

Tabulation-Based Hashing

Mihai Pătrașcu Mikkel Thorup

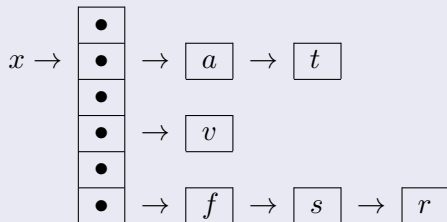


April 23, 2010

Applications of Hashing

Hash tables:

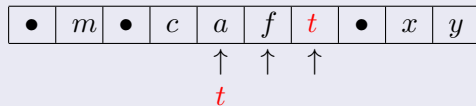
- chaining



Applications of Hashing

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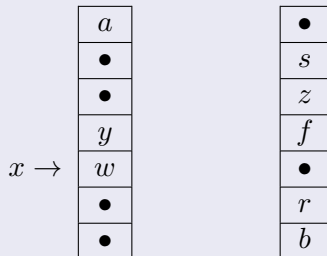
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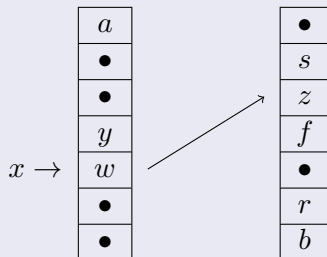
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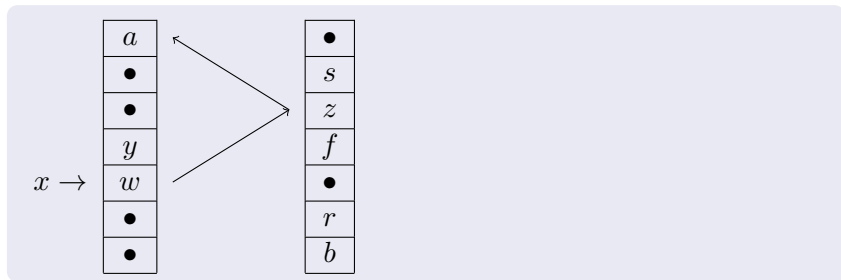
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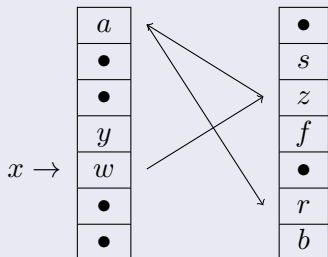
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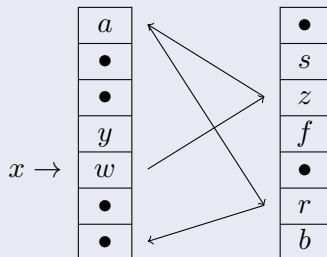
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Sketching and streaming:

- moment estimation: $F_2(\bar{x}) = \sum_i x_i^2$

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- etc, etc.

Minwise independence

Hash each set through h , keep the minimum

$$\frac{|A \cap B|}{|A \cup B|} = \Pr_h[\min h(A) = \min h(B)]$$

- repeat with k different h ;
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NB: for weighted A, B the generalization is priority sampling

Carter & Wegman (1977)

A family $\mathcal{H} = \{h : [u] \rightarrow [b]\}$ is **k -independent** iff:

- $(\forall)x \in u$, $h(x)$ is uniform in $[b]$;
- $(\forall)x_1, \dots, x_k \in [u]$, $h(x_1), \dots, h(x_k)$ are independent.

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Prototypical example: degree k polynomial

- u prime;
- choose a_0, a_1, \dots, a_{k-1} randomly in $[u]$;
- $h(x) = (a_0 + a_1x + \dots + a_{k-1}x^{k-1}) \bmod b$.

How much independence?

Chaining	2	
Linear probing	≤ 5 [Pagh ² , Ružić'07]	≥ 5 [PT'10]
Cuckoo hashing	$O(\lg n)$	≥ 6 [Cohen, Kane'05]
F_2 estimation	4 [Thorup, Zhang'04]	
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Minwise independence:
 k -level inclusion/exclusion estimates probabilities to $\pm 2^{-k}$.

Linear probing

Implementing k -independence

Goals:

- constant time for $\omega(1)$ independence
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Tabulation hashing:

- q basic characters: $x \mapsto (x_1, \dots, x_q)$
- d derived characters: $y_i = f_i(x_1, \dots, x_q)$
- store $q + d$ random tables $T_i[u^{1/q}]$
- $h(x) = T_1[x_1] \oplus \dots \oplus T_q[x_q] \oplus T_{q+1}[y_1] \oplus \dots$

Tabulation-Based Hashing

	Independence	# characters
[Carter, Wegman'77]	3	q (★)
[Siegel'90s]	$n^{\Omega(1)}$	$q^{O(q)}$
[Dietzf., Woelfel'03]	k	$k \cdot q$
[Thorup, Zhang'04]	k	$(k - 1)(q - 1)$
[Thorup, Zhang'10]	5	$2q - 1$
recent	$\omega(1)$	$O(q^2)$

(★) simple tabulation (no derived characters)

Peeling $(q = 2, k = 3)$

$$(x_1, x_2) \mapsto T_1[x_1] \oplus T_2[x_2]$$

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Any set of ≤ 3 keys is peelable, thus independent.

Peeling $(q = 2, k = 4)$

$$(x_1, x_2) \mapsto T_1[x_1] \oplus T_2[x_2] \oplus T_3[x_1 + x_2]$$

Let's prove $\{a, b, c, d\}$ are independent.

- if we can peel, reduce to 3-independence.
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Only possible equalities: $x + s = y + t$ or $x + t = y + s$.

Both cannot hold, so we have peeling in derived character.

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- two columns have Hamming distance = 4

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- all columns at Hamming distance = 2

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NB: $h(a) = h(b) \oplus h(c) \oplus h(d)$

If e independent of b, c, d , also independent of $f(b, c, d)$.

Putting it together

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chaining	2
F_2 estimation	4
linear probing	5
ε -minwise	$\Theta(\lg \frac{1}{\varepsilon})$
cuckoo hashing	$O(\lg n)?$

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What exactly are we doing here?

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- Chernoff concentration $\Rightarrow O(\lg n)$ query time w.h.p.
- preserve moments in linear probing, chaining:
 F_p w/ simple tabulation = F_p w/ truly random + $o(1)$

Simple tabulation as a PRG

Pseudorandom numbers $\approx h(0), h(1), h(1), \dots$

	$f(0)$	$f(1)$	\dots	$f(S-1)$
$g(0)$	$h(0)$	$h(1)$	\dots	$h(S-1)$
$g(1)$	$h(S)$	$h(S+1)$	\dots	$h(2S-1)$
$g(2)$	$h(2S)$	$h(2S+1)$	\dots	$h(3S-1)$
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\dots				

Use $S = O(\lg n)$ truly random numbers.
Compute $\lg n$ independent $g(\cdot)$, but rarely.

PRG has:

- concentration (load balancing, \dots)
- minwise independence (treaps, \dots)

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→ for $k \geq \lg n$, any k operations work in $O(k)$ time w.h.p.

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We only get $k = n^\varepsilon$ for any $\varepsilon > 0$.

Counterexample for $n^{o(1)}$.

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Simple++:

- $h_1 : [u] \rightarrow [b]$, $h_2 : [u] \rightarrow [u^{1/q}]$. Just simple tabulation...
- $h(x) = h_1(x) \oplus T[h_2(x)]$.

All previous properties, plus:

- minwise independence with $\varepsilon(u) = o(1)$.
- linear probing/chaining with buffer $k = O(\lg n)$.

THE END