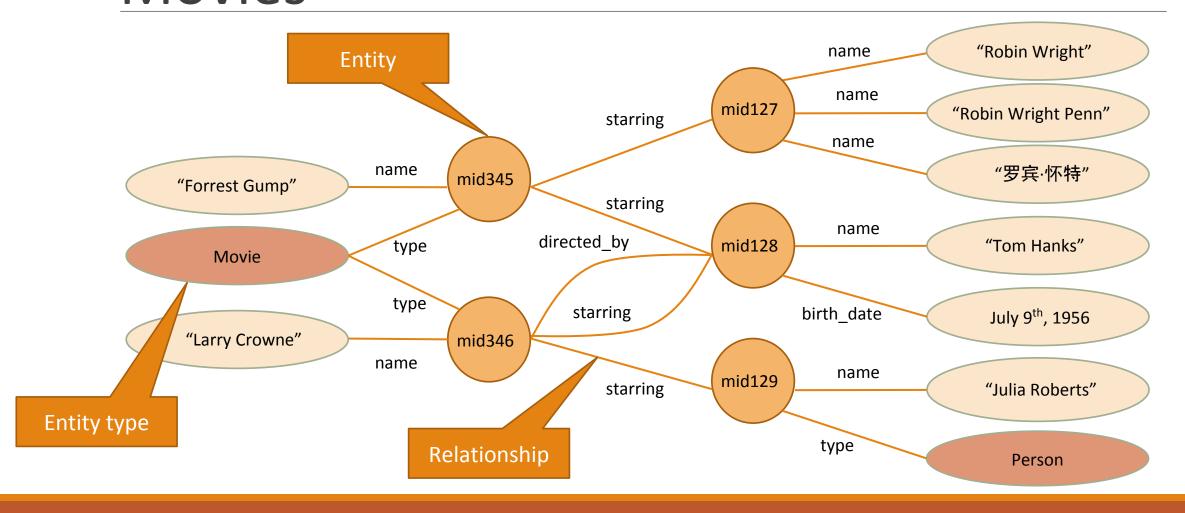
Challenges and Innovations in Building a Product Knowledge Graph

XIN LUNA DONG, AMAZON JANUARY, 2018

Product Graph vs. Knowledge Graph

Knowledge Graph Example for 2 Movies

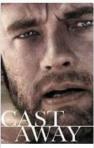


Knowledge Graph in Search

Tom Hanks > Movies



Forrest Gump 1994



Cast Away 2000



Saving Private Ryan 1998



Captain Phillips 2013



Big

1988



Sully

2016

Toy Story 1995



ory Catch Me If You Can 2002



Apollo 13 1995

List of Tom Hanks performances - Wikipedia

https://en.wikipedia.org/wiki/List_of_Tom_Hanks_performances ▼ Jump to Film - The Simpsons Movie, 2007, Yes, Himself, Cameo Voice role. Mamma Mia! 2008, Yes, −, Executive producer. City of Ember, 2008, Yes, −.

A Hologram for the King (film) · Big (film) · Larry Crowne · He Knows You're Alone

Tom Hanks (@tomhanks) · Twitter

https://twitter.com/tomhanks

And don't miss this songstress at the famous Cafe Carlyle. Through Saturday nite! Hanx @RitaWilson pic.twitter.com/J70XJbf...

12 hours ago · Twitter

Beware! Crass self-serving Social Media Post! This book goes on sale tomorrow! Hanx pic.twitter.com/V2EqPKL...

16 hours ago · Twitter

Lost (g)love. Looking for a mate. Good luck. Hanx. pic.twitter.com/ApH7rEG...

1 day ago · Twitter

Tom Hanks

American actor



Thomas Jeffrey Hanks is an American actor and filmmaker. He is known for his various comedic and dramatic film roles, including Splash, Big, Turner & Hooch, A League of Their Own, Sleepless in Seattle, ... Wikipedia

Born: July 9, 1956 (age 61), Concord, CA

Awards: Academy Award for Best Actor, MORE

Spouse: Rita Wilson (m. 1988), Samantha Lewes (m. 1978-1987)

TV shows: Bosom Buddies, Celebrity Jeopardy!, MORE

Knowledge Graph in Personal Assistant

Alexa, play the music by Michael Jackson





List of officially released compilations and

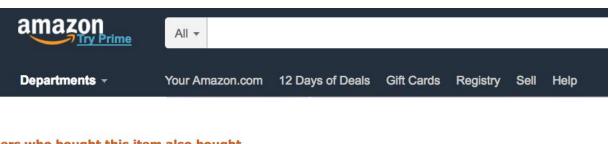
[92][93][94][95]

- Portrait of Michael Jackson / Portrait of Jackson 5 (1973)
- Os Grandes Sucessos, Vol. 2 (1980)
- Motown Superstar Series, Vol. 7 (1980)
- Superstar (1980)
- Michael Jackson & The Jackson 5 (1983)
- Ain't No Sunshine (1984)
- The Great Love Songs of Michael Jackson (1984)
- Ben / Got to Be There (1986)
- Looking Back to Yesterday (1986)
- The Original Soul of Michael Jackson (1987)
- Rockin' Robin (1993)
- Dangerous The Remix Collection (1993)
- Michael Jackson Story (1996)
- Master Series (1997)
- Ghosts Deluxe Collector Box Set (1997)
- Got to Be There / Forever, Michael (1999)
- Bad / Thriller (2000)
- Forever, Michael / Music & Me / Ben (2000)
- Classic The Universal Masters Collection (2001)

Product Graph

■Mission: To answer any question about products and related knowledge in the world





Customers who bought this item also bought







Step2 Push Around Sport Buggy ★★★★☆ 18 \$49.99 \prime

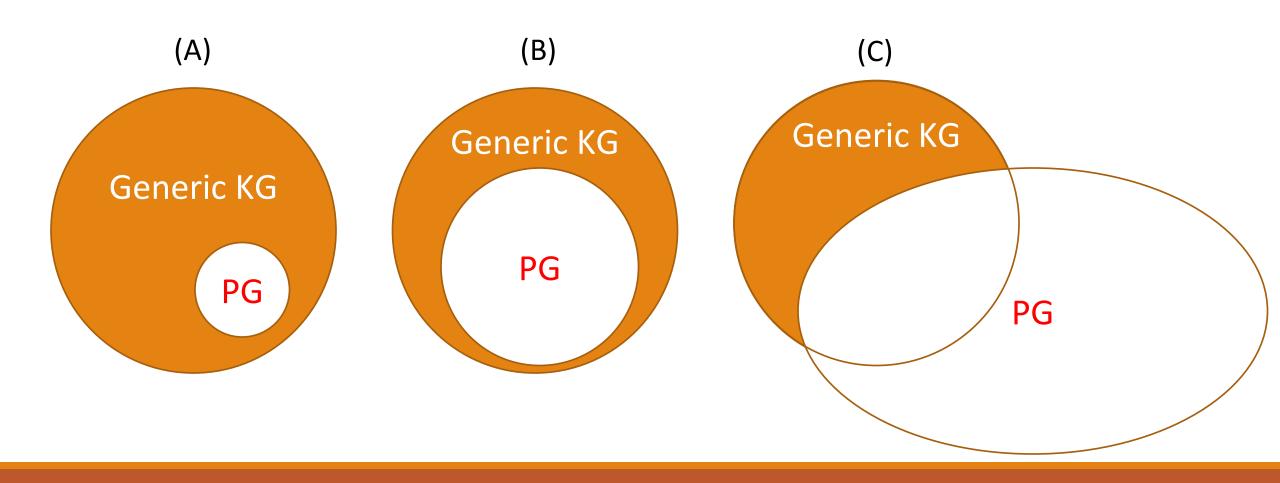


GOOD NIGHT, LIGHTNIN > RH Disney **全全全**章 299 Board book \$7.70 vprime

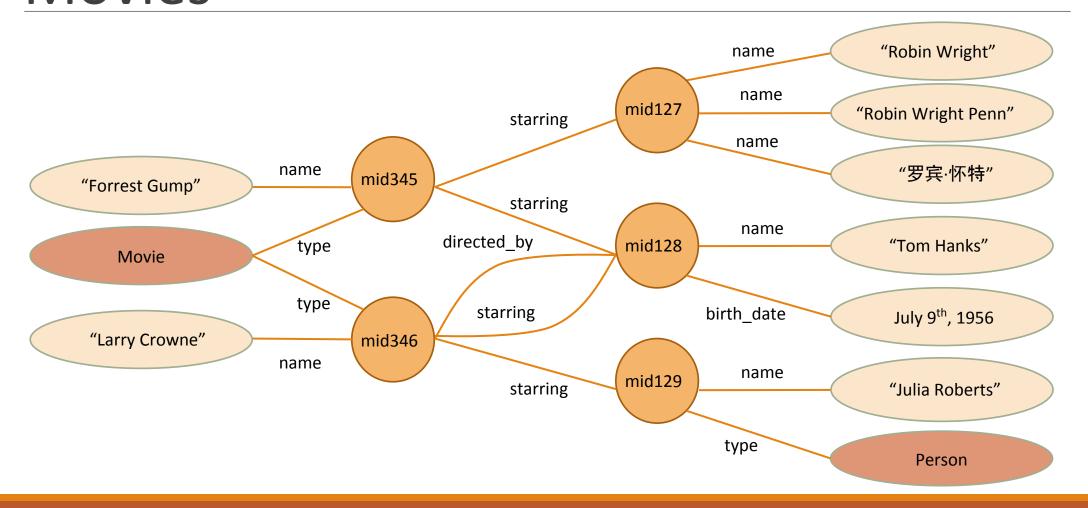


Prime Day?"

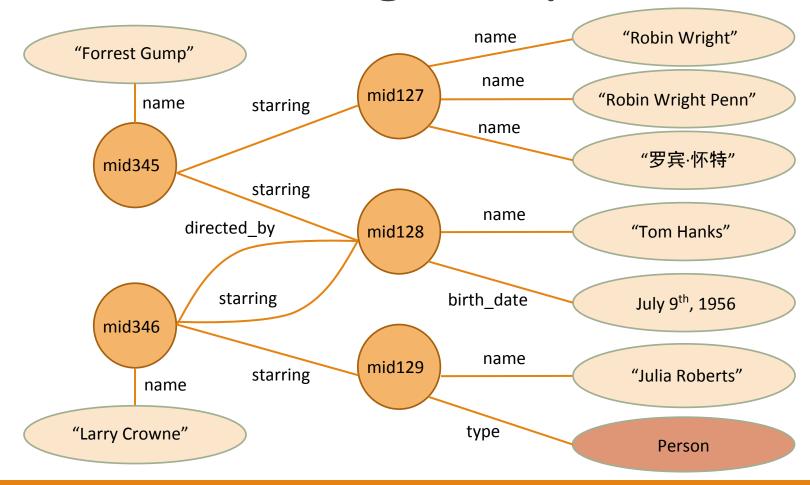
Product Graph vs. Knowledge Graph



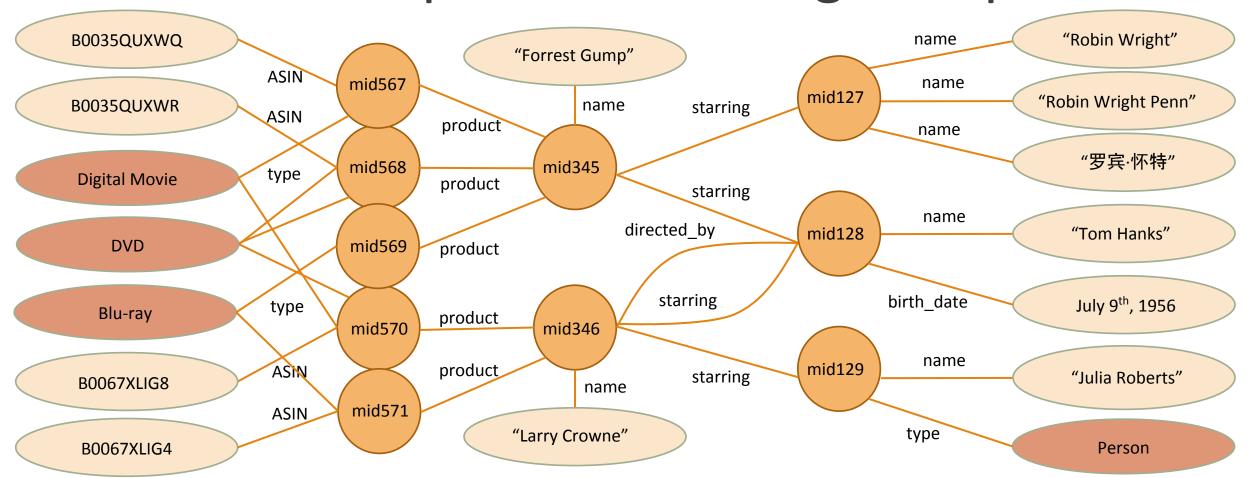
Knowledge Graph Example for 2 Movies



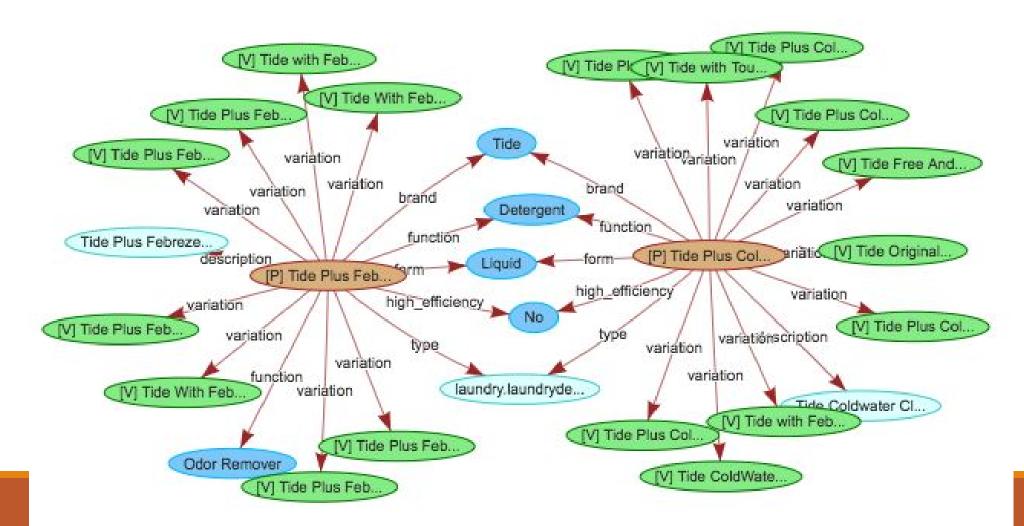
Product Graph vs. Knowledge Graph



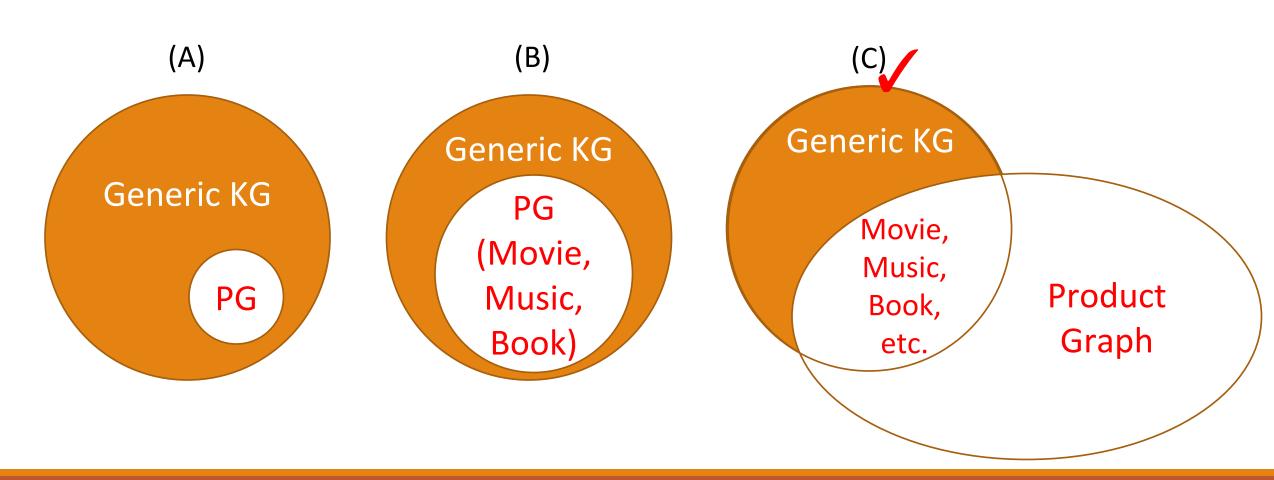
Product Graph vs. Knowledge Graph

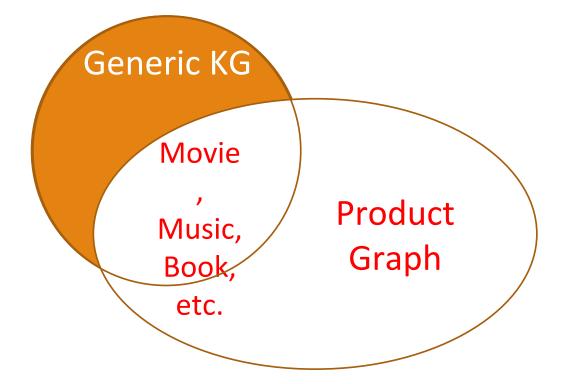


Another Example of Product Graph



Knowledge Graph vs. Product Graph





But, Is The Problem Harder?

Challenges in Building Product Graph I

- ■No major sources to curate product knowledge from
 - Wikipedia does not help too much
 - ☐ A lot of structured data buried in text descriptions in Catalog
 - Retailers gaming with the system so noisy data

Challenges in Building Product Graph II

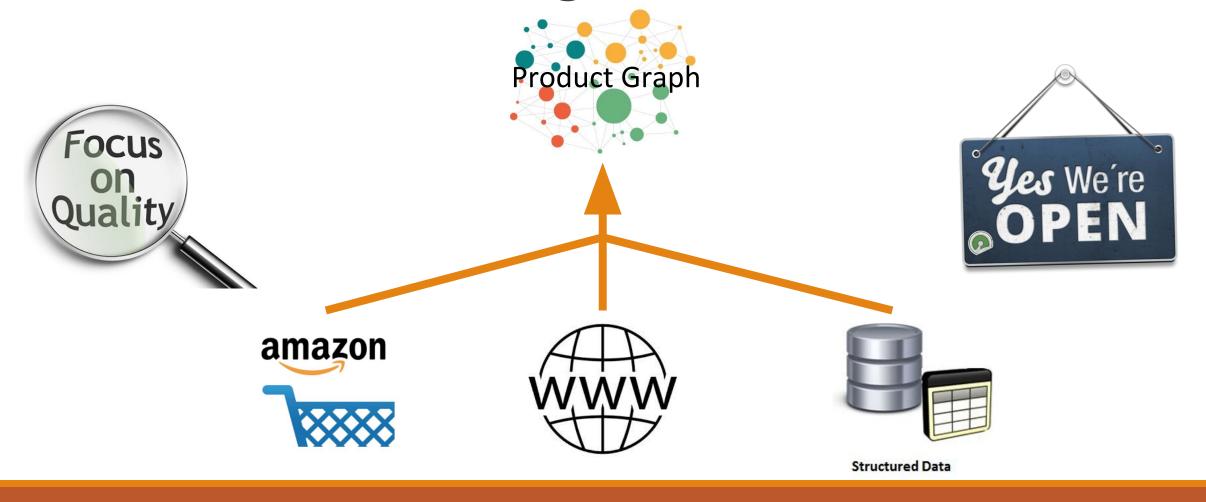
- Large number of new products everyday
 - Curation is impossible
 - ☐ Freshness is a big challenge

Challenges in Building Product Graph III

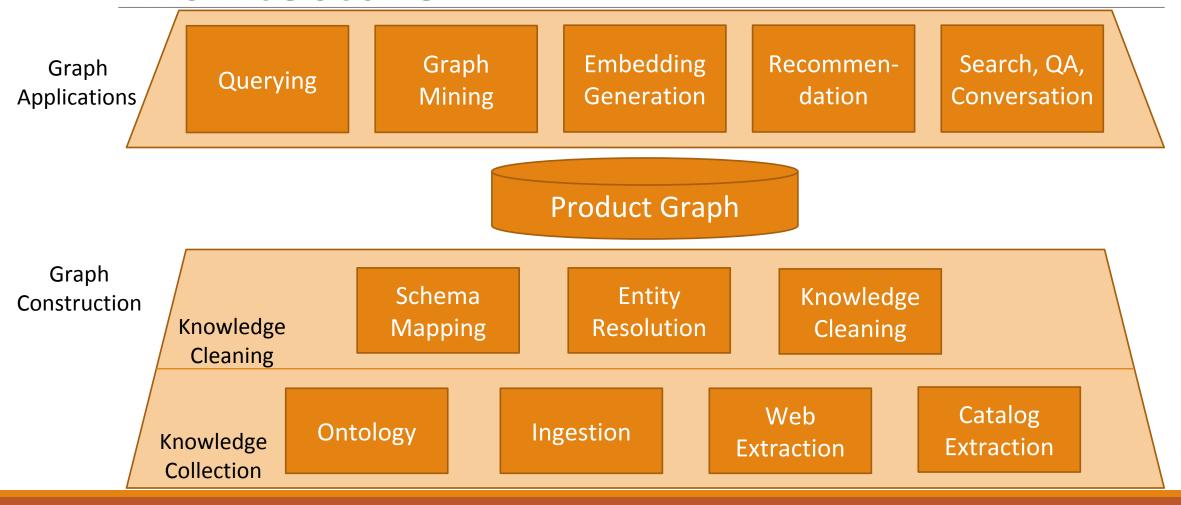
- ☐ Large number of product categories
 - ■A lot of work to manually define ontology
 - Hard to catch the trend of new product categories and properties

How to Build a Product Graph?

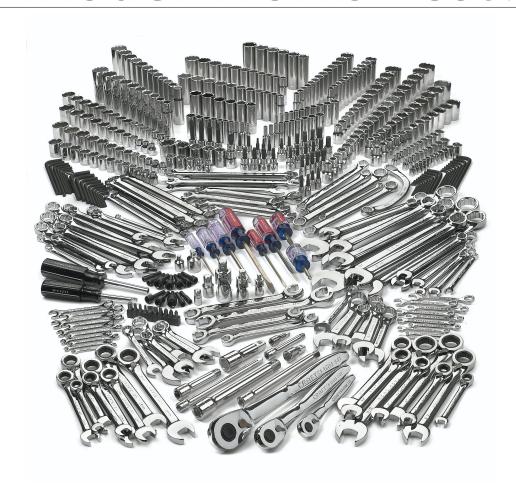
Where is Knowledge from?



Architecture



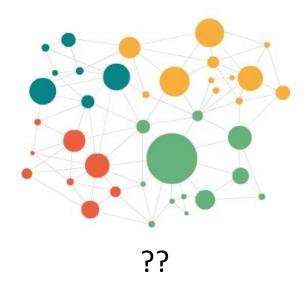
Which ML Model Works Best?



Which ML Model Works Best?

ID	NAME	CLASS	MARK	SEX	
1	John Deo	Four	75	female	
2	Max Ruin	Three	85	male	
3	Arnold	Three	55	male	
4	Krish Star	Four	60	female	
5	John Mike	Four	60	female	
6	Alex John	Four	55	male	
7	My John Rob	Fifth	78	male	
8	Asruid	Five	85	male	
9	Tes Qry	Six	78	male	
10	Big John	Four	55	female	

Tree-based models



SCENE FROM "DAN'L DRUCE."

This interesting domestic drama, by Mr. W. S. Gilbert, has continued to engage the sympathies of a nightly sufficient audience at the Haymarket Theatre, where it has now been represented more than sixty times. Its subject and character were described by us, in the ordinary report of theatrical novelties, about two months ago. Our readers will probably not need to be reminded that the hero of the story, Dan'l Druce, the blacksmith, is a so'livary recluse dwelling on the coast of Norfolk, where his lone cottage is visited by fugitives from party v. ngeance during the civil wars of the Commonwealth. His hoard of money is stolen; but a different sort of treasure, a helpless female infant, is left by some mysterious agency, and may be accepted, as in George Eliot's tale of "Silas Marner," for a Divine gift to the sad-hearted misanthrope, far better than riches. In this spirit, at least, he is content to receive the precious human charge; and so to those who would remove it from his home, Dan'l Druce here makes answer with the solemn exclamation, "Touch not the Lord's gift!" This character is well acted by Mr. Hermann Vezin.







Neural network

Research Philosophy



Moonshots: Strive to apply and invent the state-of-the-art

Roofshots: Deliver incrementally and make production impacts

I. Extracting Knowledge from Semi-Structured Data on the Web





Aamir Khan is receiving rave reviews for Dangal.

Rating: 4/5

f Dangal Cast: Aamir Khan, Sakshi Tanwar, Far Khurrana, Sanya Malhotra Director: Nitesh Tiwari

卧虎藏龙 臥虎藏龍 (2000)



导演: 李安

编剧: 王蕙玲 / 詹姆斯·夏慕斯 / 蔡国荣

主演: 周润发 / 杨紫琼 / 章子怡 / 张震 / 郑佩佩 /

更多...

类型: 剧情 / 动作 / 爱情 / 武侠 / 古装

制片国家/地区: 台湾 / 香港 / 美国 / 中国大陆

语言: 汉语普通话

上映日期: 2000-10-13(中国大陆) / 2000-05-16 (戛纳电影节) / 2000-07-07(台湾) / 2000-07-13

(香港) / 2001-01-12(美国)

片长: 120 分钟

又名: Crouching Tiger, Hidden Dragon

IMDb链接: tt0190332



想看 看过

评价: 公公公公公

♡ 写短评 ⊘ 写影评 + 提问题 分享到 ▼

推荐

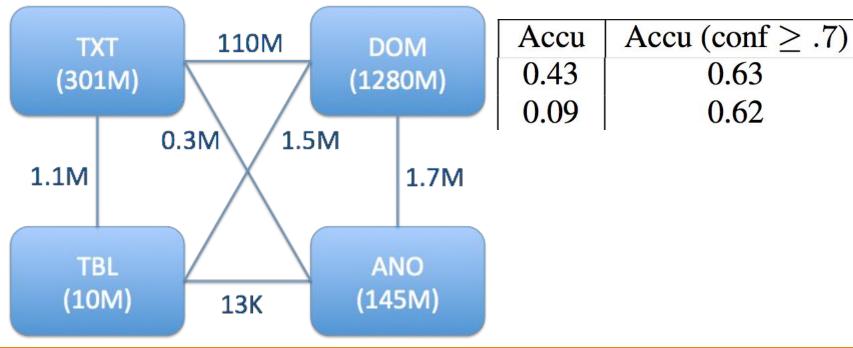
卧虎藏龙的剧情简介 · · · · · ·

一代大侠李慕白(周润发饰)有退出江湖之意,托付红颜知己俞秀莲(杨紫琼饰)将青冥剑转交给贝勒爷(郎雄饰)收藏,不料当夜遭玉娇龙(章子怡)窃取。俞秀莲暗中查访也大约知道是玉府小姐玉蛟龙所为,她想办法迫使玉蛟龙归还宝剑,免伤和气。但李慕白发现了害死师傅的碧眼狐狸(郑佩佩饰)的踪迹,她隐匿于玉府并收玉蛟龙为弟子。而玉蛟龙欲以青冥剑来斩断阻碍罗小虎(张震饰)的枷锁,他们私定终身。关系变得错综复杂,俞秀莲和李慕白爱惜玉蛟龙人才难得,苦心引导,但玉蛟龙却使性任气不听劝阻…… ⑤豆瓣

I. Extracting Knowledge from Semi-Structured Data on the Web

Knowledge Vault @ Google showed big potential from DOM-tree extraction [Dong et al., KDD'14][Dong et al., VLDB'14]

Accu	Accu (conf \geq .7)
0.36	0.52



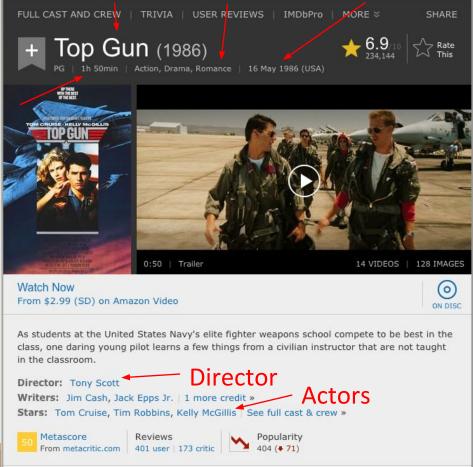
0.63

0.62

I. Extracting Knowledge from Web—Annotation-Based DOM Extraction

Title Genre Release Date





Extracted relationships

- (Top Gun, type.object.name, "Top Gun")
- (Top Gun, film.film.genre, Action)
- (Top Gun, film.film.directed_by, Tony Scott)
- (Top Gun, film.film.starring, Tom Cruise)
- (Top Gun, film.film.runtime, "1h 50min")
- film.film.release_Date_s,
 "16 May 1986")



Annotation-based knowledge extraction

I. Extracting Knowledge from Web—Annotation-Based DOM Extraction

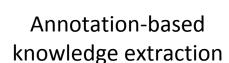


Annotation-based knowledge extraction



Alexa, When did Padme Amidala die? What model is R2D2? Who is Luke Skywalker's master? Where is Boba Fett from? Who is Darth Vader's apprentice?



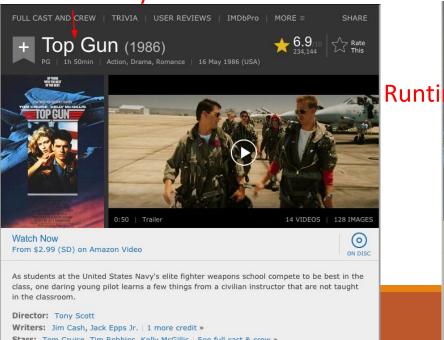


Entity Identification Automatic Annotation

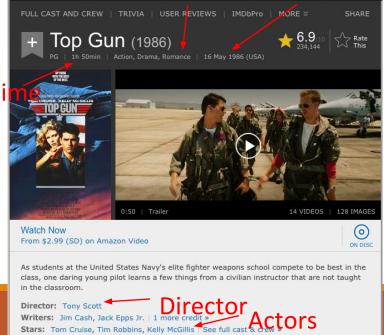
Training

Automatic Label Generation

Movie entity



Genre Release Date



Extracted triples

- (Top Gun, type.object.name, "Top Gun")
- (Top Gun, film.film.genre, Action)
- (Top Gun, film.film.directed_by, Tony Scott)
- (Top Gun, film.film.starring, Tom Cruise)
- (Top Gun, film.film.runtime, "1h 50min")
- (Top Gun, film.film.release_Date_s, "16 May 1986")

- Extraction on IMDb
- 1. Very high extraction precision
- 2. Extracting triples with new entities

Predicate	Precision	Recall
Type.object.name ("name")	1	1
People.person.place_of_birth	1	1
Common.topic.alias	1	1
Film.actor.film	0.98	0.47
Film.director.film	0.98	0.91
Film.producer.film	0.89	0.57
Film.writer.film	0.96	0.60

Predicate	Precision	Recall
Type.object.name ("title")	0.97*	0.97*
Tv.tv_series_episode.episode_number	1	1
Tv.tv_series_episode.season_number	1	1
Film.film.directed_by	0.99	1
Film.film.written_by	1	0.98
Film.film.genre	0.90*	1
Film.film.starring	1	0.97
Tv.tv_series_episode.series	1	1

^{*}Ground truth is incomplete. Manual inspection suggests close to 100% accuracy.

Extraction experiments on http://swde.codeplex.com/ (2011)

	Title		Director(s		Genre(s)	
Site	Р	R	P	R	P	R
allmovies	1	1	1	1	0.71	0.96
amctv	1	1	0.98	0.97	0.95	0.91
boxofficemojo	1	1	1	0.98	0.67*	0.91
hollywood	1	1	0.94	1	1	0.97
iheartmovies	1	1	1	1	1	1
IMDB	1	1	1	0.98	1	1
metacritic	1	1	1	1	1	1
MSN	1	1	1	1	1	1
rottentomatoes	1	1	1	1	1	0.91
yahoo	1	1	1	0.99	0.99	0.94

I. Distantly Supervised DOM Extraction Which ML Model Works Best?

- □Logistic regression: Best results (20K features on one website)
- Random forest: lower precision and recall

I. Extracting Knowledge from Semi-Structured Data on the Web











OpenIE

DOM extraction

Nearly-automatic interactive extraction on any new vertical





Annotation-based knowledge extraction

II. Extracting Knowledge from Product Profiles in Amazon Catalog

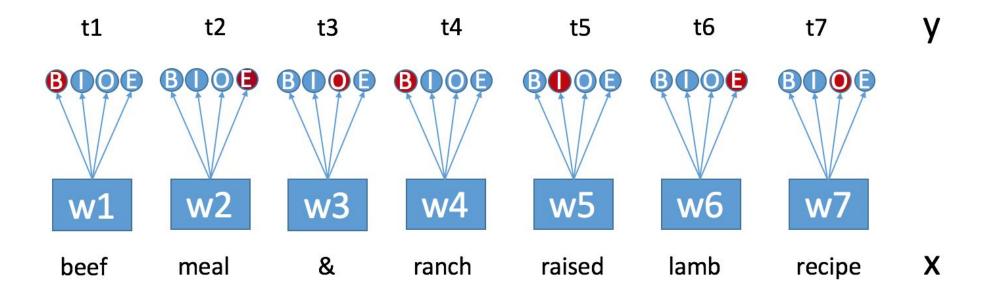
name	form	scent
Tide Detergent with Febreze Freshness		
Gain Apple Mango Tango Liquid Laundry Detergent		
Gain Joyful Expression Powder Detergent		
Tide PODS Original Scent HE Turbo Laundry Detergent Pacs 81-load Tub		
Tide PODS Free & Gentle HE Turbo Laundry Detergent Pacs 35-load Bag		

II. Open Attribute Extraction by Named Entity Recognition

- Beginning of entity
- Inside of entity
- Outside of entity
- End of entity

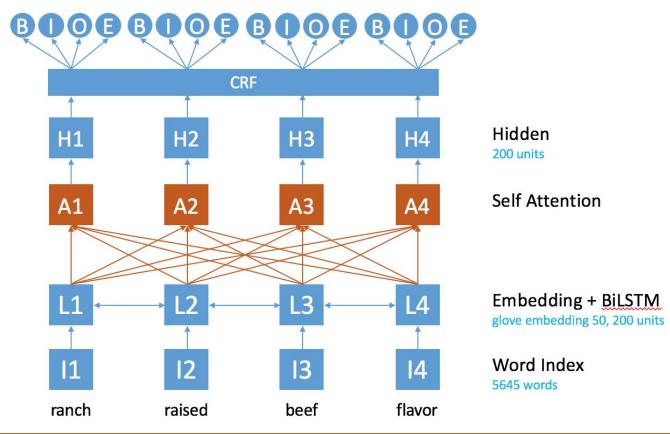
x={w1,w2,...,wn} input sequence

y={t1,t2,...,tn} tagging decision



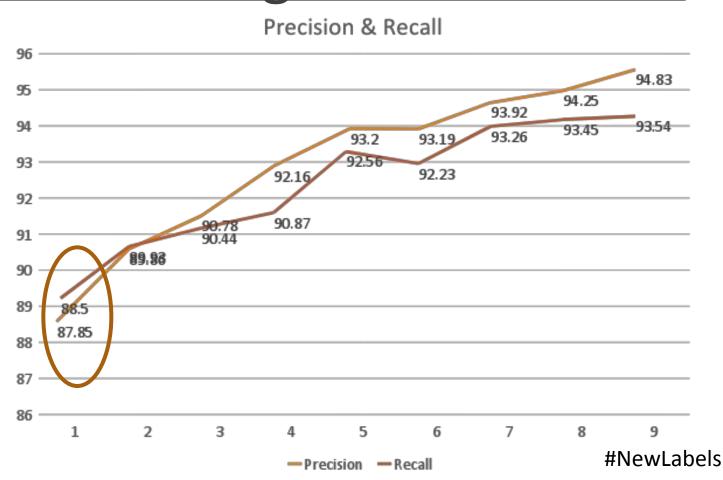
II. Open Attribute Extraction by NER—Which ML Model Works Best?

Recurrent Neural Network, CRF, Attention

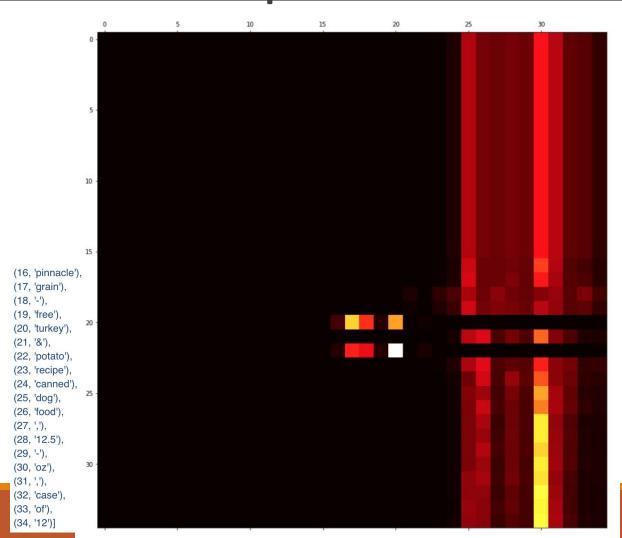


II. Open Attribute Extraction by NER—Adding Active Learning

500 Sentences **7927 Words Training** 944 Flavors 600 Sentences **7896 Words Testing** 786 Flavors Different flavors from Training data



II. Open Attribute Extraction by NER—Attentions Help Find Contexts



II. Extracting Knowledge from Product Profiles in Amazon Catalog







Open aspect extraction



Review extraction & sentiment analysis

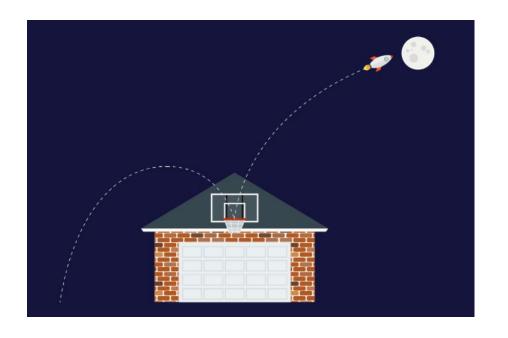


Product profile extraction



Take Aways

- ■We aim at building an authoritative knowledge graph for all products in the world
- ■We shoot for roofshot and moonshot goals to realize our mission
- There are many exciting research problems that we are tackling



Thank You!