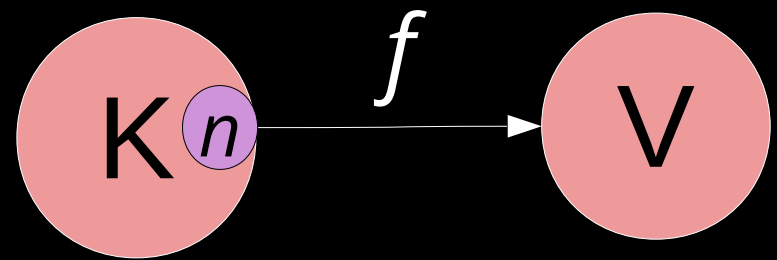
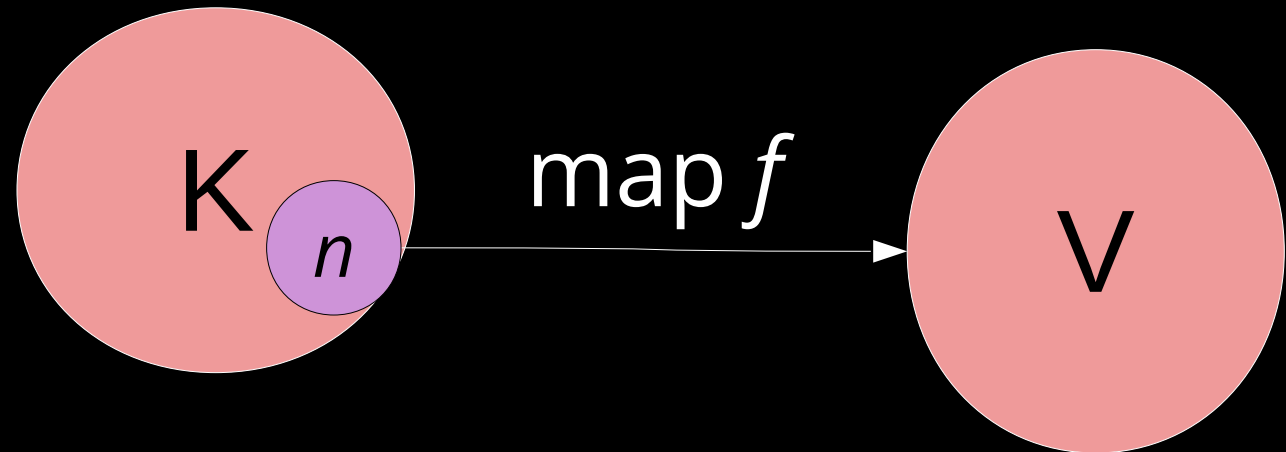


Fast and Simple Compact Hashing via Bucketing

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dynamic associative map



- K, V : sets
- f maps a *dynamic* subset of size n of K to V
- common representations of f
 - search tree
 - hash table

setting

- $K = [1..|2^\omega|]$

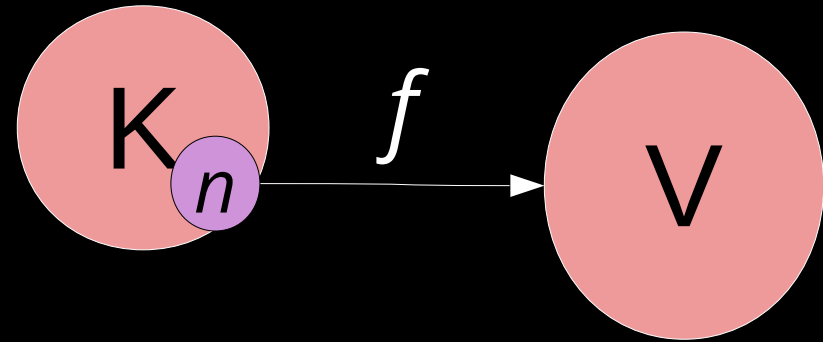
- $V = [1..|V|]$

- in case that $\omega \leq 20$

- use plain array to represent f

- space: $\lg |V| / 8$ MiB

- for larger ω not feasible



MiB = 1024^2

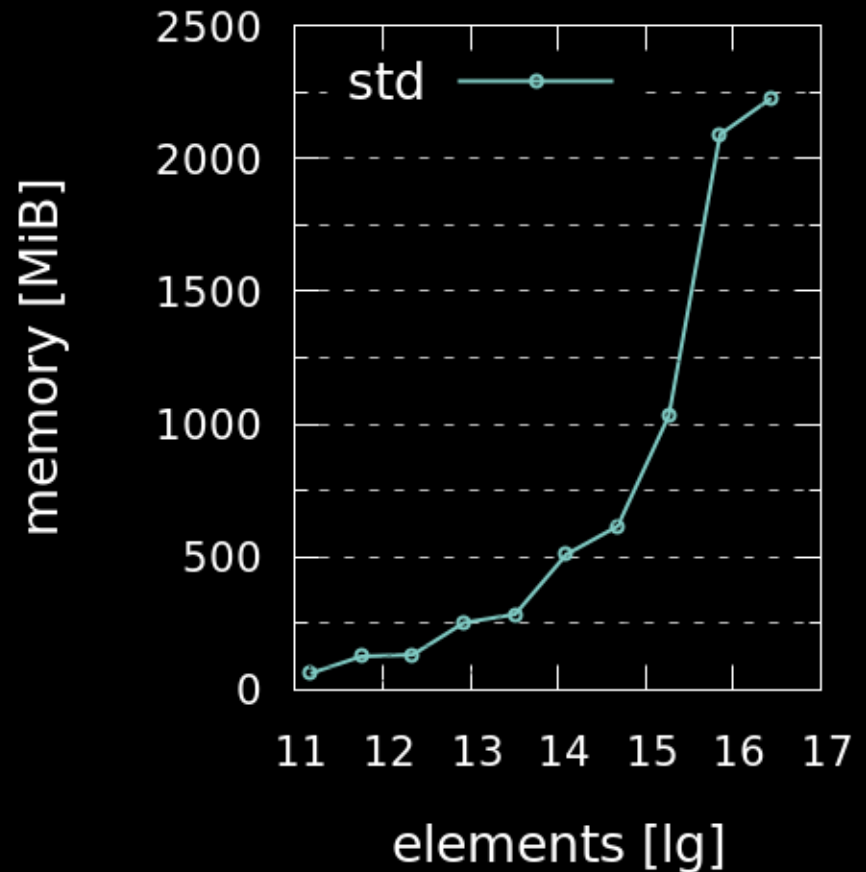
example:

- $|K| = 2^{32}$

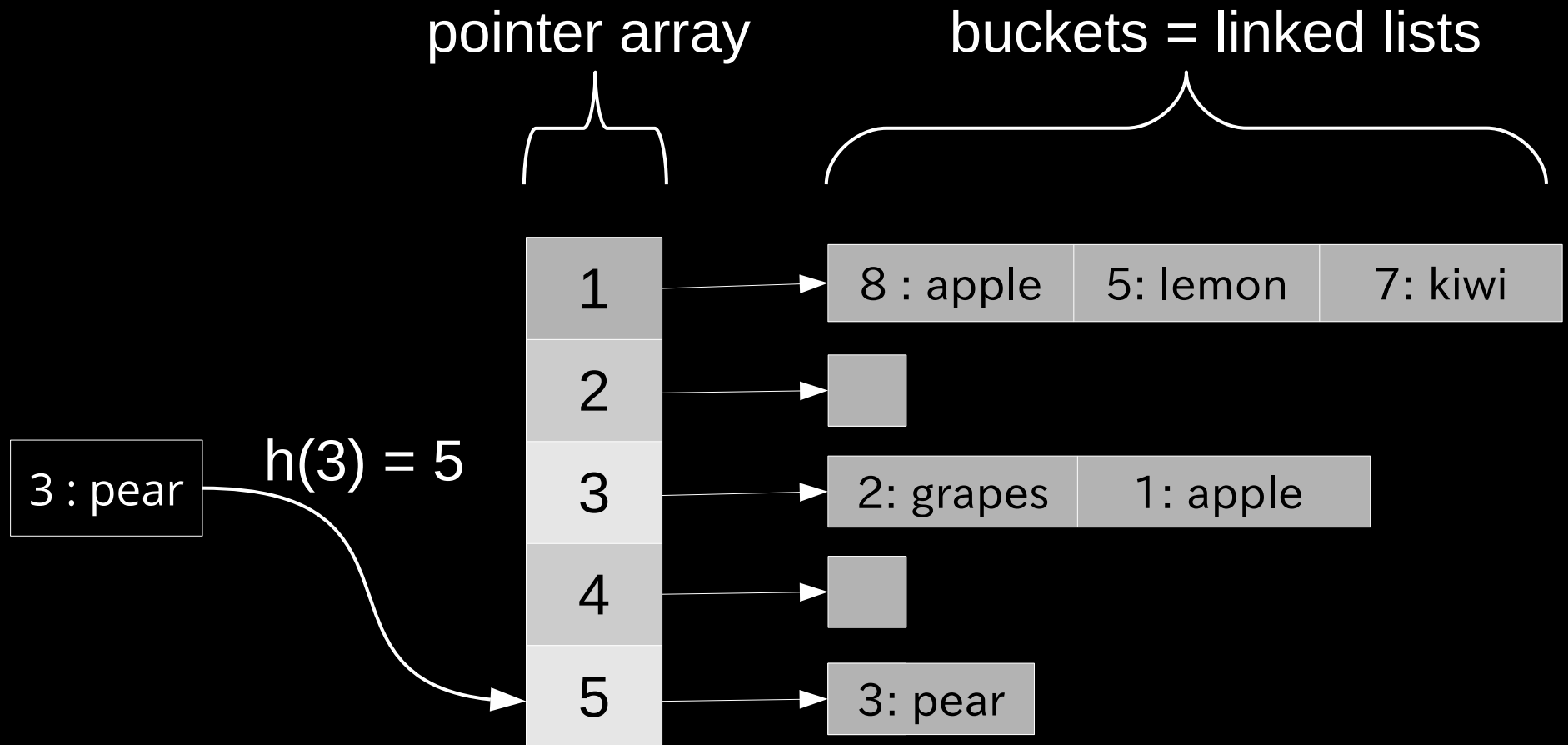
- $|V| = 2^{32}$

memory benchmark

- setting :
 - 32 bit keys
 - 32 bit values
 - randomly generated
- `std`: C++ STL hash table `unordered_map`
 - closed addressing
 - $n = 2^{16} = 65536$: more than 2 GiB RAM needed!



closed addressing

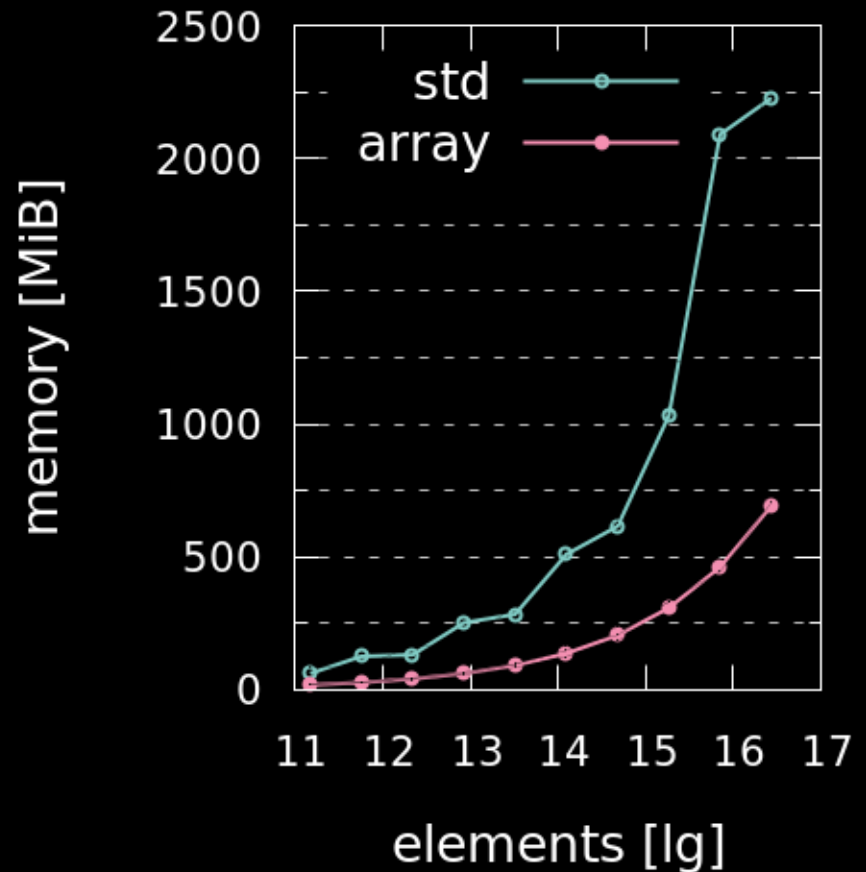


h: hash function

array list

array:

- key and values stored in a list
- ordered by insertion time

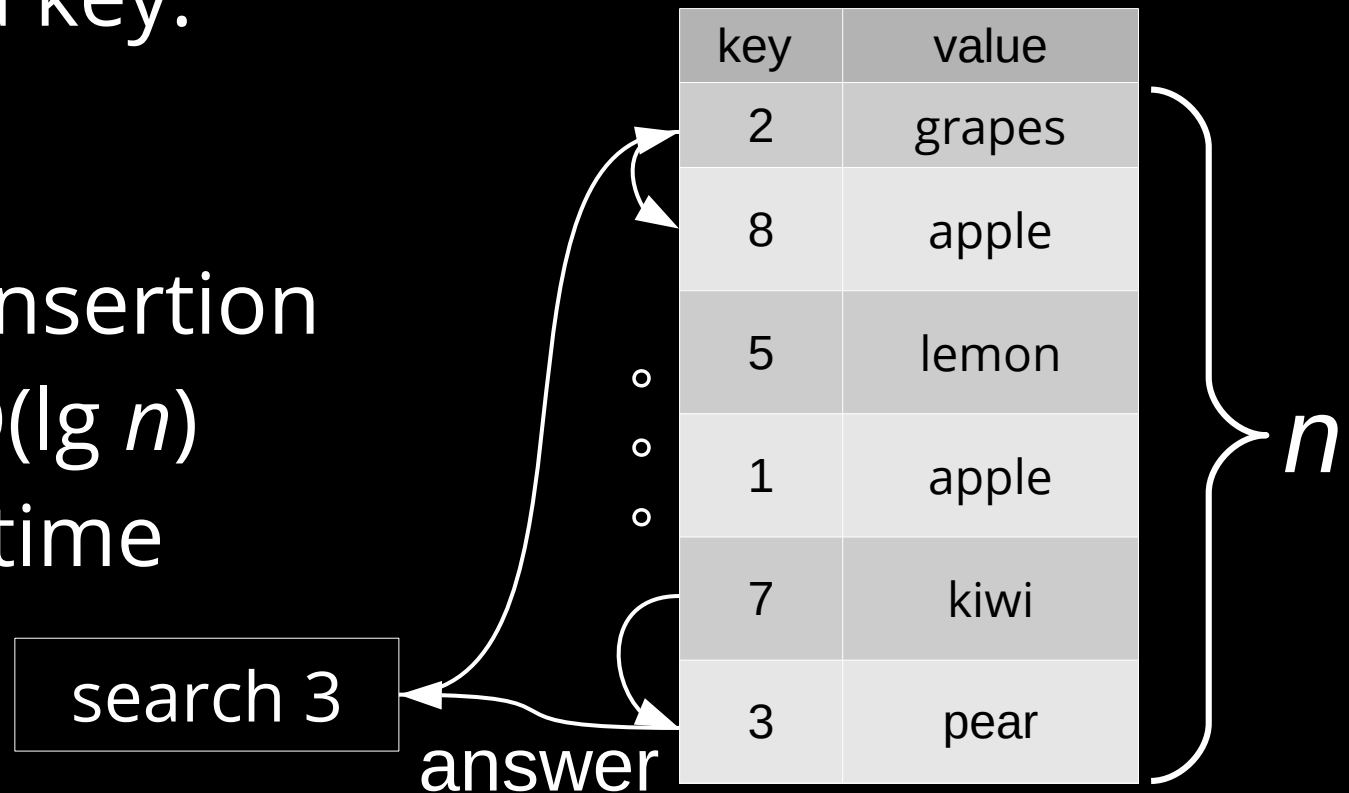


array list

searching a key:

- $O(n)$ time
- if we sort, insertion becomes $O(\lg n)$ amortized time

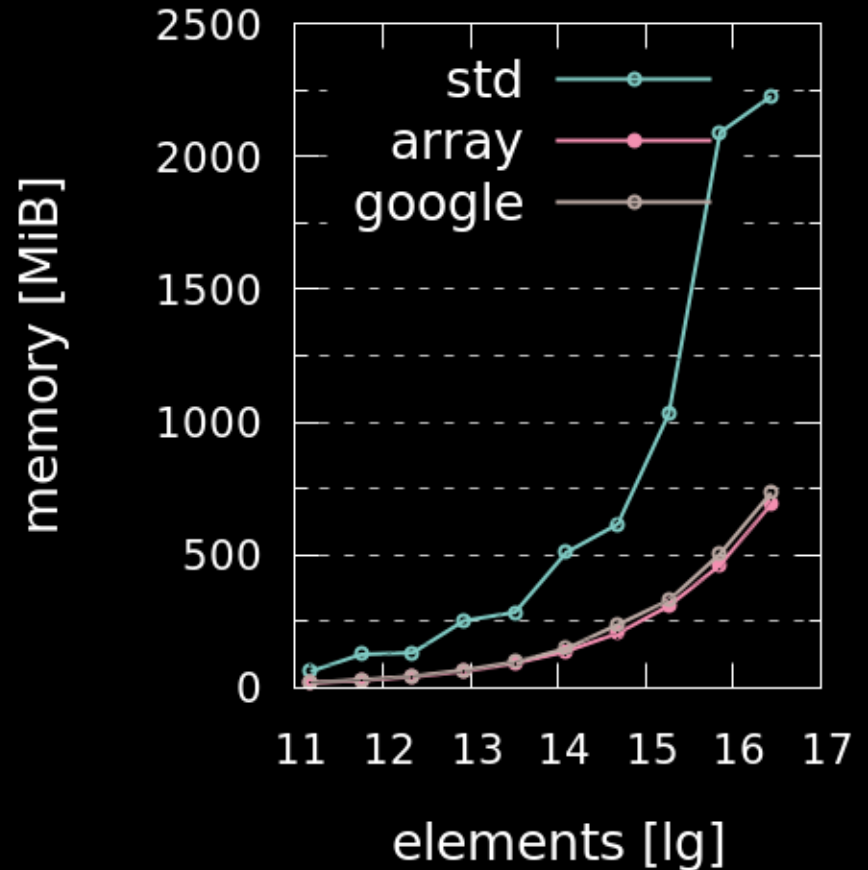
(not fast)



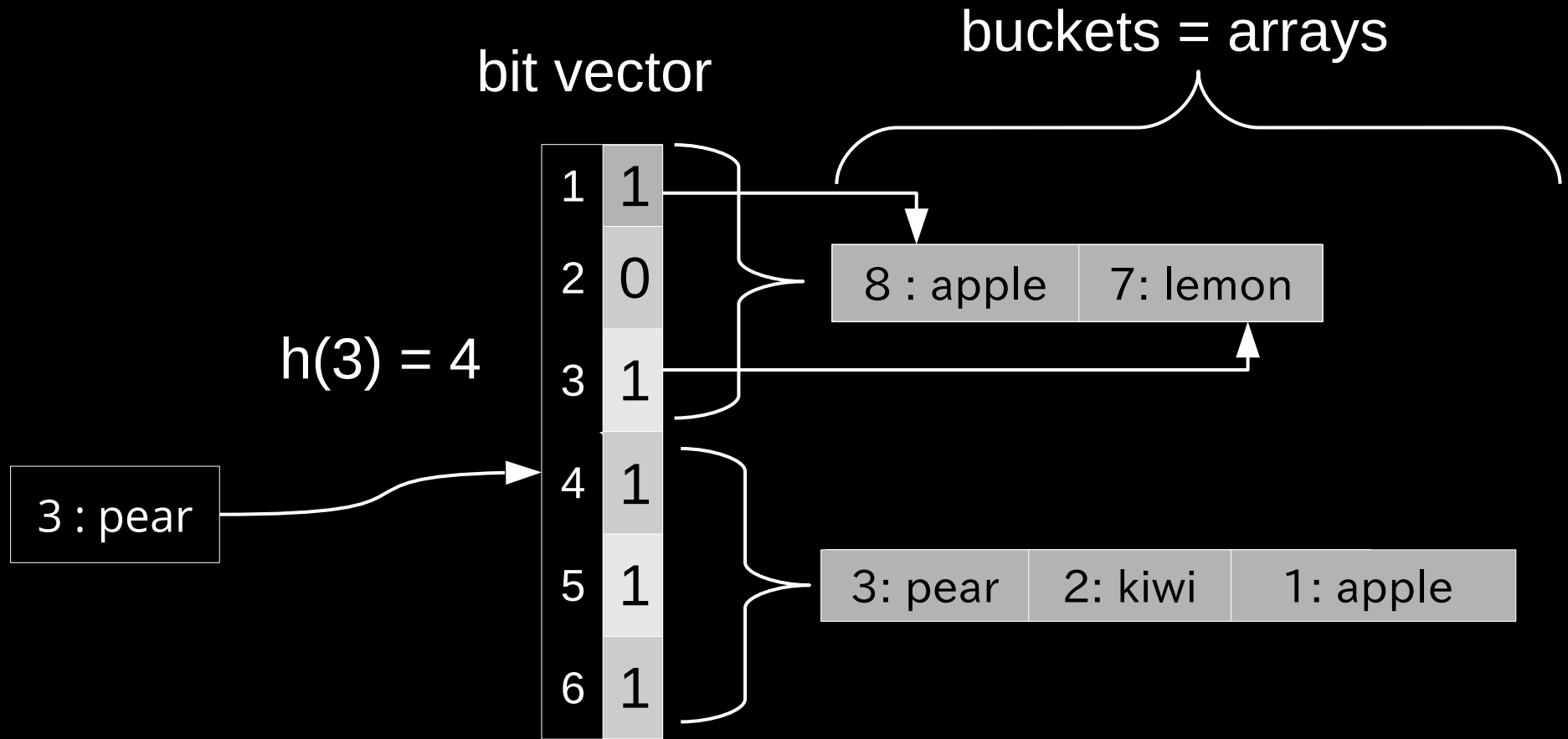
google sparse hash

google:

- open addressing
- grouped into *dynamic* buckets
- a bit vector addresses buckets



sparse hash table



compact hashing

Cleary '84:

- open addressing
- $\varphi : K \rightarrow \varphi(K)$ bijection
 - $\varphi(k) = (h(k), r(k))$
 - $\varphi^{-1}(h(k), r(k)) = k$
- instead of k store $r(k)$
(may need less space than k)

compact hashing

$$\varphi(k) = (h(k), r(k))$$

$h(k)$ ($r(k)$, value)

1	2: kiwi
2	1: apple
3	2: lemon
4	3: apple
5	

$$\varphi(5) = (3, 2)$$

5 : lemon

$$\varphi^{-1}(3, 2) = 5$$

Clearly: linear probing

$$\varphi(k) = (h(k), r(k))$$

$h(k)$ ($r(k)$, value)

displacement

info

$$\varphi(4) = (3, 1)$$

4 : pear

1	2: kiwi
2	1: apple
3	2: lemon
4	3: apple
5	1: pear

collision

3

$$\varphi^{-1}(5, 1) = 8 \neq 4$$

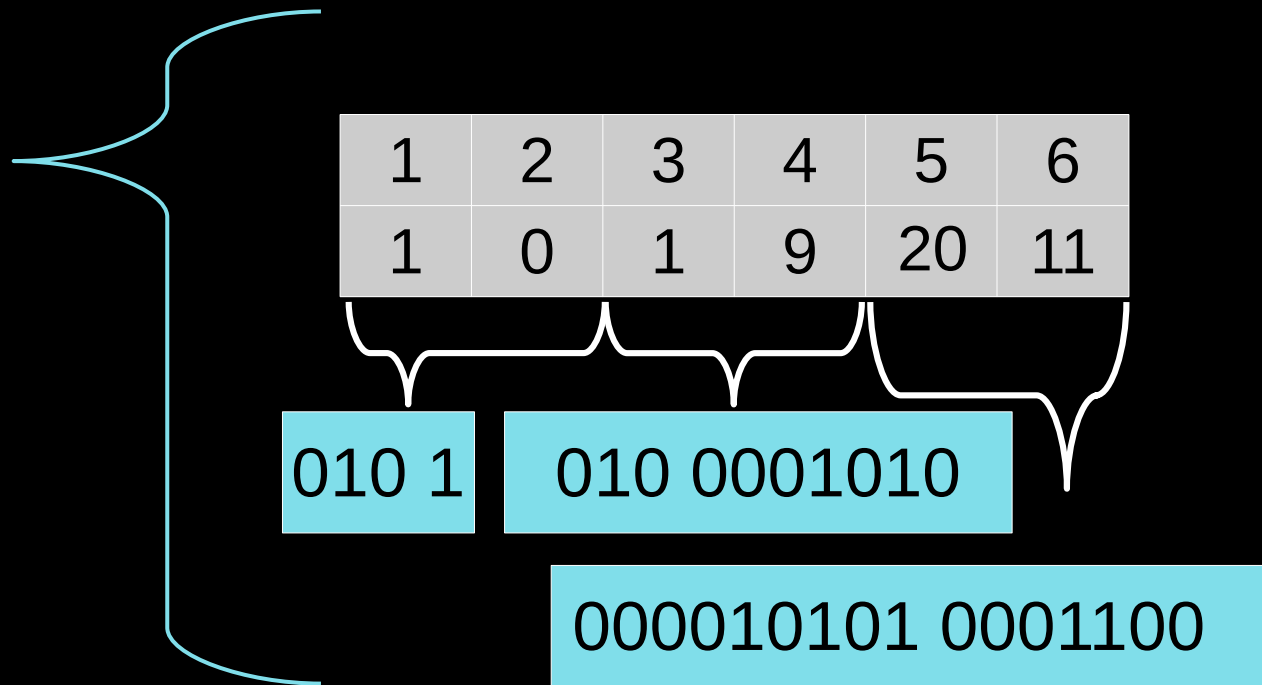
as a plain array:
costs too much space!

displacement info

representations :

- Cleary '84: $2m$ bits
- Poyias+ '15:
 - Elias γ code
 - layered array

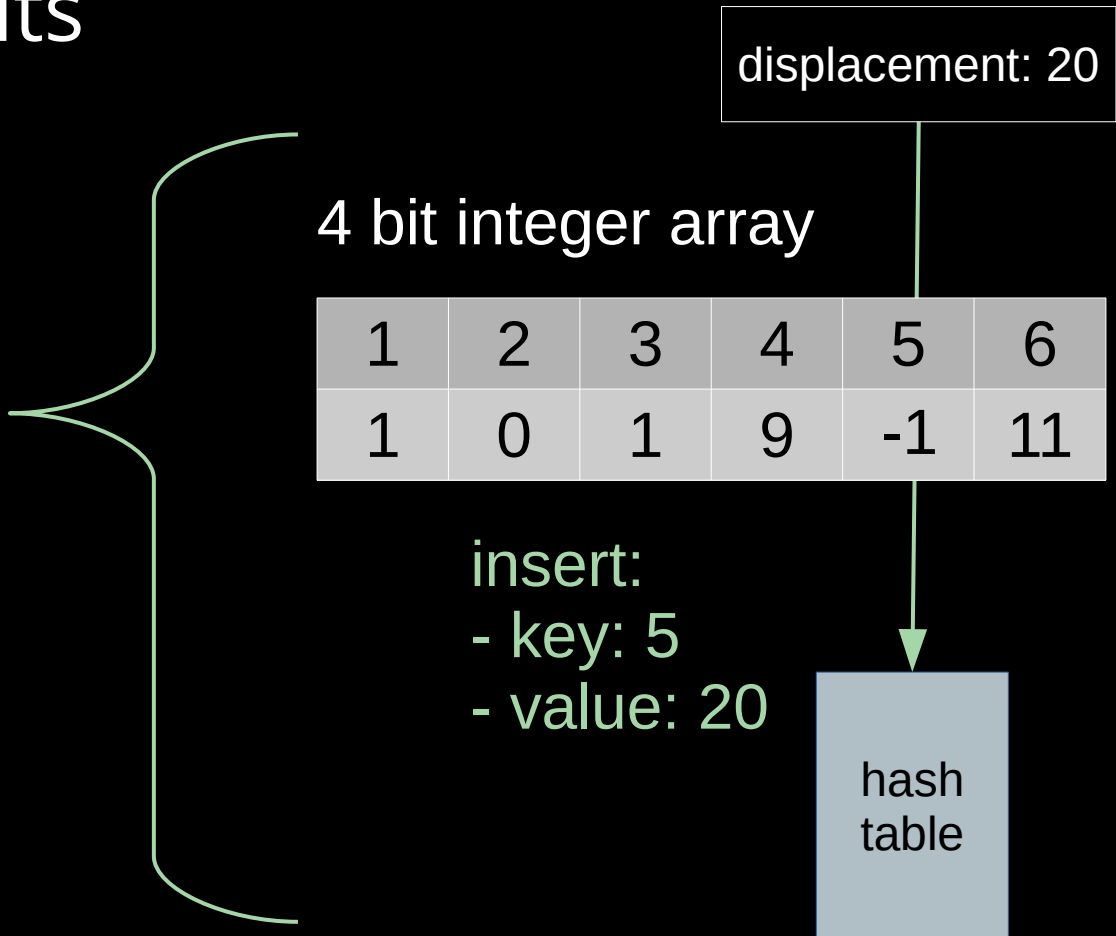
m : image size of h
= # cells in H



displacement info

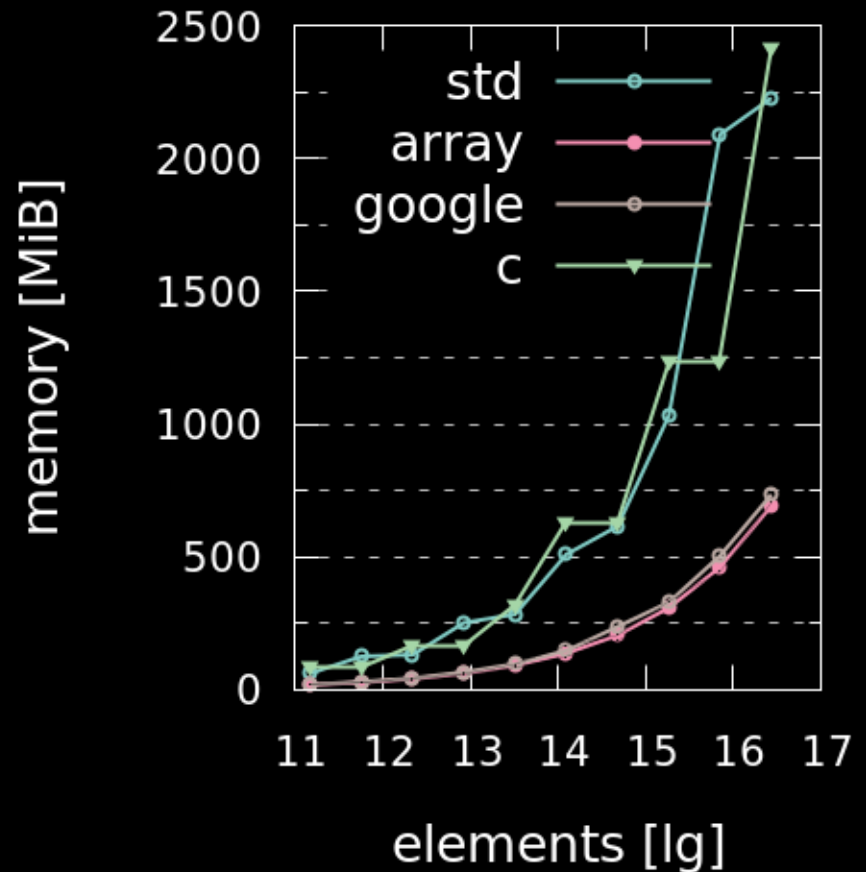
representations :

- Cleary '84: $2m$ bits
- Poyias+ '15:
 - Elias γ code
 - layered array



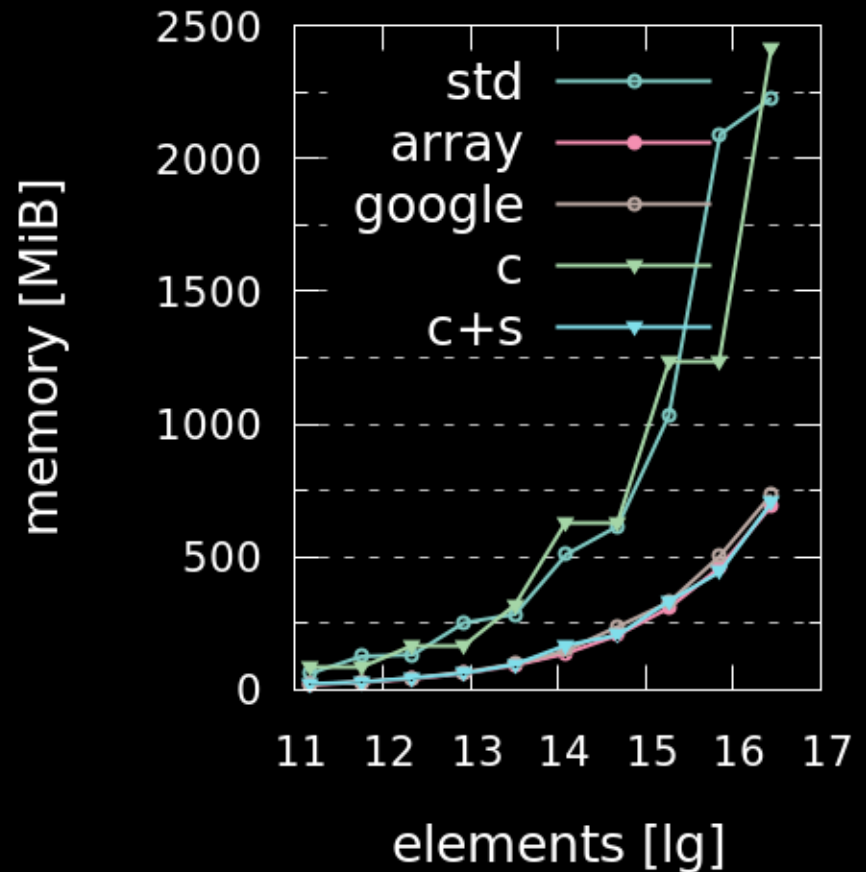
memory benchmark

- **c: compact**
 - layered
 - max. load factor 0.5
- not space efficient!



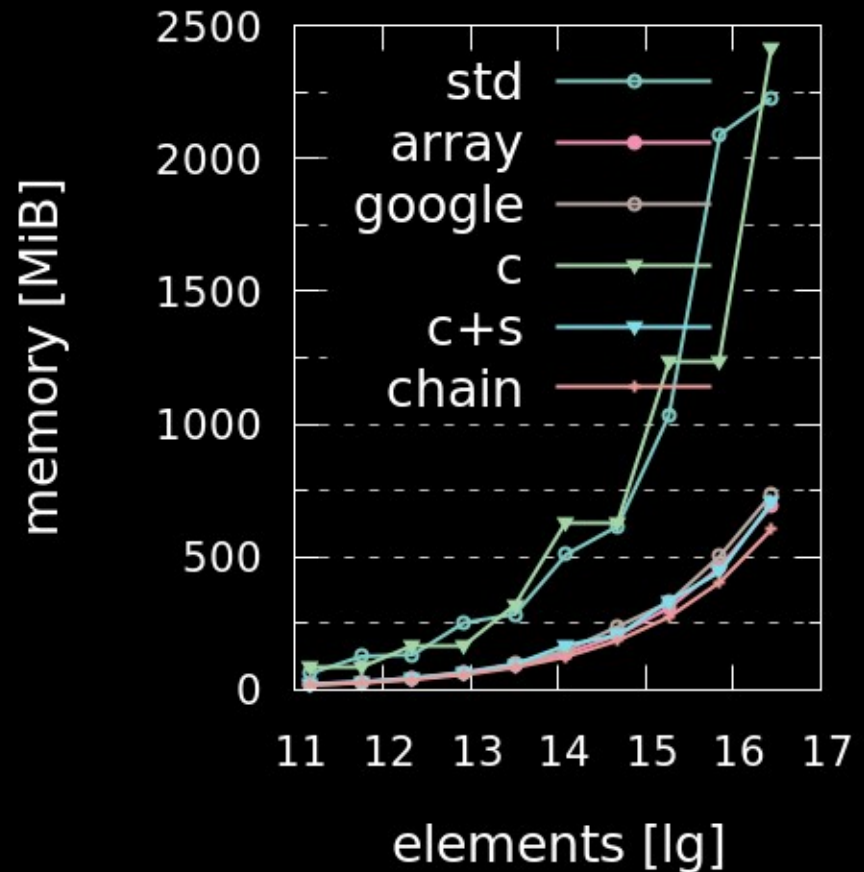
memory benchmark

- **c+s**: composition of
 - compact with
 - sparse
- competitive with **array**



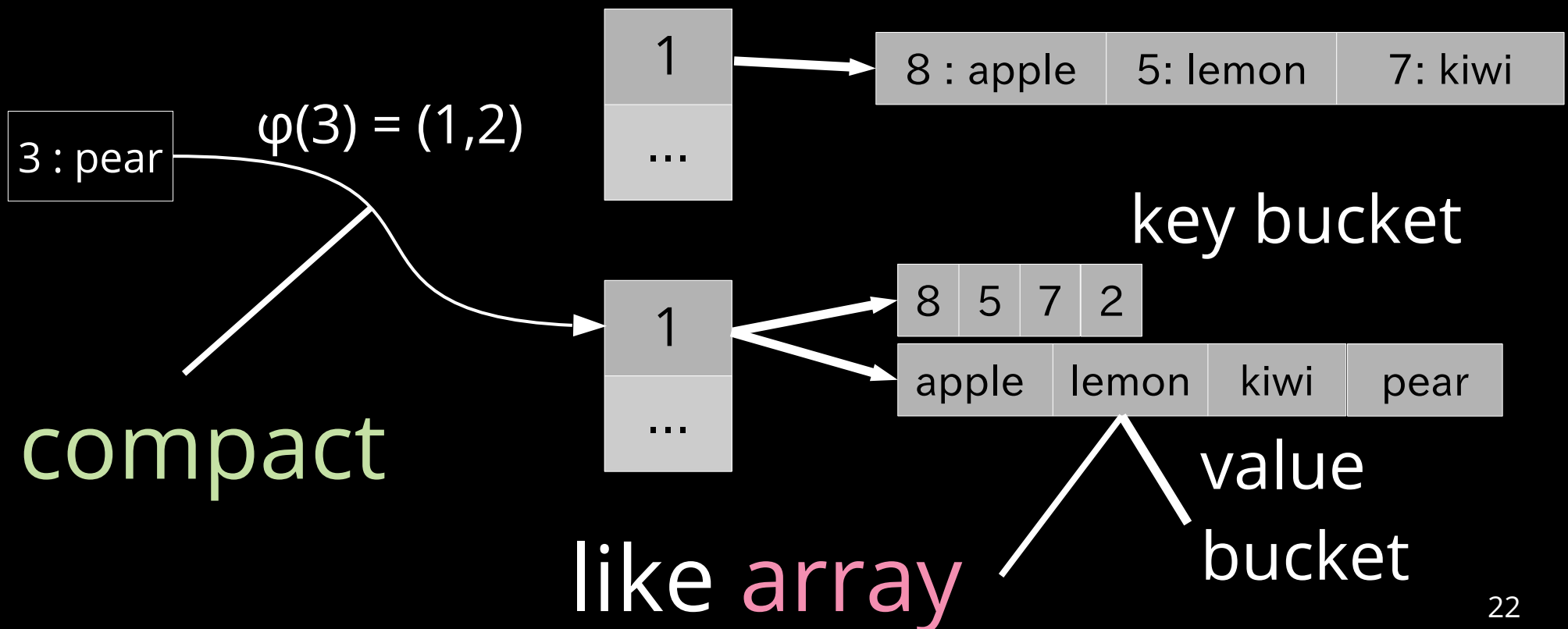
chain

- composition of
 - closed addressing
 - array
 - compact
- most space efficient (our contribution)



chain

- closed addressing
- buckets: instead of lists use two arrays



chain: space analysis

- a bucket costs $O(\omega)$ bits (pointer + length)
- want $O(n \lg n)$ bits
 \Rightarrow # buckets: $O(n / \omega)$

space for improvement!

- then $m = n / \omega$ (image size of h)

- $r(k)$ uses $\sim \omega - \lg(n / \omega) = \omega - \lg n + \lg \omega$ bits

- $K = [1..2^\omega]$
- n : #elements

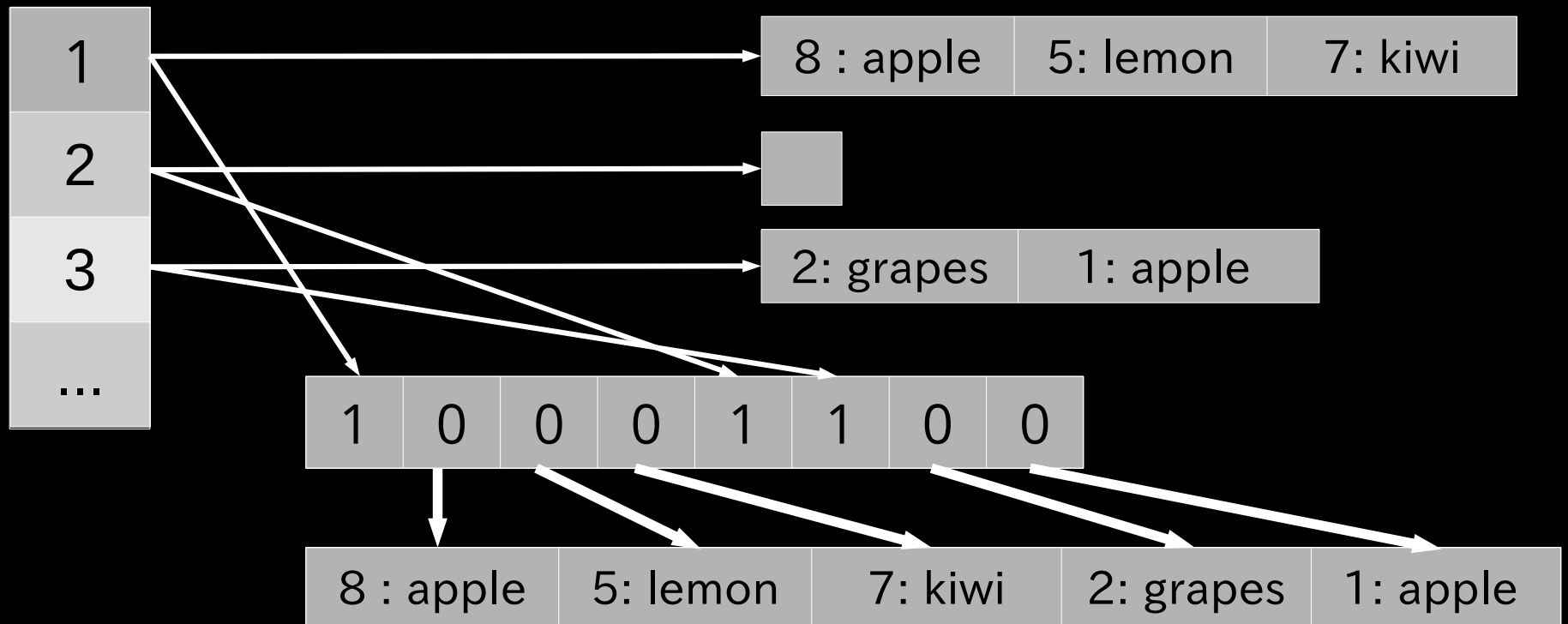
$r(k)$ of compact

improve space

- want n buckets such that $m = n$
- but each bucket costs $O(\omega)$ bits!
- idea: maintain buckets in a group
(similar to sparse)

chain → grp

- **chain** represents each bucket separately
- **grp** uses bit vector to mark bucket boundaries



rehashing

chain

- if a bucket reaches $O(\omega)$ elements

grp

- if a group reaches $O(\omega)$ elements
- group bit vector has $O(\omega)$ bits,
- scan bit vector naively

we set this maximum bucket / group size to 255
in practice (\Rightarrow length costs a byte)

insertion time

chain

- bucket has $O(\omega)$ elements

grp

- group has $O(\omega)$ elements

⇒ $O(\omega)$ worst-case time
(assuming that we do not need to rehash)

query time

chain

- bucket has $O(\omega)$ elements
 $\Rightarrow O(\omega)$ worst-case time

assume that $\Omega(\omega)$ bits fit into a machine word

grp

- bit vector has $O(\omega)$ bits
 \Rightarrow find respective bucket in $O(1)$ expected time
- bucket size is $O(1)$ expected
 $\Rightarrow O(1)$ expected time

theoretic space bounds

to store n keys from $K = [1..2^\omega]$

we need at least

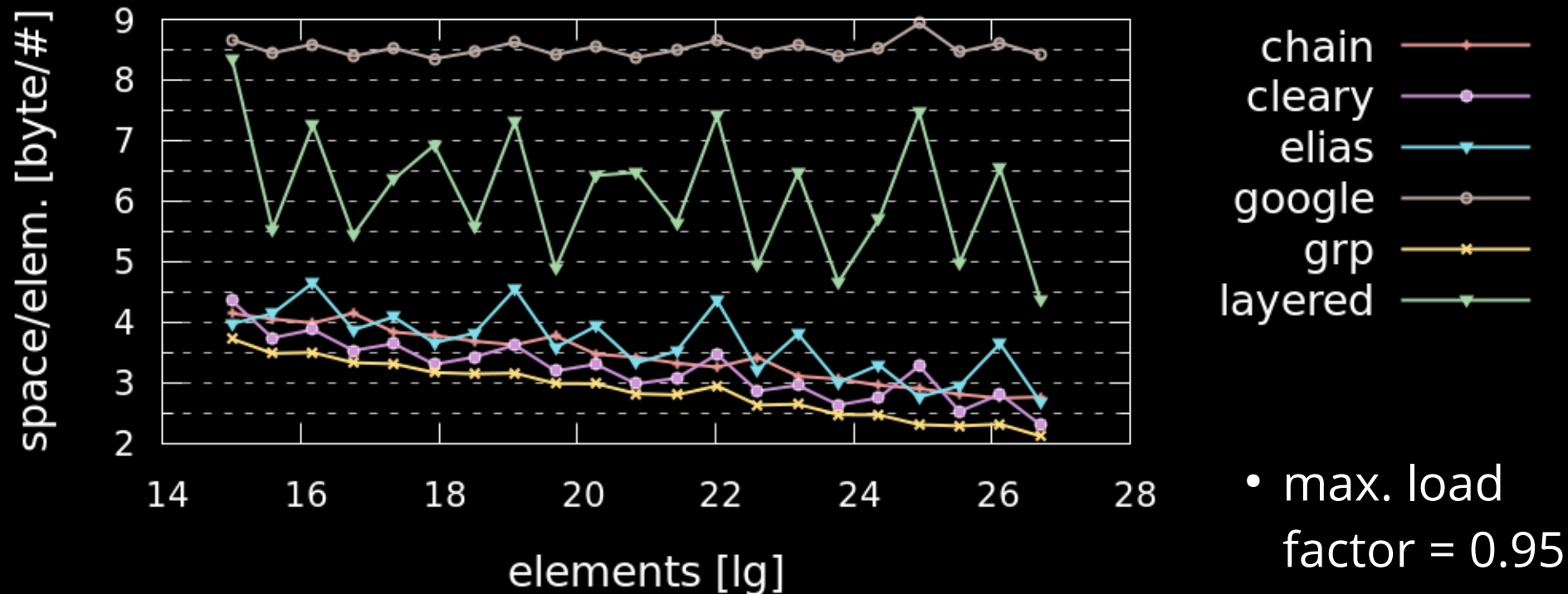
$$B := \lg \binom{2^\omega}{n} = n\omega - n \lg n + O(n) \text{ bits}$$

theoretic space bounds

$\varepsilon \in (0,1]$ constant

	construction		query
	space in bits	time	expected time
cleary	$(1+\varepsilon) B + O(n)$	$O(1/\varepsilon^3)$ exp.	$O(1/\varepsilon^2)$
elias	$(1+\varepsilon) B + O(n)$	$O(1/\varepsilon)$ exp.	$O(1/\varepsilon)$
layered	$(1+\varepsilon) B + O(n \lg \lg \lg n)$	$O(1/\varepsilon)$ exp.	$O(1/\varepsilon)$
chain	$B + O(n \lg \omega)$	$O(\omega)$ worst	$O(\omega)$ worst
grp	$B + O(n)$	$O(\omega)$ worst	$O(1)$

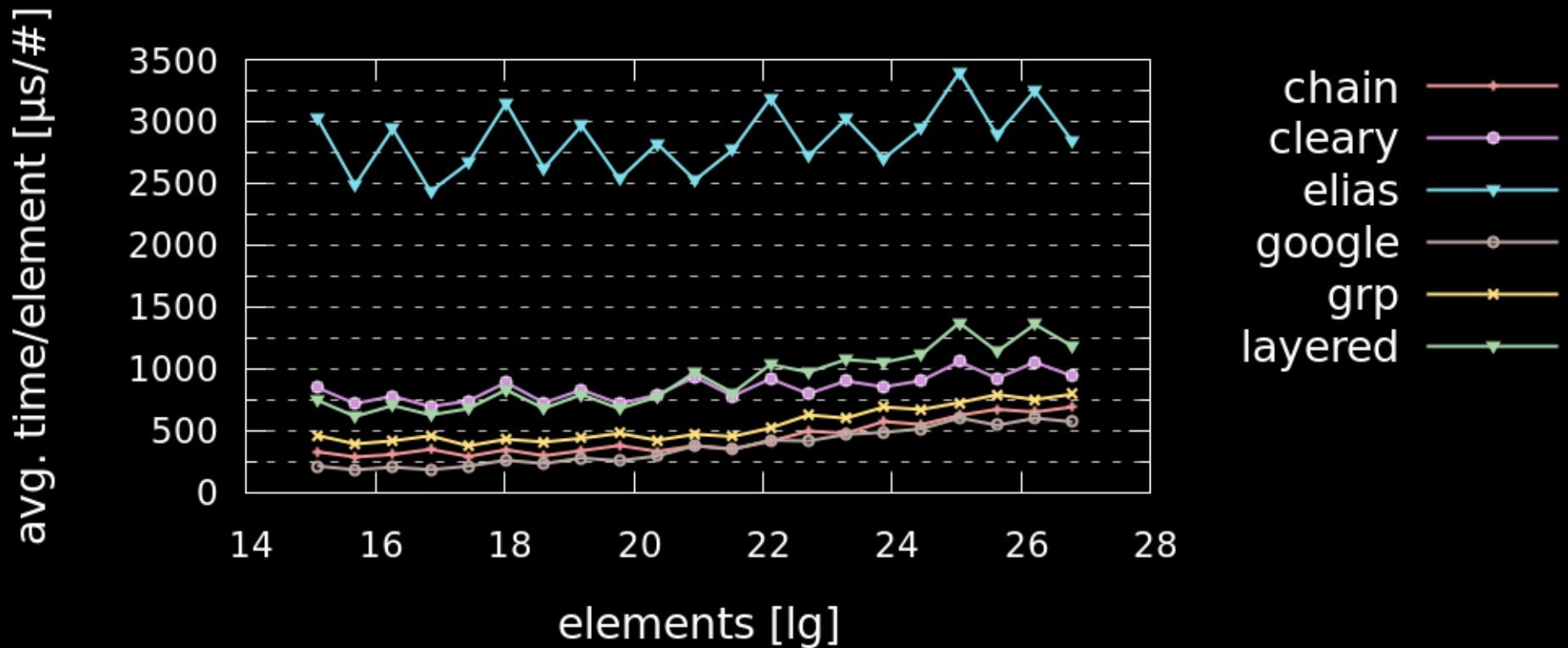
average space per element



- **grp** has the smallest space requirements
- **cleary**, **chain**, and **elias** are roughly equal
- **google** and **layered** are not as space economic

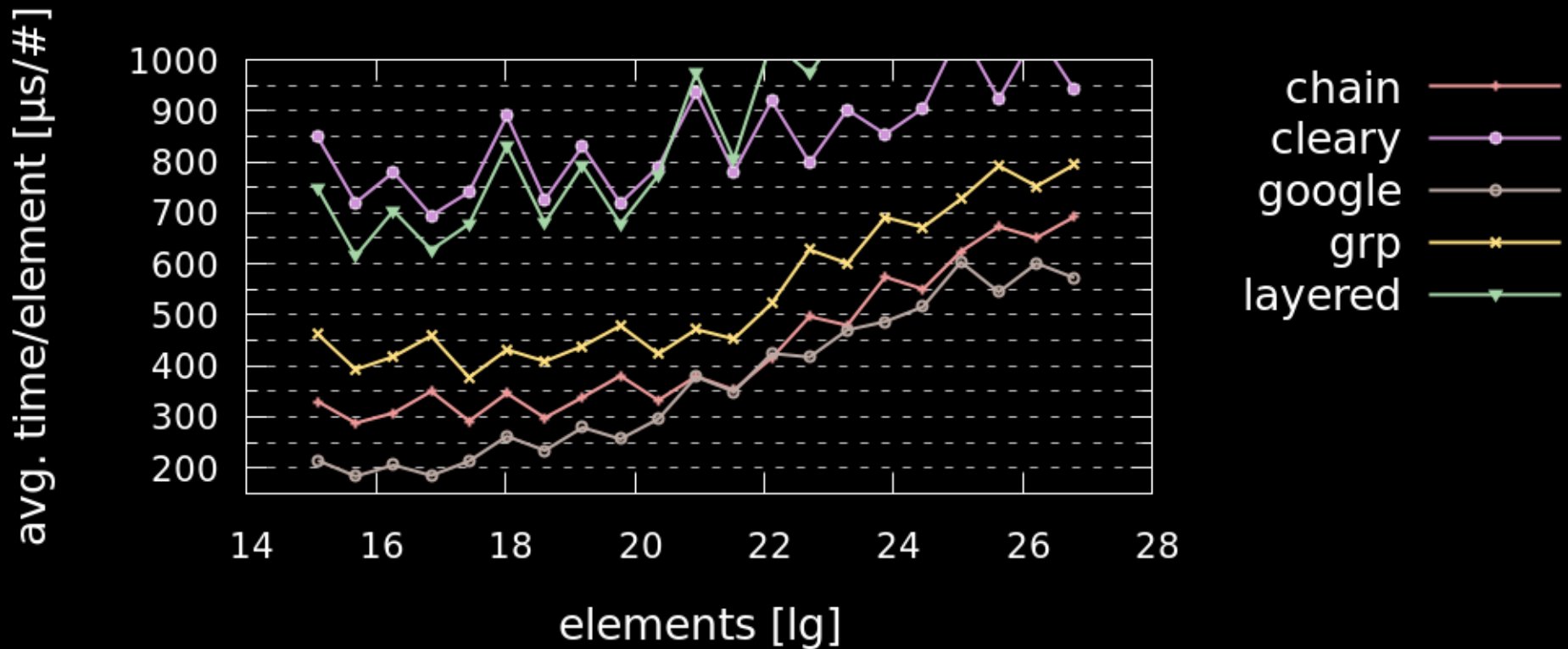
- max. load factor = 0.95
- use sparse layout
- 32 bit keys
- 8 bit values

construction time



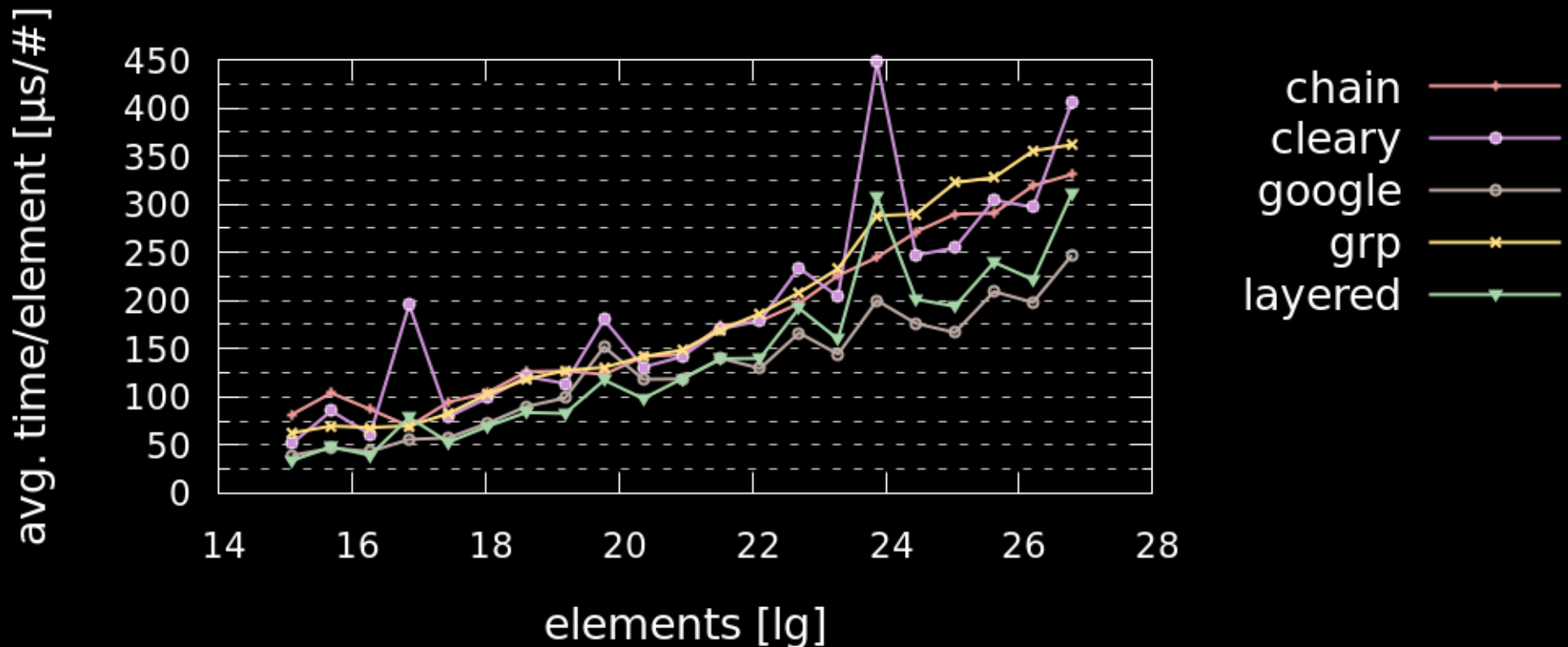
elias is very slow → omit it

construction time



- google is fastest
- grp is always slower than chain
- cleary and layered are slow

query time



- **grp** is mostly slower than **chain**
- **google** is fastest. **cleary** and **layered** have spikes (happening at high load factors)

experimental summary

	construction		query
hash table	space	time	time
google	bad	fast	fast
cleary	good	slow	slow
elias	good	very slow	very slow
layered	average	slow	fast
chain	good	fast	slow
grp	best	fast	slow

but sometimes slower than grp at high loads

proposed two hash tables

- techniques are combination of
 - closed addressing
 - bucketing [Askitis'09]
 - compact hashing [Cleary'84]
 - bit vector like in google's sparse table
- characteristics:
 - no displacement info
 - memory-efficient
 - fast construction but
 - slow query times
- current research:
 - speed up queries with SIMD
 - overflow table for averaging the loads of the buckets

thank you for watching!