

移动计算时代NLP研究的 挑战和机遇

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Outline

- 移动计算时代的来临
 - 摘自Mary Meeker 的Internet Trends报告
- 移动计算的技术特点
- NLP相关的应用分析
 - SIRI
 - Word Lens
 - Google Now
- Compact Data Structures for Mobiles
- 结束语

移动计算时代的来临

INTERNET TRENDS

D10 CONFERENCE
5/30/2012

Mary Meeker



<http://www.kpcb.com/file/kpcb-internet-trends-2012>

移动设备的普及

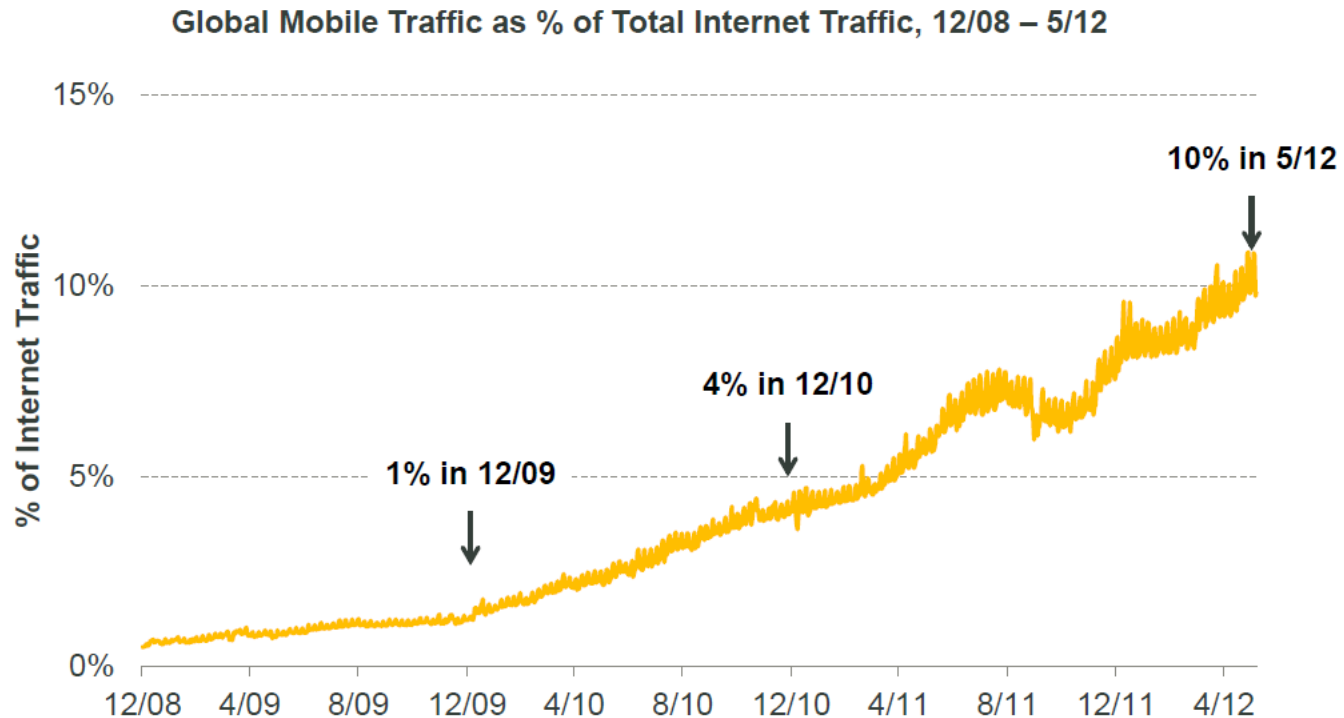
1.1B Global Mobile 3G Subscribers, 37% Growth, Q4 – @ Only 18% of Mobile Subscribers

Rank	Country	CQ4:11 3G Subs (MM)	3G Penetr ation	3G Sub Y/Y Growth	Rank	Country	CQ4:11 3G Subs (MM)	3G Penetr ation	3G Sub Y/Y Growth
1	USA	208	64%	31%	16	Canada	16	62%	34%
2	Japan	122	95	9	17	Taiwan	14	48	17
3	China	57	6	115	18	South Africa	13	21	49
4	Korea	45	85	10	19	Turkey	13	20	62
5	Italy	44	51	25	20	Portugal	13	78	19
6	UK	42	53	25	21	Vietnam	12	11	358
7	Brazil	41	17	99	22	Mexico	11	11	55
8	India	39	4	841	23	Malaysia	10	27	7
9	Germany	38	36	23	24	Sweden	10	73	25
10	Spain	33	57	21	25	Philippines	10	11	45
11	France	30	45	35	26	Saudi Arabia	10	19	17
12	Indonesia	29	11	27	27	Netherlands	9	44	34
13	Poland	28	57	17	28	Egypt	8	10	60
14	Australia	22	76	21	29	Austria	7	58	24
15	Russia	17	8	45	30	Nigeria	6	6	51

Global 3G Stats: Subscribers = 1,098MM Penetration = 18% Growth = 37%

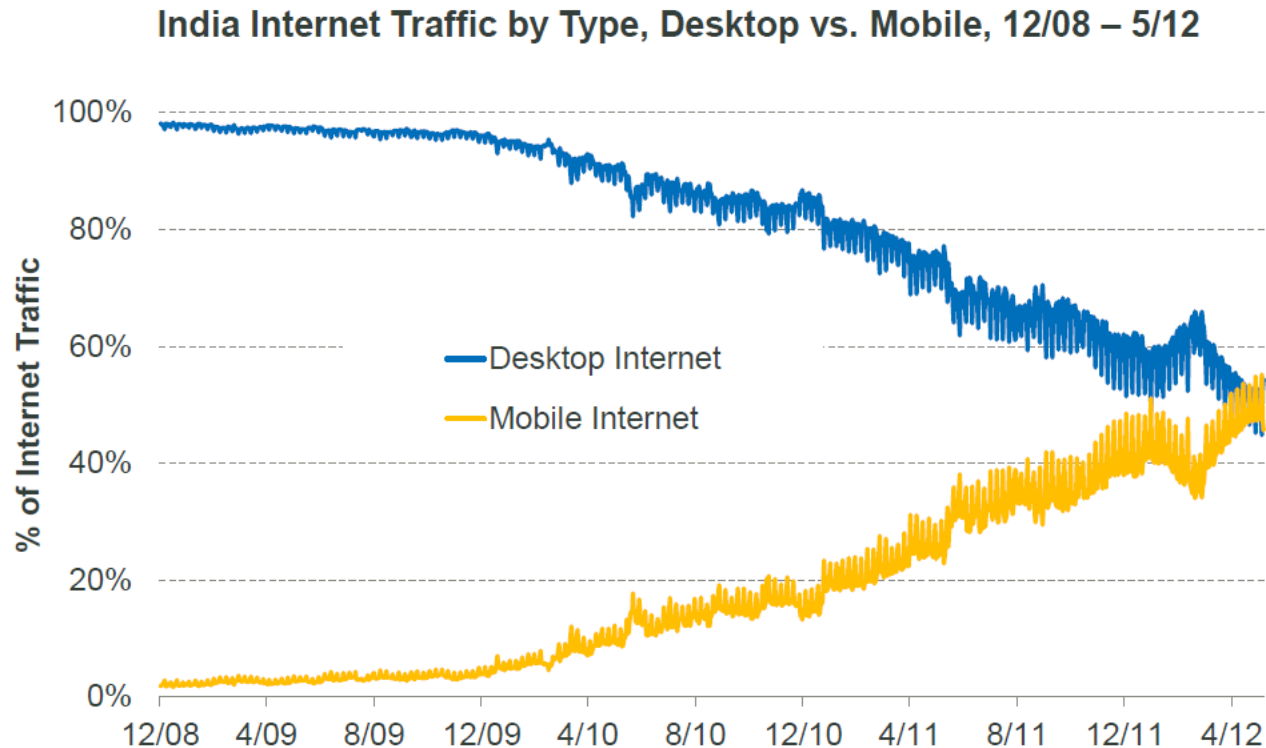
移动流量快速增长

Global Mobile Traffic Growing Rapidly to 10% of Internet Traffic



移动流量快速增长

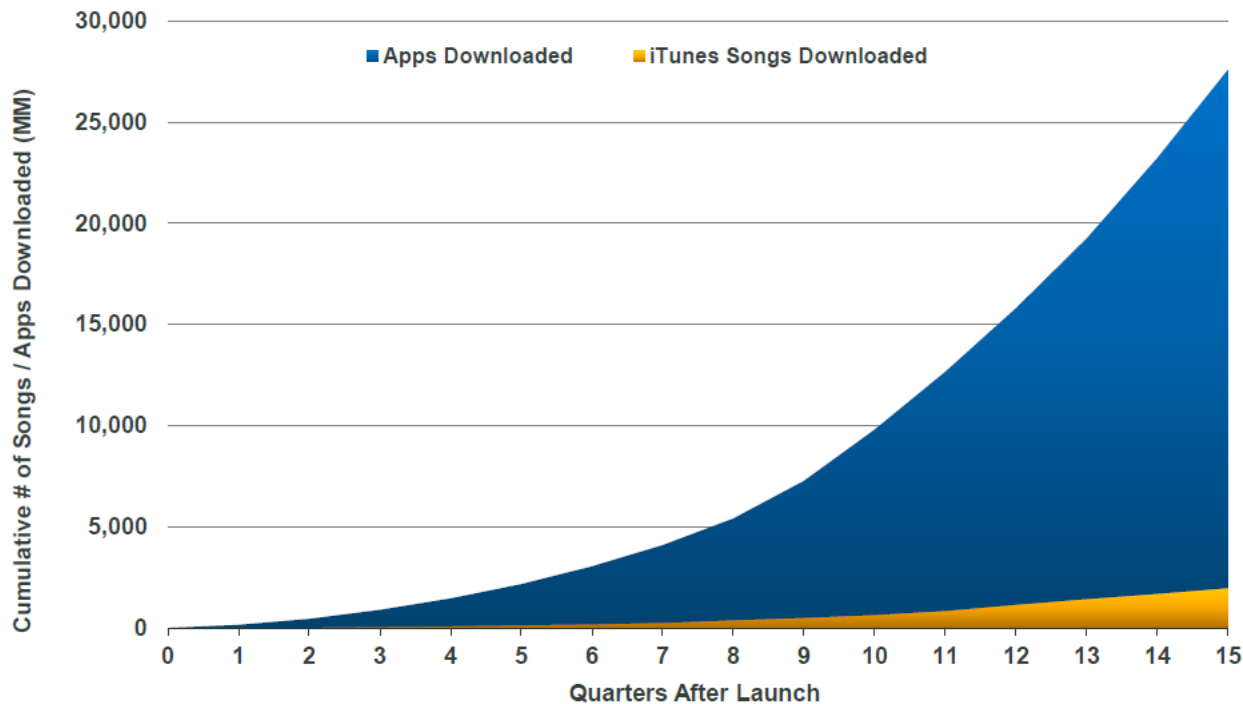
Good / Bad News – Rapidly Growing Mobile Internet Usage Surpassed More Highly Monetized Desktop Internet Usage in May, 2012, in India



移动应用的快速增长

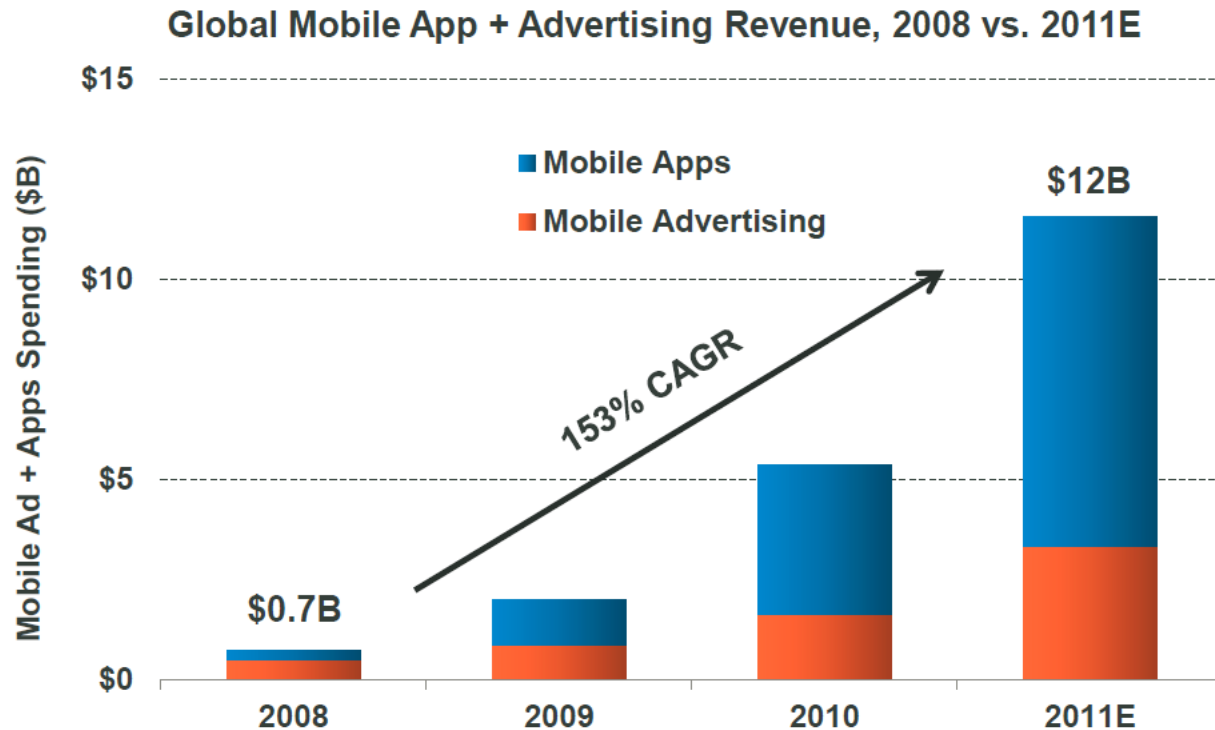
Apple App Store Distribution –
iTunes App Store Driving 46MM+* Downloads per Day

First 15 Quarters Cumulative # of Downloads, iTunes Music vs. Apps



移动计算的商业成功

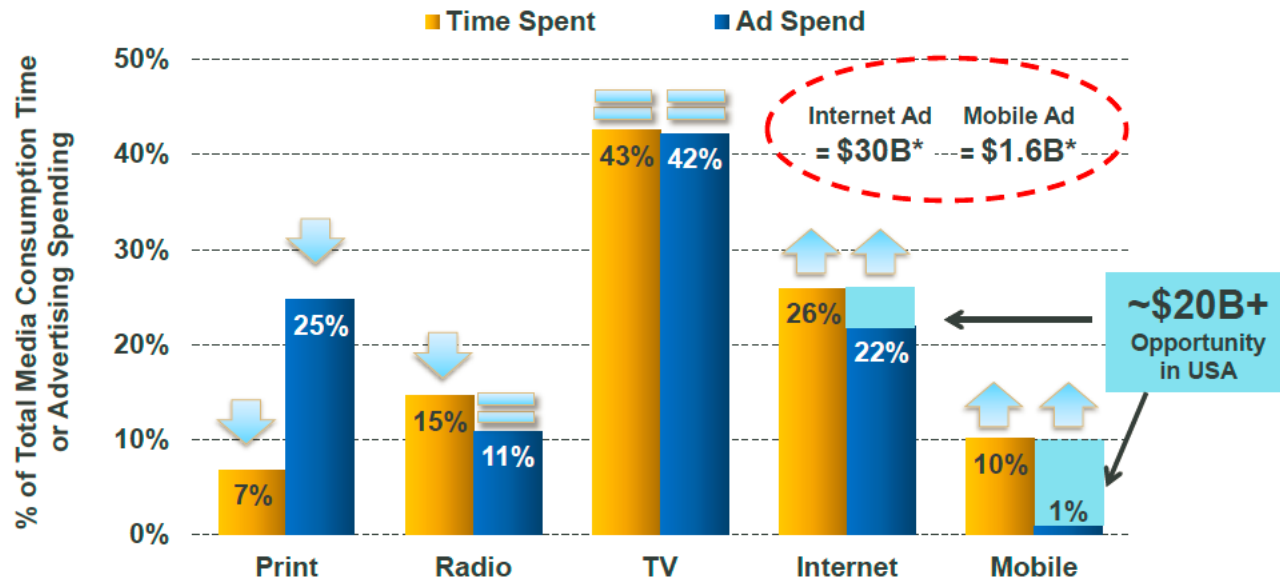
Mobile Monetization Growing Rapidly (71% Apps, 29% Ads)



移动计算的商业前景

Material Upside for Mobile Ad Spend vs. Mobile Usage

% of Time Spent in Media vs. % of Advertising Spending, USA 2011



移动计算时代的 重新思考 (Re-imagination)

计算设备的变化

Re-Imagination of Computing Devices...

THEN...
(Desktops / Notebooks)



NOW...
(Tablets / Smartphones)



电话通信的变化

Re-Imagination of Connectivity...

THEN...



NOW...



杂志的变化

Re-Imagination of Magazines...

THEN...

Piles of Print Copies



NOW...

(Flipboard)

More Content / Always Up-To-Date /
Personalized / Access Everywhere /
Interactive (Video + Audio) / Share



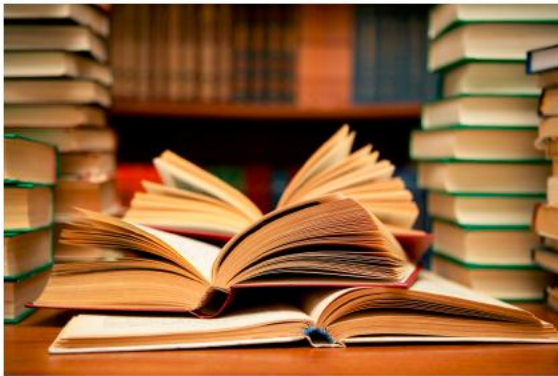
Your new  Flipboard

Instagram. Social search. Speed.

书的变化

Re-Imagination of Books...

THEN...



NOW...

(Amazon Kindle / Apple iBooks)



音乐的变化

Re-Imagination of Music...

THEN...

Buy Albums + CDs in Stores /
Playback via Dedicated Players



KPCB

NOW...

(Spotify...)

Discovery of Music Through Friends + Experts /
Instant On-Demand Streaming on Internet-
Enabled Devices



录音的变化

Re-Imagination of Sound...

THEN...

Tape Recorder / Hard to Edit / Share



NOW...

(SoundCloud)

Record / Edit / Upload / Playback Anywhere /
Anytime / On Any Device / Playlist sharing /
Discovery



视频的变化

Re-Imagination of Video...

THEN...

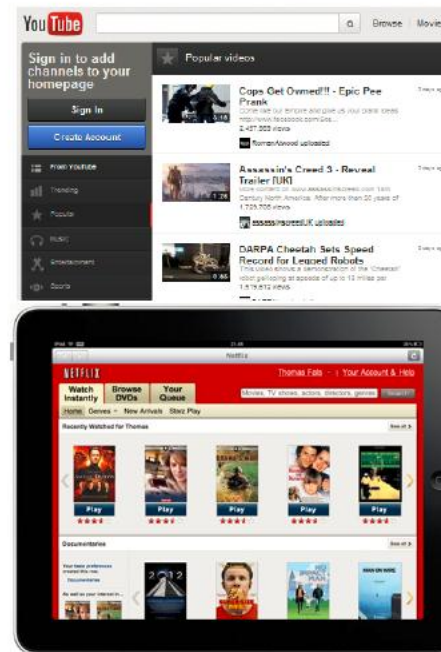
Physical Retail / Rental Stores



KPCB

NOW...

(YouTube / Netflix...)
On-Demand / Instant Streaming /
Accessible Everywhere



电话的变化

Re-Imagination of TV...

THEN...

Linear Programming / Pre-Set Channels /
Little Control Over Content



KPCB

NOW...

(YouTube Channels / Bleacher Team Stream...)
On Demand Personalized Content on Big Screen



通信的变化

Re-Imagination of Communication...

THEN...

Dedicated Devices / Limited
Function & Range / Intrusive



NOW...

(Voxer...)

Push-To-Talk / Voice Message /
Picture / Text / Location / Group Chat



绘画的变化

Re-Imagination of Drawing...

THEN...

Dedicated Canvas / Paint Supplies / Studios
/ Limited Distribution



KPCB

NOW...

(Paper by Fiftythree...)
Reusable Canvas (Screen) / Creating Art
Anywhere Anytime / Digitally Enhanced
Creation Tools / Instant Sharing



摄影的变化

Re-Imagination of Photography...

THEN...

Dedicated Camera / Manually
Transfer Digital Files / Develop Films



NOW...

(Instagr.am / Camera+ / Hipstamatic...)
Always With You Camera (Smartphone) /
Instant Digital Effects / Share / Sync / Discover



地图的变化

Re-Imagination of Navigation + Live Traffic Info...

THEN...

Physical Copies of Map in Car /
TV, Radio Reporting of Traffic Info

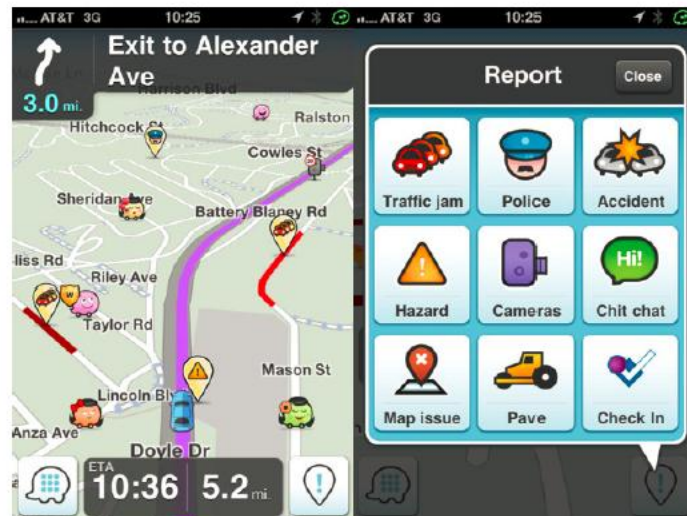


KPCB

NOW...

(Waze)

User-Generated Digital Map /
Live Crowd-Sourced Traffic Data

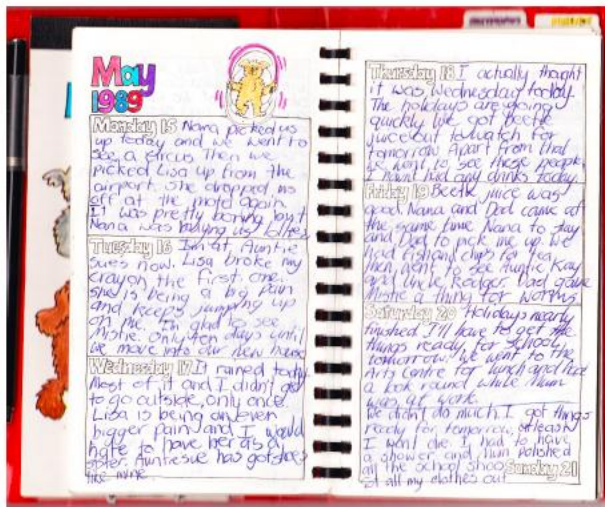


日记的变化

Re-Imagination of Diaries...

THEN...

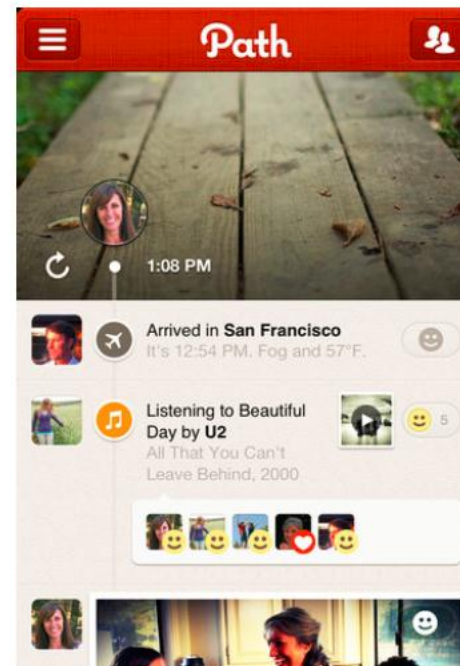
Hand-Written / Drawn



NOW...

(Path)

One-Tap to Add Entry / Multimedia /
Location-Aware / Share / Search



签名的变化

Re-Imagination of Signatures...

THEN...

Scan / Fax / Mail to Return
Signature Page



NOW...

(DocuSign)

Electronic Documents / Secure Audit
Trail / Instant E-Signature

18. Addenda: 22D(Opt. Clauses); 22J(Lead Disci); 22K
35(Inspection); 41C(SB Commission);

2FEF11E53C5944F...

John Hancock

Buyer's Signature

DocuSigned By: John Hancock

Date

Buyer's Signature

Date

1234 1st Avenue

Buyer's Address



学习的变化

Re-Imagination of Learning...

THEN...



NOW...



*From learning by listening to learning by doing...
Education and learning will become as much fun as
videogames. And we call it 'full body learning.'*

- Bing Gordon
Partner, KPCB

信息量的变化

Re-Imagination of Data...

THEN...

Store Everything Because We Can Do It
Inexpensively



SOON...

Data Obesity / Data Quality Issues
How To Find a Needle in a Haystack?



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移动设备的优点

- 移动设备配有各种传感器，就像人的眼睛，嘴巴，耳朵等。这为实现人工智能 (AI) 提供了前所未有的机遇
 - 地点定位 (GPS)
 - 摄像头，麦克风，耳机
 - 温度传感器，加速传感器
 - 触摸屏
- 随身携带性
 - 成为你生活的一部分
 - 知道你的
 - 地点 (Location)
 - 网络访问记录 (History)
 - 个人爱好 (Interests)
 - 社会关系网络 (Social Networks)

移动设备的局限性

- 有限的计算资源
 - CPU, Memory
- 输出输入不方便
 - 键盘输入不方便
 - 显示屏幕比较小
- 电池持续时间比较短
- **3G** 网络还没有普及，带宽比较窄

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SIRI: 智能个人助理

- 苹果的智能个人助理
- 播放**SIRI**的视频

<http://www.apple.com/ios/siri/>



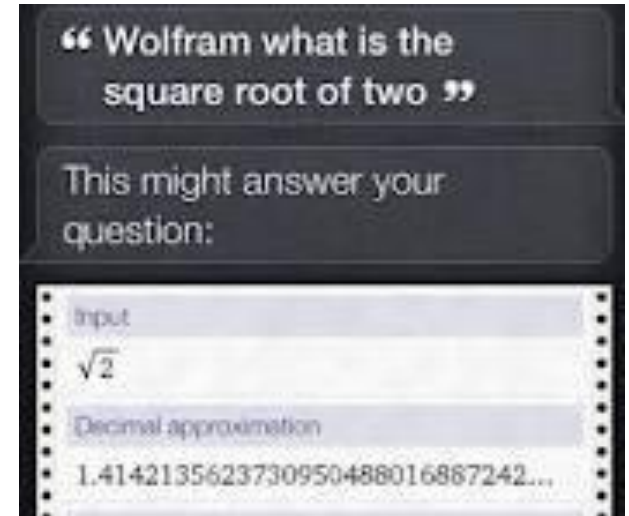
SIRI: 语音控制命令



SIRI: 信息访问



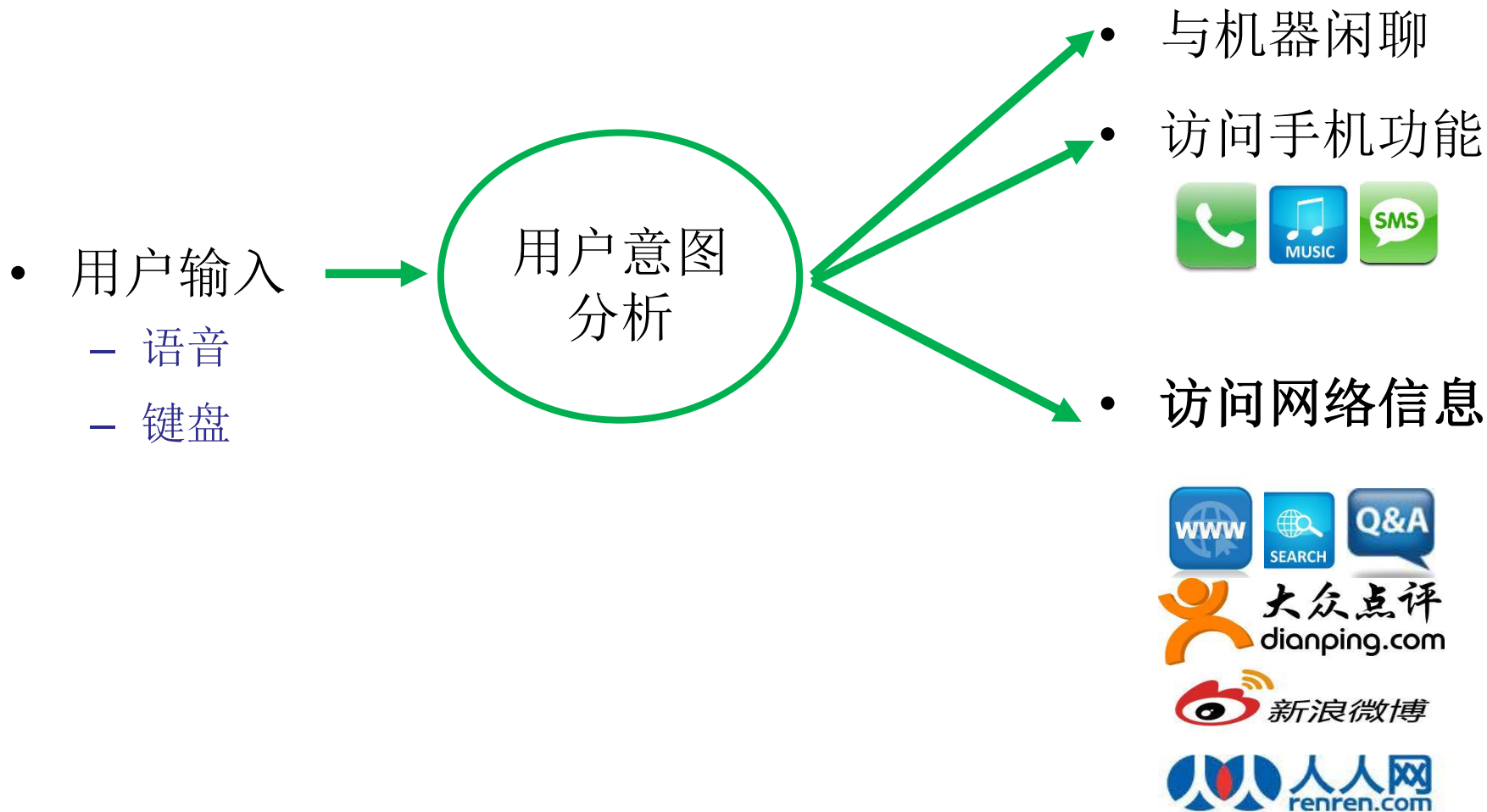
SIRI: 问题回答



技术上真的成熟了吗？



智能个人助理：Big Picture



智能网络信息访问

- 系统自动分析用户的意图，然后做出不同的反应

	用户输入	执行动作
问题回答	让子弹飞的导演是谁	回答“姜文”
	周杰伦的生日是哪天	回答“1979年1月18日”
	2012年奥运会在哪里举行	回答“伦敦”
网络搜索	北京大学	提供关于“北京大学”搜索结果
	百度一下故宫	提供关于“故宫”搜索结果
	生命的意义是什么	提供关于“生命的意义是什么”搜索结果
网络浏览	打开新浪财经的网站	浏览finance.sina.com.cn
	www.sohu.com	浏览www.sohu.com

传统搜索和SIRI

- 传统搜索 (Search)

- 优点: 大部分情况下works (i.e. 提供相关链接)
- 缺点:
 - 不能直接提供答案 (比如直接回答问题: *现在walmart还开门吗?*)
 - 不能直接在搜索结果上做Action (比如搜索餐馆然后直接订餐)
 - 不能利用自然语言输入做更精准的回答



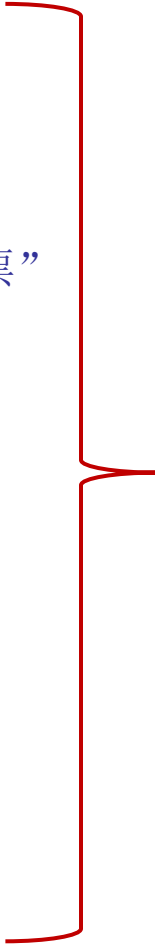
- SIRI (智能个人助理)

- 优点: 能直接提供答案和直接做动作 (如订餐, 导航等), 能自然语言输入
- 缺点: 不可靠 (30%成功, 70%失败)



相关的NLP技术

- 语音识别
- 机器学习
 - classification
- 信息提取
 - “帮我买一张从西安到北京明天晚上东航的飞机票”
- 命名实体识别
- 语义分析 (semantic parsing)
 - 对用户的输入进行分析
- 对话管理 (dialog management)
 - 明天北京天气怎么样？
 - 上海呢？
- 问题回答 (question answering)



Hard-core NLP problems!

Factoid Question Answering

- 答案是非常**明确并简短**的问题
 - 比如答案是**时间，地点，或人名**。问题的例子有：
 - 毛泽东是哪天出生的？刘德华的生日是哪天？
 - 湖南省的省会在哪里？中南工业大学在哪个城市？长沙是哪个省的省会？希腊的首都在哪里？毛泽东出生在哪个省？巴尔的摩在美国哪个州？
 - 让子弹飞的导演是谁？谢霆锋的前妻是谁？湖南省岳阳市市委书记是谁？

为什么**精准**问题回答有意义？

- 解决用户在移动环境下急需知道答案的需求
 - 比如在开车或旅游时，没法慢慢浏览
- 显示更多的相关内容 (见下一页例子)
 - 比如当用户问李亚鹏的老婆是谁时，系统不但可以回答“王菲”，更进一步可以显示王菲的近照，最近的歌曲和八卦等相关内容
- 回答需要推理的复杂问题的必须步骤
 - 后天是中秋节吗？(系统必须先知道今年中秋节是哪天)
 - 毛泽东的故乡今天天气怎么样？(系统必须先知道毛泽东的故乡在哪里)
 - 显示几张王菲的老公的照片(系统必须先知道王菲的老公是谁)
- **Empower interactive**的人机交互方式
 - 问：李亚鹏的老婆是谁？
 - 答：王菲
 - 问：她最近有什么歌曲啊？
 - 答：因为爱你
 - 问：放给我听听
 - 答：播放歌曲因为爱你

更相关的搜索结果



[李亚鹏的老婆是谁](#) 百度知道

李亚鹏的老婆...王菲啊~...就是就是 王菲啊...王菲...这你也不知道? 王菲嘛!...王菲...王菲...王菲啊都有个小孩了, 你还不知道啊!...王菲...

[李亚鹏老婆是谁呀](#) 4个回答

[李亚鹏版《笑傲江湖》里的岳不群老婆宁中则是谁演的?好..](#) 2个回答

[网络春晚《法制故事》中扮演李亚鹏妻子的是谁?](#) 3个回答

[更多知道相关问题»](#)

[zhidao.baidu.com](#)

[李亚鹏的老婆是谁](#) 在线视频观看 土豆网视频
[李亚鹏的老婆是谁](#)

原创 [李亚鹏的老婆是谁](#) [李亚鹏的老婆是谁](#) 转贴到土豆推 更多加豆单私藏下载... 老婆! 这个男的是谁? 01:38 老婆, 这个男的是谁? 老婆, 这个男的是谁?...

[www.tudou.com](#) 33k - 优化阅读

[李亚鹏的老婆是谁-娱乐动态-东南电影网](#)

家庭状况: 父亲王佑林、母亲夏桂影(去世)、哥哥王弋、



[王菲](#) 百度百科

王菲, 中国著名女歌手、影视演员, 是1990年代初期至今华语乐坛最出色的女歌手之一, 被公认为乐坛天后。她是极少数能同时被华语歌坛同行一致赞美、甚...

[baike.baidu.com](#)

[王菲](#) 百度MP3

《王菲2cd精选》 发行时间: 2009-07-14 曲目数: 28 语种: 中文 唱片公司: 环球唱片 1 我愿意 2 容易受伤的女人 3 但愿人长久 4 执迷不悔 5 棋子...

[mp3.baidu.com](#) 127k - 优化阅读

[王菲](#) 维基百科, 自由的百科全书

王菲推出的多张唱片均有百万销量佳绩。截至2000年3月, 20张专辑合计台湾、日本与香港有据可考的销量达970万张, 获金氏世界纪录大全认证为华语乐坛粤语专辑销量最高的...

[zh.wikipedia.org](#) 127k - 电脑版

[王菲](#) 的最新相关信息

窦靖童王菲刘嘉玲风格迥异 姐妹儿个个都是范儿

Re-imagination of NLP for Search

- 输入

键盘 → 语音, 手写, OCR

关键词 → 自然语言

Pattern Match → Semantic Parsing

- 输出

大屏幕 → 小屏幕

文字 → 声音 (TTS)

Links → Answers

IR → QA

IR → QA



[李亚鹏的老婆是谁_百度知道](#)

李亚鹏的老婆...王菲啊~...就是就是 王菲啊...王菲...这你也不知道? 王菲嘛!...王菲...王菲...王菲啊都有个小孩了, 你还不知道啊!...王菲...

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[更多知道相关问题»](#)

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[李亚鹏的老婆是谁_在线视频观看_土豆网视频](#)
[李亚鹏的老婆是谁](#)

原创 [李亚鹏的老婆是谁](#) [李亚鹏的老婆是谁](#) 转贴到土豆推 更多加豆单私藏下载... 老婆! 这个男的是谁? 01:38 老婆, 这个男的是谁? 老婆, 这个男的是谁?...

[www.tudou.com](#) 33k - [优化阅读](#)

[李亚鹏的老婆是谁-娱乐动态-东南电影网](#)

家庭状况: 父亲王佑林、母亲夏桂影(去世)、哥哥王弋、



李亚鹏老婆是谁

王菲



Re-imagination of NLP Models

- **Context-aware Model**

Context = time, location, person, interests ...

- **Source Model**

$P(\text{current word} \mid \text{history}) \longrightarrow P(\text{current word} \mid \text{history}, \text{context})$

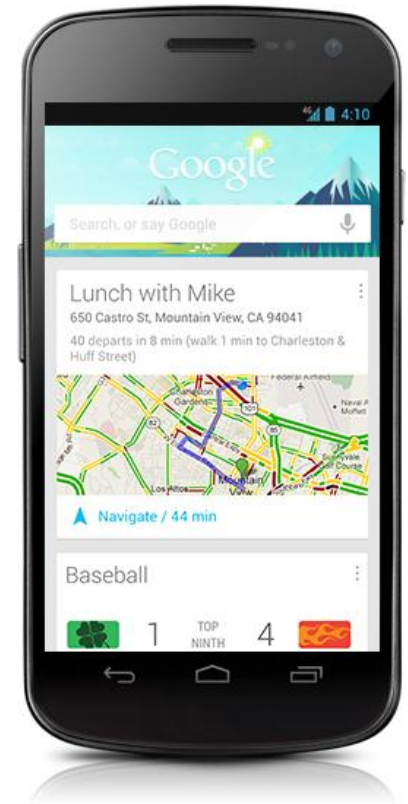
- **Channel Model**

$P(\text{output} \mid \text{input}) \longrightarrow P(\text{output} \mid \text{input}, \text{context})$

Google Now

- **Google**的最新智能应用
 - Predictive UI
 - Factoid Question Answering
- 播放**Google Now**的视频

<http://www.google.com/landing/now/>



Word Lens

- 基于**OCR**的机器翻译应用
- 播放**Word Lens**的宣传视频

<http://questvisual.com/us/>



Joint decoding for OCR and translation?

Offline OCR and Translation?

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Summary of Results

Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0
Sorted array	5	1	1	6	60	$O(\log(N))$	0
MPH	5	1	1.054	6.325	63.25	$O(1)$	0
Context encoding	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0
Implicit trie	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0
Bloomier	2	1	1.23	3.69	36.9	$O(1)$	$1/(2^{16})$
MPH	2	1	1.108	3.325	33.25	$O(1)$	$1/(2^{16})$

- Assumptions:
 - Each word ID consumes 2.5 bytes (i.e., $V=2^{20}$)
 - Probability value is quantized to one byte
 - Need to store 10M bigrams (i.e., $N=10M$)

Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0

- Java String object
 - each java object has 8 bytes overhead
 - the size of an object has to be a multiple of 8
 - an empty string in java will take 40 bytes!
- `HashMap<String, Integer>`:
 - Key is a String, which needs 72 bytes.
 - assuming 17 chars (each char consumes 2 bytes)
 - $72 \text{ bytes} = 8 + 4 + 4 + 4 + 4 + (8 + 4 + 17*2 + 2)$
 - Value is an Integer
 - $16 \text{ bytes} = (8 + 1 + 7)$
 - Overhead ratio is 1.6
 - assuming a load factor 0.625 (so overhead ratio is $1/0.625=1.6$)

```
public class String {
    char value[];
    int length;
    int offset;
    int hash;
}
```

Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0

- HashMap uses an array to store the actual keys and values
 - The naive implementation store the keys and values as objects, which have high memory overhead
- To save memory:
 - Do not store the keys and values as java objects
 - Instead, store the keys and values in a bit array
 - So, a key (i.e., a bigram) will consume 5 bytes as each word Id consumes 2.5 bytes.

Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0
Sorted array	5	1	1	6	60	$O(\log(N))$	0

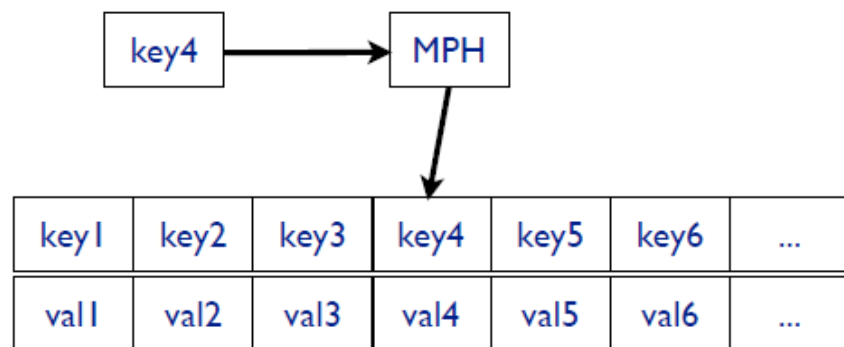
- To deal with hash collisions, HashMap needs to use an array that is much bigger than needed to store the actual keys and values
 - The benefit of this is constant lookup time.
- To save memory, we can:
 - Store the list of bigrams as a sorted array
 - Use binary search to look up a key
 - However, the look-up time will be $O(\log N)$, where N is the total number of bigrams

Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0
Sorted array	5	1	1	6	60	$O(\log(N))$	0
MPH	5	1	1.054	6.325	63.25	$O(1)$	0

- Use minimum perfect hash (MPH):
 - Store key and values in arrays
 - Use a MPH function to index the keys
- Lookup procedure:
 - First compute MPH of key
 - Then match **key** in the array
- MPH function consumes 2.6 extra bits for each key. This explains the overhead ratio 1.054

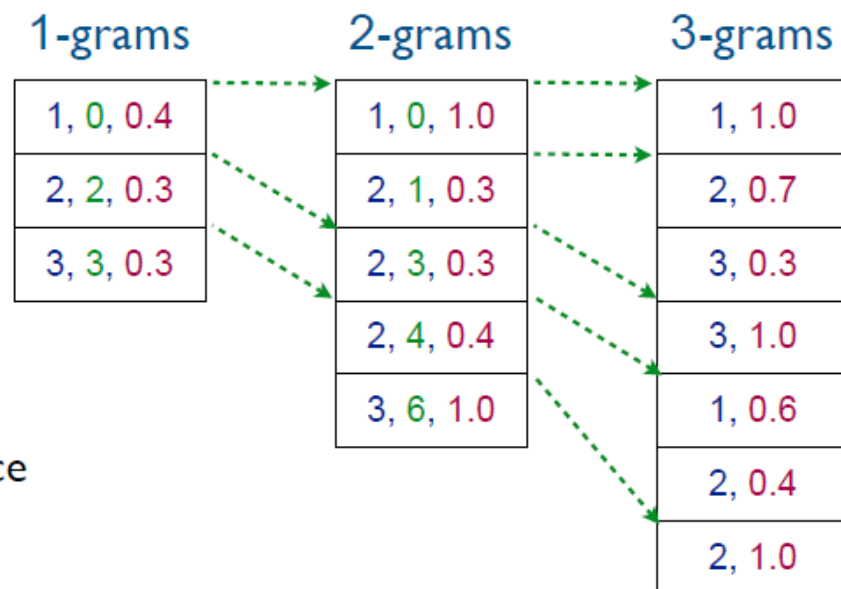
perfect: no collision

minimum: MPH outputs numbers in $[1, N]$



Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0
Sorted array	5	1	1	6	60	$O(\log(N))$	0
MPH	5	1	1.054	6.325	63.25	$O(1)$	0
Implicit trie	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0

- Store the keys and values in arrays of Trie nodes
- Each Trie node includes:
 - word id
 - offset in the next layer
 - the value (if any)
- The overhead ratio (1.285) considers space for storing offsets in the unigram table



Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0
Sorted array	5	1	1	6	60	$O(\log(N))$	0
MPH	5	1	1.054	6.325	63.25	$O(1)$	0
Implicit trie	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0
Context encoding	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0

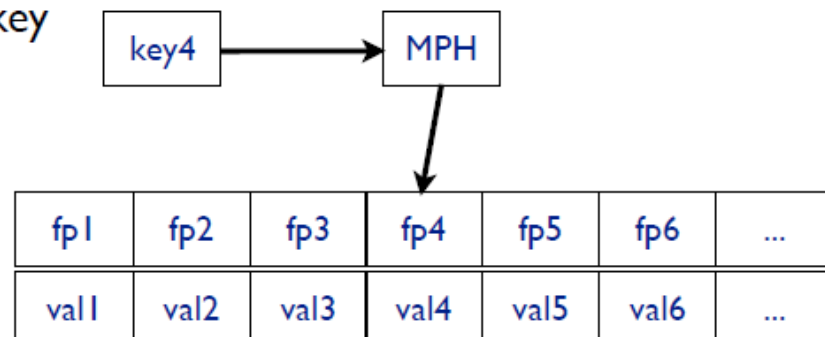
See (Pauls and Klein 2011) for details.

Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0
Sorted array	5	1	1	6	60	$O(\log(N))$	0
MPH	5	1	1.054	6.325	63.25	$O(1)$	0
Context encoding	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0
Implicit trie	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0
Bloomier	2	1	1.23	3.69	36.9	$O(1)$	$1/(2^{16})$

- In comparison with MPH, Bloomier is perfect, but not minimum
- The overhead is 1.23.

Method	Key	Val	Overhead ratio	Bytes per bigram	Total (MB)	Time	P(Error)
Map (naive java)	72	16	1.6	141	1410	$O(1)$	0
Map	5	1	1.6	9.6	96	$O(1)$	0
Sorted array	5	1	1	6	60	$O(\log(N))$	0
MPH	5	1	1.054	6.325	63.25	$O(1)$	0
Context encoding	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0
Implicit trie	2.5	1	1.285	3.95	39.5	$O(2\log V)$	0
Bloomier	2	1	1.23	3.69	36.9	$O(1)$	$1/(2^{16})$
MPH	2	1	1.108	3.325	33.25	$O(1)$	$1/(2^{16})$

- Store a finger print of the key, not the actual key
- Lookup procedure:
 - First compute MPH of key
 - Then match **fingerprint** (fp) in the array
- MPH function consumes 2.6 extra bits for each key, explaining the overhead ratio 1.108



Compression of Arrays

- All the data structures presented require to use **array(s)** to store keys (or their fingerprints) and values
- So, compression of arrays will save memory
 - Note that the keys in the approximate maps (e.g., based on bloomier filter and MPH) can not be compressed as they are random
- Variable-length coding
 - Huffman code
 - Golomb codes
 - Boldi and Vigna (2005)
 - Elias-Fano codes for monotone sequences

Element Access of Compressed Array

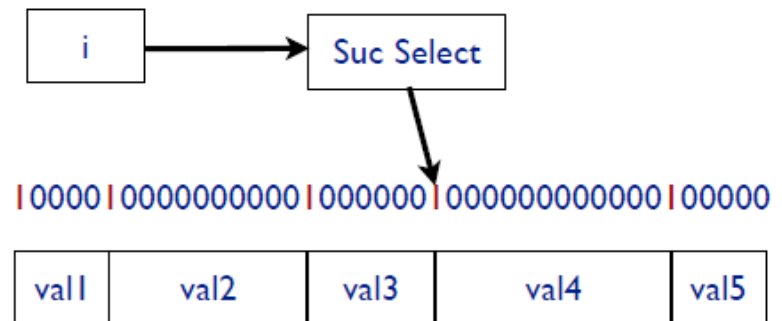
- A common operation is to access the i -th element in the array (e.g. during the binary search)
 - Need to be fast
- Naive methods require to linear scan of the array in order to access the i -th element
- Common practices use (e.g. in Pauls and Klein, 2011) block encoding
 - still slow and space non-optimal
- New method:
 - use a bit array together with succinct data structures

Constant Access with Succinct Data Structures

- Use whatever variable-length coding methods to encode elements
- Store the elements in a bit array
- Have an **additional** binary array to indicate the element boundaries
- Use succinct data structures to quickly identify the boundary (i.e. using **select** operation)
 - This requires about 13% additional space.

Two fundamental operations for a succinct data structure:

- **select**: find the position of the i-th one in the binary array
- **rank**: find how many ones before a given position in the binary array



Outline

- 移动计算时代的来临
 - 主要摘自Mary Meeker 的Internet Trend报告
- 移动计算的技术特点
- NLP相关的应用分析
 - SIRI
 - Word Lens
 - Google Now
- Compact Data Structures for Mobiles
- 结束语

Conclusions

- 移动计算时代已来临
- 移动计算下NLP研究面临新的机遇和挑战
- 移动智能应用将会越来越流行
 - SIRI, Google Now, Google Glass, Word Lens, etc
- 相关的NLP技术将会有巨大的商业应用
 - Efficient data structures for mobiles
 - Location/time-aware models
 - Semantic parsing
 - Natural language search
 - Question answering
 - Predictive models for personalization
 - Joint Decoding (for multi-modal inputs)

Google Glass

- **Google X**的项目
- 播放**Google Glass**的视频

<https://plus.google.com/+projectglass>



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