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Dictionary:

- ► *S*.insert(*x*): Insert an element *x*.
- ► *S*.delete(*x*): Delete the element pointed to by *x*.
- S.search(k): Return a pointer to an element *e* with key[*e*] = k in S if it exists; otherwise return null.

So far we have implemented the search for a key by carefully choosing split-elements.

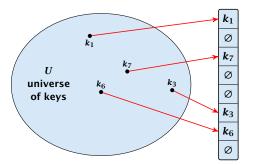
Then the memory location of an object x with key k is determined by successively comparing k to split-elements.

Hashing tries to directly compute the memory location from the given key. The goal is to have constant search time.

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Ideally the hash function maps all keys to different memory locations.



This special case is known as Direct Addressing. It is usually very unrealistic as the universe of keys typically is quite large, and in particular larger than the available memory.

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Definitions:

- ▶ Universe *U* of keys, e.g., $U \subseteq \mathbb{N}_0$. *U* very large.
- Set $S \subseteq U$ of keys, $|S| = m \le |U|$.
- Array $T[0, \ldots, n-1]$ hash-table.
- ▶ Hash function $h: U \rightarrow [0, ..., n-1]$.

The hash-function *h* should fulfill:

- Fast to evaluate.
- Small storage requirement.
- Good distribution of elements over the whole table.

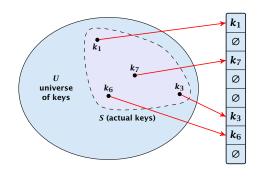
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7.7 Hashing

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Suppose that we know the set S of actual keys (no insert/no delete). Then we may want to design a simple hash-function that maps all these keys to different memory locations.



Such a hash function h is called a perfect hash function for set S.

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If we do not know the keys in advance, the best we can hope for is that the hash function distributes keys evenly across the table.

Problem: Collisions

Usually the universe U is much larger than the table-size n.

Hence, there may be two elements k_1, k_2 from the set S that map to the same memory location (i.e., $h(k_1) = h(k_2)$). This is called a collision.

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Proof.

Let $A_{m,n}$ denote the event that inserting m keys into a table of size n does not generate a collision. Then

$$\Pr[A_{m,n}] = \prod_{\ell=1}^{m} \frac{n-\ell+1}{n} = \prod_{j=0}^{m-1} \left(1 - \frac{j}{n}\right)$$
$$\leq \prod_{j=0}^{m-1} e^{-j/n} = e^{-\sum_{j=0}^{m-1} \frac{j}{n}} = e^{-\frac{m(m-1)}{2n}}$$

Here the first equality follows since the ℓ -th element that is hashed has a probability of $\frac{n-\ell+1}{n}$ to not generate a collision under the condition that the previous elements did not induce collisions.

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Typically, collisions do not appear once the size of the set *S* of actual keys gets close to *n*, but already when $|S| \ge \omega(\sqrt{n})$.

Lemma 1

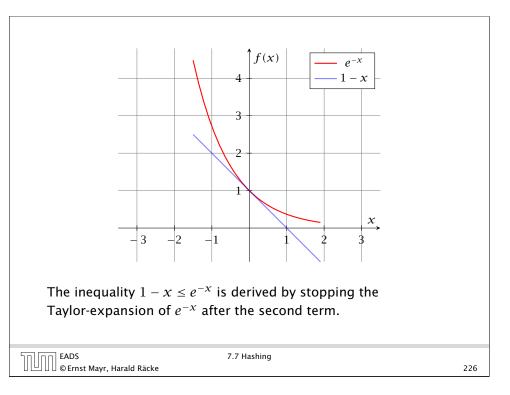
The probability of having a collision when hashing m elements into a table of size n under uniform hashing is at least

 $1 - e^{-\frac{m(m-1)}{2n}} \approx 1 - e^{-\frac{m^2}{2n}}$.

Uniform hashing:

Choose a hash function uniformly at random from all functions $f: U \rightarrow [0, ..., n-1].$

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Resolving Collisions

The methods for dealing with collisions can be classified into the two main types

- open addressing, aka. closed hashing
- hashing with chaining, aka. closed addressing, open hashing.

There are applications e.g. computer chess where you do not resolve collisions at all.

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Let A denote a strategy for resolving collisions. We use the following notation:

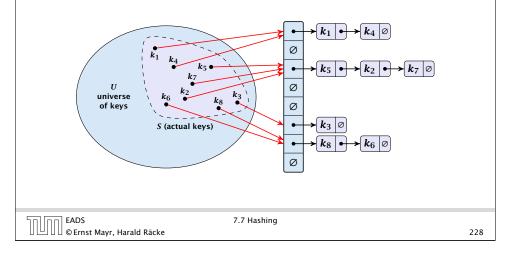
- A⁺ denotes the average time for a successful search when using A;
- A⁻ denotes the average time for an unsuccessful search when using A;
- We parameterize the complexity results in terms of $\alpha := \frac{m}{n}$, the so-called fill factor of the hash-table.

We assume uniform hashing for the following analysis.

Hashing with Chaining

Arrange elements that map to the same position in a linear list.

- Access: compute h(x) and search list for key[x].
- Insert: insert at the front of the list.



Hashing with Chaining

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The time required for an unsuccessful search is 1 plus the length of the list that is examined. The average length of a list is $\alpha = \frac{m}{n}$. Hence, if *A* is the collision resolving strategy "Hashing with Chaining" we have

$$A^- = 1 + \alpha$$

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Hashing with Chaining

For a successful search observe that we do **not** choose a list at random, but we consider a random key k in the hash-table and ask for the search-time for k.

This is 1 plus the number of elements that lie before k in k's list.

Let k_ℓ denote the ℓ -th key inserted into the table.

Let for two keys k_i and k_j , X_{ij} denote the indicator variable for the event that k_i and k_j hash to the same position. Clearly, $\Pr[X_{ij} = 1] = 1/n$ for uniform hashing.

The expected successful search cost is

keys before k_i $\mathbf{E}\left[\frac{1}{m}\sum_{i=1}^{m}\left(1+\sum_{j=i+1}^{m}X_{ij}\right)\right]$

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Hashing with Chaining

Disadvantages:

- pointers increase memory requirements
- pointers may lead to bad cache efficiency

Advantages:

- no à priori limit on the number of elements
- deletion can be implemented efficiently
- by using balanced trees instead of linked list one can also obtain worst-case guarantees.

Hashing with Chaining

$$\begin{split} \operatorname{E}\left[\frac{1}{m}\sum_{i=1}^{m}\left(1+\sum_{j=i+1}^{m}X_{ij}\right)\right] &= \frac{1}{m}\sum_{i=1}^{m}\left(1+\sum_{j=i+1}^{m}\operatorname{E}\left[X_{ij}\right]\right) \\ &= \frac{1}{m}\sum_{i=1}^{m}\left(1+\sum_{j=i+1}^{m}\frac{1}{n}\right) \\ &= 1+\frac{1}{mn}\sum_{i=1}^{m}(m-i) \\ &= 1+\frac{1}{mn}\left(m^{2}-\frac{m(m+1)}{2}\right) \\ &= 1+\frac{m-1}{2n} = 1+\frac{\alpha}{2}-\frac{\alpha}{2m} \end{split}$$

Hence, the expected cost for a successful search is $A^{+} \leq 1+\frac{\alpha}{2}$

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Open Addressing

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All objects are stored in the table itself.

Define a function h(k, j) that determines the table-position to be examined in the *j*-th step. The values $h(k, 0), \ldots, h(k, n - 1)$ must form a permutation of $0, \ldots, n - 1$.

Search(k): Try position h(k, 0); if it is empty your search fails; otw. continue with h(k, 1), h(k, 2),

Insert(x): Search until you find an empty slot; insert your element there. If your search reaches h(k, n - 1), and this slot is non-empty then your table is full.

Open Addressing

Choices for h(k, j):

- Linear probing: $h(k,i) = h(k) + i \mod n$ (sometimes: $h(k, i) = h(k) + ci \mod n$).
- Quadratic probing: $h(k,i) = h(k) + c_1 i + c_2 i^2 \mod n.$
- Double hashing: $h(k, i) = h_1(k) + ih_2(k) \mod n.$

For quadratic probing and double hashing one has to ensure that the search covers all positions in the table (i.e., for double hashing $h_2(k)$ must be relatively prime to *n* (teilerfremd); for quadratic probing c_1 and c_2 have to be chosen carefully).

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Quadratic Probing

- Not as cache-efficient as Linear Probing.
- Secondary clustering: caused by the fact that all keys mapped to the same position have the same probe sequence.

Lemma 3

Let Q be the method of quadratic probing for resolving collisions:

$$Q^+ pprox 1 + \ln\left(rac{1}{1-lpha}
ight) - rac{lpha}{2}$$
 $Q^- pprox rac{1}{1-lpha} + \ln\left(rac{1}{1-lpha}
ight) - lpha$

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Linear Probing

- Advantage: Cache-efficiency. The new probe position is very likely to be in the cache.
- Disadvantage: Primary clustering. Long sequences of occupied table-positions get longer as they have a larger probability to be hit. Furthermore, they can merge forming larger sequences.

Lemma 2

Let *L* be the method of linear probing for resolving collisions:

$$L^+ \approx \frac{1}{2} \left(1 + \frac{1}{1 - \alpha} \right)$$

$$L^{-} \approx \frac{1}{2} \left(1 + \frac{1}{(1-\alpha)^2} \right)$$

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Double Hashing

Any probe into the hash-table usually creates a cache-miss.

Lemma 4

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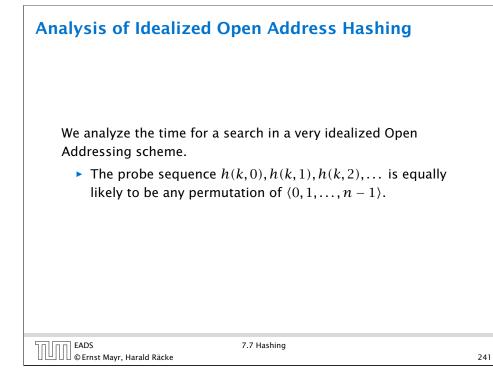
Let A be the method of double hashing for resolving collisions:

$$D^+ pprox rac{1}{lpha} \ln\left(rac{1}{1-lpha}
ight)$$
 $D^- pprox rac{1}{1-lpha}$

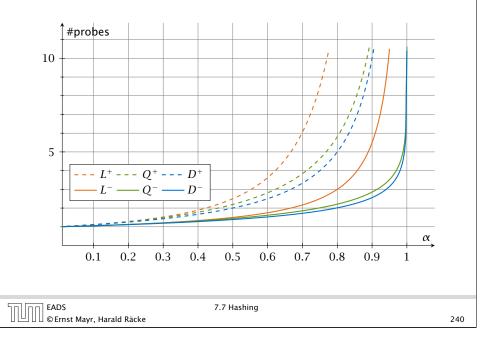
Open Addressing

Some values:

| L+ L- Q+ Q- D+ D- 0.5 1.5 2.5 1.44 2.19 1.39 2 0.9 5.5 50.5 2.85 11.40 2.55 10 0.95 10.5 200.5 3.52 22.05 3.15 20 | α | Linear | Probing | Quadrati | c Probing | Double | Hashing |
|---|------|--------|---------|----------|-----------|--------|---------|
| 0.9 5.5 50.5 2.85 11.40 2.55 10 | | L^+ | L^- | Q^+ | Q^- | D^+ | D^- |
| | 0.5 | 1.5 | 2.5 | 1.44 | 2.19 | 1.39 | 2 |
| 0.95 10.5 200.5 3.52 22.05 3.15 20 | 0.9 | 5.5 | 50.5 | 2.85 | 11.40 | 2.55 | 10 |
| | 0.95 | 10.5 | 200.5 | 3.52 | 22.05 | 3.15 | 20 |
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Open Addressing



Analysis of Idealized Open Address Hashing

Let X denote a random variable describing the number of probes in an unsuccessful search.

Let A_i denote the event that the *i*-th probe occurs and is to a non-empty slot.

$$\Pr[A_1 \cap A_2 \cap \dots \cap A_{i-1}]$$

$$= \Pr[A_1] \cdot \Pr[A_2 \mid A_1] \cdot \Pr[A_3 \mid A_1 \cap A_2] \cdot \dots \cdot \Pr[A_{i-1} \mid A_1 \cap \dots \cap A_{i-2}]$$

$$\Pr[X \ge i] = \frac{m}{n} \cdot \frac{m-1}{n-1} \cdot \frac{m-2}{n-2} \cdot \dots \cdot \frac{m-i+2}{n-i+2}$$

$$\leq \left(\frac{m}{n}\right)^{i-1} = \alpha^{i-1} .$$

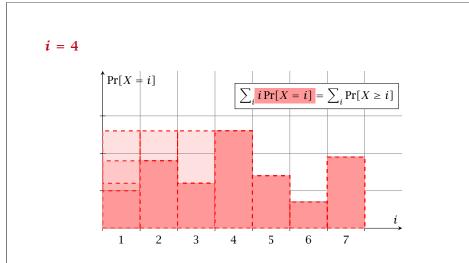
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Analysis of Idealized Open Address Hashing

$$\mathbb{E}[X] = \sum_{i=1}^{\infty} \Pr[X \ge i] \le \sum_{i=1}^{\infty} \alpha^{i-1} = \sum_{i=0}^{\infty} \alpha^{i} = \frac{1}{1-\alpha}$$

$$\frac{1}{1-\alpha} = 1 + \alpha + \alpha^2 + \alpha^3 + \dots$$

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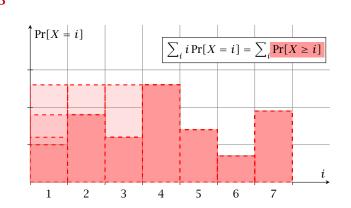


The *j*-th rectangle appears in both sums *j* times. (*j* times in the first due to multiplication with *j*; and *j* times in the second for summands i = 1, 2, ..., j)

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i = 3



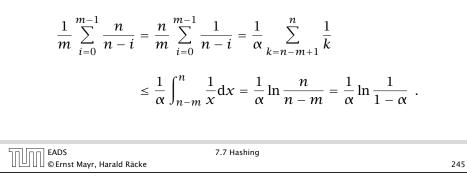
The *j*-th rectangle appears in both sums *j* times. (*j* times in the first due to multiplication with *j*; and *j* times in the second for summands i = 1, 2, ..., j)

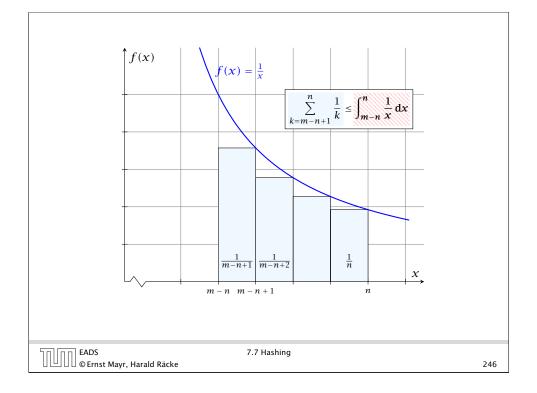
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Analysis of Idealized Open Address Hashing

The number of probes in a successful search for k is equal to the number of probes made in an unsuccessful search for k at the time that k is inserted.

Let *k* be the *i* + 1-st element. The expected time for a search for *k* is at most $\frac{1}{1-i/n} = \frac{n}{n-i}$.





Deletions

- Simply removing a key might interrupt the probe sequence of other keys which then cannot be found anymore.
- One can delete an element by replacing it with a deleted-marker.
 - During an insertion if a deleted-marker is encountered an element can be inserted there.
 - During a search a deleted-marker must not be used to terminate the probe sequence.
- The table could fill up with deleted-markers leading to bad performance.
- If a table contains many deleted-markers (linear fraction of the keys) one can rehash the whole table and amortize the cost for this rehash against the cost for the deletions.

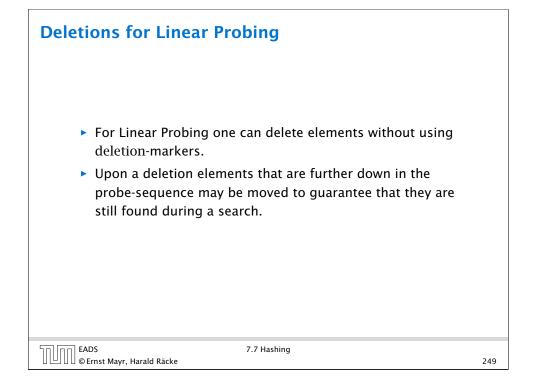
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Deletions

How do we delete in a hash-table?

- For hashing with chaining this is not a problem. Simply search for the key, and delete the item in the corresponding list.
- For open addressing this is difficult.

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| Deletions f | or Linear Probing |
|-------------|---|
| | |
| | Algorithm 16 delete(p) |
| | 1: $T[p] \leftarrow \text{null}$ |
| | 2: $p \leftarrow \operatorname{succ}(p)$ |
| | 3: while $T[p] \neq \text{null } \mathbf{do}$ |
| | 4: $y \leftarrow T[p]$ |
| | 5: $T[p] \leftarrow \text{null}$ |
| | 6: $p \leftarrow \operatorname{succ}(p)$ |
| | Algorithm 16 delete(p)1: $T[p] \leftarrow$ null2: $p \leftarrow$ succ(p)3: while $T[p] \neq$ null do4: $y \leftarrow T[p]$ 5: $T[p] \leftarrow$ null6: $p \leftarrow$ succ(p)7: insert(y) |

 \boldsymbol{p} is the index into the table-cell that contains the object to be deleted.

Pointers into the hash-table become invalid.

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7.7 Hashing

Universal Hashing

Definition 5

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A class \mathcal{H} of hash-functions from the universe U into the set $\{0, \ldots, n-1\}$ is called universal if for all $u_1, u_2 \in U$ with $u_1 \neq u_2$

$$\Pr[h(u_1) = h(u_2)] \le \frac{1}{n}$$

where the probability is w.r.t. the choice of a random hash-function from set \mathcal{H} .

Note that this means that the probability of a collision is at most $\frac{1}{n}$.

7.7 Hashing

Universal Hashing

Regardless, of the choice of hash-function there is always an input (a set of keys) that has a very poor worst-case behaviour.

Therefore, so far we assumed that the hash-function is random so that regardless of the input the average case behaviour is good.

However, the assumption of uniform hashing that h is chosen randomly from all functions $f: U \rightarrow [0, ..., n-1]$ is clearly unrealistic as there are $n^{|U|}$ such functions. Even writing down such a function would take $|U| \log n$ bits.

Universal hashing tries to define a set \mathcal{H} of functions that is much smaller but still leads to good average case behaviour when selecting a hash-function uniformly at random from \mathcal{H} .

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Universal Hashing

Definition 6

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A class \mathcal{H} of hash-functions from the universe U into the set $\{0, \ldots, n-1\}$ is called 2-independent (pairwise independent) if the following two conditions hold

- ▶ For any key $u \in U$, and $t \in \{0, ..., n-1\}$ $\Pr[h(u) = t] = \frac{1}{n}$, i.e., a key is distributed uniformly within the hash-table.
- For all u₁, u₂ ∈ U with u₁ ≠ u₂, and for any two hash-positions t₁, t₂:

$$\Pr[h(u_1) = t_1 \wedge h(u_2) = t_2] \le \frac{1}{n^2} .$$

This requirement clearly implies a universal hash-function.

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Universal Hashing

Definition 7

A class \mathcal{H} of hash-functions from the universe U into the set $\{0, \ldots, n-1\}$ is called *k*-independent if for any choice of $\ell \leq k$ distinct keys $u_1, \ldots, u_\ell \in U$, and for any set of ℓ not necessarily distinct hash-positions t_1, \ldots, t_ℓ :

 $\Pr[h(u_1) = t_1 \wedge \cdots \wedge h(u_\ell) = t_\ell] \leq \frac{1}{n^\ell} ,$

where the probability is w.r.t. the choice of a random hash-function from set \mathcal{H} .

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Universal Hashing

Let $U := \{0, ..., p - 1\}$ for a prime p. Let $\mathbb{Z}_p := \{0, ..., p - 1\}$, and let $\mathbb{Z}_p^* := \{1, ..., p - 1\}$ denote the set of invertible elements in \mathbb{Z}_p .

Define

 $h_{a,b}(x) := (ax + b \mod p) \mod n$

Lemma 9

The class

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 $\mathcal{H} = \{h_{a,b} \mid a \in \mathbb{Z}_p^*, b \in \mathbb{Z}_p\}$

is a universal class of hash-functions from U to $\{0, ..., n-1\}$.

7.7 Hashing

Universal Hashing

Definition 8

A class \mathcal{H} of hash-functions from the universe U into the set $\{0, \ldots, n-1\}$ is called (μ, k) -independent if for any choice of $\ell \leq k$ distinct keys $u_1, \ldots, u_\ell \in U$, and for any set of ℓ not necessarily distinct hash-positions t_1, \ldots, t_ℓ :

$$\Pr[h(u_1) = t_1 \wedge \cdots \wedge h(u_\ell) = t_\ell] \leq \frac{\mu}{n^\ell} ,$$

7.7 Hashing

where the probability is w. r. t. the choice of a random hash-function from set \mathcal{H} .

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Universal Hashing

Proof.

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Let $x, y \in U$ be two distinct keys. We have to show that the probability of a collision is only 1/n.

 $\bullet \ ax + b \not\equiv ay + b \pmod{p}$

If $x \neq y$ then $(x - y) \not\equiv 0 \pmod{p}$.

Multiplying with $a \not\equiv 0 \pmod{p}$ gives

 $a(x-y) \not\equiv 0 \pmod{p}$

where we use that \mathbb{Z}_p is a field (Körper) and, hence, has no zero divisors (nullteilerfrei).

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The hash-function does not generate collisions before the (mod n)-operation. Furthermore, every choice (a, b) is mapped to a different pair (t_x, t_y) with t_x := ax + b and t_y := ay + b.

This holds because we can compute *a* and *b* when given t_x and t_y :

| $t_x \equiv a$ | x + b | $(\mod p)$ | |
|-----------------------------------|---------------------------|------------|--|
| $t_{\mathcal{Y}} \equiv a$ | xy + b | $(\mod p)$ | |
| $t_x - t_y \equiv a$ | (x - y) | \pmod{p} | |
| $t_{\mathcal{Y}} \equiv a$ | xy + b | $(\mod p)$ | |
| $a \equiv ($ | $(t_x - t_y)(x - y)^{-1}$ | \pmod{p} | |
| $b \equiv t$ | y - ay | \pmod{p} | |
| | | | |
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Universal Hashing

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As $t_y \neq t_x$ there are

$$\left\lceil \frac{p}{n} \right\rceil - 1 \le \frac{p}{n} + \frac{n-1}{n} - 1 \le \frac{p-1}{n}$$

possibilities for choosing $t_{\mathcal{Y}}$ such that the final hash-value creates a collision.

This happens with probability at most $\frac{1}{n}$.

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Universal Hashing

There is a one-to-one correspondence between hash-functions (pairs (a, b), $a \neq 0$) and pairs (t_x, t_y) , $t_x \neq t_y$.

Therefore, we can view the first step (before the mod *n*-operation) as choosing a pair (t_x, t_y) , $t_x \neq t_y$ uniformly at random.

What happens when we do the mod n operation?

Fix a value t_x . There are p - 1 possible values for choosing t_y .

From the range 0, ..., p - 1 the values $t_x, t_x + n, t_x + 2n, ...$ map to t_x after the modulo-operation. These are at most $\lceil p/n \rceil$ values.

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Universal Hashing

It is also possible to show that ${\mathcal H}$ is an (almost) pairwise independent class of hash-functions.

$$\frac{\left\lfloor \frac{p}{n} \right\rfloor^2}{p(p-1)} \leq \Pr_{t_x \neq t_y \in \mathbb{Z}_p^2} \left[\begin{array}{c} t_x \mod n = h_1 \\ t_y \mod n = h_2 \end{array} \right] \leq \frac{\left\lceil \frac{p}{n} \right\rceil^2}{p(p-1)}$$

Note that the middle is the probability that $h(x) = h_1$ and $h(y) = h_2$. The total number of choices for (t_x, t_y) is p(p-1). The number of choices for t_x (t_y) such that $t_x \mod n = h_1$ $(t_y \mod n = h_2)$ lies between $\lfloor \frac{p}{n} \rfloor$ and $\lceil \frac{p}{n} \rceil$.

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Definition 10 Let $d \in \mathbb{N}$; $q \ge (d + 1)n$ be a prime; and let $\vec{a} \in \{0, ..., q - 1\}^{d+1}$. Define for $x \in \{0, ..., q\}$ $h_{\vec{a}}(x) := \left(\sum_{i=0}^{d} a_i x^i \mod q\right) \mod n$. Let $\mathcal{H}_n^d := \{h_{\vec{a}} \mid \vec{a} \in \{0, ..., q\}^{d+1}\}$. The class \mathcal{H}_n^d is (e, d + 1)-independent. Note that in the previous case we had d = 1 and chose $a_d \neq 0$.

Fix $\ell \leq d+1$; let $x_1, \ldots, x_\ell \in \{0, \ldots, q-1\}$ be keys, and let t_1, \ldots, t_ℓ denote the corresponding hash-function values. Let $A^{\ell} = \{h_{\tilde{a}} \in \mathcal{H} \mid h_{\tilde{a}}(x_i) = t_i \text{ for all } i \in \{1, \dots, \ell\}\}$ Then $h_{\bar{a}} \in A^{\ell} \Leftrightarrow h_{\bar{a}} = f_{\bar{a}} \mod n$ and $f_{\bar{a}}(x_i) \in \underbrace{\{t_i + \alpha \cdot n \mid \alpha \in \{0, \dots, \lceil \frac{q}{n} \rceil - 1\}\}}_{=:B_i}$ In order to obtain the cardinality of A^{ℓ} we choose our polynomial by fixing d + 1 points. • A^{ℓ} denotes the set of hash-We first fix the values for inputs x_1, \ldots, x_ℓ . functions such that every x_i hits its pre-defined position We have ti. $|B_1|\cdot\ldots\cdot|B_\ell|$ • B_i is the set of positions that $f_{\bar{a}}$ can hit so that $h_{\bar{a}}$ still hits possibilities to do this (so that $h_{\bar{a}}(x_i) = t_i$). EADS 7.7 Hashing 🛛 🗌 © Ernst Mayr, Harald Räcke 264

For the coefficients $\bar{a} \in \{0, \dots, q-1\}^{d+1}$ let $f_{\bar{a}}$ denote the polynomial

$$f_{\bar{a}}(x) = \left(\sum_{i=0}^{a} a_i x^i\right) \mod q$$

The polynomial is defined by d + 1 distinct points.

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Now, we choose $d - \ell + 1$ other inputs and choose their value arbitrarily. We have $q^{d-\ell+1}$ possibilities to do this.

Therefore we have

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$$|B_1| \cdot \ldots \cdot |B_\ell| \cdot q^{d-\ell+1} \le \lceil \frac{q}{n} \rceil^\ell \cdot q^{d-\ell+1}$$

possibilities to choose \bar{a} such that $h_{\bar{a}} \in A_{\ell}$.

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Therefore the probability of choosing $h_{\tilde{a}}$ from A_{ℓ} is only

$$\frac{\lceil \frac{q}{n} \rceil^{\ell} \cdot q^{d-\ell+1}}{q^{d+1}} \le \frac{(\frac{q+n}{n})^{\ell}}{q^{\ell}} \le \left(\frac{q+n}{q}\right)^{\ell} \cdot \frac{1}{n^{\ell}}$$
$$\le \left(1 + \frac{1}{\ell}\right)^{\ell} \cdot \frac{1}{n^{\ell}} \le \frac{e}{n^{\ell}} \quad .$$

This shows that the \mathcal{H} is (e, d + 1)-universal.

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7.7 Hashing

Perfect Hashing

Let m = |S|. We could simply choose the hash-table size very large so that we don't get any collisions.

Using a universal hash-function the expected number of collisions is

 $\mathbf{E}[\texttt{\#Collisions}] = \binom{m}{2} \cdot \frac{1}{n} \ .$

If we choose $n = m^2$ the expected number of collisions is strictly less than $\frac{1}{2}$.

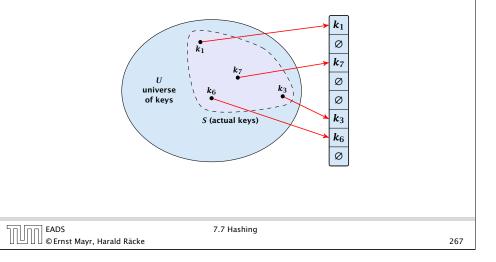
Can we get an upper bound on the probability of having collisions?

The probability of having 1 or more collisions can be at most $\frac{1}{2}$ as otherwise the expectation would be larger than $\frac{1}{2}$.

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Perfect Hashing

Suppose that we **know** the set *S* of actual keys (no insert/no delete). Then we may want to design a **simple** hash-function that maps all these keys to different memory locations.



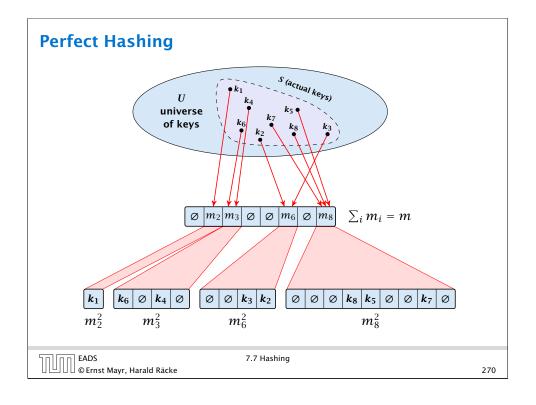
Perfect Hashing

We can find such a hash-function by a few trials.

However, a hash-table size of $n = m^2$ is very very high.

We construct a two-level scheme. We first use a hash-function that maps elements from S to m buckets.

Let m_j denote the number of items that are hashed to the *j*-th bucket. For each bucket we choose a second hash-function that maps the elements of the bucket into a table of size m_j^2 . The second function can be chosen such that all elements are mapped to different locations.



Perfect Hashing

We need only $\mathcal{O}(m)$ time to construct a hash-function *h* with $\sum_{i} m_{i}^{2} = O(4m)$, because with probability at least 1/2 a random function from a universal family will have this property.

Then we construct a hash-table h_i for every bucket. This takes expected time $\mathcal{O}(m_i)$ for every bucket. A random function h_i is collision-free with probability at least 1/2. We need $\mathcal{O}(m_i)$ to test this.

We only need that the hash-functions are chosen from a universal family!!!

Perfect Hashing

The total memory that is required by all hash-tables is $\mathcal{O}(\sum_{i} m_{i}^{2})$. Note that m_{j} is a random variable.

$$E\left[\sum_{j} m_{j}^{2}\right] = E\left[2\sum_{j} \binom{m_{j}}{2} + \sum_{j} m_{j}\right]$$
$$= 2E\left[\sum_{j} \binom{m_{j}}{2}\right] + E\left[\sum_{j} m_{j}\right]$$

The first expectation is simply the expected number of collisions, for the first level. Since we use universal hashing we have

$$=2\binom{m}{2}\frac{1}{m}+m=2m-1$$

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Cuckoo Hashing

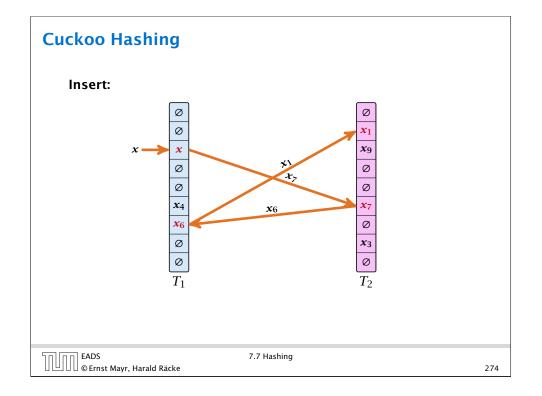
Goal:

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Try to generate a hash-table with constant worst-case search time in a dynamic scenario.

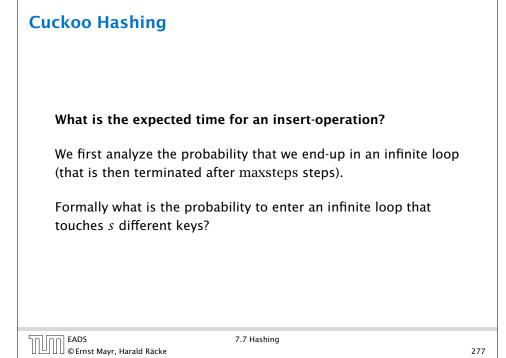
- ▶ Two hash-tables $T_1[0, \ldots, n-1]$ and $T_2[0, \ldots, n-1]$, with hash-functions h_1 , and h_2 .
- An object x is either stored at location $T_1[h_1(x)]$ or $T_2[h_2(x)].$
- A search clearly takes constant time if the above constraint is met.

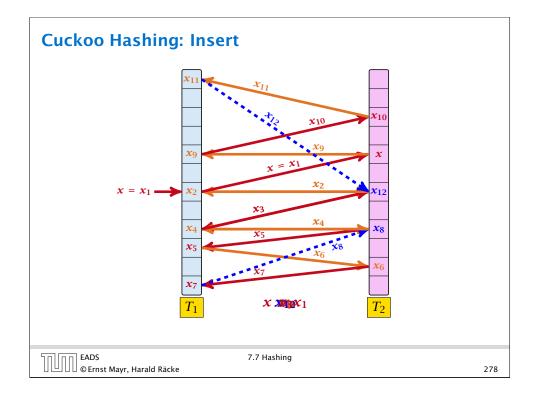


- We call one iteration through the while-loop a step of the algorithm.
- We call a sequence of iterations through the while-loop without the termination condition becoming true a phase of the algorithm.
- We say a phase is successful if it is not terminated by the maxstep-condition, but the while loop is left because x =null.

Cuckoo Hashing

| Algorithm 17 Cuckoo-Insert(x) | |
|--|-----|
| 1: if $T_1[h_1(x)] = x \lor T_2[h_2(x)] = x$ then return | |
| 2: steps ← 1 | |
| 3: while steps ≤ maxsteps do | |
| 4: exchange x and $T_1[h_1(x)]$ | |
| 5: if $x =$ null then return | |
| 6: exchange x and $T_2[h_2(x)]$ | |
| 7: if <i>x</i> = null then return | |
| 8: steps \leftarrow steps +1 | |
| 9: rehash() // change hash-functions; rehash everything | |
| 10: Cuckoo-Insert(x) | |
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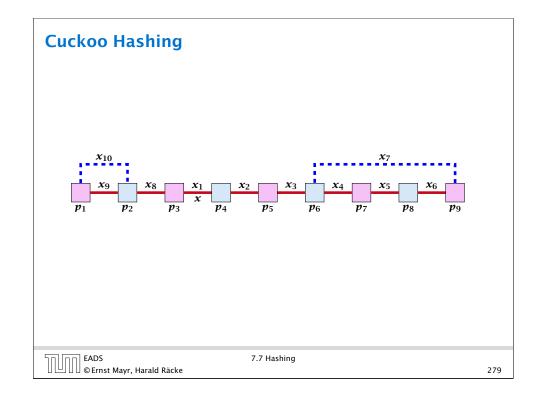


A cycle-structure is active if for every key x_{ℓ} (linking a cell p_i from T_1 and a cell p_i from T_2) we have

 $h_1(x_{\ell}) = p_i$ and $h_2(x_{\ell}) = p_i$

Observation:

If during a phase the insert-procedure runs into a cycle there must exist an active cycle structure of size $s \ge 3$.



Cuckoo Hashing

What is the probability that all keys in a cycle-structure of size *s* correctly map into their T_1 -cell?

This probability is at most $\frac{\mu}{n^s}$ since h_1 is a (μ, s) -independent hash-function.

What is the probability that all keys in the cycle-structure of size s correctly map into their T_2 -cell?

This probability is at most $\frac{\mu}{n^s}$ since h_2 is a (μ, s) -independent hash-function.

These events are independet.

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The probability that a given cycle-structure of size *s* is active is at most $\frac{\mu^2}{n^{2s}}$.

What is the probability that there exists an active cycle structure of size *s*?

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Cuckoo Hashing

The probability that there exists an active cycle-structure is therefore at most

$$\begin{split} \sum_{s=3}^{\infty} s^3 \cdot n^{s-1} \cdot m^{s-1} \cdot \frac{\mu^2}{n^{2s}} &= \frac{\mu^2}{nm} \sum_{s=3}^{\infty} s^3 \left(\frac{m}{n}\right)^s \\ &\leq \frac{\mu^2}{m^2} \sum_{s=3}^{\infty} s^3 \left(\frac{1}{1+\epsilon}\right)^s \leq \mathcal{O}\left(\frac{1}{m^2}\right) \end{split}$$

Here we used the fact that $(1 + \epsilon)m \le n$.

Hence,

$$\Pr[\mathsf{cycle}] = \mathcal{O}\left(\frac{1}{m^2}\right)$$
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Cuckoo Hashing

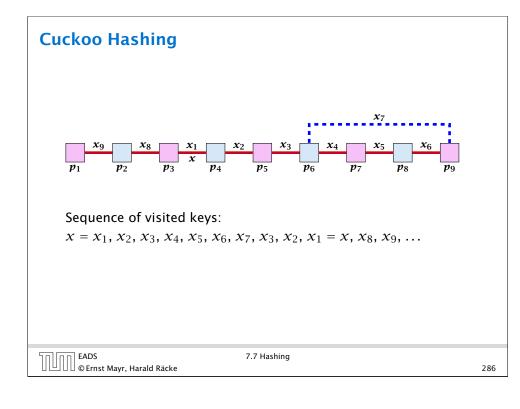
The number of cycle-structures of size *s* is at most

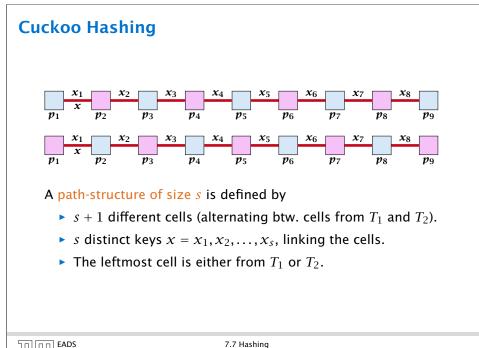
 $s^3 \cdot n^{s-1} \cdot m^{s-1}$.

- There are at most s² possibilities where to attach the forward and backward links.
- There are at most s possibilities to choose where to place key x.
- There are m^{s-1} possibilities to choose the keys apart from x.
- There are n^{s-1} possibilities to choose the cells.

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| Cuckoo Hashing | |
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| Now, we analyze the probability that a phase is not successful without running into a closed cycle. | |
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Cuckoo Hashing

Consider the sequence of not necessarily distinct keys starting with x in the order that they are visited during the phase.

Lemma 11

If the sequence is of length p then there exists a sub-sequence of at least p/3 keys starting with x of distinct keys.

Proof.

x is contained at most twice in the sequence.

Either the sub-sequence starting from x until right before the first repeated key, or the sub-sequence starting from the repetition of x until the end must contain at least p/3 distinct keys.

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7.7 Hashing

Cuckoo Hashing A path-structure is active if for every key x_{ℓ} (linking a cell p_i from T_1 and a cell p_j from T_2) we have $h_1(x_{\ell}) = p_i$ and $h_2(x_{\ell}) = p_j$ Observation:

If a phase takes at least t steps without running into a cycle there must exist an active path-structure of size (2t - 1)/3.

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The probability that a given path-structure of size *s* is active is at most $\frac{\mu^2}{n^{2s}}$.

The probability that there exists an active path-structure of size s is at most

$$\begin{aligned} 2 \cdot n^{s+1} \cdot m^{s-1} \cdot \frac{\mu^2}{n^{2s}} \\ &\leq 2\mu^2 \left(\frac{m}{n}\right)^{s-1} \leq 2\mu^2 \left(\frac{1}{1+\epsilon}\right)^{s-1} \\ &\leq 2\mu^2 \left(\frac{1}{1+\epsilon}\right)^{(2t-1)/3-1} = 2\mu^2 \left(\frac{1}{1+\epsilon}\right)^{2(t-2)/3} \end{aligned}$$

7.7 Hashing

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Cuckoo Hashing

The expected number of steps in the successful phase of an insert operation is:

E[number of steps | phase successful]

 $= \sum_{t \ge 1} \Pr[\text{search takes at least } t \text{ steps } | \text{ phase successful}]$

We have

Pr[search at least t steps | successful]

```
= \Pr[\text{search at least } t \text{ steps } \land \text{successful}] / \Pr[\text{successful}]
```

 $\leq \frac{1}{c} \Pr[\text{search at least } t \text{ steps } | \text{ no cycle}] ,$

where we use the fact that for a suitable ${\bf constant} \; c \geq 0$

```
Pr[successful] = Pr[no cycle] - Pr[unsuccessful | no cycle]
 \geq c \cdot Pr[no cycle]
```

Cuckoo Hashing

We choose maxsteps $\geq \frac{3\ell}{2} + 1$. Then the probability that a phase is terminated unsuccessfully without running into a cycle is at most

Pr[unsuccessful | no cycle]

 $\leq \Pr[\exists active path-structure of size at least \frac{2(massteps + 1) - 1}{3}]$

 $\leq \Pr[\exists active path-structure of size at least <math>\ell + 1]$

$$\leq 2\mu^2 \Big(\frac{1}{1+\epsilon}\Big)^\ell \leq \frac{1}{m^2}$$

by choosing $\ell \geq \log{(\frac{1}{2\mu^2m^2})}/\log{(\frac{1}{1+\epsilon})} = \log{(2\mu^2m^2)}/\log{(1+\epsilon)}$

Note that this gives maxsteps = $\Theta(\log m)$.

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Hence,

E[number of steps | phase successful] = \frac{1}{c} \sum_{t \ge 1} Pr[search at least t steps | no cycle]
\leq \frac{1}{c} \left[ 1 + \sum_{t \ge 2} 2\mu^2 \left( \frac{1}{1+\epsilon} \right)^{2(t-2)/3} \right]
= \frac{1}{c} + \frac{2\mu^2}{c} \sum_{t \ge 0} \left( \frac{1}{(1+\epsilon)^{2/3}} \right)^t = \mathcal{O}(1) .
```

A phase that is not successful induces $\cot O(m)$ for doing a complete rehash (this dominates the cost for the steps in the phase).

The probability that a phase is not successful is $p = O(1/m^2)$ (probability $O(1/m^2)$ of running into a cycle and probability $O(1/m^2)$ of reaching maxsteps without running into a cycle).

The expected number of unsuccessful phases is $\sum_{i\geq 1} p^i = \frac{1}{1-p} - 1 = \frac{p}{1-p} = \mathcal{O}(p).$

Therefore the expected cost for re-hashes is $\mathcal{O}(m) \cdot \mathcal{O}(p) = \mathcal{O}(1/m).$

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Cuckoo Hashing

How do we make sure that $n \ge (1 + \epsilon)m$?

- Let $\alpha := 1/(1 + \epsilon)$.
- Keep track of the number of elements in the table. When $m \ge \alpha n$ we double n and do a complete re-hash (table-expand).
- Whenever *m* drops below $\alpha n/4$ we divide *n* by 2 and do a rehash (table-shrink).
- Note that right after a change in table-size we have $m = \alpha n/2$. In order for a table-expand to occur at least $\alpha n/2$ insertions are required. Similar, for a table-shrink at least $\alpha n/4$ deletions must occur.
- Therefore we can amortize the rehash cost after a change in table-size against the cost for insertions and deletions.

Cuckoo Hashing

What kind of hash-functions do we need?

Since maxsteps is $\Theta(\log m)$ the largest size of a path-structure or cycle-structure contains just $\Theta(\log m)$ different keys.

Therefore, it is sufficient to have $(\mu, \Theta(\log m))$ -independent hash-functions.

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