


# ONE-PROBE SEARCH

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## Arrays

The simplest RAM data structure is the *array*.

An array of size  $s$  has the following properties:

- Given an index  $i \in \{1, \dots, s\}$  it returns the word written at position  $i$  in the array.
- Uses just one memory access.
- The space usage is  $s$  words.

We consider *sparse arrays* where all but  $n \ll s$  entries are empty:

How close can we get to the performance of an array using less space?

# Hashing

## Hashing in a nutshell:

- Read the description of a hash function  $h$ .
- Search for entry  $i$  (starting) at memory location  $h(i)$ .

## Properties:

- Uses space  $O(n)$ .
- Uses at least two memory probes (three in the worst case).
- Good hash functions are hard to maintain deterministically.

What can be done using *one* memory probe?

## One-probe sparse arrays

### Impossibility:

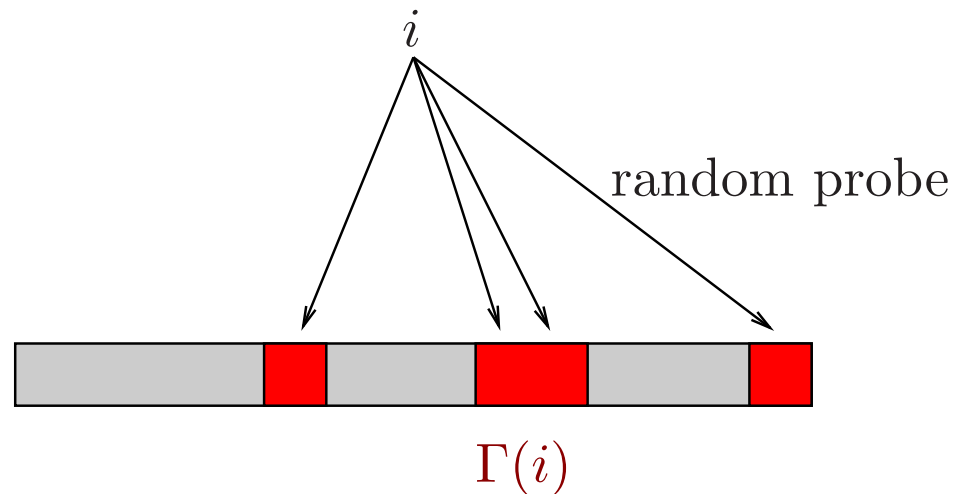
- One-probe search is impossible if the (approximate) size of the data structure is unknown.
- The space usage of an array cannot be improved by any *deterministic* one-probe implementation of a sparse array.

Deterministic lower bounds for membership queries are broken by randomized (Monte Carlo) schemes [Buhrman et al. '00].

# Randomized one-probe sparse arrays

## Randomized (Las Vegas) one-probe scheme:

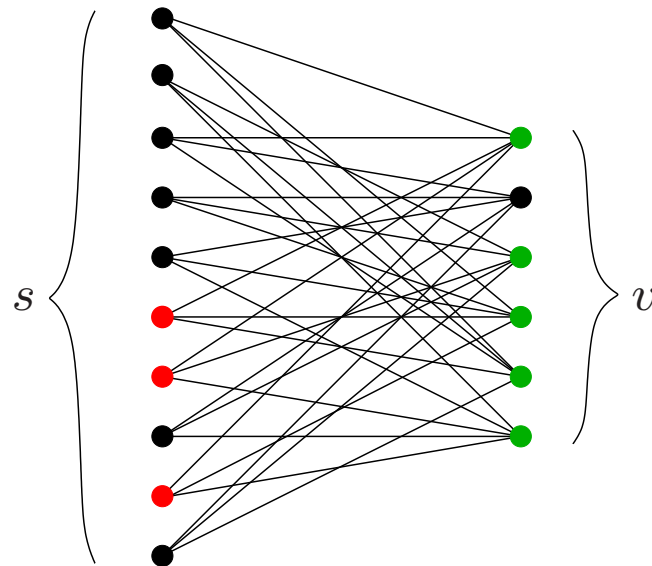
- Knows the size of the data structure.
- Looks up one word chosen at random from the data structure.
- Returns the correct answer with probability  $1 - \epsilon$ , and otherwise returns *don't know*.
- By iterating, it always returns the correct answer.



## Bipartite expander graphs

Bipartite  $(n, d, \epsilon)$ -expander graph:

Any set of  $k \leq n$  nodes (of degree  $d$ ) on the left have at least  $(1 - \epsilon)dk$  neighbors.



For any  $s$  and constant  $\epsilon > 0$  there exists an  $(n, d, \epsilon)$ -expander graph with  $d = O(\log s)$ ,  $v = O(n \log s)$ .

## Lookup procedure

Suppose we have an  $(2n + 1, d, \epsilon/2)$ -expander with neighbor function  $\Gamma : \{1, \dots, s\} \rightarrow \mathcal{P}(\{1, \dots, v\})$ ,  $v = O(n \log s)$ .

### Data structure:

- Array  $T$  of length  $v$ .
- Each entry has the fields *index* and *info*.

```
procedure lookup( $i$ )  
  choose  $v \in \Gamma(i)$  uniformly at random;  
  if  $T[v].\text{index} = i$  then return  $T[v].\text{info}$   
  else if  $T[v].\text{index} \in \{\neg i, \perp\}$  then return empty  
  else return don't know;  
end;
```

Equality-based!

## Requirements

Let  $S \subseteq \{1, \dots, s\}$ ,  $|S| = n$ , be the subset of nonempty array entries.

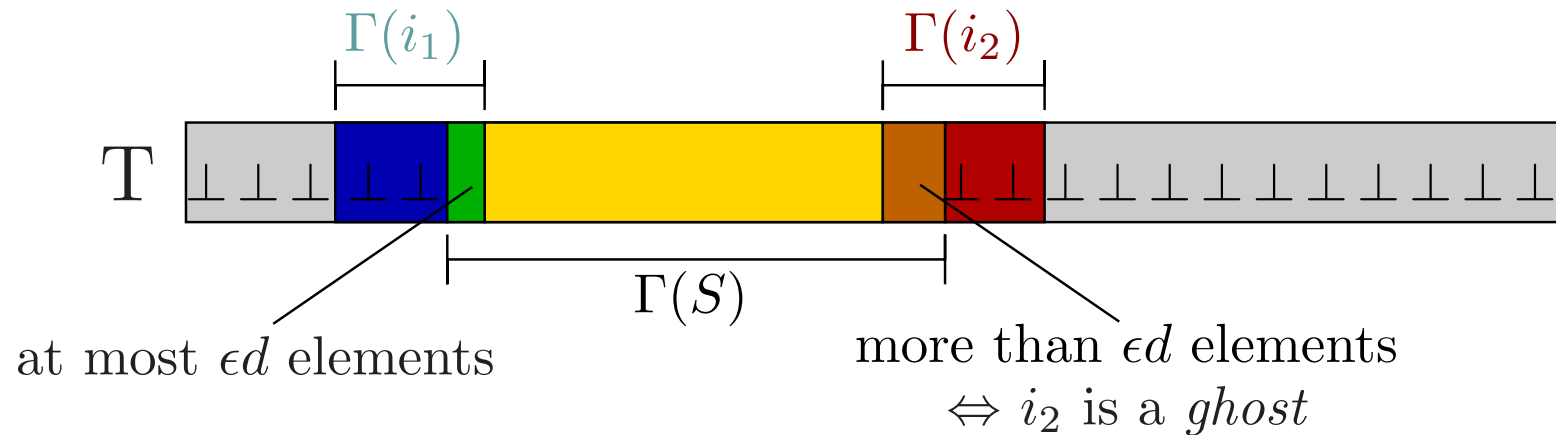
For each index  $i \in \{1, \dots, s\}$  our data structure must satisfy:

- When  $i \in S$ :
  - $T[v].\text{index} = i$  for a fraction  $1 - \epsilon$  of  $v \in \Gamma(i)$ .
  - $T[v].\text{index} \notin \{\neg i, \perp\}$  for  $v \in \Gamma(i)$ .
- When  $i \notin S$ :
  - $T[v].\text{index} \in \{\neg i, \perp\}$  for a fraction  $1 - \epsilon$  of  $v \in \Gamma(i)$ .
  - $T[v].\text{index} \neq i$  for  $v \in \Gamma(i)$ .



# Ghosts

Indices outside of  $S$  are easy to handle, except for a small set  $G$  of indices called the *ghosts* for  $S$ .



**Lemma** (Buhrman et al.):

In a  $(2n + 1, d, \epsilon/2)$ -expander there are at most  $n$  ghosts for  $S$ .

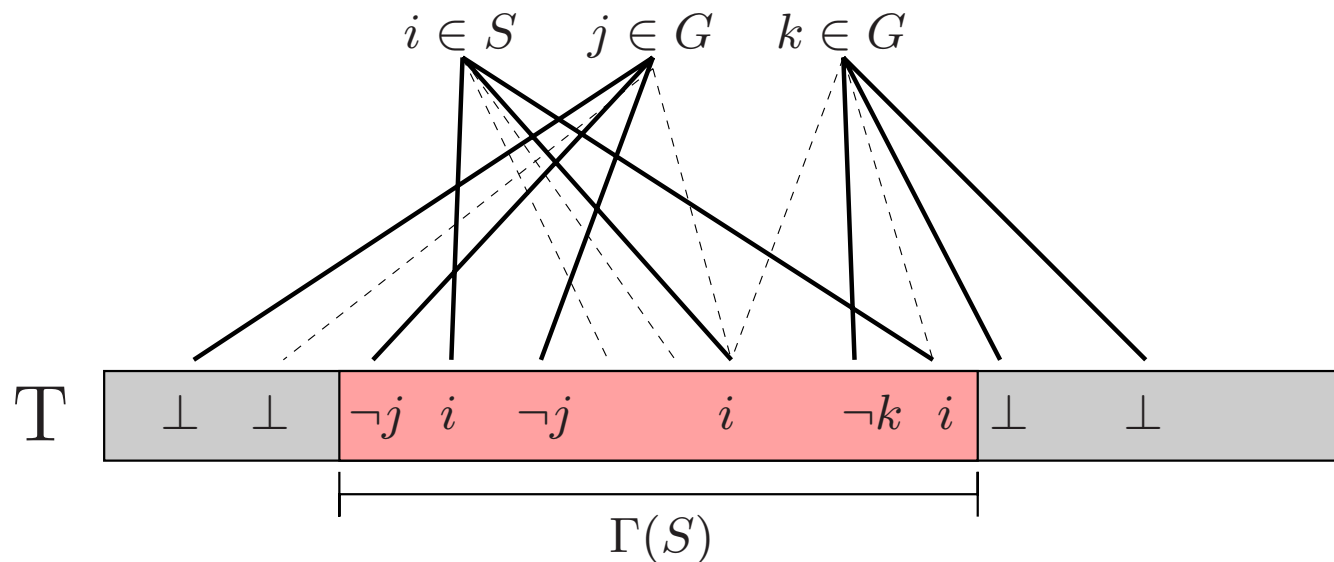
## Assignments

### Lemma:

In a  $(2n, d, \epsilon)$ -expander the indices in  $S \cup G$  can be assigned  $(1 - \epsilon)d$  unique neighbors each.

### Data structure initialization:

- For  $i \in S$  put  $i$  in assigned neighbors.
- For  $i \in G$  put  $\neg i$  in assigned neighbors in  $\Gamma(S)$ .



# Dynamization

## Greedly maintaining the assignment:

- Whenever an index in  $S \cup G$  has too few neighbors assigned to it, *all* its neighbors are assigned to it.
- Deterministic!

## Analysis (similar to [Brodal-Fagerberg '99]):

- The greedy algorithm is  $O(1)$ -competitive with the optimal off-line number of reassignments.
- The assignment can be maintained off-line with  $O(d)$  reassignments per insertion in  $S$ , amortized.

## Busting ghosts

Big problem:



How are ghosts identified?

Solution:

- We care about ghosts only when they are looked up (since the expected lookup time is too large).
- More than  $\log_{1/\epsilon} d$  random probes needed to find  $\perp$  in  $T[\Gamma(i)]$ 
  - $\Rightarrow i$  is a ghost with probability  $1 - O(1/d)$
  - $\Rightarrow$  We check  $T[\Gamma(i)]$  to see if  $i$  is a ghost.

## Overview of results

### Static data structure:

For any constant  $\epsilon > 0$  there is a nonexplicit one-probe sparse array with success probability  $1 - \epsilon$ , using space  $O(n \log s)$  words.  
(An explicit construction is possible in space  $n \cdot 2^{O((\log \log s)^3)}$  words.)

### Dynamization:

In the expander-augmented RAM model, we can perform a sequence of  $U$  updates and  $L$  lookups in the one-probe sparse array in time  $O(U(\log s)^{1+o(1)} + L + t)$  with probability  $1 - 2^{-\Omega(t/(\log s)^{1+o(1)})}$ .

## Lower bounds and open problems

### Space lower bound:

Yao's  $\Omega(\log n)$  lower bound on lookup time in implicit data structures implies that any equality-based approach must use space that grows with  $s$ .

### Some open problems:

- Is the space usage optimal for one-probe schemes?
- Linear space usage by allowing a succinct hash function?
- Can Yao's lower bound be broken using space, say,  $O(n \log^* s)$ ?