probabilistic algorithms to process MASSIVE data

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April 10th, 2012

0. DATA EXPLOSION

- > 340 million Tweets a day, 294 billion emails a day
- ▶ 35 hours of video uploaded to YouTube per minute
- one human genome: 3.2 billion letters
- NSA wants to build 150 Petabytes (150 million GB) to store personal data on people

Moore's law: processing power doubles every 18 months

Sedgewick's principle: Volumes and complexity of data increase faster than processing speed. We need ever better algorithms to keep pace.



- ► April 2012: Google's index contains 55 billion pages
- Google processes 24 petabytes every day
- ▶ = only fraction of 1 trillion existing web pages (in 2008)

data stream model

Stream: a (very large) sequence S over (also very large) domain \mathcal{D}

$$S = s_1 \ s_2 \ s_3 \ \cdots \ s_\ell, \qquad s_j \in \mathcal{D}$$

consider S as a multiset

$$\mathcal{M}=m_1^{f_1} m_2^{f_2} \cdots m_n^{f_n}$$

Ex.: $S = \text{run sally run see sally run} \Rightarrow M = \text{run}^3 \text{ sally}^2 \text{ see}^1$

Interested in estimating the following quantitive statistics:

- **A.** Length $:= \ell$
- **B.** Cardinality := $card(m_i) \equiv n$ (distinct values)
- C. Icebergs := # elem. with relative frequency $f_v/\ell > \theta$

[where θ is any fixed threshold, like 50%]

Constraints:

- very little processing memory
- on the fly (single pass + simple main loop)
- no statistical hypothesis
- accuracy within a few percentiles

Prelude: you need $\log_2 N$ bits to count up to N

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bit: smallest unit of information, either 0 or 1



with 10 fingers/bits you can count up to $2^{10}-1=512+256+\ldots+1$





1 + 8 + 16 = 25

1. Approximate Counting (count length ℓ)

With 8 bits, can count up to $2^8 - 1 = 255$ elements.

Question: is it possible to count more?? \Rightarrow **YES**, with coin flips!

First idea: increment every other time

- Initialize: C := 0
- Increment: with probability 1/2, C := C + 1
- Output: 2 · C

 $\mathbb{E}[2 \cdot C] = n$ and only 3% error

Limitation: only save 1 bit (with 8 bits count to $2^{8+1} - 1 = 511$) [not very interesting, be honest!]

Second idea: generalize, and increment 1 out of 2^k [prob $1/2^k$ = flip k coins, and all equal to 1]

- Initialize: C := 0
- Increment: with probability $1/2^k$, C := C + 1
- **•** Output: $2^k \cdot C$

better: saves **k** bits, i.e., count up to 2^{8+k} with 8 bits

Limitations:

- only saves linear number of bits
- for k = 8, error is 55%
- worst: always inaccurate for small values $< 2^k$

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[ because smallest value returned is 2^k \cdot C ]
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the GOOD idea

Third idea: probability of increment depends on value of counter C

- Initialize: C := 0
- Increment: with probability $1/2^{\mathsf{C}}$, $\mathbf{C} := \mathbf{C} + 1$
- ▶ **Output:** 2^C 1

Gets "harder" to increment:

$$\begin{array}{c} C=1 \\ \hline 1/2 \\ C=2 \\ \hline 1/4 \\ C=3 \\ \hline 1/8 \\ \hline C=4 \\ \hline 1/16 \\ C=5 \end{array}$$

Finally:

- accurate for small values
- ▶ with 8 bits count up to 2¹⁶ with **15% error**

Morris 1978, Flajolet 1985



application to genetics: finding patterns in genomes

Genome: long sequence of letters $\{A, C, G, T\}$

count occurrences of all subwords of size k





Π	AA	AC	AG	AT	CA	сс	CG	СТ	GA	GC	GG	GT	TA	тс	ΤG	ТТ
Π	4	3	-	-	-	-	1	1	-	-	-	1	2	-	-	-

AA occurs 4 times

▶ GA is absent, and is called a *nullomer* (as are AG, AT, etc.) [significant: because in total randomness all patterns would appear]



Limitations of exact count:

- for k = 13, requires **2 GB** of memory
- ▶ k > 14 requires Approx. Counting!



patterns in anthrax bacteria genome (5.23 M)



distribution of sequences of nucleotides of size k = 7, 8, 9, 10 source: Csűrös 2007, http://www.iro.umontreal.ca/~csuros/spectrum/

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distribution of 5.23 M random strings

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2. **DISTINCT** elements

Back to our stream:

 $S = s_1 \ s_2 \ s_3 \ \cdots \ s_\ell, \qquad s_j \in \mathcal{D}$

We want the number n of **distinct elements** used.

Ex.: S = run sally run see sally run (3 distinct elements)

- idea 1: sort data; then same elements next to each other; scan sorted data and count distinct elements
- ▶ idea 2: have bag, for each s_j, if not in bag, add it; then count number of elements in bag

Bad ideas: too much memory is used; at minimum O(n).

A weird way to use hash functions!!

Definition: a hash function h is defined as $h: \mathcal{A}^* \to [0, 1].$

Main idea. With "good enough" hash functions, our data is uniformized.



Two neat things about the minimum

Fact 1: the minimum not sensitive to repetitions

 $\min\{0.83, 0.32, 0.83, 0.83, 0.95, 0.74\} = \min\{0.83, 0.32, 0.95, 0.74\} = 0.32$

Fact 2: *n* uniform random variables in [0,1] have min. $M \approx 1/(n+1)$

$$\mathbb{E}_n[M] = \int_0^1 x \cdot n(1-x)^{n-1} \mathrm{d}x = \boxed{\frac{1}{n+1}}$$

Minimum-counting algorithm (Bar-Yossef et al. 2002, L. 2010):

- ▶ hash elements of stream to [0,1] values
- take the minimum M
- ▶ return 1/M 1

With some optimizations, you can obtain 3% error with only 4kB!

AltaVista: remove near-duplicates

Broder (1997) uses similarity measure AltaVista The most powerful and useful guide to the Net $S(A,B) = \frac{|A \cap B|}{|A \cup B|}$ Ask Alta VistaTM a question. Or enter a few words in any language Help - Advanced Search Example: Where can I download mp3 files for instrumental music? AV Family Filter - AV Photo Finder - AV Tools & Gadgets Specialty Entertainment - Health - Online Shopping - Careers - Maps Searches People Finder - Stock Quotes - Travel - Usenet - Yellow Pages S(A, B) = 0 : sets disjoint CATEGORIES NEWS BY ABCNEWS.com $\frac{Featured Sponsors}{50\% Savings!} S(A, B) = 1 : sets overlap$ Automotive Lewinsky Talks Olympic House-cleaning Business & Finance Jasper Trial Begins Quality DutyFree **Computers & Internet** al Mass Draws 1 Million Mexican Jewelry! Health & Fitness ALTAVISTA HIGHLIGHTS Great Gifts from Hobbies & Interests Search Clinton Video Footage: Home & Family New State of The Union Save on mpeachment Trial Media & Amusements bestsellers if S(A, B) > 0.99, consider docu-Clinton Testimon People & Chat everyday at ideo courtesy of C-SPAN Amazon Reference & Education ments A and B are same OTHER SERVICES Shopping & Services PC Flowers and Gifts Valentines AltaVista Discovery - Video Search Demi Society & Politics Specials FREE Email - AV Translation Services Sports & Recreation Make Us Your Homepage - Create A Card Photo Albums! - Asian Language Travel & Vacations

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 \Rightarrow eliminate near duplicates!

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15/17

with Minimum-counting algorithm... easy

$$S(A, B) = \frac{|A \cap B|}{|A \cup B|} = \frac{|A \cup B| - |A| - |B|}{|A \cup B|}$$

and, if you note h(A) and h(B) the streams where you apply the hash function h to A and B,

▶
$$|A| = 1/\min(h(A)) - 1$$

▶
$$|B| = 1/\min(h(B)) - 1$$

►
$$|A \cup B| = 1/\min(h(A), h(B)) - 1$$

so only need to keep the minimum for each document, then only O(1) operations to compare to documents!

compare 10^5 documents of size 10^5 with each-other in only minutes instead of days



Many other applications:

Network security:

detect attacks (denial of service), or the spreading of worms/spam,...



- Data mining: document classification, ...
- Databases: query optimization
- Distributed: censor networks