

Cache-Oblivious Algorithms

Cache-Oblivious Model

The Unknown Machine

Algorithm



C program

↓ gcc

Object code

↓ linux

Execution

Can be executed on machines with a specific class of CPUs

Algorithm



Java program

↓ javac

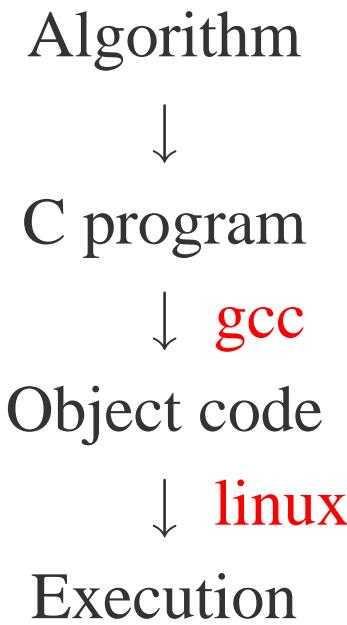
Java bytecode

↓ java

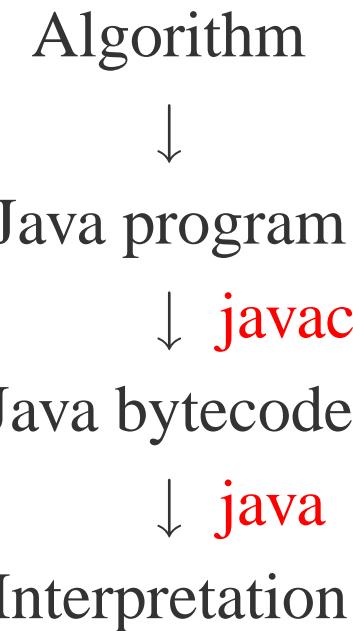
Interpretation

Can be executed on any machine with a Java interpreter

The Unknown Machine



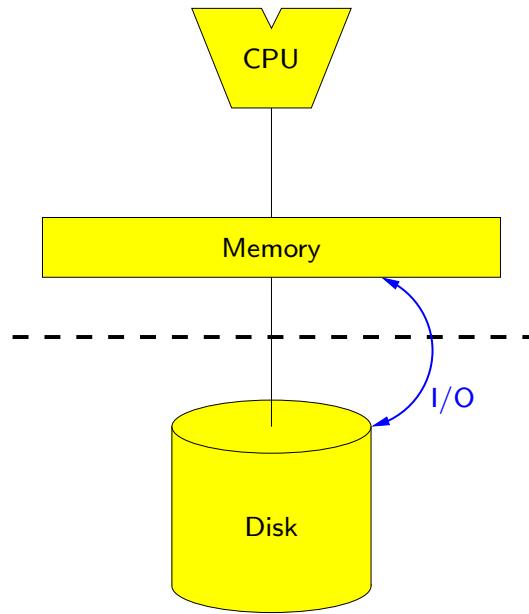
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Can be executed on any machine with a Java interpreter

Goal Develop algorithms that are optimized w.r.t. memory hierarchies without knowing the parameters

Cache-Oblivious Model



- I/O model
- Algorithms do not know the parameters B and M
- Optimal off-line cache replacement strategy

Frigo et al. 1999

Justification of the ideal-cache model

Optimal replacement

$\text{LRU} + 2 \times \text{cache size} \Rightarrow \text{at most } 2 \times \text{cache misses}$

Sleator an Tarjan, 1985

Corollary

$T_{M,B}(N) = O(T_{2M,B}(N)) \Rightarrow \#\text{cache misses using LRU is } O(T_{M,B}(N))$

Two memory levels

Optimal cache-oblivious algorithm satisfying $T_{M,B}(N) = O(T_{2M,B}(N))$
 \Rightarrow optimal #cache misses on each level of a **multilevel** cache using LRU

Fully associativity cache

Simulation of LRU

- Direct mapped cache
- Explicit memory management
- Dictionary (2-universal hash functions) of cache lines in memory
- Expected $O(1)$ access time to a cache line in memory

Matrix Multiplication

Matrix Multiplication

Problem

$$C = A \cdot B , \quad c_{ij} = \sum_{k=1..N} a_{ik} \cdot b_{kj}$$

Layout of matrices

0	1	2	3	4	5	6	7
8	9	10	11	12	13	14	15
16	17	18	19	20	21	22	23
24	25	26	27	28	29	30	31
32	33	34	35	36	37	38	39
40	41	42	43	44	45	46	47
48	49	50	51	52	53	54	55
56	57	58	59	60	61	62	63

Row major

0	8	16	24	32	40	48	56
1	9	17	25	33	41	49	57
2	10	18	26	34	42	50	58
3	11	19	27	35	43	51	59
4	12	20	28	36	44	52	60
5	13	21	29	37	45	53	61
6	14	22	30	38	46	54	62
7	15	23	31	39	47	55	63

Column major

0	1	2	3	16	17	18	19
4	5	6	7	20	21	22	23
8	9	10	11	24	25	26	27
12	13	14	15	28	29	30	31
32	33	34	35	48	49	50	51
36	37	38	39	52	53	54	55
40	41	42	43	56	57	58	59
44	45	46	47	60	61	62	63

4 × 4-blocked

0	1	4	5	16	17	20	21
2	3	6	7	18	19	22	23
8	9	12	13	24	25	28	29
10	11	14	15	26	27	30	31
32	33	36	37	48	49	52	53
34	35	38	39	50	51	54	55
40	41	44	45	56	57	60	61
42	43	46	47	58	59	62	63

Bit interleaved

Matrix Multiplication

Algorithm 1: Nested loops

- Row major
- Reading a **column** of B uses N I/Os
- Total $O(N^3)$ I/Os

```
for  $i = 1$  to  $N$ 
    for  $j = 1$  to  $N$ 
         $c_{ij} = 0$ 
        for  $k = 1$  to  $N$ 
             $c_{ij} = c_{ij} + a_{ik} \cdot b_{kj}$ 
```

Matrix Multiplication

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Algorithm 2: Blocked algorithm (cache-aware)

- Partition A and B into blocks of size $s \times s$ where $s = \Theta(\sqrt{M})$
- Apply Algorithm 1 to the $\frac{N}{s} \times \frac{N}{s}$ matrices where elements are $s \times s$ matrices

0	1	2	3	4	5	6	7
8	9	10	11	12	13	14	15
16	17	18	19	20	21	22	23
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$$O\left(\left(\frac{N}{s}\right)^3 \cdot \frac{s^2}{B}\right) = O\left(\frac{N^3}{s \cdot B}\right) = O\left(\frac{N^3}{B\sqrt{M}}\right) \text{ I/Os}$$

Matrix Multiplication

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- Optimal

Hong & Kung, 1981

Matrix Multiplication

Algorithm 3: Recursive algorithm (cache-oblivious)

$$\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix} = \begin{pmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{pmatrix}$$

- 8 recursive $\frac{N}{2} \times \frac{N}{2}$ matrix multiplications + 4 $\frac{N}{2} \times \frac{N}{2}$ matrix sums

Matrix Multiplication

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- # I/Os if bit interleaved or (row major and $M = \Omega(B^2)$)

$$T(N) \leq \begin{cases} O\left(\frac{N^2}{B}\right) & \text{if } N \leq \varepsilon\sqrt{M} \\ 8 \cdot T\left(\frac{N}{2}\right) + O\left(\frac{N^2}{B}\right) & \text{otherwise} \end{cases}$$

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Matrix Multiplication

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$$T(N) \leq O\left(\frac{N^3}{B\sqrt{M}}\right)$$

- Optimal Hong & Kung, 1981
- Non-square matrices Frigo et al., 1999

Matrix Multiplication

Algorithm 4: Strassen's algorithm (cache-oblivious)

- 7 recursive $\frac{N}{2} \times \frac{N}{2}$ matrix multiplications + $O(1)$ matrix sums

$$\begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}$$

$$\begin{aligned} m_1 &:= (a_{21} + a_{22} - a_{11})(b_{22} - b_{12} + b_{11}) & c_{11} &:= m_2 + m_3 \\ m_2 &:= a_{11}b_{11} & c_{12} &:= m_1 + m_2 + m_5 + m_6 \\ m_3 &:= a_{12}b_{21} & c_{21} &:= m_1 + m_2 + m_4 - m_7 \\ m_4 &:= (a_{11} - a_{21})(b_{22} - b_{12}) & c_{22} &:= m_1 + m_2 + m_4 + m_5 \\ m_5 &:= (a_{21} + a_{22})(b_{12} - b_{11}) \\ m_6 &:= (a_{12} - a_{21} + a_{11} - a_{22})b_{22} \\ m_7 &:= a_{22}(b_{11} + b_{22} - b_{12} - b_{21}) \end{aligned}$$

Matrix Multiplication

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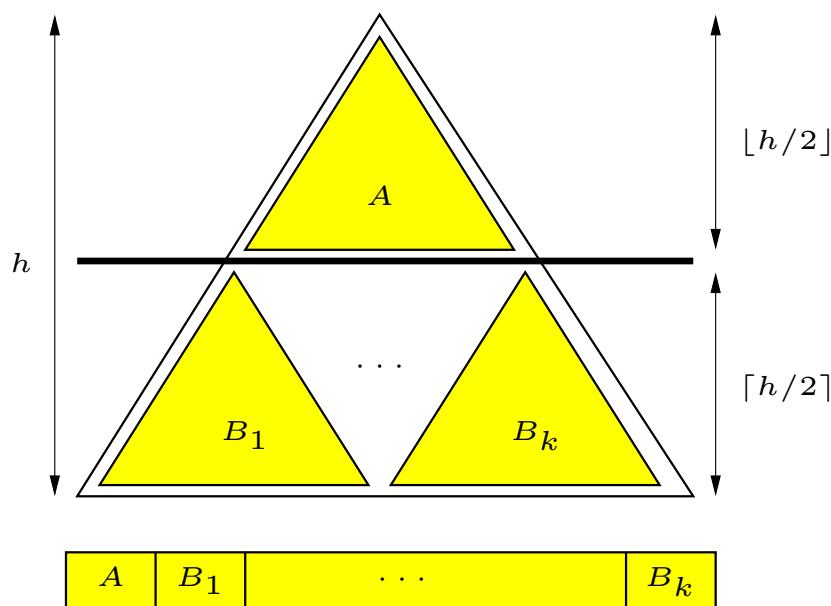
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$$T(N) \leq O\left(\frac{N^{\log_2 7}}{B\sqrt{M}}\right) \quad \log_2 7 \approx 2.81$$

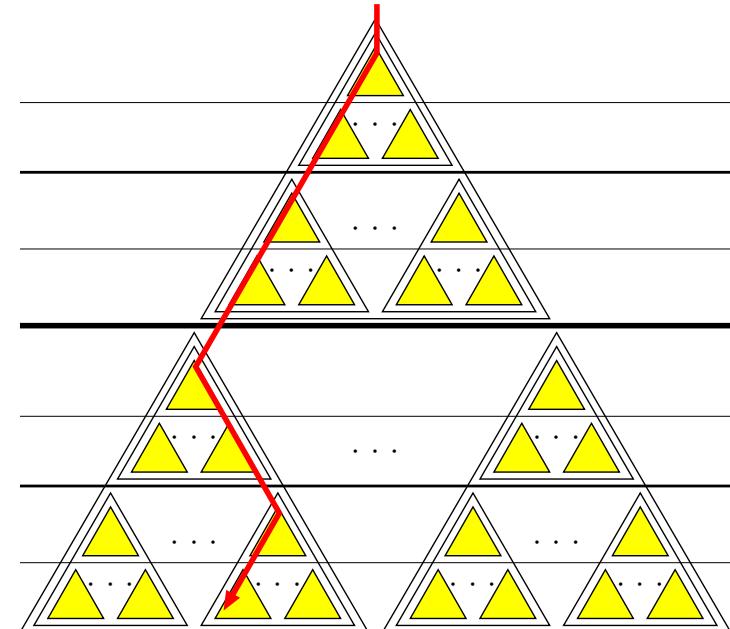
Cache-Oblivious Search Trees

Static Cache-Oblivious Trees

Recursive memory layout \equiv van Emde Boas layout



Degree $O(1)$

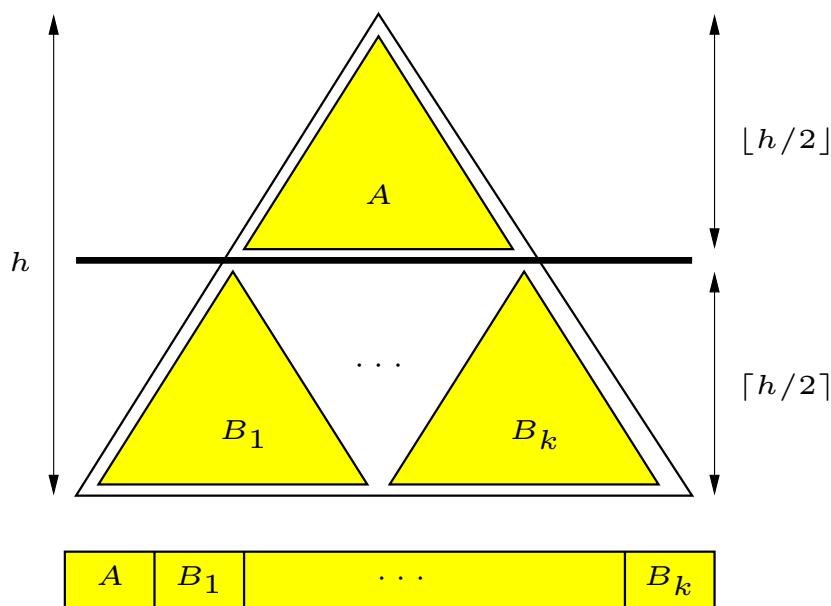


Searches use $O(\log_B N)$ I/Os

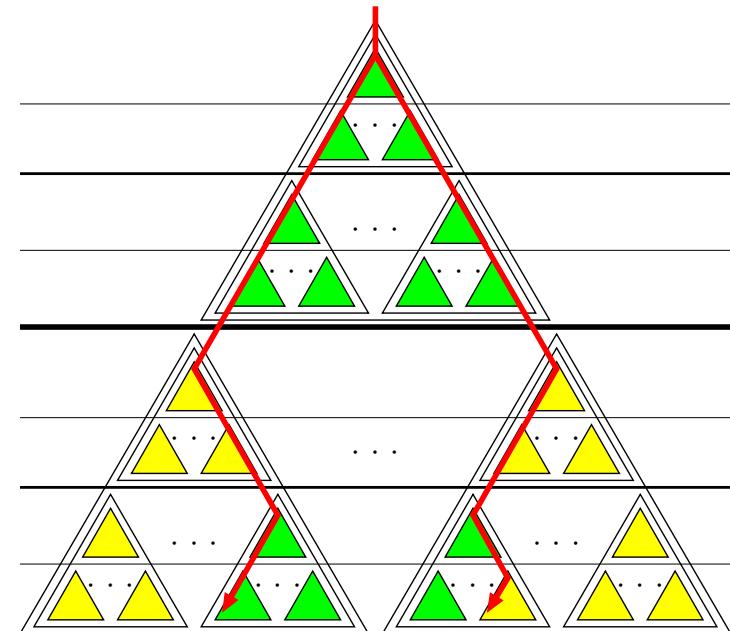
Prokop 1999

Static Cache-Oblivious Trees

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Degree $O(1)$



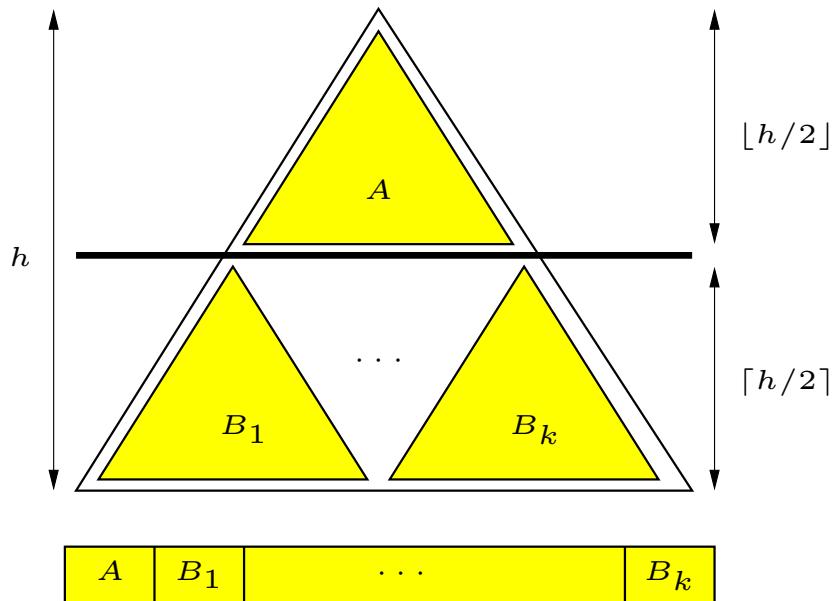
Searches use $O(\log_B N)$ I/Os

Range reportings use
 $O\left(\log_B N + \frac{k}{B}\right)$ I/Os

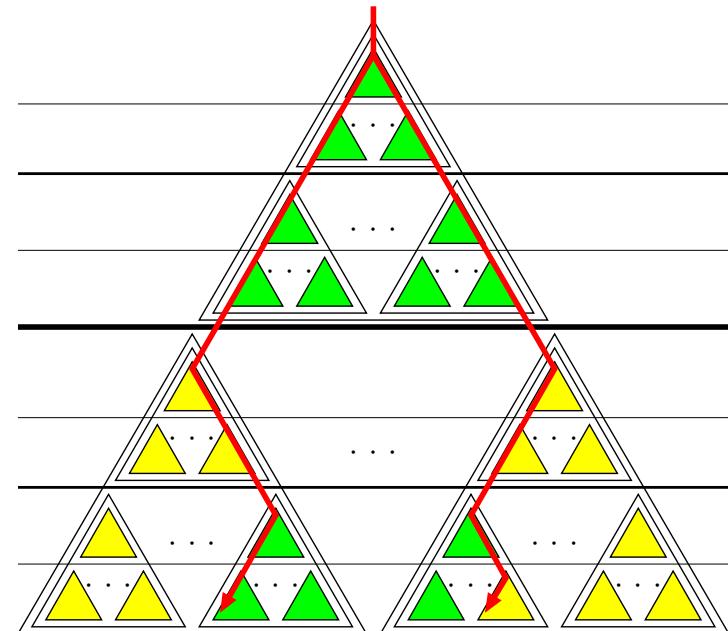
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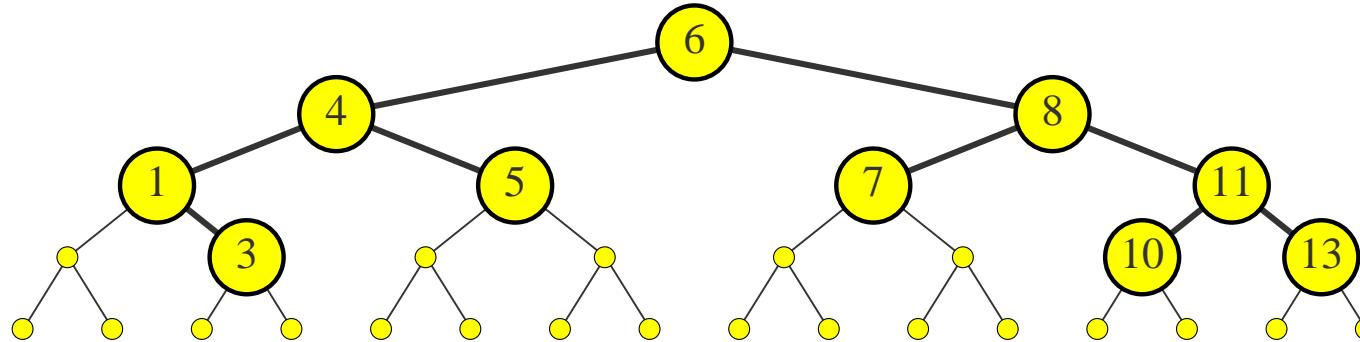
Prokop 1999

Best possible $(\log_2 e + o(1)) \log_B N$

Bender, Brodal, Fagerberg, Ge, He, Hu
Iacono, López-Ortiz 2003

Dynamic Cache-Oblivious Trees

- Embed a dynamic tree of small height into a complete tree
- Static van Emde Boas layout
- Rebuild data structure whenever N doubles or halves



Search

$O(\log_B N)$

Range Reporting

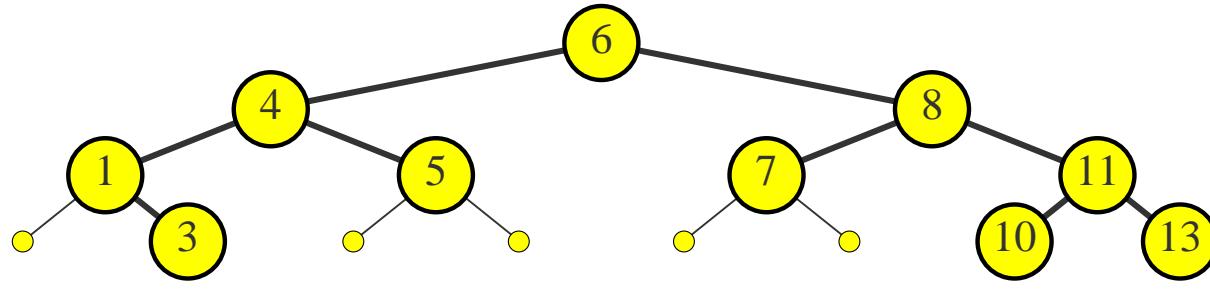
$O\left(\log_B N + \frac{k}{B}\right)$

Updates

$O\left(\log_B N + \frac{\log^2 N}{B}\right)$

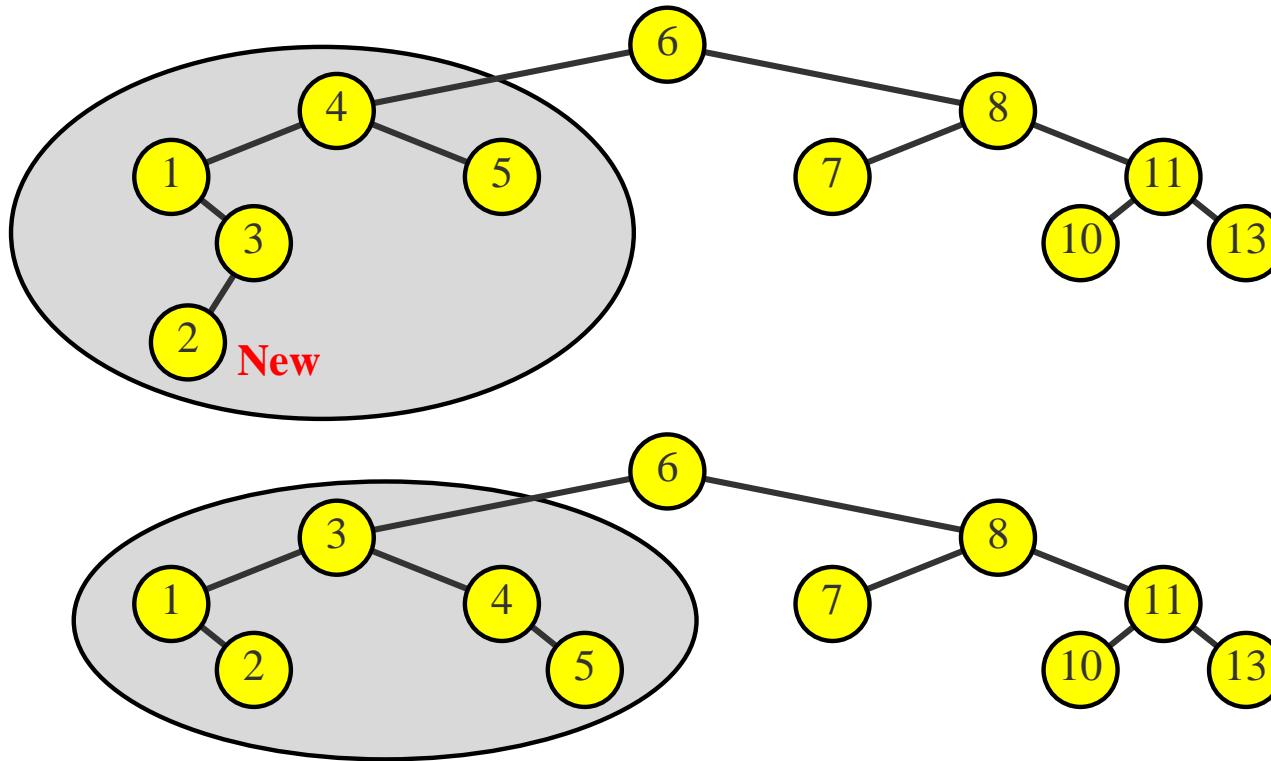
Brodal, Fagerberg, Jacob 2001

Example



6	4	8	1	-	3	5	-	-	7	-	-	11	10	13
---	---	---	---	---	---	---	---	---	---	---	---	----	----	----

Binary Trees of Small Height



- If an insertion causes non-small height then rebuild subtree at nearest ancestor with sufficient few descendants
- Insertions require amortized time $O(\log^2 N)$

Andersson and Lai 1990

Binary Trees of Small Height

- For each level i there is a threshold $\tau_i = \tau_L + i\Delta$, such that $0 < \tau_L = \tau_0 < \tau_1 < \dots < \tau_H = \tau_U < 1$
- For a node v_i on level i define **the density**

$$\rho(v_i) = \frac{\text{\# nodes below } v_i}{m_i}$$

where $m_i = \#$ possible nodes below v_i with depth at most H

Insertion

- Insert new element
- If depth $> H$ then locate nearest ancestor v_i with $\rho(v_i) \leq \tau_i$ and **rebuild subtree** at v_i to have minimum height and elements evenly distributed between left and right subtrees

Binary Trees of Small Height

Theorem Insertions require amortized time $O(\log^2 N)$

Proof Consider two redistributions of v_i

- After the first redistribution $\rho(v_i) \leq \tau_i$
- Before second redistribution a child v_{i+1} of v_i has $\rho(v_{i+1}) > \tau_{i+1}$
- Insertions below v_i : $m(v_{i+1}) \cdot (\tau_{i+1} - \tau_i) = m(v_{i+1}) \cdot \Delta$
- Redistribution of v_i costs $m(v_i)$, i.e. per insertion below v_i

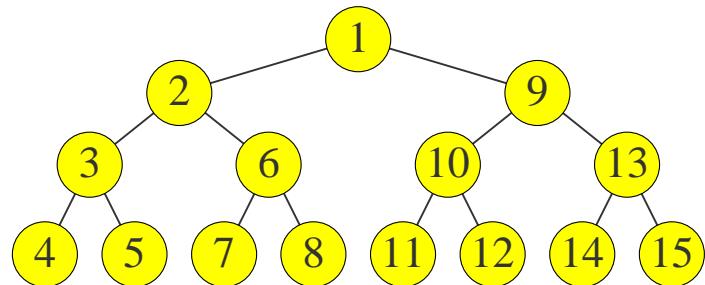
$$\frac{m(v_i)}{m(v_{i+1}) \cdot \Delta} \leq \frac{2}{\Delta}$$

- Total insertion cost per element

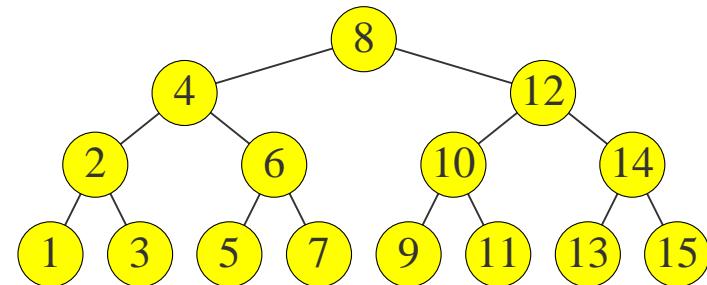
$$\sum_{i=0}^H \frac{2}{\Delta} = O(\log^2 N)$$

□

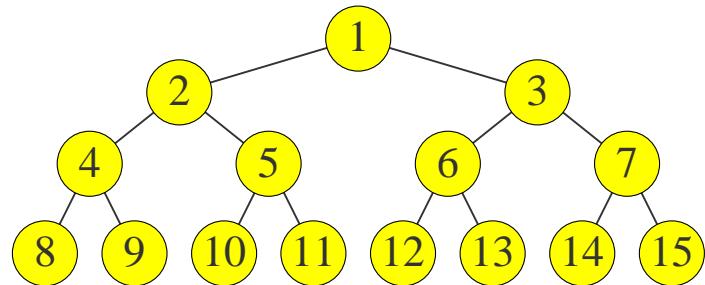
Memory Layouts of Trees



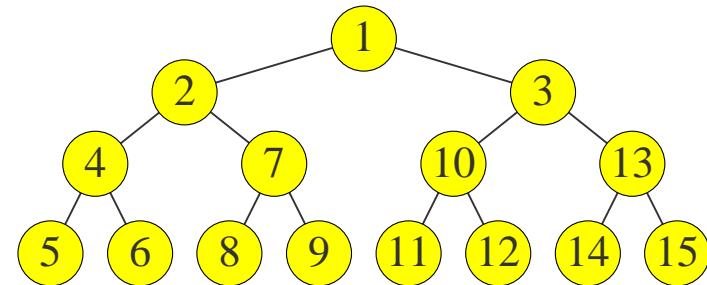
DFS



inorder

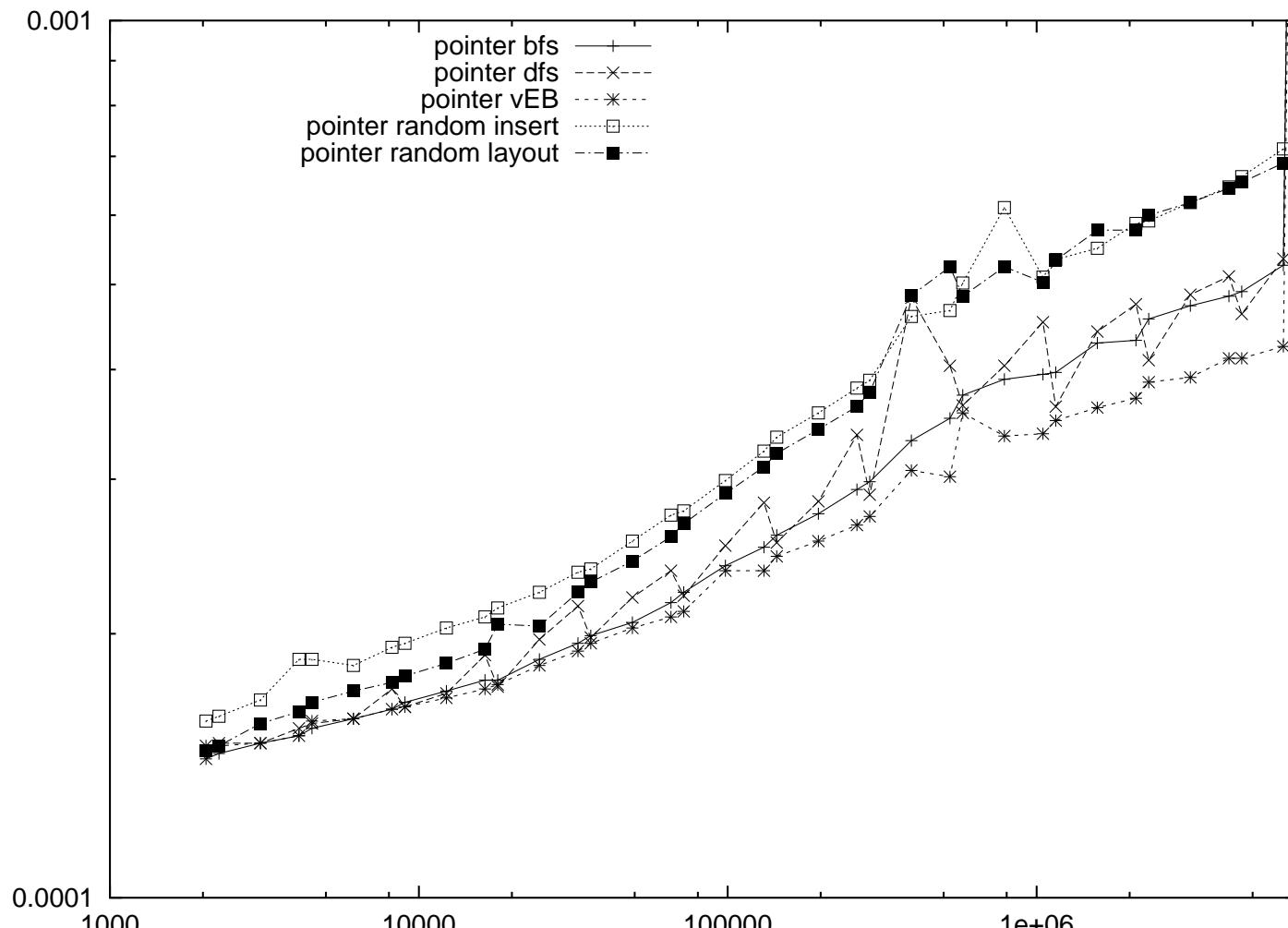


BFS



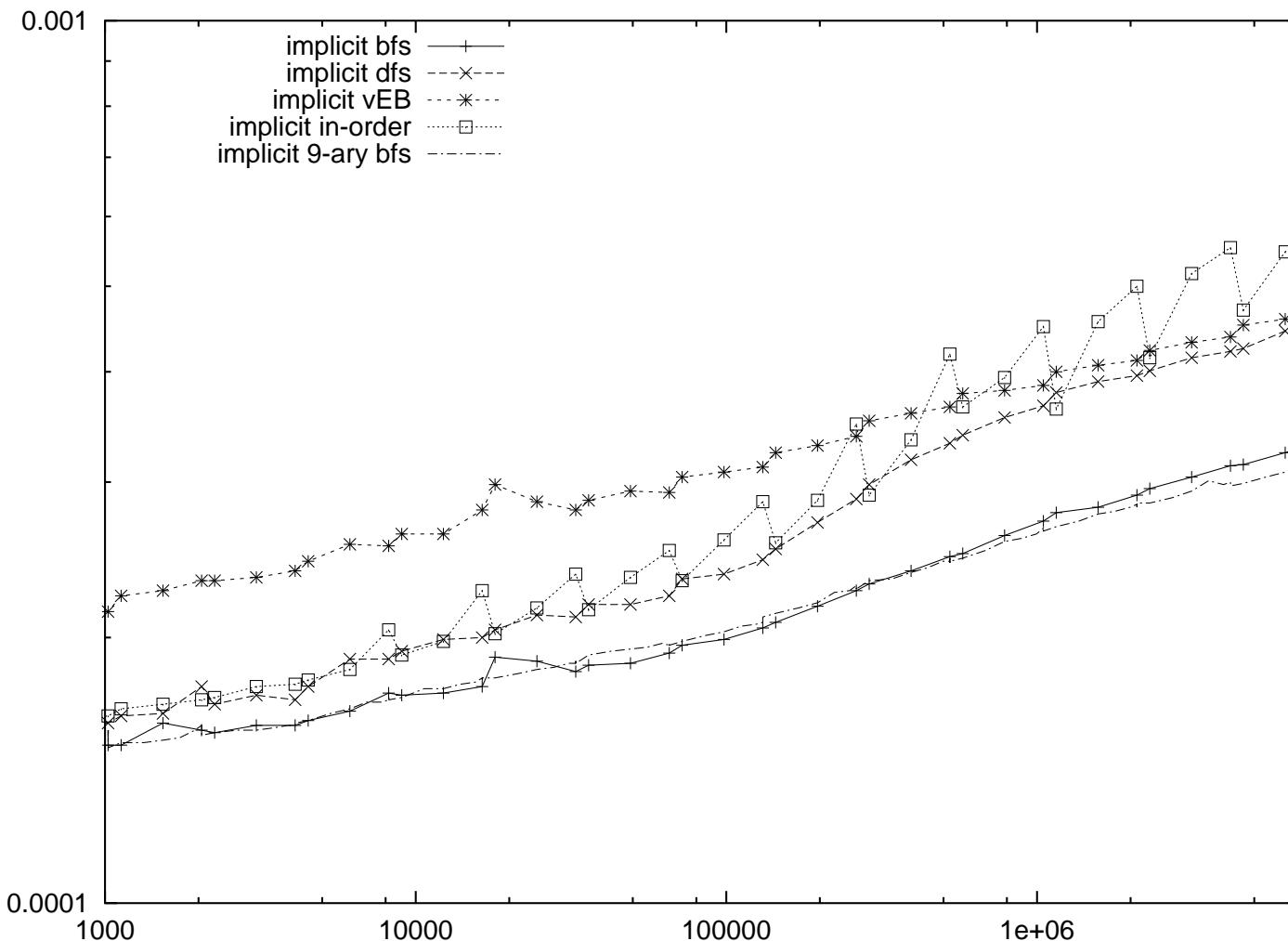
van Emde Boas
(in theory best)

Searches in Pointer Based Layouts



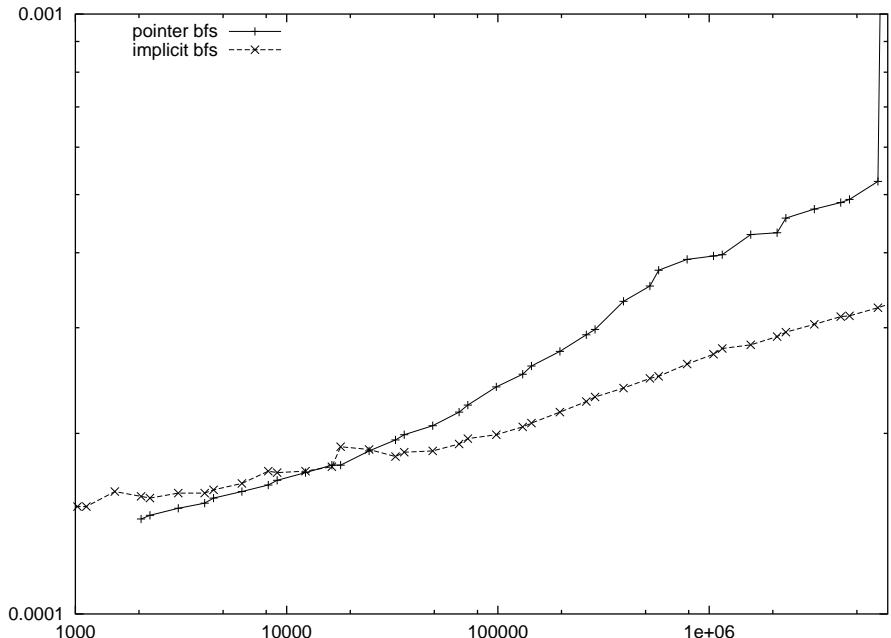
- van Emde Boas layout wins, followed by the BFS layout

Searches with Implicit Layouts

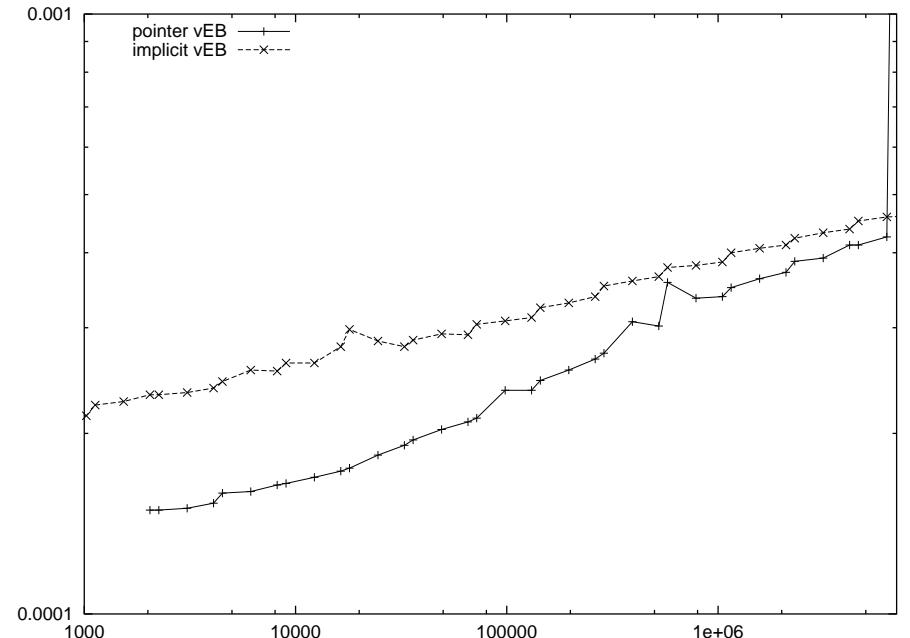


- BFS layout wins due to simplicity and caching of topmost levels
- van Emde Boas layout requires quite complex index computations

Implicit vs Pointer Based Layouts



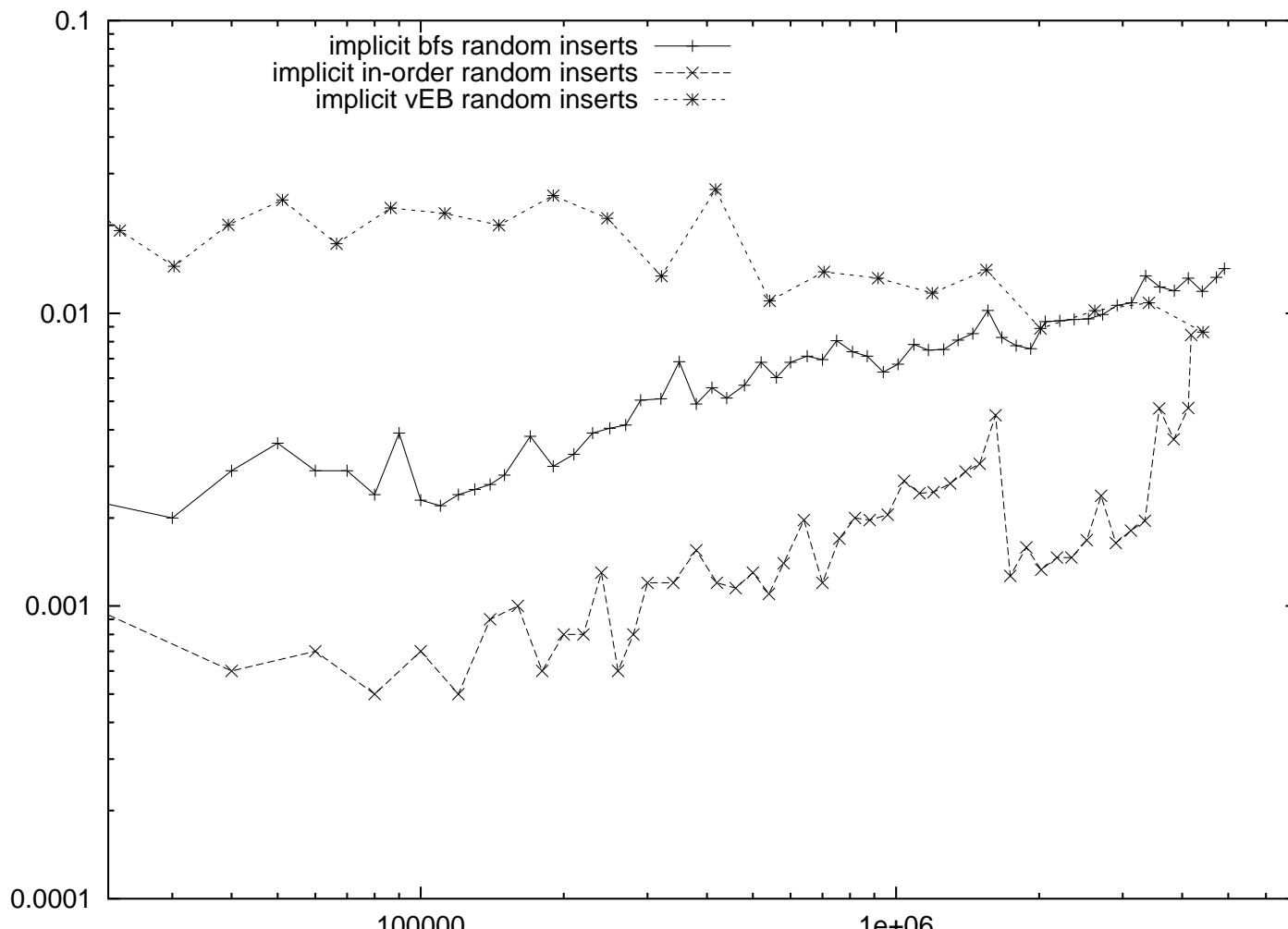
BFS layout



van Emde Boas layout

- Implicit layouts become competitive as n grows

Insertions in Implicit Layouts



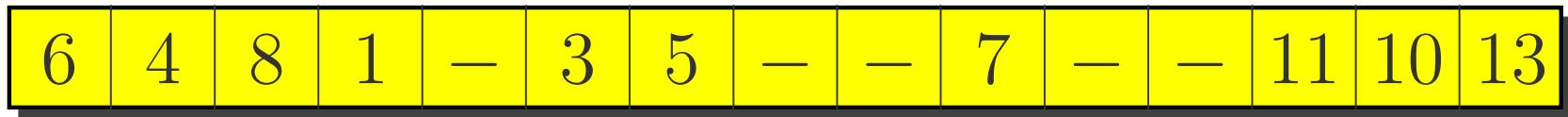
- Insertions are rather slow (factor 10-100 over searches)

Summary

- Dynamic cache-oblivious search trees

Search	$O(\log_B N)$
Range Reporting	$O\left(\log_B N + \frac{k}{B}\right)$
Updates	$O\left(\log_B N + \frac{\log^2 N}{B}\right)$

- Update time $O(\log_B N)$ by one level of indirection
(implies sub-optimal range reporting)
- Importance of memory layouts
- van Emde Boas layout gives good cache performance
- Computation time is important when considering caches



Cache-Oblivious Sorting

Sorting Problem

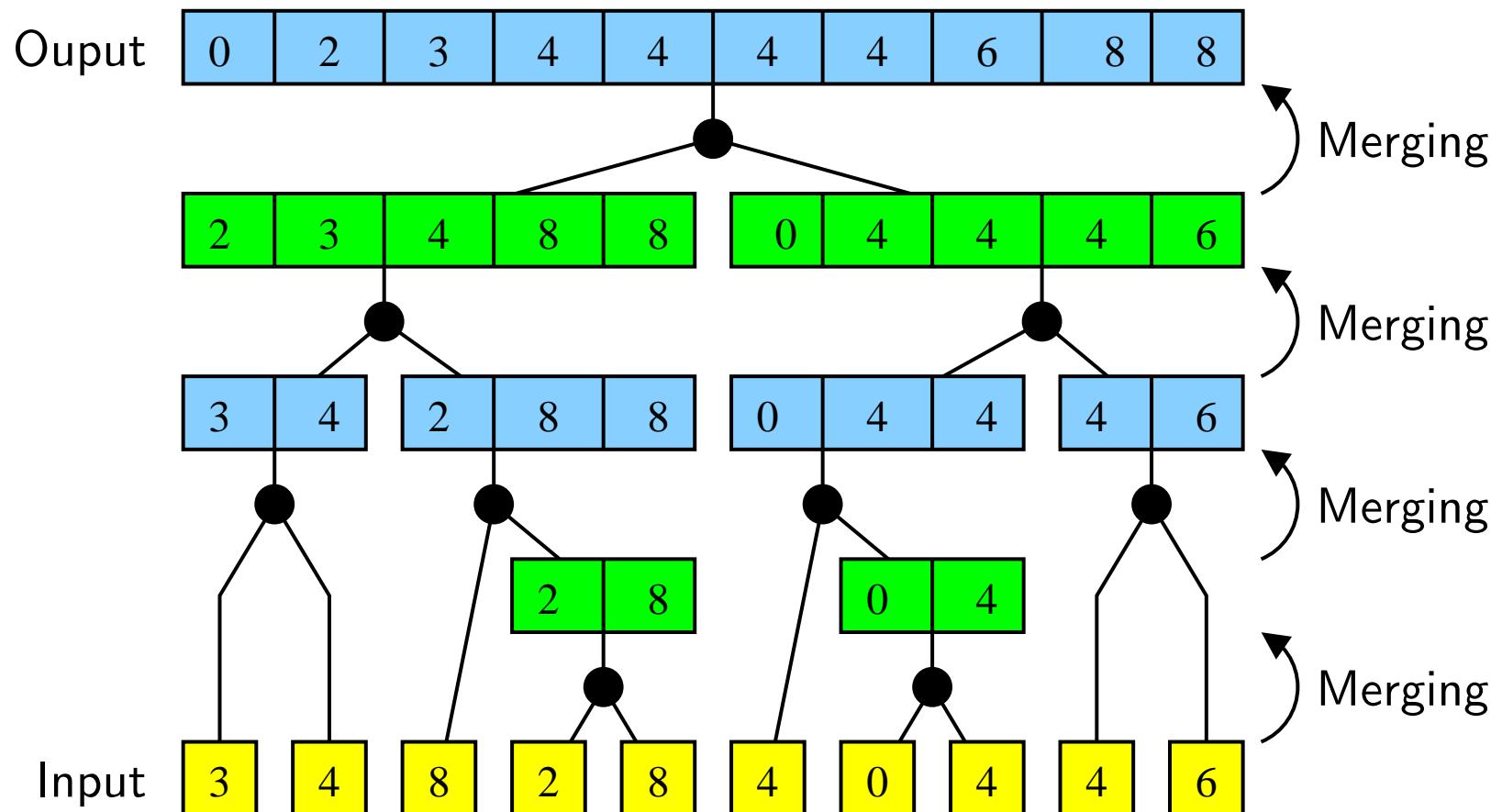
- Input : array containing x_1, \dots, x_N
- Output : array with x_1, \dots, x_N in sorted order
- Elements can be compared and copied

3	4	8	2	8	4	0	4	4	6
---	---	---	---	---	---	---	---	---	---

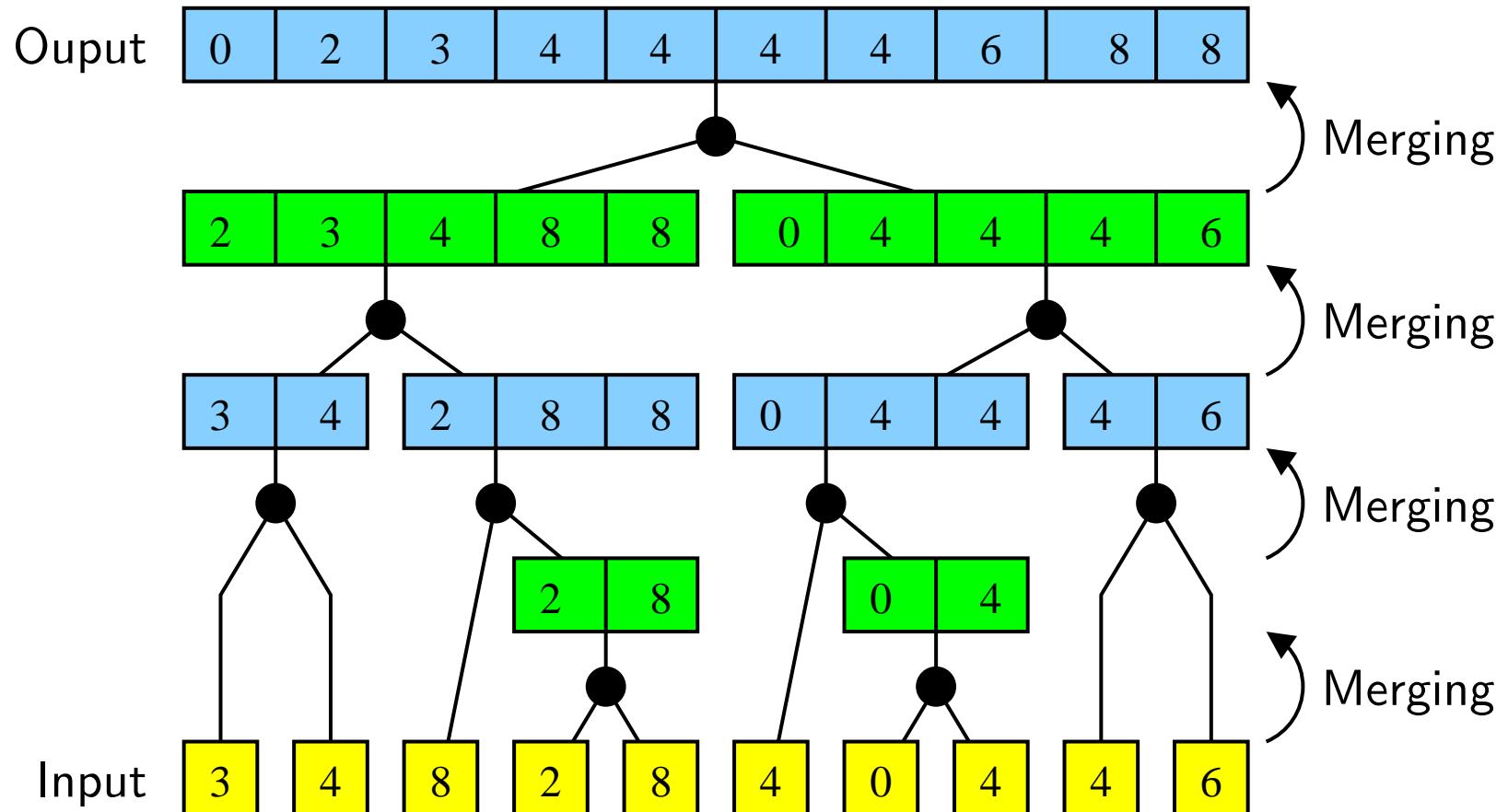


0	2	3	4	4	4	4	6	8	8
---	---	---	---	---	---	---	---	---	---

Binary Merge-Sort



Binary Merge-Sort



- Recursive; two arrays; size $O(M)$ internally in cache
- $O(N \log N)$ comparisons
- $O\left(\frac{N}{B} \log_2 \frac{N}{M}\right)$ I/Os

Merge-Sort

Degree	I/O
2	$O\left(\frac{N}{B} \log_2 \frac{N}{M}\right)$
d $(d \leq \frac{M}{B} - 1)$	$O\left(\frac{N}{B} \log_d \frac{N}{M}\right)$
$\Theta\left(\frac{M}{B}\right)$	$O\left(\frac{N}{B} \log_{M/B} \frac{N}{M}\right) = O(\text{Sort}_{M,B}(N))$

Aggarwal and Vitter 1988

Funnel-Sort

2 $(M \geq B^{1+\varepsilon})$	$O\left(\frac{1}{\varepsilon} \text{Sort}_{M,B}(N)\right)$
	Frigo, Leiserson, Prokop and Ramachandran 1999
	Brodal and Fagerberg 2002

Lower Bound

Brodal and Fagerberg 2003

	Block Size	Memory	I/Os
Machine 1	B_1	M	t_1
Machine 2	B_2	M	t_2

One algorithm, two machines, $B_1 \leq B_2$

Trade-off

$$8t_1B_1 + 3t_1B_1 \log \frac{8Mt_2}{t_1B_1} \geq N \log \frac{N}{M} - 1.45N$$

Lower Bound

	Assumption	I/Os
Lazy Funnel-sort	$B \leq M^{1-\varepsilon}$	(a) $B_2 = M^{1-\varepsilon}$: Sort _{B_2, M} (N) (b) $B_1 = 1$: Sort _{B_1, M} (N) · $\frac{1}{\varepsilon}$
Binary Merge-sort	$B \leq M/2$	(a) $B_2 = M/2$: Sort _{B_2, M} (N) (b) $B_1 = 1$: Sort _{B_1, M} (N) · log M

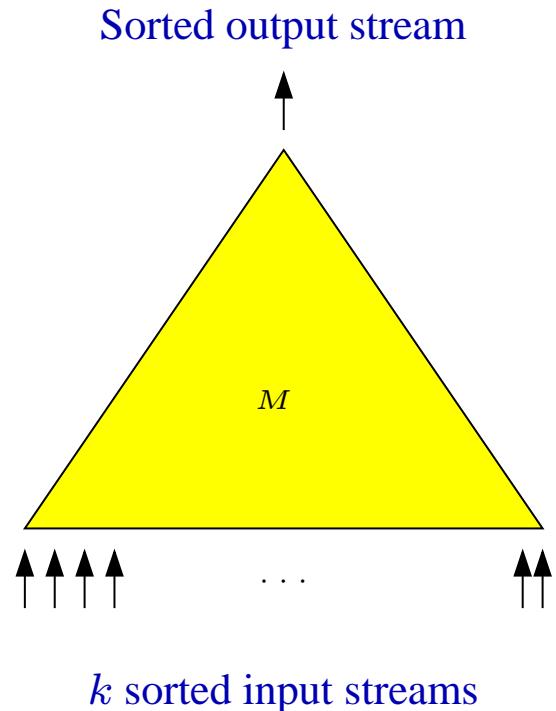
Corollary (a) \Rightarrow (b)

Funnel-Sort



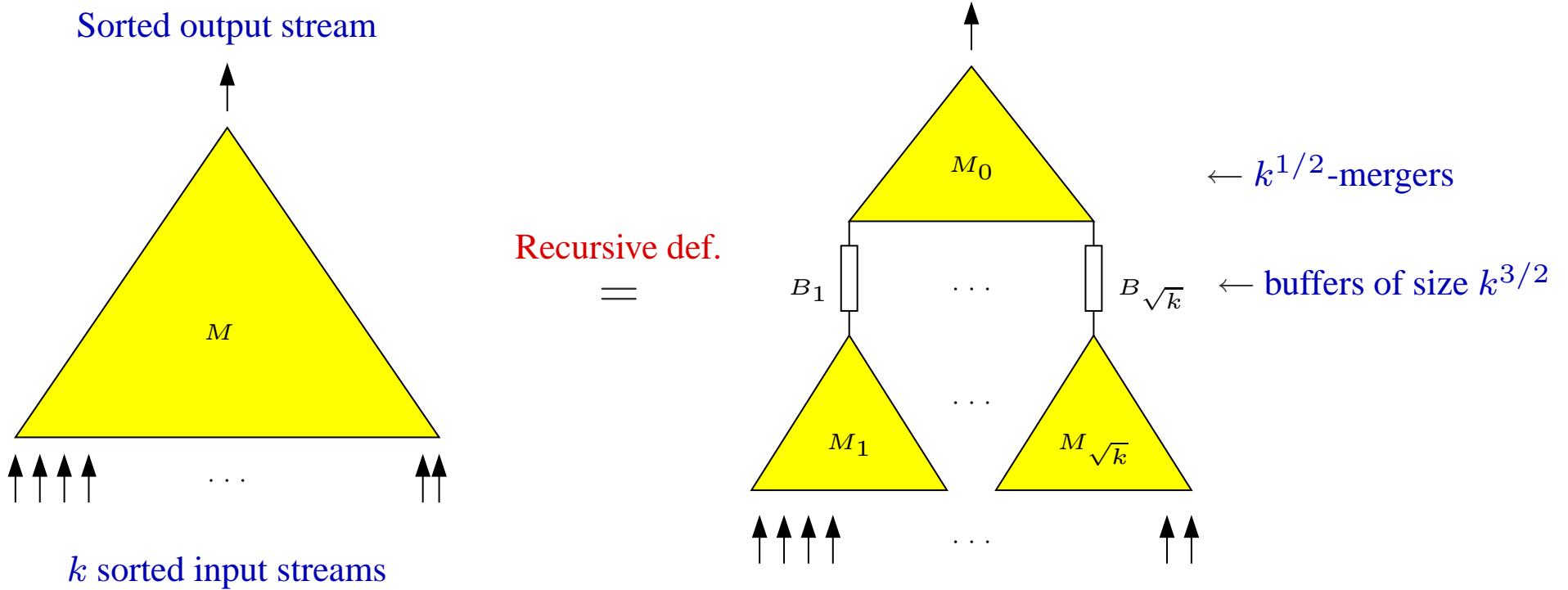
k -merger

Frigo et al., FOCS'99



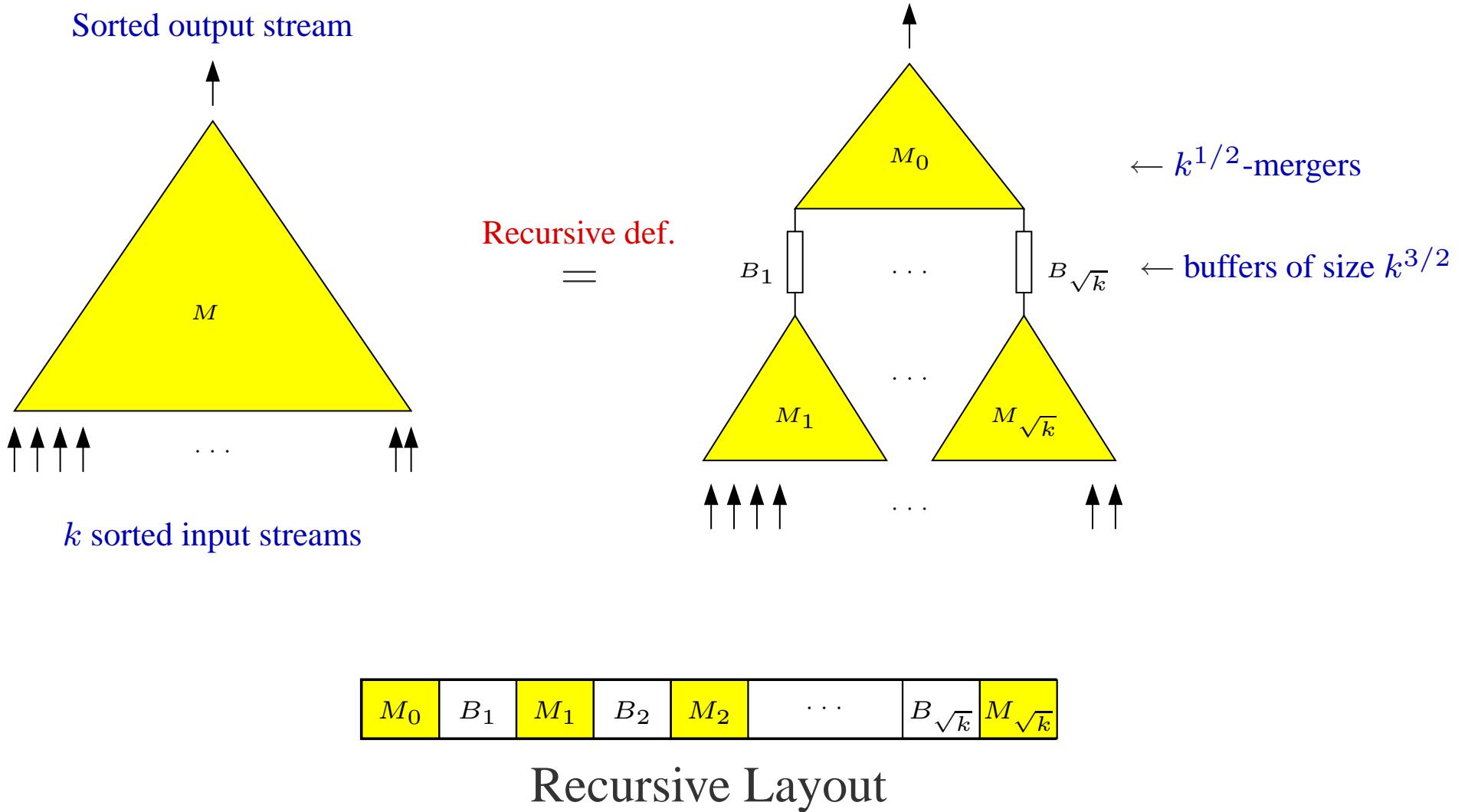
k -merger

Frigo et al., FOCS'99



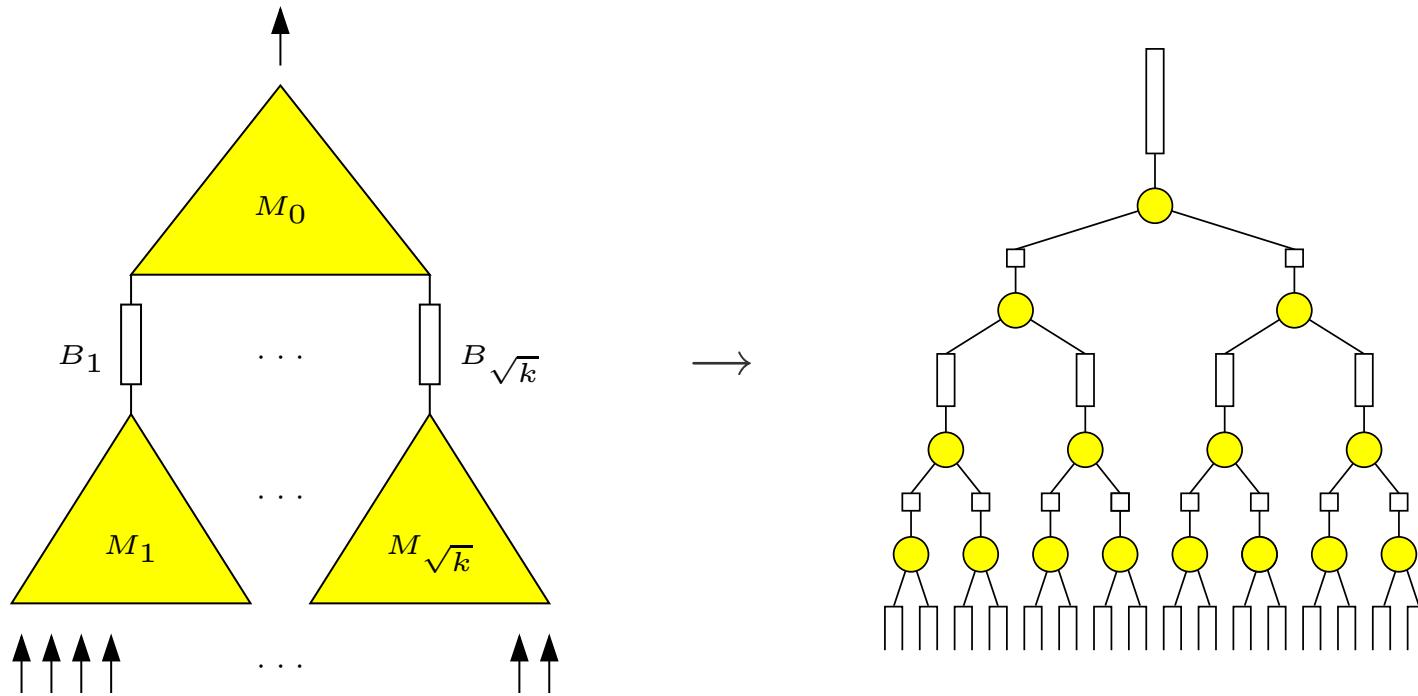
k -merger

Frigo et al., FOCS'99



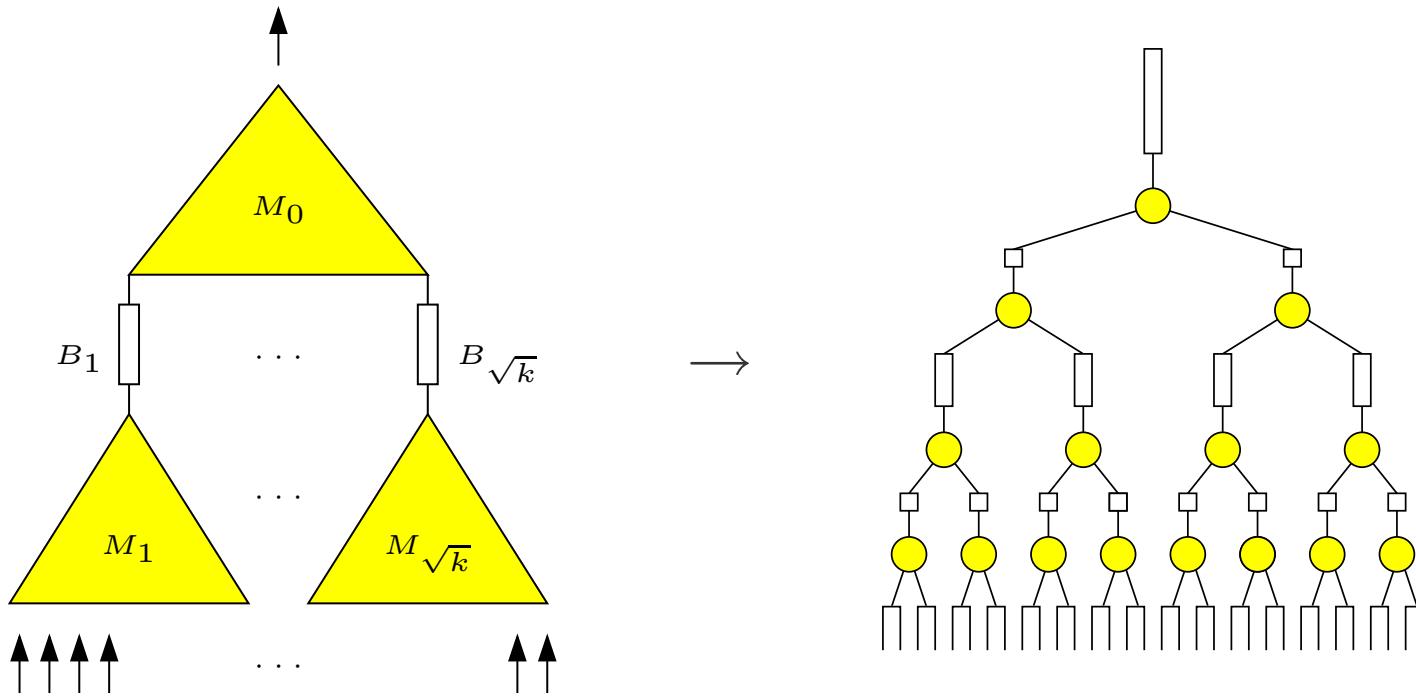
Lazy k -merger

Brodal and Fagerberg 2002



Lazy k -merger

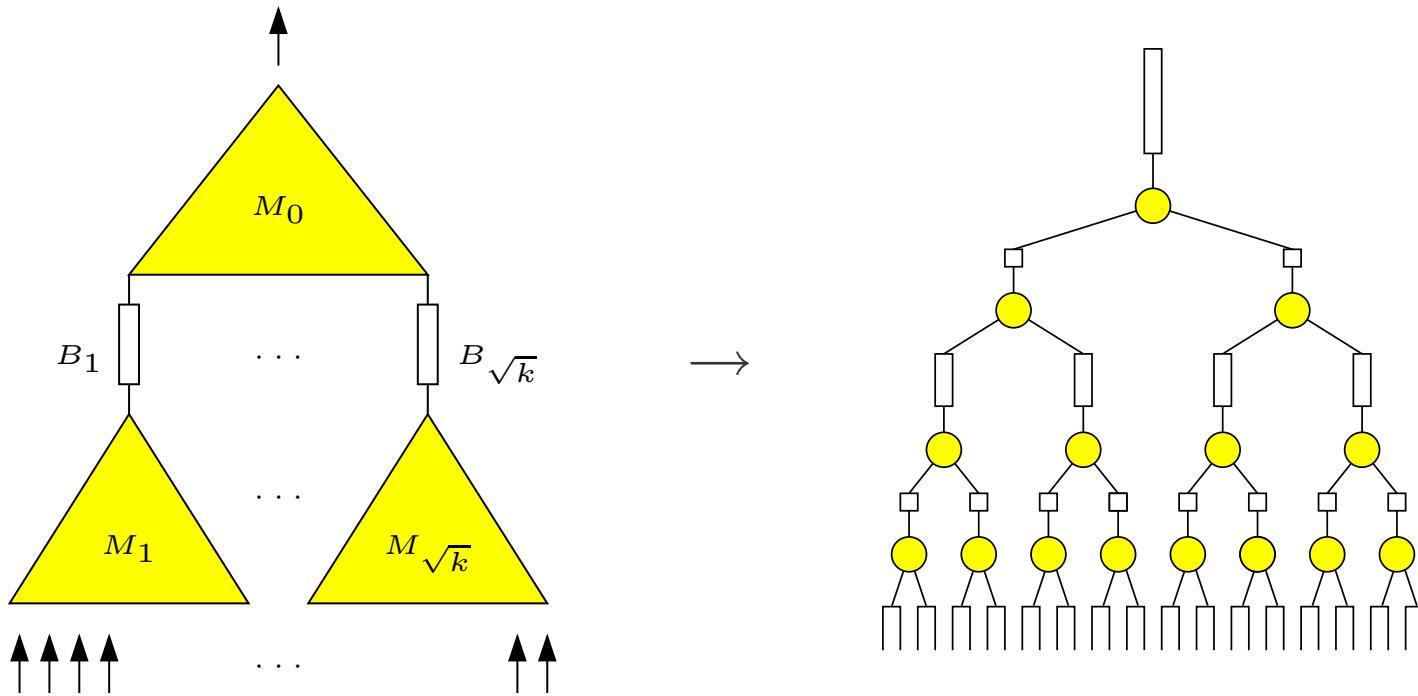
Brodal and Fagerberg 2002



```
Procedure Fill( $v$ )
    while out-buffer not full
        if left in-buffer empty
            Fill(left child)
        if right in-buffer empty
            Fill(right child)
        perform one merge step
```

Lazy k -merger

Brodal and Fagerberg 2002



```

Procedure Fill( $v$ )
  while out-buffer not full
    if left in-buffer empty
      Fill(left child)
    if right in-buffer empty
      Fill(right child)
    perform one merge step
  
```

Lemma

If $M \geq B^2$ and output buffer has size k^3 then $O(\frac{k^3}{B} \log_M(k^3) + k)$ I/Os are done during an invocation of **Fill**(root)

Funnel-Sort

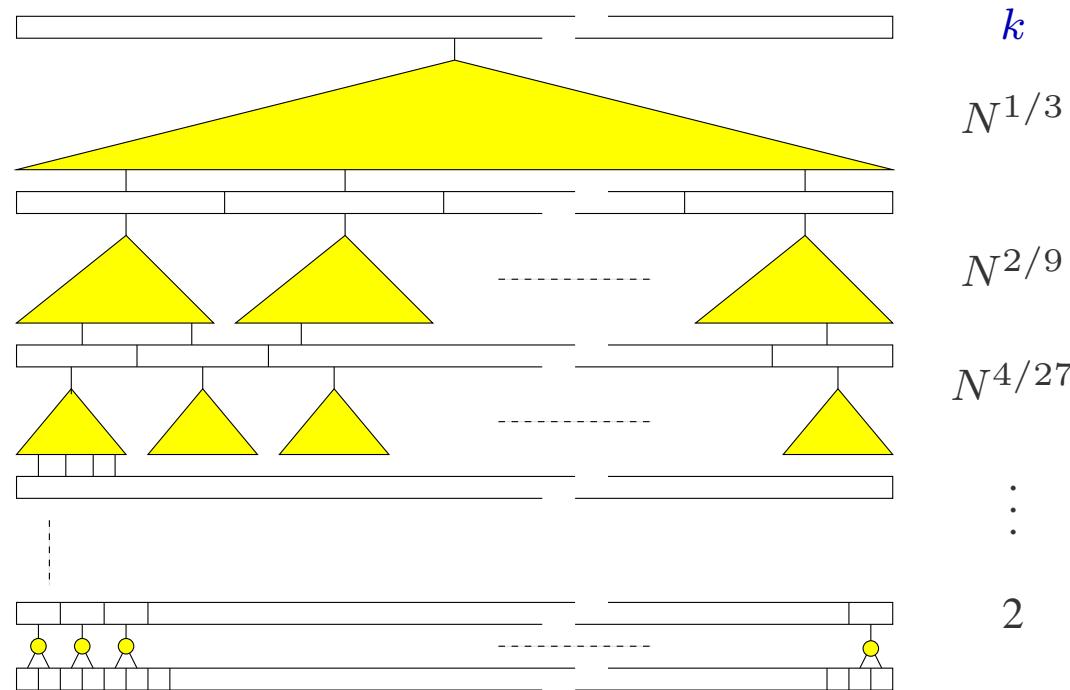
Brodal and Fagerberg 2002

Frigo, Leiserson, Prokop and Ramachandran 1999

Divide input in $N^{1/3}$ segments of size $N^{2/3}$

Recursively **Funnel-Sort** each segment

Merge sorted segments by an $N^{1/3}$ -merger



Funnel-Sort

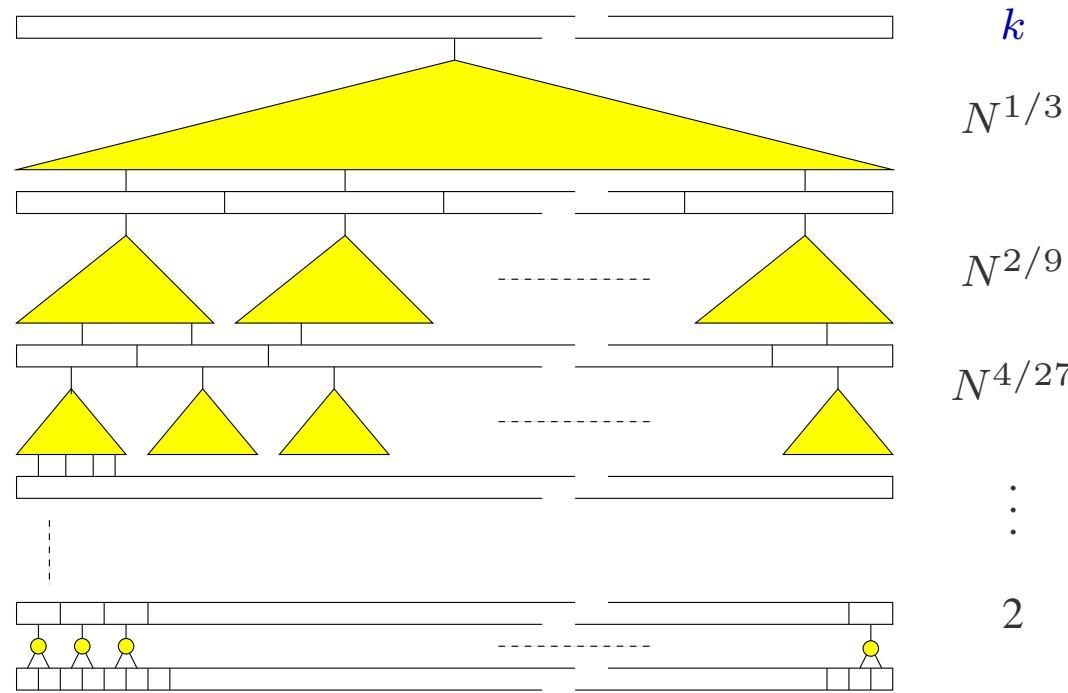
Brodal and Fagerberg 2002

Frigo, Leiserson, Prokop and Ramachandran 1999

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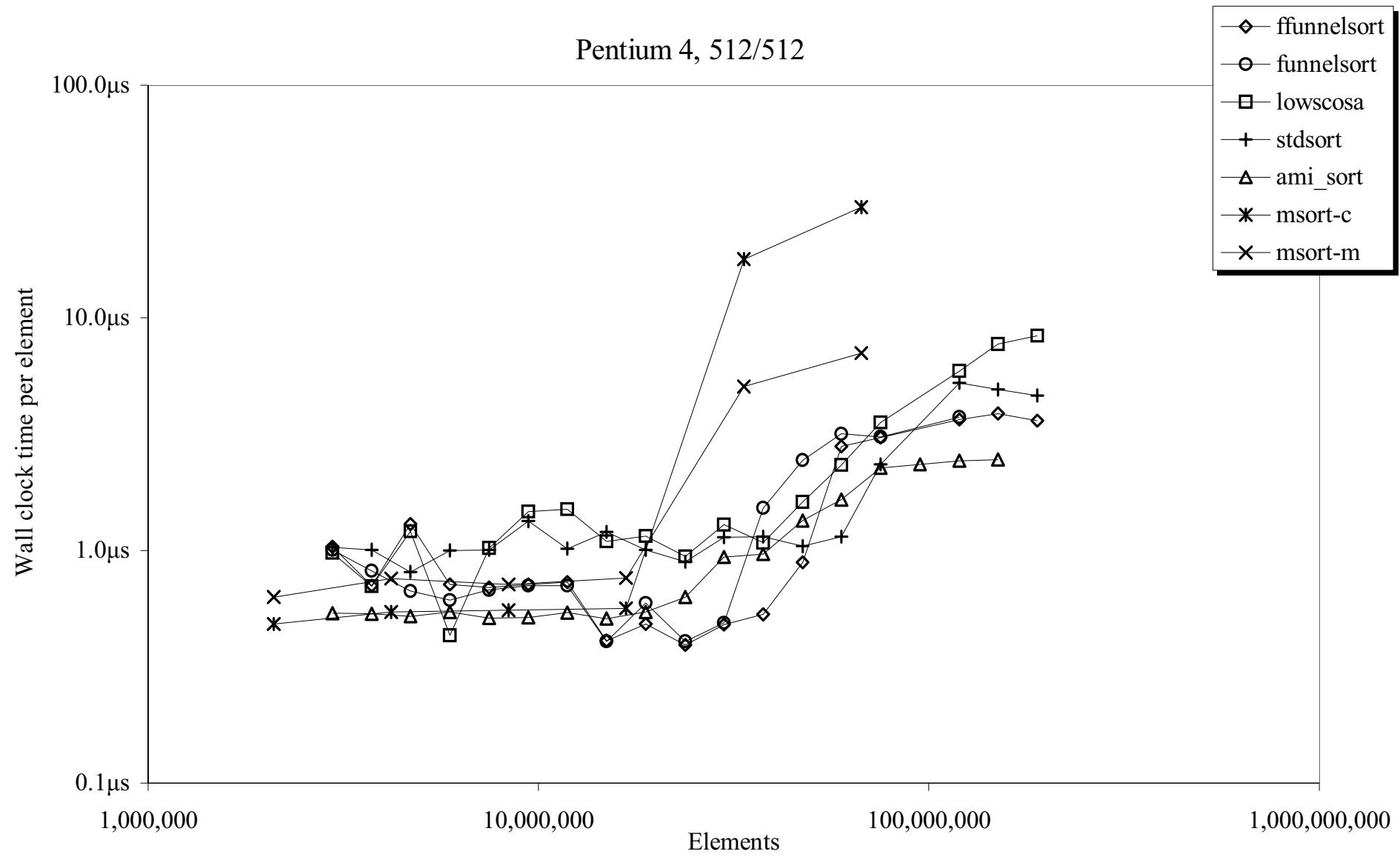


Theorem Funnel-Sort performs $O(\text{Sort}_{M,B}(N))$ I/Os for $M \geq B^2$

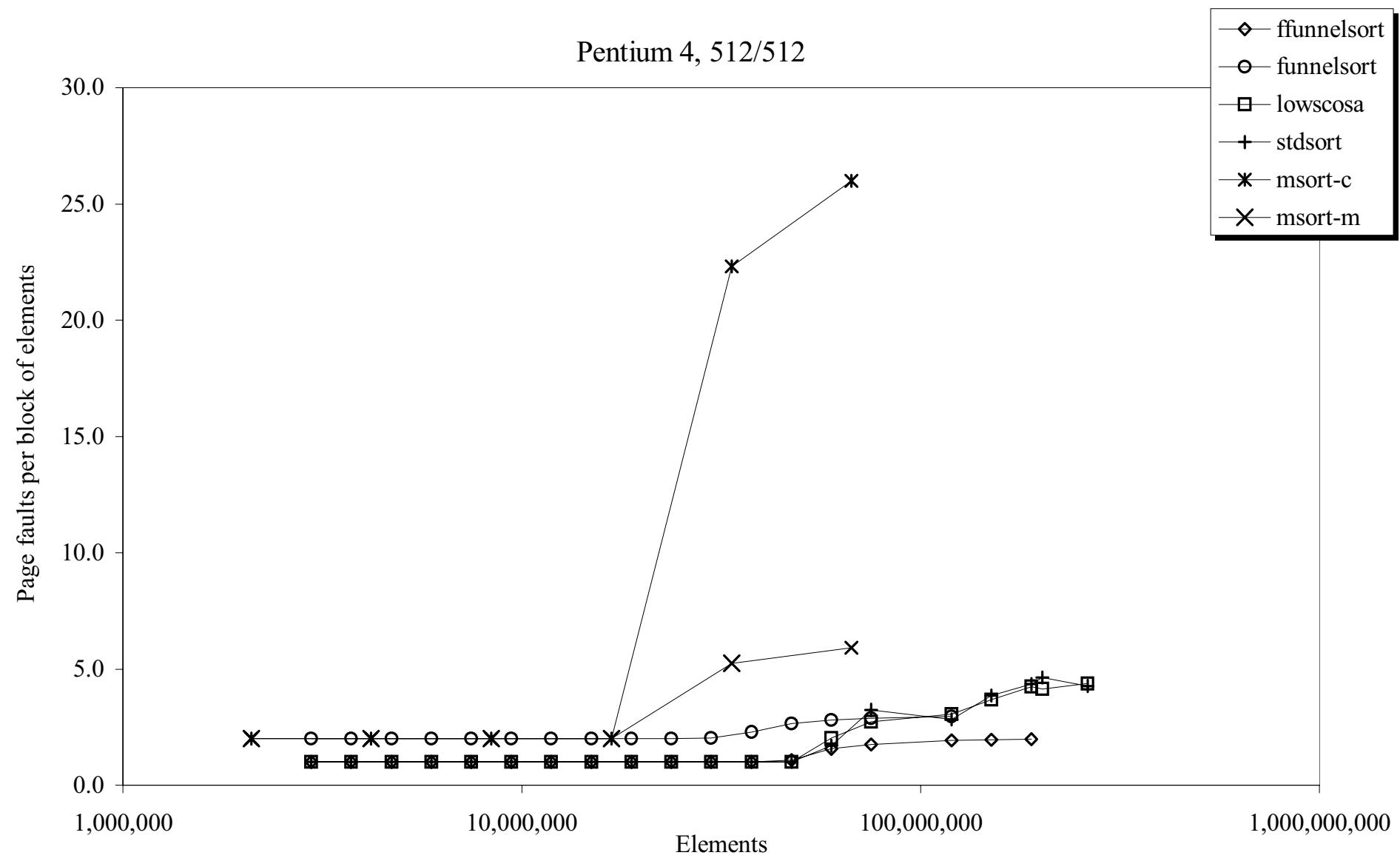
Hardware

Processor type	Pentium 4	Pentium 3	MIPS 10000
Workstation	Dell PC	Delta PC	SGI Octane
Operating system	GNU/Linux Kernel version 2.4.18	GNU/Linux Kernel version 2.4.18	IRIX version 6.5
Clock rate	2400 MHz	800 MHz	175 MHz
Address space	32 bit	32 bit	64 bit
Integer pipeline stages	20	12	6
L1 data cache size	8 KB	16 KB	32 KB
L1 line size	128 Bytes	32 Bytes	32 Bytes
L1 associativity	4 way	4 way	2 way
L2 cache size	512 KB	256 KB	1024 KB
L2 line size	128 Bytes	32 Bytes	32 Bytes
L2 associativity	8 way	4 way	2 way
TLB entries	128	64	64
TLB associativity	Full	4 way	64 way
TLB miss handler	Hardware	Hardware	Software
Main memory	512 MB	256 MB	128 MB

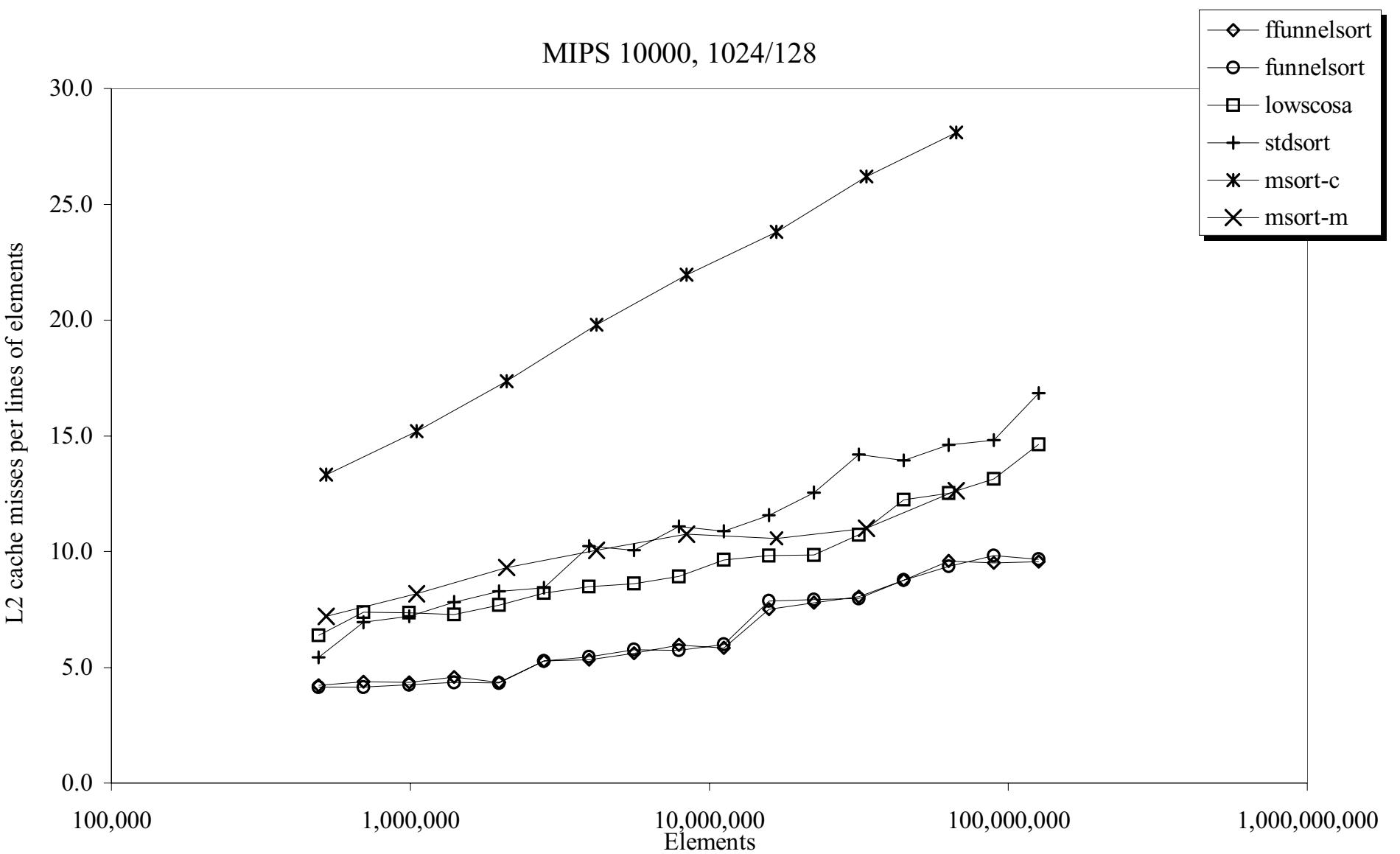
Wall Clock



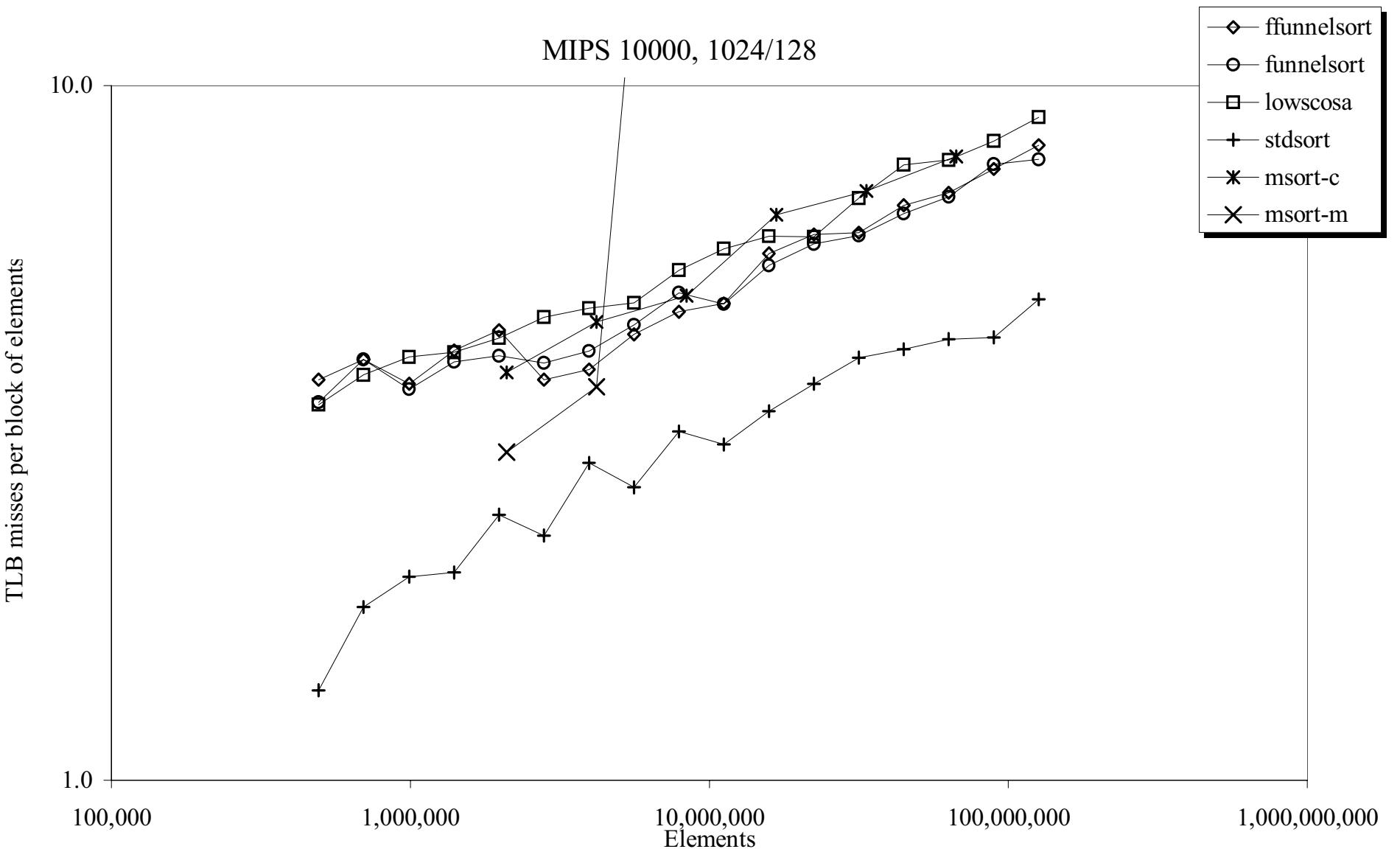
Page Faults



Cache Misses



TLB Misses



Conclusions

Cache oblivious sorting

- is possible
- requires a tall cache assumption $M \geq B^{1+\varepsilon}$
- comparable performance with cache aware algorithms

Future work

- more experimental justification for the cache oblivious model
- limitations of the model — time space trade-offs ?
- tool-box for cache oblivious algorithms

