# CS 473: Algorithms 

Ruta Mehta

University of Illinois, Urbana-Champaign
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## CS 473: Algorithms, Spring 2018

## Universal Hashing

Lecture 10
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Most slides are courtesy Prof. Chekuri

## Part I

## Hash Tables

## Dictionary Data Structure

(1) $\mathcal{U}$ : universe of keys with total order: numbers, strings, etc.
(2) Data structure to store a subset $S \subseteq \mathcal{U}$
(3) Operations:
(1) Search/look up: given $x \in \mathcal{U}$ is $x \in S$ ?
(2) Insert: given $x \notin S$ add $x$ to $S$.
(3) Delete: given $x \in S$ delete $x$ from $S$
(9) Static structure: $S$ given in advance or changes very infrequently, main operations are lookups.
(5) Dynamic structure: $S$ changes rapidly so inserts and deletes as important as lookups.

Can we do everything in $O(1)$ time?

## Hashing and Hash Tables

Hash Table data structure:
(1) A (hash) table/array $\boldsymbol{T}$ of size $\boldsymbol{m}$ (the table size).
(2) A hash function $h: \mathcal{U} \rightarrow\{0, \ldots, m-1\}$.
(3) Item $x \in \mathcal{U}$ hashes to slot $h(x)$ in $T$.

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## Ideal situation:

(1) Each element $x \in S$ hashes to a distinct slot in $T$. Store $x$ in slot $h(x)$
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Collisions unavoidable if $|\boldsymbol{T}|<|\mathcal{U}|$. Several techniques to handle them.

## Handling Collisions: Chaining

Collision: $h(x)=h(y)$ for some $x \neq y$.
Chaining/Open hashing to handle collisions:
(1) For each slot $i$ store all items hashed to slot $i$ in a linked list. $T[i]$ points to the linked list
(2) Lookup: to find if $y \in \mathcal{U}$ is in $T$, check the linked list at $T[h(y)]$. Time proportion to size of linked list.


Does hashing give $O(1)$ time per operation for dictionaries?

## Hash Functions

Parameters: $N=|\mathcal{U}|$ (very large), $m=|T|, n=|S|$
Goal: $O(1)$-time lookup, insertion, deletion.

## Single hash function

If $N \geq \boldsymbol{m}^{2}$, then for any hash function $\boldsymbol{h}: \mathcal{U} \rightarrow \boldsymbol{T}$ there exists $\boldsymbol{i}<\boldsymbol{m}$ such that at least $\boldsymbol{N} / \boldsymbol{m} \geq \boldsymbol{m}$ elements of $\mathcal{U}$ get hashed to slot $i$.

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## Lesson:

- Consider a family $\mathcal{H}$ of hash functions with good properties and choose $h$ uniformly at random.
- Guarantees: small \# collisions in expectation for a given $S$.
- $\mathcal{H}$ should allow efficient sampling.


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(1) Uniform: Consider any element $x \in \mathcal{U}$. Then if $\boldsymbol{h} \in \mathcal{H}$ is picked randomly then $x$ should go into a random slot in $\boldsymbol{T}$. In other words $\operatorname{Pr}[h(x)=i]=\mathbf{1} / \boldsymbol{m}$ for every $\mathbf{0} \leq \boldsymbol{i}<\boldsymbol{m}$.

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(2) Universal: Consider any two distinct elements $x, y \in \mathcal{U}$. Then if $\boldsymbol{h} \in \mathcal{H}$ is picked randomly then the probability of a collision between $x$ and $y$ should be at most $\mathbf{1 / m}$. In other words $\operatorname{Pr}[h(x)=h(y)]=1 / m$ (cannot be smaller).

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(0) Second property is stronger than the first and the crucial issue.

## Definition

A family of hash function $\mathcal{H}$ is (2-)universal if for all distinct $x, y \in \mathcal{U}, \operatorname{Pr}_{h \sim \mathcal{H}}[h(x)=h(y)]=1 / m$ where $m$ is the table size.

## Analyzing Universal Hashing

Question: What is the expected time to look up $x$ in $T$ using $h$ assuming chaining used to resolve collisions?

Answer: $O(n / m)$.

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Answer: $O(n / m)$.
Comments:
(1) $O(1)$ expected time also holds for insertion.
(2) Analysis assumes static set $S$ but holds as long as $S$ is a set formed with at most $O(m)$ insertions and deletions.
(0) Worst-case: look up time can be large! How large? $\Omega(\log n / \log \log n)$

## Compact Universal Hash Family

Parameters: $N=|\mathcal{U}|, m=|T|, n=|S|$
(1) Choose a prime number $p>N$. Define function $h_{a, b}(x)=((a x+b) \bmod p) \bmod m$.
(2) Let $\mathcal{H}=\left\{h_{a, b} \mid a, b \in \mathbb{Z}_{p}, a \neq 0\right\}\left(\mathbb{Z}_{p}=\{0,1, \ldots, p-1\}\right)$.

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## Theorem

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## $\mathcal{H}$ is a universal hash family.

Comments:
(1) $\boldsymbol{h}_{\mathrm{a}, \mathrm{b}}$ can be evaluated in $O(\mathbf{1})$ time.
(2) Easy to store, i.e., just store $\boldsymbol{a}, \boldsymbol{b}$. Easy to sample.

## Some math required...

## Lemma (LemmaUnique)

Let $\boldsymbol{p}$ be a prime number, and $\mathbb{Z}_{\boldsymbol{p}}=\{\mathbf{0}, \mathbf{1}, \ldots, p-1\}$.
$\boldsymbol{x}$ : an integer number in $\mathbb{Z}_{\boldsymbol{p}}, \boldsymbol{x} \neq \mathbf{0}$
$\Longrightarrow$ There exists a unique $y \in \mathbb{Z}_{p}$ s.t. $x y=1 \bmod p$.

In other words: For every element there is a unique inverse.
$\Longrightarrow$ set $\mathbb{Z}_{\boldsymbol{p}}=\{0,1, \ldots, \boldsymbol{p}-1\}$ when working modulo $p$ is a field.

## Proof of LemmaUnique

## Claim

Let $p$ be a prime number. For any $x, y, z \in\{1, \ldots, p-1\}$ s.t. $y \neq z$, we have that $x y \bmod p \neq x z \bmod p$.

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## Proof.

Assume for the sake of contradiction $x y \bmod p=x z \bmod p$. Then

$$
\begin{aligned}
& x(y-z)=0 \quad \bmod p \\
& \Longrightarrow p \text { divides } x(y-z) \\
& \Longrightarrow p \text { divides } x \text { OR } p \text { divides }(y-x) \\
& \Longrightarrow y-z=0 \Longrightarrow y=z
\end{aligned}
$$

And that is a contradiction.

## Proof of LemmaUnique

## Lemma (LemmaUnique)

Let $p$ be a prime number, $x$ : an integer number in $\{\mathbf{1}, \ldots, p-1\}$.
$\Longrightarrow$ There exists a unique $y$ s.t. $x y=1 \bmod p$.

## Proof.

By the above claim if $x y=1 \bmod p$ and $x z=1 \bmod p$ then $y=z$. Hence uniqueness follows.

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Existence. For any $x \in\{1, \ldots, p-1\}$ we have that $\{x * 1 \bmod p, x * 2 \bmod p, \ldots, x *(p-1) \bmod p\}=$

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Existence. For any $x \in\{1, \ldots, p-1\}$ we have that $\{x * 1 \bmod p, x * 2 \bmod p, \ldots, x *(p-1) \bmod p\}=$ $\{1,2, \ldots, p-1\}$.
$\Longrightarrow$ There exists a number $y \in\{1, \ldots, p-1\}$ such that $x y=1 \bmod p$.

## Proof of the Theorem: Outline

$$
\left.h_{a, b}(x)=((a x+b) \bmod p) \bmod m\right)
$$

## Theorem

$$
\mathcal{H}=\left\{h_{a, b} \mid a, b \in \mathbb{Z}_{\boldsymbol{p}}, a \neq 0\right\} \text { is universal. }
$$

## Proof.

$\operatorname{Fix} x, y \in \mathcal{U}$. Show that $\operatorname{Pr}_{\boldsymbol{h}_{\mathbf{a}, \boldsymbol{b}} \sim \mathcal{H}}\left[\boldsymbol{h}_{\mathbf{a}, \boldsymbol{b}}(\boldsymbol{x})=\boldsymbol{h}_{\mathbf{a}, \boldsymbol{b}}(\boldsymbol{y})\right] \leq \mathbf{1} / \boldsymbol{m}$. Note that $|\mathcal{H}|=p(p-1)$.

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(1) Let $(\boldsymbol{a}, \boldsymbol{b})$ (equivalently $\boldsymbol{h}_{\boldsymbol{a}, \boldsymbol{b}}$ ) be bad for $\boldsymbol{x}, \boldsymbol{y}$ if

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(2) Claim: Number of bad $(a, b)$ is at most $p(p-1) / m$.

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(1) Let $(\boldsymbol{a}, \boldsymbol{b})$ (equivalently $\boldsymbol{h}_{\mathbf{a}, \boldsymbol{b}}$ ) be bad for $\boldsymbol{x}, \boldsymbol{y}$ if $\boldsymbol{h}_{a, b}(x)=\boldsymbol{h}_{a, b}(y)$. At most howmany bad $\boldsymbol{h}$ is ok?
(2) Claim: Number of bad $(a, b)$ is at most $p(p-1) / m$.
(3) Total number of hash functions is $p(p-1)$ and hence probability of a collision is $\leq \mathbf{1 / m}$.

## Intuition for the Claim

$$
g_{a, b}(x)=(a x+b) \bmod p, h_{a, b}(x)=\left(g_{a, b}(x)\right) \bmod m
$$

$$
\text { First } \operatorname{map} x \neq y \text { to } r=g_{a, b}(x) \text { and } s=g_{a, b}(y)
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LemmaUnique proof $\Longrightarrow r \neq s$


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First map $x \neq y$ to $r=g_{a, b}(x)$ and $s=g_{a, b}(y)$.
LemmaUnique proof $\Longrightarrow r \neq s$


As $(a, b)$ varies, $(r, s)$ takes all possible $p(p-1)$ values. Since $(a, b)$ is picked u.a.r., every value of $(r, s)$ has equal probability.

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$$
g_{a, b}(x)=(a x+b) \bmod p, h_{a, b}(x)=\left(g_{a, b}(x)\right) \bmod m
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$$
\Longrightarrow_{\text {mod } m}
$$



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g_{a, b}(x)=(a x+b) \bmod p, \quad h_{a, b}(x)=\left(g_{a, b}(x)\right) \bmod m
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## Intuition for the Claim

For a fixed $\boldsymbol{a} \in\{\mathbf{0}, \ldots, \boldsymbol{m} \mathbf{- 1}\}$ what is an upper bound on the size of set $\{s \in\{0, \ldots,(p-1)\} \mid a=s \bmod m\}$ ?
(A) $m$.
(B) $m^{2}$.
(C) $p$.
(D) $p / m$.
(E) Many. At least two.

## Intuition for the Claim

$g_{a, b}(x)=(a x+b) \bmod p, h_{a, b}(x)=\left(g_{a, b}(x)\right) \bmod m$
(1) First part of mapping maps $(x, y)$ to a random location $\left(g_{a, b}(x), g_{a, b}(y)\right)$ in the "matrix".
(2) $\left(g_{a, b}(x), g_{a, b}(y)\right)$ is not on main diagonal.
(3) All blue locations are "bad" map by $\bmod \boldsymbol{m}$ to a location of collision.
(4) But... at most $\mathbf{1} / \boldsymbol{m}$ fraction of allowable locations in the matrix are bad.


## We need

to show at most $1 / \mathrm{m}$ fraction of bad $\boldsymbol{h}_{a, b}$

$$
h_{a, b}(x)=(((a x+b) \bmod p) \operatorname{modm})
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2 lemmas ...
Fix $x \neq y \in \mathbb{Z}_{p}$, and let $r=(a x+b) \bmod p$ and $s=(a y+b) \bmod p$.

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(1) 1-to-1 correspondence between $p(p-1)$ pairs of $(a, b)$ (equivalently $h_{a, b}$ ) and $p(p-1)$ pairs of $(r, s)$.

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(1) 1-to-1 correspondence between $p(p-1)$ pairs of $(a, b)$ (equivalently $h_{a, b}$ ) and $p(p-1)$ pairs of $(r, s)$.
(2) Out of all possible $p(p-1)$ pairs of $(r, s)$, at most $p(p-1) / m$ fraction satisfies $r \bmod m=s \bmod m$.

## Some Lemmas

## Lemma

If $x \neq y$ then for any $a, b \in \mathbb{Z}_{\boldsymbol{p}}$ such that $\boldsymbol{a} \neq \mathbf{0}$, we have $a x+b \bmod p \neq a y+b \bmod p$.

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Suppose not
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Suppose not
$a x+b \bmod p=a y+b \bmod p \Rightarrow a(x-y) \bmod p=0$
Since $\boldsymbol{p}$ is a prime, $\boldsymbol{p}$ divides either $\boldsymbol{a}$ or $(\boldsymbol{x}-\boldsymbol{y})$. But $\boldsymbol{a}<\boldsymbol{p}$ and $(x-y)<p$, and hence $a=0$ or $(x-y)=0$. Contradiction!

## Some Lemmas

## Lemma

If $x \neq y$ then for each $(r, s)$ such that $r \neq s$ and $0 \leq r, s \leq p-1$ there is exactly one $a, b$ such that $a x+b \bmod p=r$ and $a y+b \bmod p=s$

## Proof.

Solve the two equations:

$$
a x+b=r \quad \bmod p \quad \text { and } \quad a y+b=s \quad \bmod p
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## Proof.

Solve the two equations:

$$
a x+b=r \quad \bmod p \quad \text { and } \quad a y+b=s \quad \bmod p
$$

We get $a=\frac{r-s}{x-y} \bmod p$ and $b=r-a x \bmod p$.
One-to-one correspondence between $(a, b)$ and $(r, s)$

## Understanding the hashing

Once we fix $\boldsymbol{a}$ and $\boldsymbol{b}$, and we are given a value $\boldsymbol{x}$, we compute the hash value of $x$ in two stages:
(1) Compute: $r \leftarrow(a x+b) \bmod p$.
(2) Fold: $r^{\prime} \leftarrow r \bmod m$

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Once we fix $\boldsymbol{a}$ and $\boldsymbol{b}$, and we are given a value $\boldsymbol{x}$, we compute the hash value of $x$ in two stages:
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(2) Fold: $r^{\prime} \leftarrow r \bmod m$

## Collision...

Given two distinct values $x$ and $y$ they might collide only because of folding.

## Lemma

\# not equal pairs $(r, s)$ of $\mathbb{Z}_{\boldsymbol{p}} \times \mathbb{Z}_{\boldsymbol{p}}$ that are folded to the same number is $p(p-1) / m$.

## Folding numbers

## Lemma

$\#$ pairs $(r, s) \in \mathbb{Z}_{\boldsymbol{p}} \times \mathbb{Z}_{\boldsymbol{p}}$ such that $r \neq s$ and $r \bmod m=s$ $\bmod m$ (folded to the same number) is $p(p-1) / m$.

## Proof.

Consider a pair $(r, s) \in\{0,1, \ldots, p-1\}^{2}$ s.t. $r \neq s$. Fix $r$ :
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(3) One of them is when $r=s$.

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(4) $\Longrightarrow \#$ of colliding pairs $(\lceil p / m\rceil-1) p \leq(p-1) p / m$

## Proof of Claim

 \# of bad pairs is $\mathrm{p}(\mathrm{p}-1) / \mathrm{m}$
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Let $a, b \in \mathbb{Z}_{\boldsymbol{p}}$ such that $\boldsymbol{a} \neq 0$ and $\boldsymbol{h}_{a, b}(x)=h_{a, b}(y)$.
(1) Let $r=a x+b \bmod p$ and $s=a y+b \bmod p$.
(2) Collision if and only if $r \bmod \boldsymbol{m}=\boldsymbol{s} \bmod \boldsymbol{m}$.
(3) (Folding error): Number of pairs $(r, s)$ such that $r \neq s$ and $0 \leq r, s \leq p-1$ and $r \bmod m=s \bmod m$ is $p(p-1) / m$.
(4) From previous lemma there is one-to-one correspondence between $(\boldsymbol{a}, \boldsymbol{b})$ and $(\boldsymbol{r}, \boldsymbol{s})$. Hence total number of bad $(\boldsymbol{a}, \boldsymbol{b})$ pairs is $p(p-1) / m$.

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Prob of $x$ and $y$ to collide: $\frac{\# \text { bad }(a, b) \text { pairs }}{\#(a, b) \text { pairs }}=\frac{\boldsymbol{p}(\boldsymbol{p}-1) / \boldsymbol{m}}{\boldsymbol{p}(\boldsymbol{p}-1)}=\frac{\mathbf{1}}{\boldsymbol{m}}$.

## Look up Time

Say $|S|=|T|=m$.
For $\mathbf{0} \leq \boldsymbol{i} \leq m-\mathbf{1}, \ell(\boldsymbol{i})$ : list of elements hashed to slot $\boldsymbol{i}$ in $\boldsymbol{T}$.

## Expected look up time

Since for $x \neq y, \operatorname{Pr}\left[h_{a, b}(x)=h_{a, b(y)}\right]=1 / m$, we get $\mathrm{E}[|\ell(i)|]=|S| / m \leq 1$.

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## What if $|T|=m^{2}\left(\#\right.$ Bins is $\left.m^{2}\right)$

Claim: If $|T|=m^{2}$, then $E\left[\max _{i=0}^{m-1}|\ell(i)|\right]=O(1)$.

## Perfect Hashing

## Two levels of hash tables

Question: Can we make look up time $\mathbf{O ( 1 )}$ in worst case?

## Perfect Hashing for Static Data

- Do hashing once.
- If $Y_{i}=|\ell(i)|>10$ then hash elements of $\ell(i)$ to a table of size $Y_{i}^{2}$.


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Expected worst case look up time is $O(1)$.
Lemma (Size)
If $|S|=O(m)$ then space usage of perfect hashing is $O(m)$.

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$\mathrm{E}\left[\sum_{i} \boldsymbol{m}_{i}^{2}\right]=m+2 \sum_{x<y} \operatorname{Pr}[h(x)=h(y)]=m+2 \frac{m(m-1)}{2} \frac{1}{m}<2 m$

## Rehashing, amortization and...

making the hash table dynamic
So far we assumed fixed $S$ of size $\simeq \boldsymbol{m}$.
Question: What happens as items are inserted and deleted?
(1) If $|S|$ grows to more than $\mathbf{c m}$ for some constant $c$ then hash table performance clearly degrades.
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(2) If $|\boldsymbol{S}|$ stays around $\simeq \boldsymbol{m}$ but incurs many insertions and deletions then the initial random hash function is no longer random enough!
Solution: Rebuild hash table periodically!
(1) Choose a new table size based on current number of elements in the table.
(2) Choose a new random hash function and rehash the elements.
(3) Discard old table and hash function.

Question: When to rebuild? How expensive?

## Rebuilding the hash table

(1) Start with table size $\boldsymbol{m}$ where $\boldsymbol{m}$ is some estimate of $|S|$ (can be some large constant).
(2) If $|S|$ grows to more than twice current table size, build new hash table (choose a new random hash function) with double the current number of elements. Can also use similar trick if table size falls below quarter the size.

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The amortize cost of rebuilding to previously performed operations. Rebuilding ensures $O(\mathbf{1})$ expected analysis holds even when $S$ changes. Hence $O(\mathbf{1})$ expected look up/insert/delete time dynamic data dictionary data structure!

## Bloom Filters

## Hashing:

(1) To insert $x$ in dictionary store $x$ in table in location $h(x)$
(2) To lookup $\boldsymbol{y}$ in dictionary check contents of location $\boldsymbol{h}(\boldsymbol{y})$
(3) Storing items in dictionary expensive in terms of memory, especially if items are unwieldy objects such a long strings, images, etc with non-uniform sizes.

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( To lookup $\boldsymbol{y}$ compute $\boldsymbol{h}_{\boldsymbol{i}}(\boldsymbol{y})$ for $\mathbf{1} \leq \boldsymbol{i} \leq \boldsymbol{k}$ and say yes only if each bit in the corresponding location is $\mathbf{1}$, otherwise say no. If probability of false positive for one hash function is $\alpha<\mathbf{1}$ then with $\boldsymbol{k}$ independent hash function it is $\boldsymbol{\alpha}^{\boldsymbol{k}}$.

## Take away points

(1) Hashing is a powerful and important technique for dictionaries. Many practical applications.
(2) Randomization fundamental to understand hashing.
(3) Good and efficient hashing possible in theory and practice with proper definitions (universal, perfect, etc).
(4) Related ideas of creating a compact fingerprint/sketch for objects is very powerful in theory and practice.

## Practical Issues

Hashing used typically for integers, vectors, strings etc.

- Universal hashing is defined for integers. To implement for other objects need to map objects in some fashion to integers (via representation)
- Practical methods for various important cases such as vectors, strings are studied extensively. See http://en.wikipedia.org/wiki/Universal_hashing for some pointers.
- Details on Cuckoo hashing and its advantage over chaining http://en.wikipedia.org/wiki/Cuckoo_hashing.
- Relatively recent important paper bridging theory and practice of hashing. "The power of simple tabulation hashing" by Mikkel Thorup and Mihai Patrascu, 2011. See http://en.wikipedia.org/wiki/Tabulation_hashing
- Cryptographic hash functions have a different motivation and

