# Cell-Probe Lower Bounds for Succinct Partial Sums

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## Succinct Data Structures

N + o(N), N + O(N/lg N), N + O(VN), ...

Given some input of N bits

=> some data structure of close to N bits

to answer useful queries

#### Why?

**Practice:** functional data compression You often want to query compressed data, ... so the data structure on top better be small, too

**Theory:** algorithmic ideas with nice information theory flavor

## Interesting Upper Bounds

[Dodis, P, Thorup '10]

Store a vector A[1..n] from alphabet  $\Sigma$ 

- Space  $[n \log_2 \Sigma]$
- Constant time to read or write A[i]

[P, FOCS'08]

Store a vector A[1..n] of bits

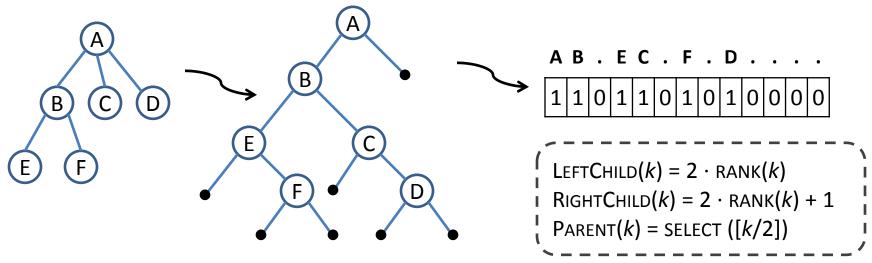
- Query time O(t) for RANK(k) := A[1] + ... + A[k]
- Space n + n / (w/t)<sup>t</sup>

## Rank / Select

RANK(*k*) = *A*[1] + ... + *A*[*k*] SELECT(*k*) = index of *k*<sup>th</sup> one in *A*[1..*n*]

A staple of succinct data structures.

Example: representing trees succinctly



## Interesting Lower Bounds

[Gál, Miltersen '03] polynomial evaluation => redundancy \* query time  $\geq \Omega(n)$ 

 $\ensuremath{\mathfrak{S}}$  nobody really expects a succinct solution

[Golynski SODA'09] store a permutation and query  $\pi(\cdot)$ ,  $\pi^{-1}(\cdot)$ With space  $2n \lg n$ , query time is 1 If space is  $(1 + \varepsilon) n \lg n =>$  query time is  $\Omega(1 / \sqrt{\varepsilon})$ 

[HERE] RANK / SELECT For query time  $t \Rightarrow$  space is  $\ge n + n / w^{O(t)}$ NB: Upper bound was  $n + n / (w/t)^{O(t)}$ 

# Models

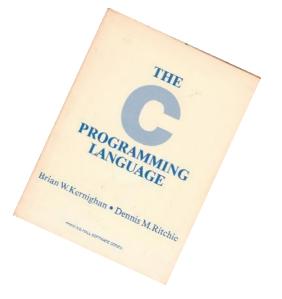
#### "Standard Model" (a.k.a. Word RAM)

- memory = words of w bits
- $w \ge \lg n$  (store pointers, indices)
- constant time ops: read/write memory,
   +, -, \*, /, %, <, >, ==, >>, <<, &, |, ^</li>

#### Lower bounds: "Cell-Probe Model"

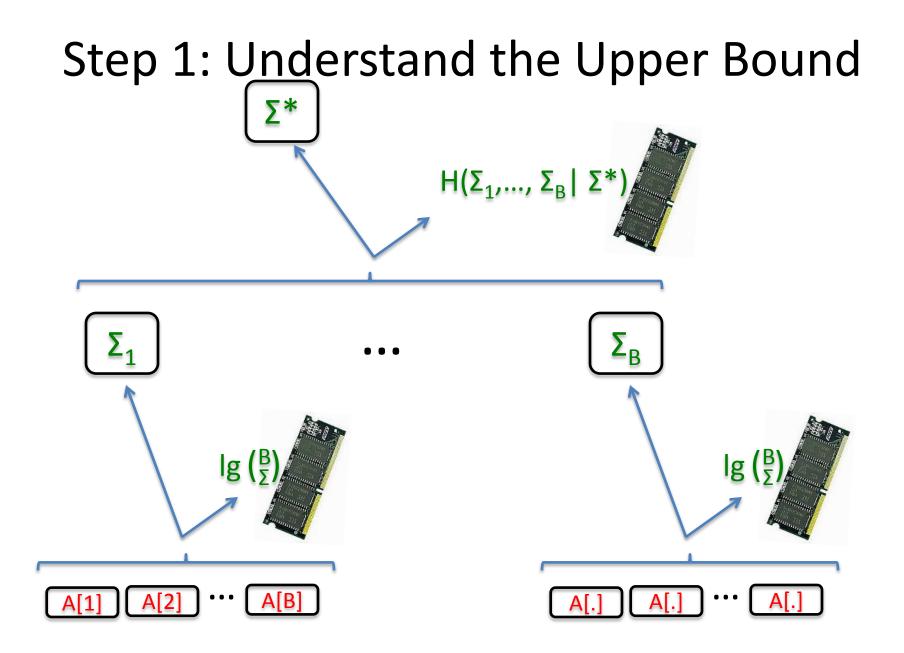
- memory = words (cells) of w bits
- time = # cell reads/writes

Nice: information theoretic, holds even with exotic instructions

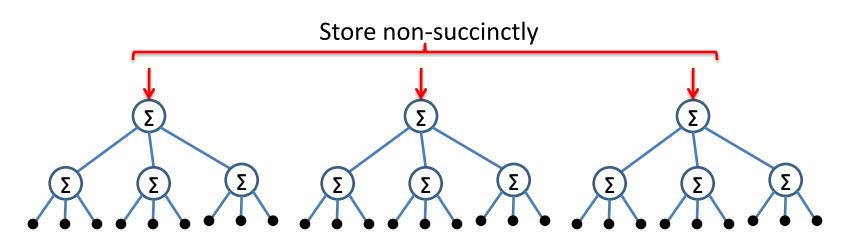


### The Lower Bound Proof





## Recursion



Spend constant time per node => decode  $\Sigma_i$  from  $\Sigma^*$  and the  $H(\Sigma_1,..., \Sigma_B | \Sigma^*)$  memory bits Set parameters such that:  $H(\Sigma_1,..., \Sigma_B | \Sigma^*) = O(w) \implies B \approx w$ Can go *t* levels up => redundancy  $\approx n / w^t$ 

## **Published Bits**

For induction, use a stronger model:

- memory = cells of *w* bits; cost 1 to read a cell
- *P* published bits, free to read (=cache)

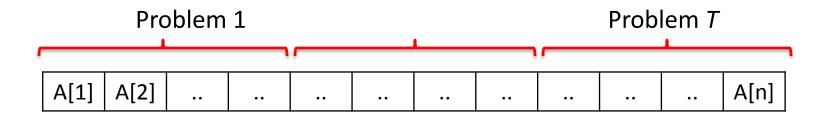
Initially, set P = redundancy -> publish some arbitrary<sup>(!)</sup> P bits

Intuitively, think of published bits ≈ sums at roots of the subtrees

## Published Bits Die by Direct Sum

Do published bits trivialize the problem?

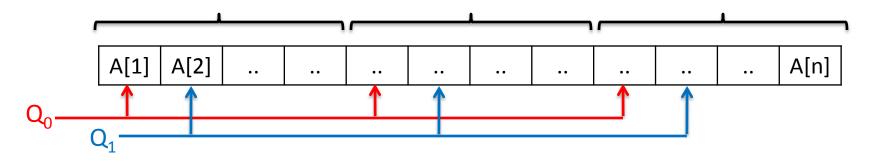
No. Consider *T*=100·*P* subproblems



For the average problem, published bits have 0.01 information => can't really help

## **Recursion Intuition**

Upper Bound		Lower Bound
Remove top level of the trees		Main Lemma:
=>	<i>t</i> -= 1	Remove one cell-probe
	redundancy *= w	Increase $P *= O(w)$



Intuition: { first cells probed by Q<sub>0</sub> } = { first cells probes by Q<sub>k</sub> } = the roots of the trees

So just publish { first cells probed by Q<sub>0</sub> } => get rid of 1 probe for all queries

## Careful for Bad Algorithms!

Upper	Bound	Lower Bound
Remove top level of the trees		Main Lemma:
=>	<i>t</i> -= 1	Remove one cell-probe
	redundancy *= w	Increase $P *= O(w)$

Can't make an argument based on **first** probe!

Intuition: { first cells probed by  $Q_0$  } = { first cells probes by  $Q_k$  } = the roots of the trees

So just publish { first cells probed by Q<sub>0</sub> } => get rid of 1 probe for all queries

## Intuition Fix

Upper	Bound	Lower Bound
Remove top level of the trees		Main Lemma:
=>	<i>t</i> -= 1	Remove one cell-probe
	redundancy *= w	Increase P *= O(poly(w))

```
Intuition: | Foot(Q_0) \cap Foot(Q_k) | = \Omega(P)
So just publish Foot(Q_0)
=> get rid of 1 probe for most queries
```

## Why are Cells Reused?

Suppose, for contradiction, that | Foot( $Q_0$ )  $\cap$  Foot( $Q_k$ ) | = o(P)

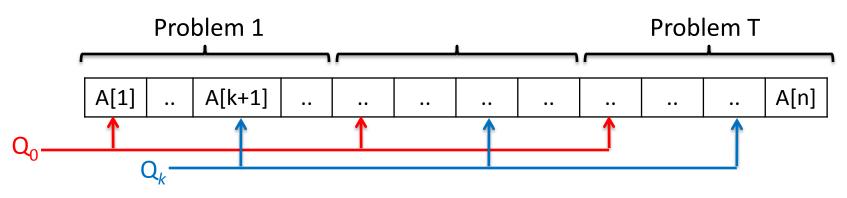
So the answers to most of  $Q_k$  can be decoded from Foot( $Q_0$ )

Ignored in this talk

But Answers( $Q_0$ ) and Answers( $Q_k$ ) are highly correlated: H(Ans( $Q_0$ )) + H(Ans( $Q_k$ )) >> H(Ans( $Q_0$ ), Ans( $Q_k$ ))

So the data structure is an inefficient encoding (non-succinct).

## **Entropy Computation**

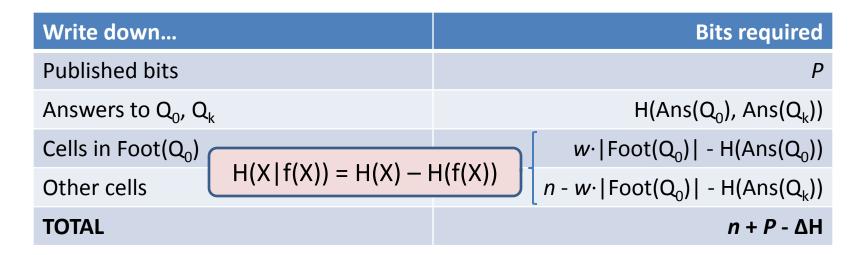


 $H(Ans(Q_0)) \approx H(Ans(Q_k)) \approx T \cdot H(binomial on n/T trials)$  $≈ T \cdot c lg(n/T) for some constant c$ 

 $\begin{aligned} \mathsf{H}(\mathsf{Ans}(\mathsf{Q}_0), \mathsf{Ans}(\mathsf{Q}_k)) &= \mathsf{H}(\mathsf{Ans}(\mathsf{Q}_0)) + \mathsf{H}(\mathsf{Ans}(\mathsf{Q}_k) \mid \mathsf{Ans}(\mathsf{Q}_0)) \\ &\approx T \cdot c \, \lg(n/T) + T \cdot H(\mathsf{binomial on } k \mathsf{ trials}) \\ &\approx T \cdot c \, \lg(n/T) + T \cdot c \, \lg k \\ &< T \cdot c \, \lg(n/T) + T \cdot c \, \lg n/(2T) \\ &= \mathsf{H}(\mathsf{Ans}(\mathsf{Q}_0)) + \mathsf{H}(\mathsf{Ans}(\mathsf{Q}_k)) - \Omega(T) \end{aligned}$ 

# **Proof by Encoding**

Claim: We will encode A with less than n bits (impossible)



 $\Delta H = H(Ans(Q_0), Ans(Q_k)) - [H(Ans(Q_0)) + H(Ans(Q_k))] = \Omega(T)$ 

Contradiction for T >> P

