Albert-Ludwigs-Universität Freiburg

#### Prof. Dr. Rolf Backofen

Bioinformatics Group / Department of Computer Science Algorithms and Datastructures, November 2017

## Structure



#### **Associative Arrays**

Introduction Hash Map

#### **Universal Hashing**

Introduction

**Probability Calculation** 

Proof

Examples

#### Reminder:

■ An associative array is like a normal array, only that the indices are not 0,1,2,..., but different, e.g. telephone numbers

#### **Problem:**

- Quickly find a element with a specific key
- Naive solution: Store pairs of key and value in a normal field
- For n keys searching requires  $\Theta(n)$  time
- With a hash map this just requires  $\Theta(1)$  in the best case, ... regardless how many elements are in the map!

#### Idea:

- Mapping the keys onto indices with a hash function
- Store the values at the calculated indices in a normal array

## Example:

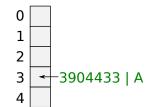
The Hash Map

- $\blacksquare$  Key set:  $x = \{3904433, 312692, 5148949\}$
- Hash function:  $h(x) = x \mod 5$ , in the range  $[0, \ldots, 4]$
- We need an array T with 5 elements. A "hashtable" with 5 "buckets"
- The element with the key x is stored in T[h(x)]

#### Storage:

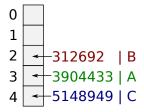
- insert(3904433,  $\ddot{A}$ "):  $h(3904433) = 3 \Rightarrow T[3] = (3904433, <math>\ddot{A}$ ")
- insert (312692, "B"):  $h(312692) = 2 \Rightarrow T[2] = (312692, "B")$
- insert (5148949, "C"):  $h(5148949) = 4 \Rightarrow T[4] = (5148949, "C")$





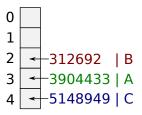
## Searching:

- search(3904433):  $h(3904433) = 3 \Rightarrow T[3] \rightarrow (3904433, Ä")$
- search(123459):  $h(123459) = 4 \Rightarrow T[4]$ 
  - ⇒ Value with key 123459 does not exist
- Search time for this example:  $\mathcal{O}(1)$



#### Further inserting:

- insert(876543, "D"): h(876543) = 3⇒  $T[3] = (876543, "D") \Rightarrow Collision$
- This happens more often than expected
  - **Birthday problem:** With 23 people we have the probability of 50 % that 2 of them have birthday at the same day

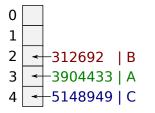


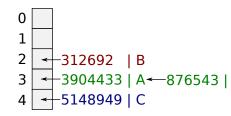
#### Problem:

Two keys are equal h(x) = h(y) but not the values  $x \neq y$ 

#### **Easiest Solution:**

- Represent each bucket as list of key value pairs
- Append new values to the end of the list

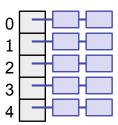




# Best case:

- We have n keys which are equally distributed over m buckets
- We have  $\approx \frac{n}{m}$  pairs per bucket
- The runtime for searching is nearly 𝒪(1) when not n ≫ m

# **Best case** (m = 5, n = 10)

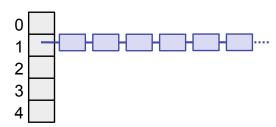


#### Worst case:

- All n keys are mapped onto the same bucket
- The runtime is  $\Theta(n)$  for searching

# Worst case

$$(m = 5, n = 10)$$



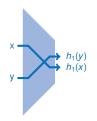
#### **Thought Experiment:**

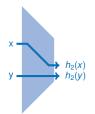
- A hash function is defined for a given key set
- Find a set of keys resulting in a degenerated hash table
  - The hash function stays fixed
  - For table size of 100: Try 100 × (99 + 1) different numbers
  - Worst case: All 100 key sets map to one bucket
- Now: Find a solution to avoid that problem

Idea

#### Solution: universal hashing

- Out of a set of hash functions we randomly choose one
- The expected result of the hash function is an equal distribution over the buckets
- This hash function stays fixed for the lifetime of table Optional: copy table with new hash when degenerated





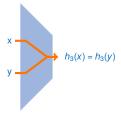


Abbildung: Hash func.

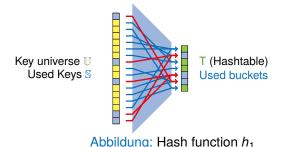
Abbildung: Hash func.

Abbildung: Hash func. coll.



#### **Definition:**

- lacktriangle We call  $\Bbb U$  the set (universum) of possible keys
- $\blacksquare$  The size m of the hash table T
- Set of hash functions  $\mathbb{H} = \{h_1, h_2, ..., h_n\}$  with  $h_i : \mathbb{U} \to \{0, ..., m-1\}$
- Idea: runtime should be  $O(1 + \frac{|S|}{m})$ , where  $\frac{|S|}{m}$  is the table load



 An average of 3 out of 15 functions produce collisions

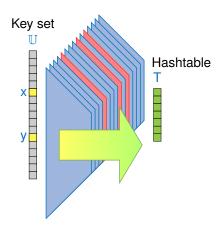


Abbildung: Set of hash functions ℍ

## **Definition:** $\mathbb{H}$ is *c*-universal if $\forall x, y \in \mathbb{U} \mid x \neq y$ :

Number of hash functions that create collisions

$$\underbrace{\left|\left\{h\in\mathbb{H}:h(x)=h(y)\right\}\right|}_{\left|\mathbb{H}\right|}$$

$$\leq c \cdot \frac{1}{m}, \quad c \in \mathbb{R}$$

Number of hash functions

■ With other words, given a arbitrary but fixed pair x, y. If  $h \in \mathbb{H}$  is chosen randomly then

$$Prob(h(x) = h(y)) \le c \cdot \frac{1}{m}$$

Note: If the hash function assigns each key *x* and *y* randomly to buckets then:

$$Prob(Collision) = \frac{1}{m} \Leftrightarrow c = 1$$

- U: Key universe
- S: Used Keys
- $S_i \subseteq S$ : Keys mapping to Bucket i ("synonyms")
- Ideal would be  $|S_i| = \frac{|S|}{m}$

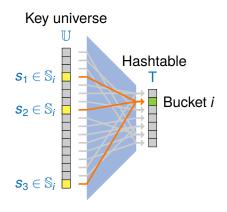


Abbildung: Hash function  $h \in \mathbb{H}$ 

# Universal Hashing Definition



- Let H be a c-universal class of hash functions
- Let  $\mathbb S$  be a set of keys and  $h \in \mathbb H$  selected randomly
- Let  $S_i$  be the key x for which h(x) = i
- The expected average number of elements to search through per bucket is

$$\mathbb{E}[|\mathbb{S}_i|] \leq 1 + c \cdot \frac{|\mathbb{S}|}{m}$$

■ Particulary: If  $(m = \Omega(|S|))$  then  $\mathbb{E}[|S_i|] = \mathcal{O}(n)$ 

- We just discuss the discrete case
- Probability space Ω with elementary (simple) events
- Events e have probabilities ...

$$\sum_{e\in\Omega}P(e)=1$$

■ The probability for a subset of events  $E \subseteq \Omega$  is

$$P(E) = \sum_{e \in E} P(e) \mid e \in E$$

#### Tabelle: Throwing a dice

e	P(e)
1	1/6
2	1/6
3	1/6
4	1/6
5	1/6
6	1/6

**Probability Calculation** 

## **Example:**

- Rolling a dice twice  $(\Omega = \{1, ..., 6\}^2)$
- Each event  $e \in \Omega$  has the probability P(e) = 1/36
- $\blacksquare$  *E* = if both results are even, then P(E) =

#### Tabelle: Throwing a dice twice

е	P(e)
(1,1)	1/36
(1,2)	1/36
(1,3)	1/36
(6,5)	1/36
(6,6)	1/36

# Example:

- Random variable
  - Assigns a number to the result of an experiment
  - For example: X = Sum of results for rolling twice
  - X = 12 and  $X \ge 7$  are regarded as events
  - Example 1: P(X = 2) =
  - Example 2: P(X = 4) =

#### Tabelle: Throwing a dice twice

е	P(e)	X	
(1,1) (1,2) (1,3)	1/36 1/36 1/36	2 3 4	
(6,5) (6,6)	1/ <sub>36</sub> 1/ <sub>36</sub>	 11 12	

# **Expected value** is defined as $\mathbb{E}(X) = \sum (k \cdot P(X = k))$

■ Intuitive: The weighted average of possible values of *X*, where the weights are the probabilities of the values

Tabelle: Throwing a dice once

Tabelle: Throwing a dice twice

X	P(X)
2	1/36
3	2/36
4	3/36
11	2/36
12	1/36

- **Example rolling once:**  $\mathbb{E}(X) = 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + \dots + 6 \cdot \frac{1}{6} = 3.5$
- Example rolling twice:  $\mathbb{E}(X) = 2 \cdot \frac{1}{36} + 3 \cdot \frac{2}{36} + \cdots + 12 \cdot \frac{1}{36} = 7$

**Sum of expected values:** For arbitrary discrete random variables  $X_1, ..., X_n$  we can write:

$$\mathbb{E}(X_1 + \cdots + X_n) = \mathbb{E}(X_1) + \cdots + \mathbb{E}(X_n)$$

Example: Throwing two dice

- $X_1$ : Expected result of dice 1:  $\mathbb{E}(X_1) = 3.5$
- $X_2$ : Expected result of dice 2:  $\mathbb{E}(X_2) = 3.5$
- $X = X_1 + X_2$ : Expected total number:

$$\mathbb{E}(X) = \mathbb{E}(X_1 + X_2) = \mathbb{E}(X_1) + \mathbb{E}(X_2) = 3.5 + 3.5 = 7$$

**Probability Calculation** 

# Corollary:

The probability of the event E is p = P(E). Let X be the occurrences of the event E and n be the number of executions of the experiment. Then  $\mathbb{E}(X) = n \cdot P(E) = n \cdot p$ 

Example (Rolling the dice 60 times:)

$$\mathbb{E}$$
(occurences of 6) =  $\frac{1}{6} \cdot 60 = 10$ 

# **Proof Corollary:**

Indicator variable:  $X_i$ 

$$X_i = \begin{cases} 1, & \text{if event occurs} \\ 0, & \text{else} \end{cases}$$

$$\Rightarrow X = \sum_{i=1}^{n} X_i$$

$$\mathbb{E}(X) = \mathbb{E}\left(\sum_{i=1}^{n} X_i\right) = \sum_{i=1}^{n} \mathbb{E}(X_i) \stackrel{\text{def. } \mathbb{E}\text{-value}}{=} \sum_{i=1}^{n} p = n \cdot p$$

Def.  $\mathbb{E}$ -value:  $\mathbb{E}(X_i) = 0 \cdot P(X_i = 0) + 1 \cdot P(X_i = 1) = P(X_i = 1)$ 



- We pick two random keys  $x, y \in \mathbb{S} \mid x \neq y$  and a random hash function  $h \in \mathbb{H}$
- We know the probability of a collision:

$$P(h(x) = h(y)) \le c \cdot \frac{1}{m}$$

To proof:

$$\mathbb{E}[|\mathbb{S}_i|] \leq 1 + c \cdot \frac{|\mathbb{S}|}{m} \quad \forall i$$

#### We know:

$$\mathbb{S}_i = \{x \in \mathbb{S} : h(x) = i\}$$

If  $\mathbb{S}_i = \emptyset \Rightarrow |\mathbb{S}_i| = 0$  otherwise, let  $x \in \mathbb{S}_i$  be any key

We define an indicator variable:

$$I_y = \begin{cases} 1, & \text{if } h(y) = i \\ 0, & \text{else} \end{cases} \quad y \in \mathbb{S} \setminus \{x\}$$

$$\Rightarrow \qquad \left| \mathbb{S}_i \right| = 1 + \sum_{y \in \mathbb{S} \setminus x} I_y$$

$$\Rightarrow \quad \mathbb{E}(|\mathbb{S}_i|) = \mathbb{E}\left(1 + \sum_{y \in \mathbb{S} \setminus X} l_y\right) = 1 + \sum_{y \in \mathbb{S} \setminus X} \mathbb{E}(l_y)$$

## **Proof**

Auxiliary calculation:  $\mathbb{E}[I_V] = P(I_V = 1)$ 

$$\mathbb{E}[I_y] = P(I_y = 1)$$

$$= P(h(y) = i)$$

$$= P(h(y) = h(x))$$

$$\leq c \cdot \frac{1}{m}$$

Hence: 
$$\mathbb{E}[|\mathbb{S}_i|] = 1 + \sum_{y \in \mathbb{S} \setminus x} \mathbb{E}[l_y] \le 1 + \sum_{y \in \mathbb{S} \setminus x} c \cdot \frac{1}{m}$$

$$= 1 + (|\mathbb{S}| - 1) \cdot c \cdot \frac{1}{m}$$

$$\le 1 + |\mathbb{S}| \cdot c \cdot \frac{1}{m}$$

$$= 1 + c \cdot \frac{|\mathbb{S}|}{m}$$

#### **Negative example:**

Examples

- The set of all h for which  $h_a(x) = (a \cdot x) \mod m$ , for a  $a \in \mathbb{U}$
- Is not c-universal.
- If universal:

$$\forall x,y \quad x \neq y \colon \frac{|\{h \in \mathbb{H} : h(x) = h(y)\}|}{|\mathbb{H}|} \leq c \cdot \frac{1}{m}$$

■ Which x,y lead to a relative collision count bigger than  $\frac{c}{m}$ ?

# Examples

#### Positive example:

- Let p be a big prime number, p > m and  $p \ge |\mathbb{U}|$
- Let  $\mathbb{H}$  be the set of all h for which:

$$h_{a,b}(x) = ((a \cdot x + b) \mod p) \mod m,$$
  
where  $1 \le a < p, \ 0 \le b < p$ 

- This is ≈ 1-universal, see Exercise 4.11 in Mehlhorn/Sanders
- E.g.:  $U = \{0, ..., 99\}, p = 101, a = 47, b = 5$
- Then  $h(x) = ((47 \cdot x + 5) \mod 101) \mod m$
- Easy to implement but hard to proof
- Exercise: show empirically that it is 2-universal

# Positive example:

Examples

■ The set of hash functions is *c*-universal:

$$h_a(x) = a \bullet x \mod m, \quad a \in \mathbb{U}$$

■ We define:

$$a = \sum_{0,\dots,k-1} a_i \cdot m^i, \qquad k = \text{ceil}(\log_m |\mathbb{U}|)$$
$$x = \sum_{0,\dots,k-1} x_i \cdot m^i$$

Intuitive: Scalar product with base m

$$a \bullet x = \sum_{0,\dots,k-1} a_i \cdot x_i$$

# Universal Hashing

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Example (
$$\mathbb{U} = \{0, ..., 999\}, m = 10, a = 348$$
)

With 
$$a = 348$$
:  $a_2 = 3$ ,  $a_1 = 4$ ,  $a_0 = 8$ 

$$h_{348}(x) = (a_2 \cdot x_2 + a_1 \cdot x_1 + a_0 \cdot x_0) \mod m$$
  
=  $(3x_2 + 4x_1 + 8x_0) \mod 10$ 

With 
$$x = 127$$
:  $x_2 = 1$ ,  $x_1 = 2$ ,  $x_0 = 7$ 

$$h_{348}(127) = (3 \cdot x_2 + 4 \cdot x_1 + 8 \cdot x_0) \mod 10$$
  
=  $(3 \cdot 1 + 4 \cdot 2 + 8 \cdot 7) \mod 10$   
= 7

#### ■ General for this Lecture

[CRL01] Thomas H. Cormen, Ronald L. Rivest, and Charles E. Leiserson. Introduction to Algorithms. MIT Press, Cambridge, Mass, 2001.

[MS08] Kurt Mehlhorn and Peter Sanders. Algorithms and data structures, 2008. https://people.mpi-inf.mpg.de/~mehlhorn/ftp/Mehlhorn-Sanders-Toolbox.pdf.

## ■ Hash Map - Theory

- [Wik] Hash table
  - https://en.wikipedia.org/wiki/Hash\_table
- Hash Map Implementations / API
  - [Cpp] C++ hash\_map
    http://www.sgi.com/tech/stl/hash\_map.html
  - [Jav] Java HashMap
     https://docs.oracle.com/javase/7/docs/api/
     java/util/HashMap.html
  - [Pyt] Python Dictionaries (Hash table)
     https://en.wikipedia.org/wiki/Hash\_table