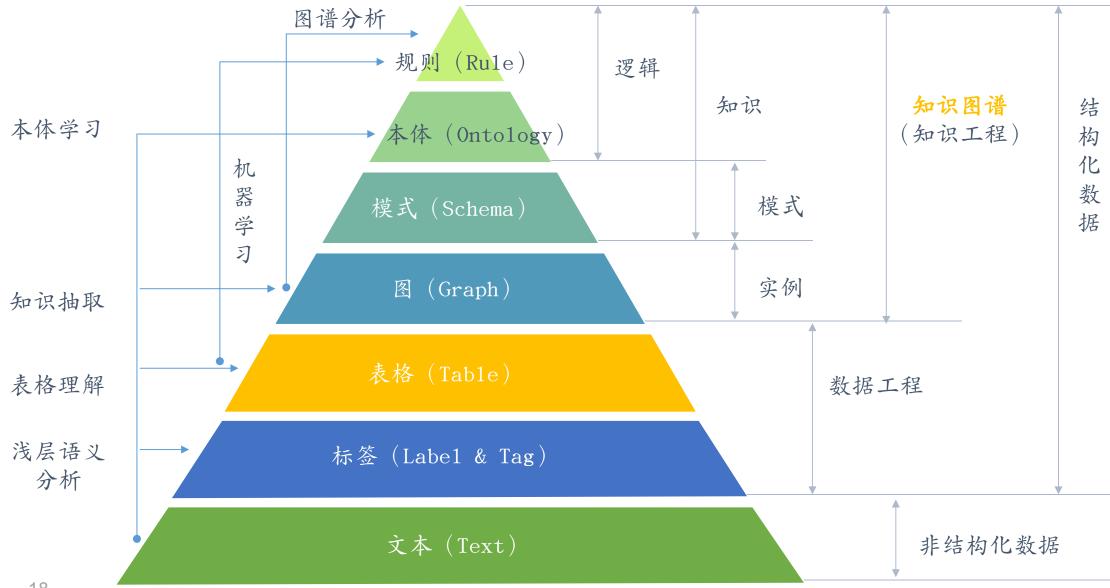
# 知识图谱的(抽象简化)层级





#### 知识图谱技术金字塔





# 知识图谱的"基本原则"



- 1. 数据与应用的分离
- 2. 数据的自描述性
- 3. 数据的弱/无模式性
- 4. 数据的可演进性



"Artificial"
Intelligence

Augmented Intelligence







机器学习(含深度学习)转移智能

人机交互创造智能

社会要富裕,光靠金融不行,要工业农业,把财富创造出来。智能系统也一样,不管是训练样本还是知识规则,都要靠人来产生。



#### RODNEY BROOKS Robots, AI, and other stuff

BLOG MIT y





Search ...

#### RECENT POSTS

A Better Lesson

Predictions Scorecard, 2019 January 01

[FoR&AI] Steps Toward Super Intelligence IV, Things to Work on Now

[FoR&AI] Steps Toward Super Intelligence III, Hard Things Today

[FoR&AI] Steps Toward Super Intelligence II, Beyond the Turing POST: A BETTER LESSON

MARCH 19, 2019 — REVIEWS

A Better Lesson J f



rodneybrooks.com/a-better-lesson/

Just last week Rich Sutton published a very short blog post titled The Bitter Lesson. I'm going to try to keep this review shorter than his post. Sutton is well known for his long and sustained contributions to reinforcement learning.

In his post he argues, using many good examples, that over the 70 year history of Al, more computation and less built in knowledge has always won out as the best way to build Artificial Intelligence systems. This resonates with a current mode of thinking among many of the newer entrants to AI that it is better to design learning networks and put in massive amounts of computer power, than to try to design a structure for computation that is specialized in any way for the task. I must say, however, that at a two day work shop on Deep Learning last week at the National Academy of Science, the latter idea was much more in vogue, something of a backlash against exactly what Sutton is arguing.

I think Sutton is wrong for a number of reasons.

1. One of the most celebrated successes of Deep Learning is image labeling.





image by neilsethlevine.com

## Intelligence at the Interface

Semantic Technology and the Consumer Internet Experience

Tom Gruber Stealth-Company.com





MIT, Computer Science and Artificial Intelligence Laboratory The Stata Center, Room G592 32 Vassar St. Cambridge, MA 02139 (617)-258-6167 (office) karger@mit.edu @karger http://people.csail.mit.edu/karger/

UROPs: My group often has openings for urops. You can find a list of ongoing projects here. If you find Students one of them interesting, take a look at the relevant papers, then send me a resume and (informal) transcript. Urops should have completed 6.01 or equivalent; superurops 6.005 or equivalent.

I am a member of the Computer Science and Artificial Intelligence Laboratory in the EECS department at MIT.

My primary interest these days is in developing tools that help individuals manage information better. This involves studying people and current tools to understand where the problems are, creating and evaluating tools that address those problems, and deploying those tools to learn how people use them and iterate the whole process. I draw on whatever fields can help: information retrieval, machine learning, databases, and algorithms, but most often human computer interaction.

- Amy Zhang
- Lea Verou
- Tarfah Alrashed
- Nadiia Chepurko
- Farnaz Jahanbakhsh
- Soya Park
- Luke Murray
- Jumana Almahmoud
- Alumni



# Web3.0的核心是一场HCI的革命

结构化数据(知识图谱)是这场革命的催化剂



知识图谱的突破点可能在HCI

不在于知识提取或者知识表现

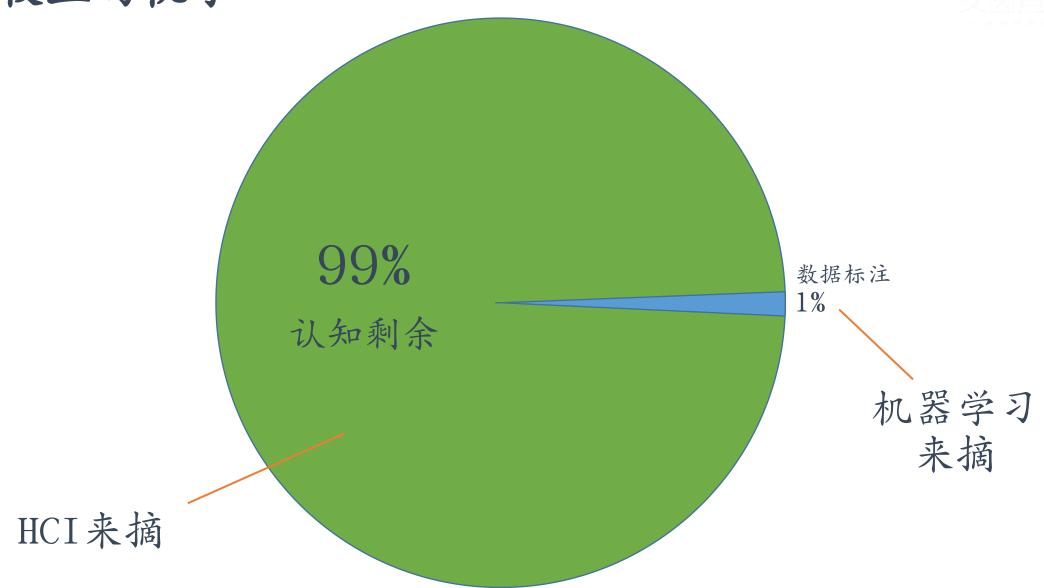


# 知识图谱人工智能的突破点可能在HCI

不在于知识提取或者知识表现

# "低枝上的桃子"





### 知识图谱的交互层



- 检索式 (结构查找)
  - 查询 (计算机语言检索)
  - 搜索 (关键词检索)
  - 问答 (自然语言检索)
  - 探索 (人机结合检索,如分面浏览和电子表格) ◀ 机会点
- 响应式 (规则激发)
  - 推送 (条件激发内容)
  - 约束检查 (如合规检查)
  - 验证执行(如智能合约)
- 图形式 (具象化)
  - 节点-边型图可视化
  - BI (图表与自动化报告)
  - 卡片式总结 (如infobox, factbox)
  - 增强现实(如智能眼镜)



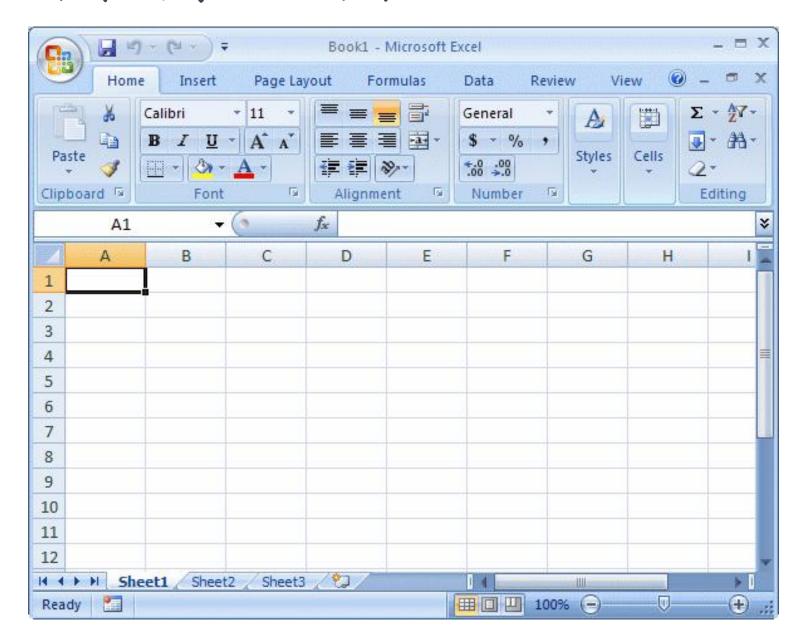
问答引擎

探索引擎



# 最成功的探索引擎:电子表格





# 探索引擎



- 人机结合解决问题
- 提升数据的可发现性
- 和用户的交互中on-demand地更新和提取知识
- 把智能放在数据而非代码中
  - 智能可以积累
  - 界面可以演化

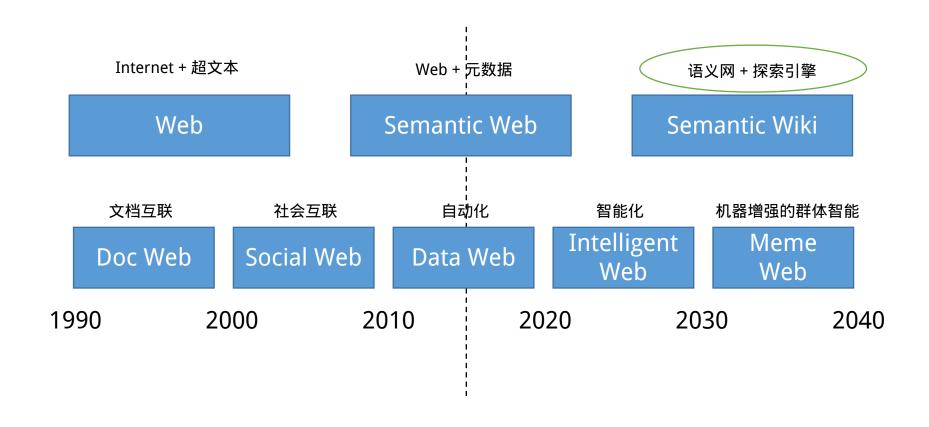
# 需要重点关注的未来探索技术



- 分面浏览/编辑器 ◀ 机会点
- · 本体编辑器 Palantir的发家技术
- 语义wiki
- · VR中的信息流

# Web的下半场





# 谢谢!

个人微信:baojie\_memect

2019年3月

文因互联