

Advanced Data Structures

Lecture 11: Minimal Perfect Hashing

Florian Kurpicz

The slides are licensed under a Creative Commons Attribution-ShareAlike 4.0 International License @) www.creativecommons.org/licenses/by-sa/4.0 | commit c70729e compiled at 2023-07-10-13:34



www.kit.edu



PINGO



https://pingo.scc.kit.edu/489786



Operations

- INSERT(t, operation): insert operation at time t
- DELETE(t): delete operation at time t
- QUERY(t, query): ask query at time t

for a priority queue updates are

- insert
- delete-min
- time is integer () for simplicity otherwise use order-maintenance data structure

insert(7) insert(2) insert(3) del-min del-min						queries			
0	1		2 3	3 4	4 no	Św	time		



Operations

- INSERT(t, operation): insert operation at time t
- DELETE(t): delete operation at time t
- QUERY(t, query): ask query at time t

for a priority queue updates are

- insert
- delete-min
- time is integer () for simplicity otherwise use order-maintenance data structure

inse	rt(7) inse	min	queries					
Ċ) -	1 2	2 3	3 4	4 n	ów	time	-

Definition: Partial Retroactivity

QUERY is only allowed for $t = \infty$ () now



Operations

- INSERT(t, operation): insert operation at time t
- DELETE(t): delete operation at time t
- QUERY(t, query): ask query at time t

for a priority queue updates are

- insert
- delete-min
- time is integer () for simplicity otherwise use order-maintenance data structure

inse	insert(7) insert(2) insert(3) del-min del-min						queries		
	0 .	1 2	2 3	3 4	4	now	time		

Definition: Partial Retroactivity

QUERY is only allowed for $t = \infty$ () now

Definition: Full Retroactivity

QUERY is allowed at any time t

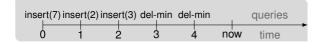


Operations

- INSERT(t, operation): insert operation at time t
- DELETE(t): delete operation at time t
- QUERY(t, query): ask query at time t

for a priority queue updates are

- insert
- delete-min
- time is integer () for simplicity otherwise use order-maintenance data structure



Definition: Partial Retroactivity

QUERY is only allowed for $t = \infty$ () now

Definition: Full Retroactivity

QUERY is allowed at any time t

Definition: Nonoblivious Retroactivity

INSERT, DELETE, and QUERY at any time *t* but also identify changed QUERY results

Karlsruhe Institute of Technology

Hashing (1/2)

- $h: \{0, \ldots, u-1\} \to \{0, \ldots, m-1\}$
- n objects
- from universe $U = \{0, \ldots, u-1\}$
- hash table of size m 1 m close to n
- *m* ≪ u

Karlsruhe Institute of Technology

Hashing (1/2)

- $h: \{0, ..., u-1\} \to \{0, ..., m-1\}$
- n objects
- from universe $U = \{0, \ldots, u-1\}$
- hash table of size m 1 m close to n
- *m* ≪ *u*

Definition: Totally Random

- $\mathbb{P}[h(x) = t] = 1/m$
- independent of h(y) for all $x \neq y \in U$
- requires ⊖(u log m) bits of space to store
 too big



- $h: \{0, \ldots, u-1\} \to \{0, \ldots, m-1\}$
- n objects
- from universe $U = \{0, \ldots, u-1\}$
- hash table of size m I m close to n
- m ≪ u

Definition: Totally Random

- $\mathbb{P}[h(x) = t] = 1/m$
- independent of h(y) for all $x \neq y \in U$
- requires ⊖(u log m) bits of space to store
 too big

Definition: Universal

- choose *h* from family *H* with $\mathbb{P}_{h \in H}[h(x) = h(y)] = O(1/m)$ for all $x \neq y \in U$
- family is small to enable efficient encoding



•
$$h: \{0, \ldots, u-1\} \to \{0, \ldots, m-1\}$$

- n objects
- from universe $U = \{0, \ldots, u-1\}$
- hash table of size m I m close to n
- *m* ≪ *u*

Definition: Totally Random

- $\mathbb{P}[h(x) = t] = 1/m$
- independent of h(y) for all $x \neq y \in U$
- requires ⊖(u log m) bits of space to store
 too big

Definition: Universal

- choose *h* from family *H* with $\mathbb{P}_{h \in H}[h(x) = h(y)] = O(1/m)$ for all $x \neq y \in U$
- family is small to enable efficient encoding
- *h*(*x*) = (*ax* mod *u*) mod *m* for 0 < *a* < *p* and *p* being prime > *u*
- h(x) = ax » (log u log m) for m, u being powers of two
- Why is this family easier to store? I PINGO



Definition: *k*-wise Independent

• choose *h* from family *H* with $\mathbb{P}[h(x_1) = t_1 \& \dots \& h(x_k) = t_k] = O(1/m^k)$ for distinct $x_1, \dots, x_k \in U$



Definition: *k*-wise Independent

• choose *h* from family *H* with $\mathbb{P}[h(x_1) = t_1 \& \dots \& h(x_k) = t_k] = O(1/m^k)$ for distinct $x_1, \dots, x_k \in U$

implies universal



Definition: *k*-wise Independent

• choose *h* from family *H* with $\mathbb{P}[h(x_1) = t_1 \& \dots \& h(x_k) = t_k] = O(1/m^k)$ for distinct $x_1, \dots, x_k \in U$

implies universal

- $h(x) = ((\sum_{i=0}^{k-1} a_i x^i) \mod p) \mod m$ for $0 \le a_i < p$ and $0 < a_{k-1} < p$
- pairwise (k = 2) independence is stronger than universal
- $h(x) = ((ax + b) \mod u) \mod m$



Definition: k-wise Independent

• choose *h* from family *H* with $\mathbb{P}[h(x_1) = t_1 \& \dots \& h(x_k) = t_k] = O(1/m^k)$ for distinct $x_1, \dots, x_k \in U$

implies universal

- $h(x) = ((\sum_{i=0}^{k-1} a_i x^i) \mod p) \mod m$ for $0 \le a_i < p$ and $0 < a_{k-1} < p$
- pairwise (k = 2) independence is stronger than universal
- $h(x) = ((ax + b) \mod u) \mod m$

Definition: Simple Tabulation Hashing

- view x as vector x_1, \ldots, x_c of characters
- totally random hash table T_i for each character

•
$$h(x) = T_1(x_1)$$
 xor ... xor $T_c(x_c)$



Definition: k-wise Independent

• choose *h* from family *H* with $\mathbb{P}[h(x_1) = t_1 \& \dots \& h(x_k) = t_k] = O(1/m^k)$ for distinct $x_1, \dots, x_k \in U$

implies universal

- $h(x) = ((\sum_{i=0}^{k-1} a_i x^i) \mod p) \mod m$ for $0 \le a_i < p$ and $0 < a_{k-1} < p$
- pairwise (k = 2) independence is stronger than universal
- $h(x) = ((ax + b) \mod u) \mod m$

Definition: Simple Tabulation Hashing

- view x as vector x_1, \ldots, x_c of characters
- totally random hash table T_i for each character

•
$$h(x) = T_1(x_1)$$
 xor ... xor $T_c(x_c)$

Why can we use totally random hash tables?
 PINGO



Definition: k-wise Independent

• choose *h* from family *H* with $\mathbb{P}[h(x_1) = t_1 \& \dots \& h(x_k) = t_k] = O(1/m^k)$ for distinct $x_1, \dots, x_k \in U$

implies universal

- $h(x) = ((\sum_{i=0}^{k-1} a_i x^i) \mod p) \mod m$ for $0 \le a_i < p$ and $0 < a_{k-1} < p$
- pairwise (k = 2) independence is stronger than universal
- $h(x) = ((ax + b) \mod u) \mod m$

Definition: Simple Tabulation Hashing

- view x as vector x_1, \ldots, x_c of characters
- totally random hash table T_i for each character
- $h(x) = T_1(x_1) \text{ xor } \dots \text{ xor } T_c(x_c)$
- Why can we use totally random hash tables?
 PINGO
- $O(cu^{1/c})$ space
- O(c) time to compute
- 3-wise independent

Minimal Perfect Hashing



Definition: Perfect Hash Function

- injective hash function
- maps n objects to m slots
- lower space bound for $m = (1 + \epsilon)n$ is

$$\log e - \epsilon \log \frac{1+\epsilon}{\epsilon}$$

for m close to n there are likely collisions

Minimal Perfect Hashing



Definition: Perfect Hash Function

- injective hash function
- maps n objects to m slots
- lower space bound for $m = (1 + \epsilon)n$ is

$$\log \boldsymbol{e} - \epsilon \log \frac{1+\epsilon}{\epsilon}$$

• for *m* close to *n* there are likely collisions

Definition: Minimal Perfect Hash Function

- bijective hash function
- maps n objects to m = n slots
- $h: N \rightarrow [0, n)$
- lower space bound as for PHF with $\epsilon = 0$:

 $\log e \approx 1.44$

no collisions

Minimal Perfect Hashing



Definition: Perfect Hash Function

- injective hash function
- maps n objects to m slots
- lower space bound for $m = (1 + \epsilon)n$ is

$$\log \boldsymbol{e} - \epsilon \log \frac{1+\epsilon}{\epsilon}$$

• for *m* close to *n* there are likely collisions

Definition: Minimal Perfect Hash Function

- bijective hash function
- maps n objects to m = n slots
- $h: N \rightarrow [0, n)$
- lower space bound as for PHF with $\epsilon = 0$:

 $\log e \approx 1.44$

no collisions

can we make PHF to MPHF? Standard PINGO



BDZ (RAM) Algorithm [BPZ13]

- for each object calculate three *potential* slots (h₀, h₁, and h₂)
- for each slot that contains only one object, remove the object from all its other slots
- one slot per object
- if that does not work use other hash functions
- use rank data structure to map slots to [0, n)
- example on the board
- 1.95 bits per object when m = 1.23n

BDZ (RAM) Algorithm [BPZ13]



- for each object calculate three *potential* slots (h₀, h₁, and h₂)
- for each slot that contains only one object, remove the object from all its other slots
- one slot per object
- if that does not work use other hash functions
- use rank data structure to map slots to [0, n)
- example on the board
- 1.95 bits per object when m = 1.23n

- how to check if hash function works
- interpret each slot as node in a hypergraph
- objects are edges
- if graph is peelable, we have a feasible mapping

Definition: Peelable

A hypergraph is peelable, if it is possible to obtain a graph without edges by iteratively taking away edges that contain a node with degree 1

example on the board ____



Compress, Hash, and Displace [BBD09a]

- partition keys into buckets
- set $m = (1 + \epsilon)n$ (1.01n
- sort partitions by size
- starting with largest bucket, find universal hash function mapping all keys to empty slots
- if key mapped to non-empty slot, try next hash function
- for each bucket store universal hash function
- use rank data structure to map slots to [0, n)

example on the board

Compress, Hash, and Displace [BBD09a]



- partition keys into buckets
- set $m = (1 + \epsilon)n$ 1.01n
- sort partitions by size
- starting with largest bucket, find universal hash function mapping all keys to empty slots
- if key mapped to non-empty slot, try next hash function
- for each bucket store universal hash function
- use rank data structure to map slots to [0, n)

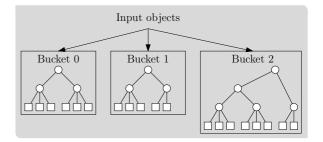
example on the board

- can be used as PHF
- there are a lot of tricks w.r.t. bucket sizes and size distributions
- requires around 2.05 bits per object



RecSplit Overview [EGV20a]

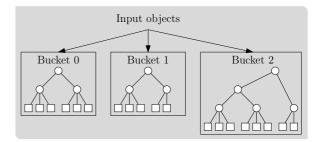
- partition keys into buckets of size b
- for each bucket compute splitting trees
- split keys into smaller sets
- stop when sets have size ℓ





RecSplit Overview [EGV20a]

- partition keys into buckets of size b
- for each bucket compute splitting trees
- split keys into smaller sets
- stop when sets have size ℓ
- upper aggregation levels have fanout 2
- Iower two aggregation levels have fanout
 - $\max\{2, \lceil 0.35\ell + 0.55 \rceil\}$
 - max{2, [0.21ℓ + 0.9]}



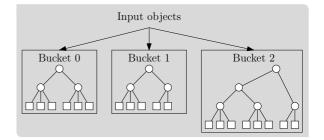


RecSplit Overview [EGV20a]

- partition keys into buckets of size b
- for each bucket compute splitting trees
- split keys into smaller sets
- stop when sets have size ℓ
- upper aggregation levels have fanout 2
- Iower two aggregation levels have fanout
 - $\max\{2, \lceil 0.35\ell + 0.55 \rceil\}$
 - max{2, [0.21ℓ + 0.9]}

last level is leaf level

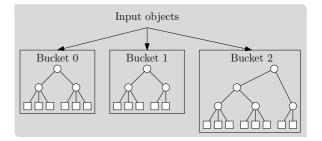
find bijections





RecSplit Splitting Tree

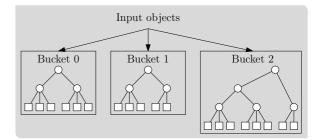
- tree structure is well defined
- store information for each node in preorder
- store hash function for each splitter
- encode function using Golomb-Rice





RecSplit Splitting Tree

- tree structure is well defined
- store information for each node in preorder
- store hash function for each splitter
- encode function using Golomb-Rice
- encodings of splitting trees stored in one bit vector
- use Elias-Fano to store
 - size of buckets
 - starting position of bucket in bit vector



Karlsruhe Institute of Technology

Golomb Encoding [Gol66]

Definition: Golomb Code

Given an integer x > 0 and a constant b > 0, the Golomb code consists of

•
$$q = \lfloor \frac{x}{b} \rfloor$$

•
$$r = x - qb = x\% b$$

with

$$(x)_{Gol(b)} = (q)_1(r)_2$$

- r < 2^{⌊lg b} −1: r requires ⌊lg b⌋ bits and starts with a 0
- r ≥ 2^{⌊lg b} −1: r requires [lg b] bits and starts with a 1 and it encodes r − 2^{⌊lg b} −1

Golomb Encoding [Gol66]



Definition: Golomb Code

Given an integer x > 0 and a constant b > 0, the Golomb code consists of

•
$$q = \lfloor \frac{x}{b} \rfloor$$

•
$$r = x - qb = x\% b$$

with

$$(x)_{Gol(b)} = (q)_1(r)_2$$

- r < 2^{⌊lg b} −1: r requires ⌊lg b⌋ bits and starts with a 0
- r ≥ 2^{⌊lg b}]⁻¹: r requires [lg b] bits and starts with a 1 and it encodes r − 2^{⌊lg b}]⁻¹

- b has to be fixed for all codes
- still variable length

Golomb Encoding [Gol66]



Definition: Golomb Code

Given an integer x > 0 and a constant b > 0, the Golomb code consists of

•
$$q = \lfloor \frac{x}{b} \rfloor$$

•
$$r = x - qb = x \% b$$

•
$$c = \lceil \lg b \rceil$$

with

$$(x)_{Gol(b)} = (q)_1(r)_2$$

- r < 2^{⌊lg b} −1: r requires ⌊lg b⌋ bits and starts with a 0
- r ≥ 2^{⌊lg b}]⁻¹: r requires [lg b] bits and starts with a 1 and it encodes r − 2^{⌊lg b}]⁻¹

- b has to be fixed for all codes
- still variable length
- Golomb-Rice is special case where r is power of two

Golomb Encoding [Gol66]



Definition: Golomb Code

Given an integer x > 0 and a constant b > 0, the Golomb code consists of

•
$$q = \lfloor \frac{x}{b} \rfloor$$

•
$$r = x - qb = x \% b$$

with

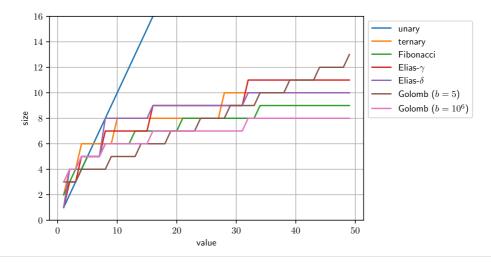
$$(x)_{\mathrm{Gol}(b)}=(q)_1(r)_2$$

- r < 2^{⌊lg b} −1: r requires ⌊lg b⌋ bits and starts with a 0
- r ≥ 2^{⌊lg b}]⁻¹: r requires [lg b] bits and starts with a 1 and it encodes r − 2^{⌊lg b}]⁻¹

- b has to be fixed for all codes
- still variable length
- Golomb-Rice is special case where r is power of two
- for b = 5, there are 4 remainders: 00,01,100,101, and 110
- $2^{\lfloor \lg 5 \rfloor 1} = 2$
- 0, 1 < 2: 00 and 01 require 2 bits</p>
- 2, 3, 4 \geq 2: require 3 bits and encode 0, 1, 2 starting with 1

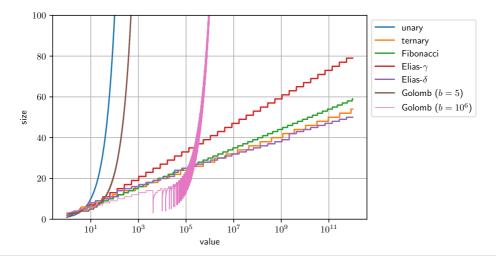


Comparison of Codes (1/2)





Comparison of Codes (2/2)

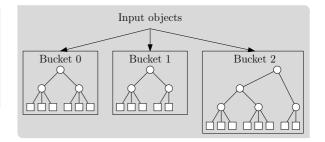


Institute of Theoretical Informatics, Algorithm Engineering

RecSplit Leaves



- find perfect hash function for keys in leaves
- test hash functions brute force
- \blacksquare use hash value modulo ℓ
- \blacksquare set bit in "bit vector" of length ℓ
- all bits set indicates bijection

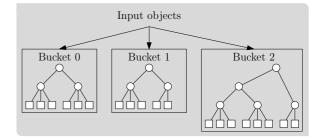


15/25 2023-07-10 Florian Kurpicz | Advanced Data Structures | 11 Minimal Perfect Hashing

Institute of Theoretical Informatics, Algorithm Engineering

RecSplit Queries

- find bucket
- follow splitting tree
- accumulate number of objects to the left
- use bijection in leaf
- result is sum of
 - objects in previous buckets
 - objects to the left in splitting tree
 - value of bijection





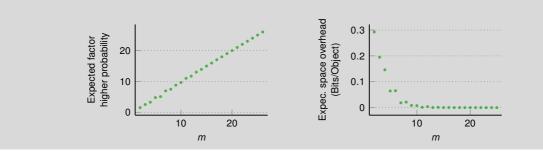
Parallel RecSplit



- Dominik Bez, Florian Kurpicz, Hans-Peter Lehmann, and Peter Sanders.
 "High Performance Construction of RecSplit Based Minimal Perfect Hash Functions". In: CoRR abs/2212.09562 (2022)
- results to be presented at ESA'23
- based on a Domink Bez' Master's thesis

- randomly distribute objects in leaf in two sets A and B
- hash objects in both set
- two "bit vectors": cyclic shift one until all bits are set when 0Red
- store hash function and rotation

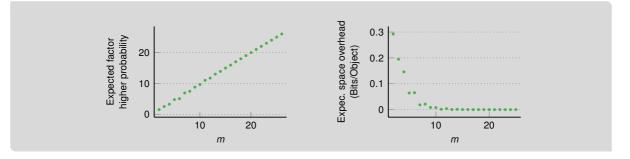
- randomly distribute objects in leaf in two sets A and B
- hash objects in both set
- two "bit vectors": cyclic shift one until all bits are set when 0Red
- store hash function and rotation



- randomly distribute objects in leaf in two sets A and B
- hash objects in both set
- two "bit vectors": cyclic shift one until all bits are set when 0Red
- store hash function and rotation

Lemma: Rotation Fitting

Let $|A| = \mathbb{A}$, $|B| = \mathbb{B}$, and $\mathbb{P}(R)$ be the probability of finding a bijection using rotation fitting. Let $\mathbb{P}(B)$ denote the probability of finding a bijection using RecSplit's brute force strategy. Then, $\mathbb{P}(R) \to m\mathbb{P}(B)$ for $m \to \infty$.



Rotation Fitting (2/3)



Proof (Sketch)

- consider number of different injective functions under cyclic shifts
- bit vector of length m with \mathbb{B} set bits
- total number of equivalence classes under rotation is $\frac{1}{m} \sum_{d \text{ divides } \gcd(\mathbb{A},\mathbb{B})} \phi(d) \binom{m/d}{\mathbb{B}/d}$
- probability of the event *I* that there is a rotation has the *m* least significant bits set is

$$\mathbb{P}(\mathcal{I}) \geq m rac{1}{\sum_{\textit{d} ext{ divides } \gcd(\mathbb{A},\mathbb{B})} \phi(\textit{d}) {m/d \choose \mathbb{B}/d}},$$

Proof (Sketch, cnt.)

- determine the probability $\mathbb{P}(R)$ using the events
 - \mathcal{A} : popcount(a)= \mathbb{A}
 - B: popcount(b)=B
 - B: found bijection using brute-force



Rotation Fitting (3/3)

Proof (Sketch, ctn.)

$$\begin{split} \mathbb{P}(R) &= \mathbb{P}(\mathcal{A})\mathbb{P}(\mathcal{B})\mathbb{P}(\mathcal{I}) \\ &\geq \frac{m!}{(m-\mathbb{A})!m^{\mathbb{A}}} \cdot \frac{m!}{(m-\mathbb{B})!m^{\mathbb{B}}} \cdot \mathbb{P}(\mathcal{I}) = \frac{m!}{m^{m}} \cdot \frac{m!}{\mathbb{A}!\mathbb{B}!} \cdot \mathbb{P}(\mathcal{I}) = \mathbb{P}(B) \cdot \frac{m!}{\mathbb{A}!\mathbb{B}!} \cdot \mathbb{P}(\mathcal{I}) \\ &\geq \mathbb{P}(B) \cdot \frac{m!}{\mathbb{A}!\mathbb{B}!} \cdot m \frac{1}{\sum_{d|\text{gcd}(\mathbb{A},\mathbb{B})} \phi(d) \binom{m/d}{b/d}} = \mathbb{P}(B) \cdot m \cdot \frac{m!}{m! + (\mathbb{A}!\mathbb{B}!) \sum_{d|\text{gcd}(\mathbb{A},\mathbb{B}), d\neq 1} \phi(d) \binom{m/d}{b/d}} \\ &= \mathbb{P}(B) \cdot m \cdot \frac{1}{1 + \sum_{d|\text{gcd}(\mathbb{A},\mathbb{B}), d\neq 1} \phi(d) \frac{(m/d)!\mathbb{A}!\mathbb{B}!}{m!(\mathbb{A}/d)!(\mathbb{B}/d)!}} \\ &\sim \mathbb{P}(B) \cdot m \text{ for } m \to \infty \end{split}$$

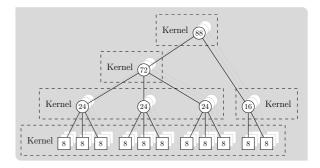


Parallel RecSplit on the GPU

Computing on the GPU

- several streaming multiprocessors (SMs)
- each SM contains many arithmetic logic units (ALUs)
- several threads operat in lock-step (warp)
- to hide latencies, each SM is oversubscribed with more threads than ALUs
- in CUDA, kernels are functions that can be executed on the GPU
- a kernel is executed on a grid of thread blocks

use GPU to determine splitting and bijections



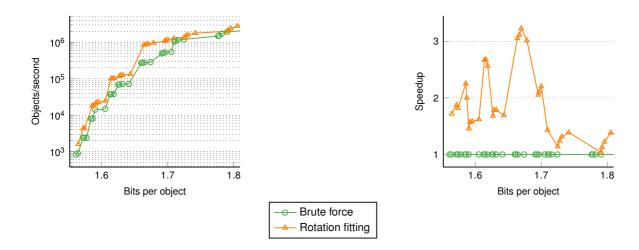
Experimental Evaluation



- Intel i7 11700 processor with 8 cores (16 hardware threads (HT)), base clock: 2.5 GHz
- AVX-512.
- Ubuntu 22.04 with Linux 5.15.0
- NVIDIA RTX 3090 GPU
- AMD EPYC 7702P processor with 64 cores (128 hardware threads), base clock: 2.0 GHz
- AVX2
- Ubuntu 20.04 with Linux 5.4.0
- GNU C++ compiler v.11.2.0 (-03
 - -march=native)

Rotation Fitting





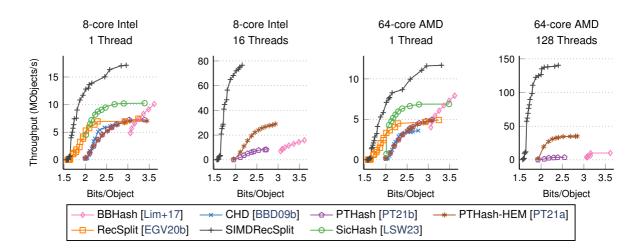
Overview Results



Configuration	Method	Bijections	Threads	B/Obj	Constr.	Speedup
$\ell = 16, b = 2000$	RecSplit [EGV20b]	Brute force	1	1.560	1175.4	1
	RecSplit	Brute force	16	1.560	206.5	5
	SIMDRecSplit	Rotation fitting	1	1.560	138.0	8
	SIMDRecSplit	Rotation fitting	16	1.560	27.9	42
	GPURecSplit	Brute force	GPU	1.560	1.8	655
	GPURecSplit	Rotation fitting	GPU	1.560	1.0	1173
$\ell = 18, b = 50$	RecSplit [EGV20b]	Brute force	1	1.707	2942.9	1
	RecSplit	Brute force	16	1.713	504.0	5
	SIMDRecSplit	Rotation fitting	1	1.709	58.3	50
	SIMDRecSplit	Rotation fitting	16	1.708	12.3	239
	GPURecSplit	Brute force	GPU	1.708	5.2	564
	GPURecSplit	Rotation fitting	GPU	1.709	0.5	5438
$\ell = 24, b = 2000$	GPURecSplit	Brute force	GPU	1.496	2300.9	_
	GPURecSplit	Rotation fitting	GPU	1.496	467.9	_



Comparison with Competitors

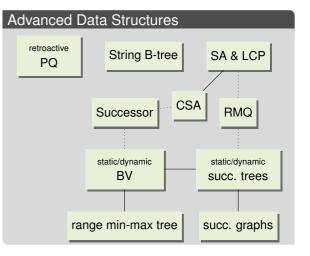


Conclusion and Outlook



This Lecture

minimal perfect hash functions



Conclusion and Outlook

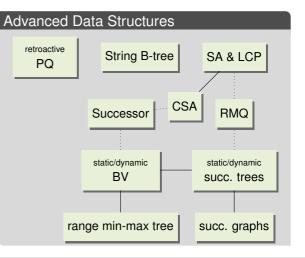


This Lecture

minimal perfect hash functions

Next Lecture

dynamic data structures



Bibliography I



- [BBD09a] Djamal Belazzougui, Fabiano C. Botelho, and Martin Dietzfelbinger. "Hash, Displace, and Compress". In: *ESA*. Volume 5757. Lecture Notes in Computer Science. Springer, 2009, pages 682–693. DOI: 10.1007/978-3-642-04128-0_61.
- [BBD09b] Djamal Belazzougui, Fabiano C. Botelho, and Martin Dietzfelbinger. "Hash, Displace, and Compress". In: *ESA*. Volume 5757. Lecture Notes in Computer Science. Springer, 2009, pages 682–693. DOI: 10.1007/978-3-642-04128-0_61.
- [Bez+22] Dominik Bez, Florian Kurpicz, Hans-Peter Lehmann, and Peter Sanders. "High Performance Construction of RecSplit Based Minimal Perfect Hash Functions". In: CoRR abs/2212.09562 (2022).
- [BPZ13] Fabiano C. Botelho, Rasmus Pagh, and Nivio Ziviani. "Practical perfect hashing in nearly optimal space". In: *Inf. Syst.* 38.1 (2013), pages 108–131. DOI: 10.1016/j.is.2012.06.002.

Bibliography II



- [EGV20a] Emmanuel Esposito, Thomas Mueller Graf, and Sebastiano Vigna. "RecSplit: Minimal Perfect Hashing via Recursive Splitting". In: ALENEX. SIAM, 2020, pages 175–185. DOI: 10.1137/1.9781611976007.14.
- [EGV20b] Emmanuel Esposito, Thomas Mueller Graf, and Sebastiano Vigna. "RecSplit: Minimal Perfect Hashing via Recursive Splitting". In: ALENEX. SIAM, 2020, pages 175–185. DOI: 10.1137/1.9781611976007.14.
- [Gol66] Solomon W. Golomb. "Run-length Encodings (Corresp.)". In: *IEEE Trans. Inf. Theory* 12.3 (1966), pages 399–401. DOI: 10.1109/TIT.1966.1053907.
- [Lim+17] Antoine Limasset, Guillaume Rizk, Rayan Chikhi, and Pierre Peterlongo. "Fast and Scalable Minimal Perfect Hashing for Massive Key Sets". In: *SEA*. Volume 75. LIPIcs. Schloss Dagstuhl -Leibniz-Zentrum für Informatik, 2017, 25:1–25:16. DOI: 10.4230/LIPICS.SEA.2017.25.

Bibliography III



- [LSW23] Hans-Peter Lehmann, Peter Sanders, and Stefan Walzer. "SicHash Small Irregular Cuckoo Tables for Perfect Hashing". In: ALENEX. SIAM, 2023, pages 176–189. DOI: 10.1137/1.9781611977561.CH15.
- [PT21a] Giulio Ermanno Pibiri and Roberto Trani. "Parallel and External-Memory Construction of Minimal Perfect Hash Functions with PTHash". In: *CoRR* abs/2106.02350 (2021).
- [PT21b] Giulio Ermanno Pibiri and Roberto Trani. "PTHash: Revisiting FCH Minimal Perfect Hashing". In: *SIGIR*. ACM, 2021, pages 1339–1348. DOI: 10.1145/3404835.3462849.