

6.172  
Performance  
Engineering  
of Software  
Systems

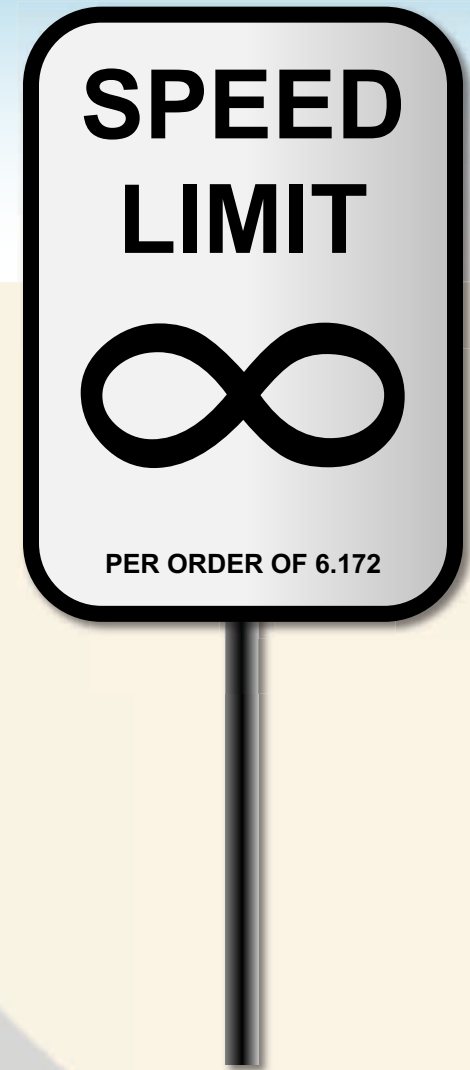


LECTURE 15  
Cache-Oblivious  
Algorithms

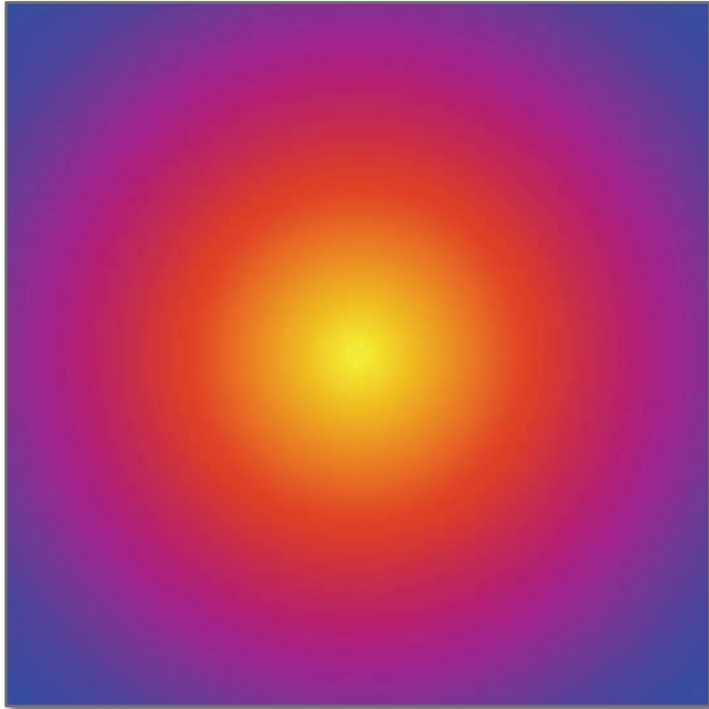
Julian Shun



# SIMULATION OF HEAT DIFFUSION



# Heat Diffusion



## 2D heat equation

Let  $u(t, x, y)$  = temperature at time  $t$  of point  $(x, y)$ .

$$\frac{\partial u}{\partial t} = \alpha \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$$

$\alpha$  is the *thermal diffusivity*.

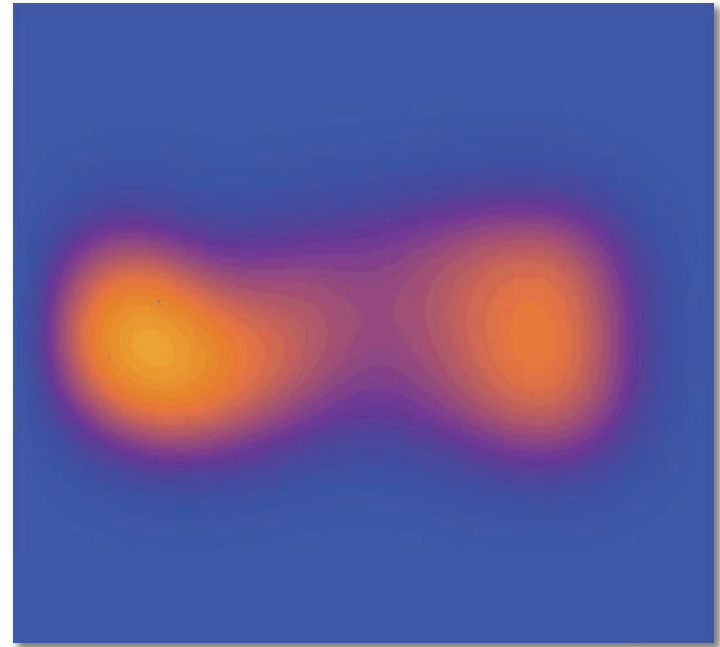
## Acknowledgment

Some of the slides in this presentation were inspired by originals due to Matteo Frigo.

# 2D Heat-Diffusion Simulation



Before



After

# 1 D Heat Equation

$$\frac{\partial u}{\partial t} = \alpha \frac{\partial^2 u}{\partial x^2}$$



# Finite-Difference Approximation

$$\frac{\partial u}{\partial t} = \alpha \frac{\partial^2 u}{\partial x^2}$$

$$\frac{\partial}{\partial t} u(t, x) \approx \frac{u(t + \Delta t, x) - u(t, x)}{\Delta t},$$

$$\frac{\partial}{\partial x} u(t, x) \approx \frac{u(t, x + \Delta x/2) - u(t, x - \Delta x/2)}{\Delta x},$$

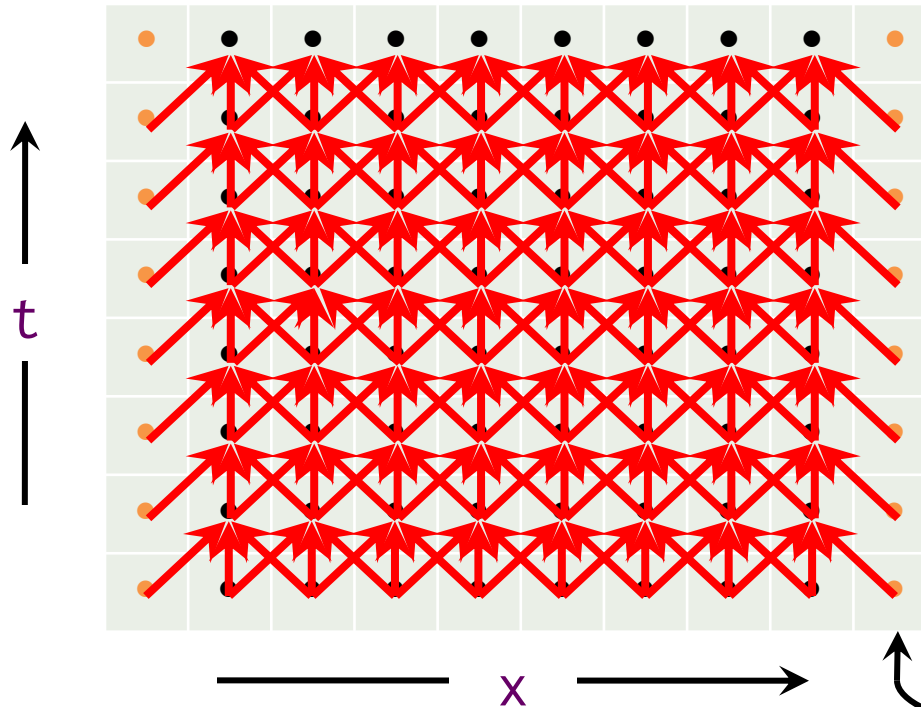
$$\frac{\partial^2}{\partial x^2} u(t, x) \approx \frac{\frac{\partial}{\partial x} u(t, x + \Delta x/2) - \frac{\partial}{\partial x} u(t, x - \Delta x/2)}{\Delta x}$$

The 1D heat equation thus reduces to

$$\frac{u(t + \Delta t, x) - u(t, x)}{\Delta t} = \alpha \left( \frac{u(t, x + \Delta x) - 2u(t, x) + u(t, x - \Delta x)}{(\Delta x)^2} \right).$$

# 3-Point Stencil

$$\frac{u(t + \Delta t, x) - u(t, x)}{\Delta t} = \alpha \left( \frac{u(t, x + \Delta x) - 2u(t, x) + u(t, x - \Delta x)}{(\Delta x)^2} \right)$$



A **stencil computation** updates each point in an array by a fixed pattern, called a **stencil**.

iteration  
space

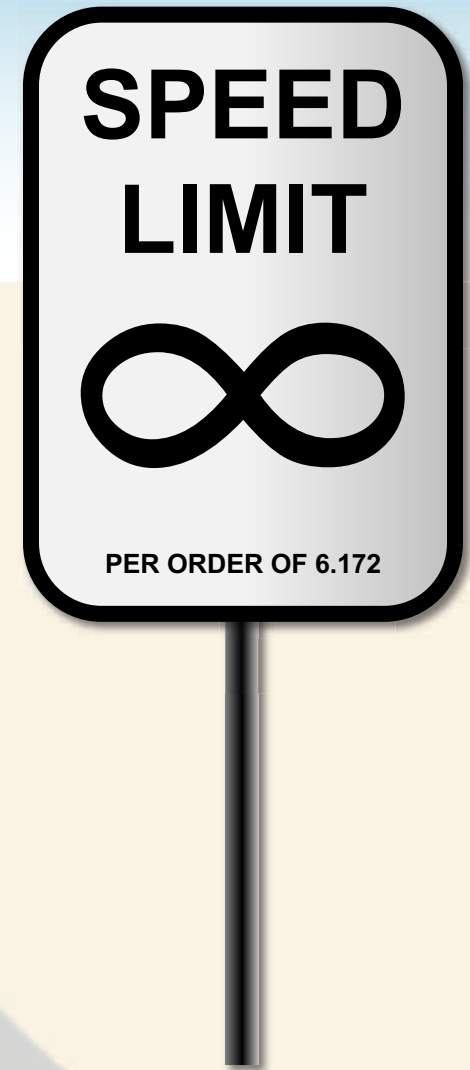
boundary

Update rule

$$\Delta t = 1, \Delta x = 1$$

```
u[t+1][x] = u[t][x] + ALPHA  
            * (u[t][x+1] - 2*u[t][x] + u[t][x-1]);
```

# CACHE-OBLIVIOUS STENCIL COMPUTATIONS

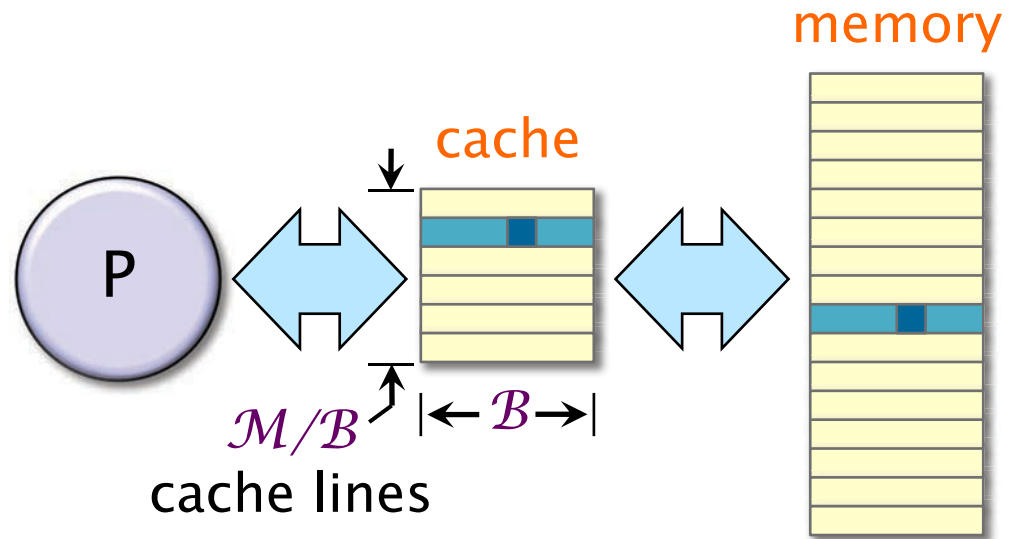




# Recall: Ideal-Cache Model

## Parameters

- Two-level hierarchy.
- Cache size of  $\mathcal{M}$  bytes.
- Cache-line length (block size) of  $\mathcal{B}$  bytes.
- Fully associative.
- Optimal omniscient replacement, or LRU.



## Performance Measures

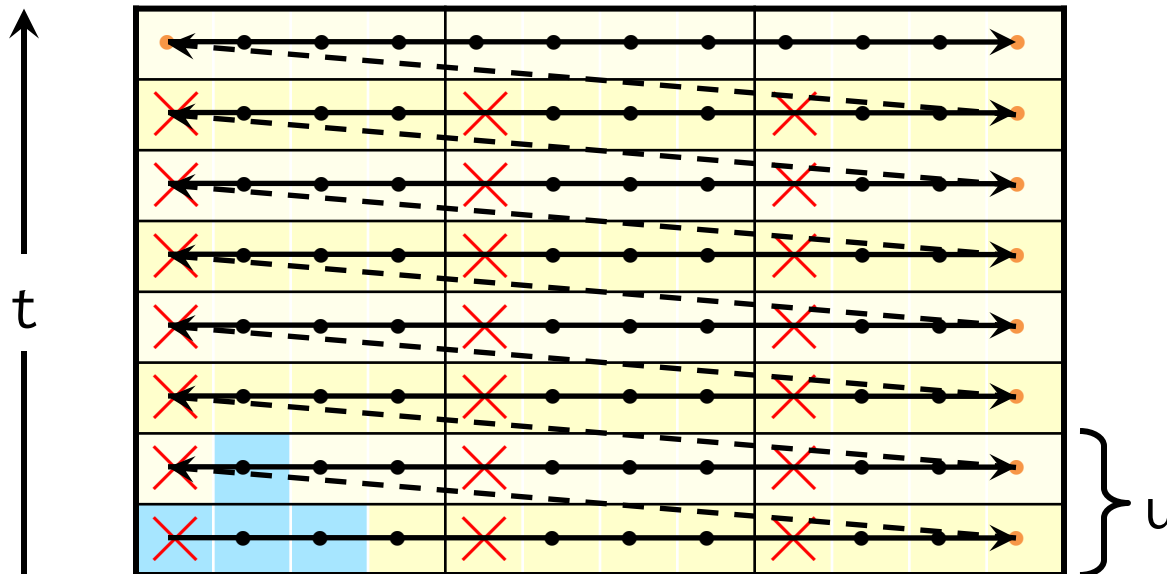
- **work**  $\mathcal{W}$  (ordinary running time)
- **cache misses**  $\mathcal{Q}$

# Cache Behavior of Looping

```
double u[2][N]; // even-odd trick

static inline double kernel(double * w) {
    return w[0] + ALPHA * (w[-1] - 2*w[0] + w[1]);
}

for (size_t t = 1; t < T-1; ++t) { // time loop
    for(size_t x = 1; x < N-1; ++x) // space loop
        u[(t+1)%2][x] = kernel( &u[t%2][x] );
```



Assuming LRU,  
if  $N > \mathcal{M}$ , then  
 $Q = \Theta(NT/\mathcal{B})$ .

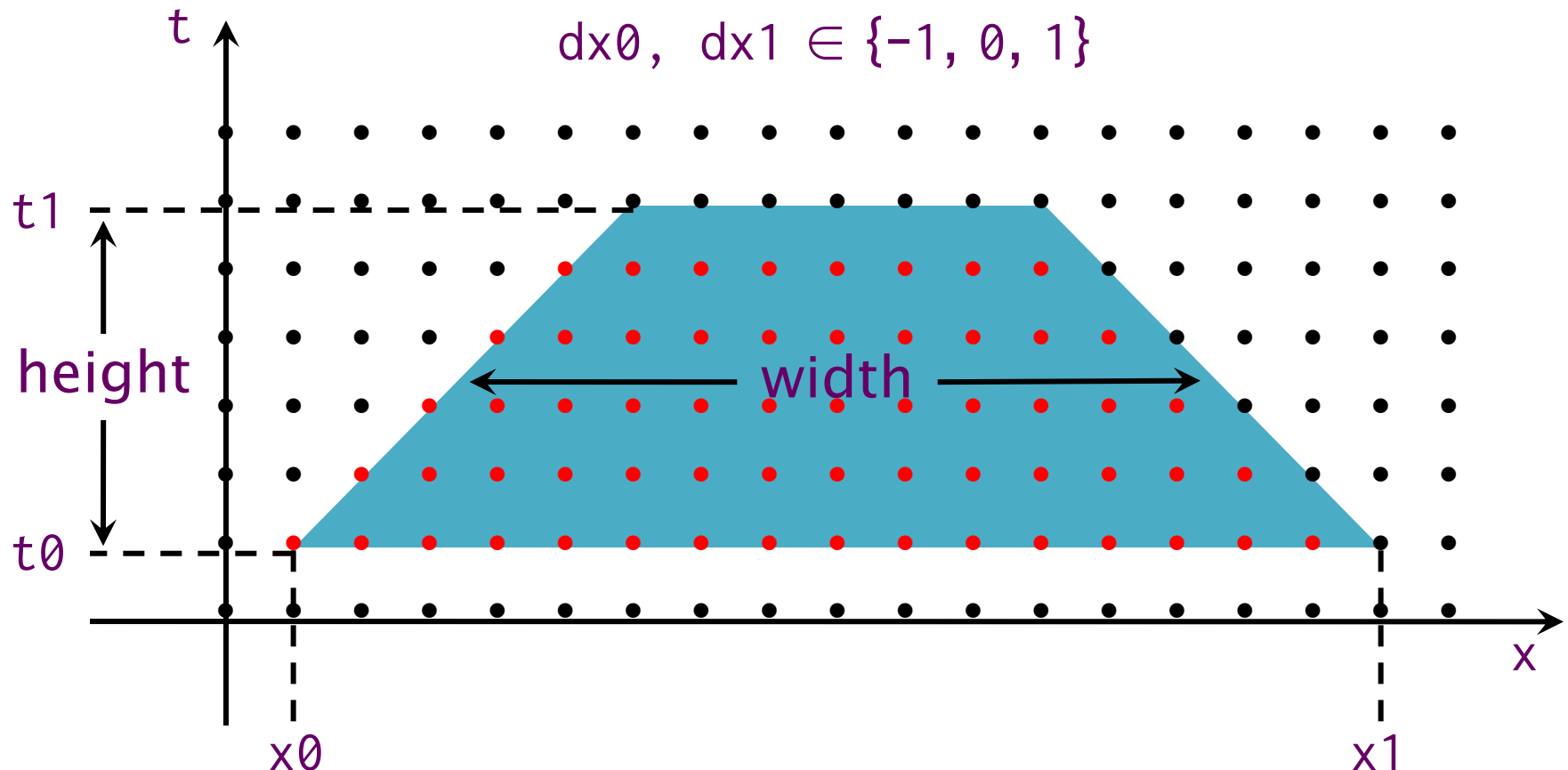
# Cache-Oblivious 3-Point Stencil

Recursively traverse trapezoidal regions of space-time points  $(t, x)$  such that

$$t_0 \leq t < t_1$$

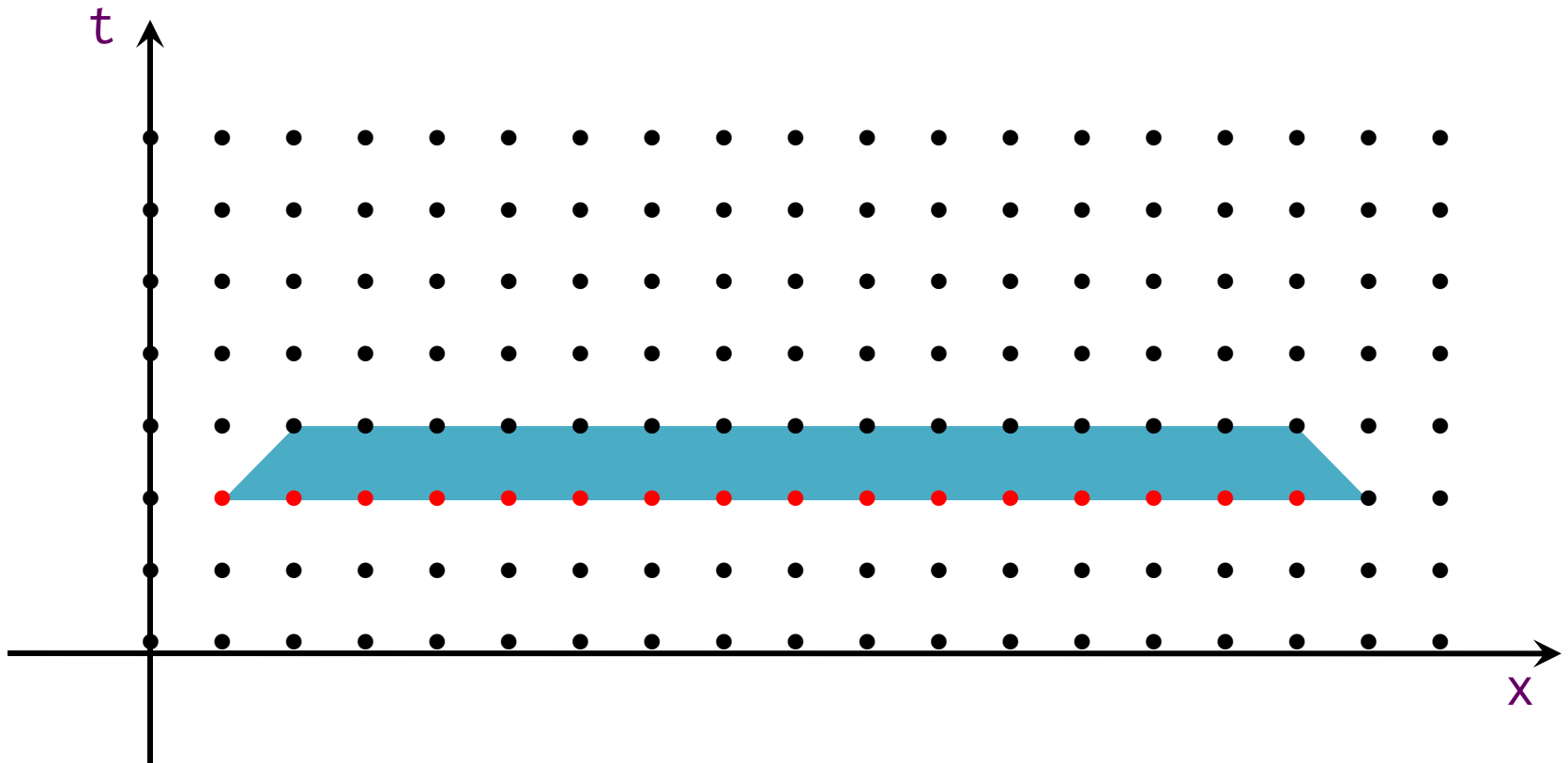
$$x_0 + dx_0(t - t_0) \leq x < x_1 + dx_1(t - t_0)$$

$$dx_0, dx_1 \in \{-1, 0, 1\}$$



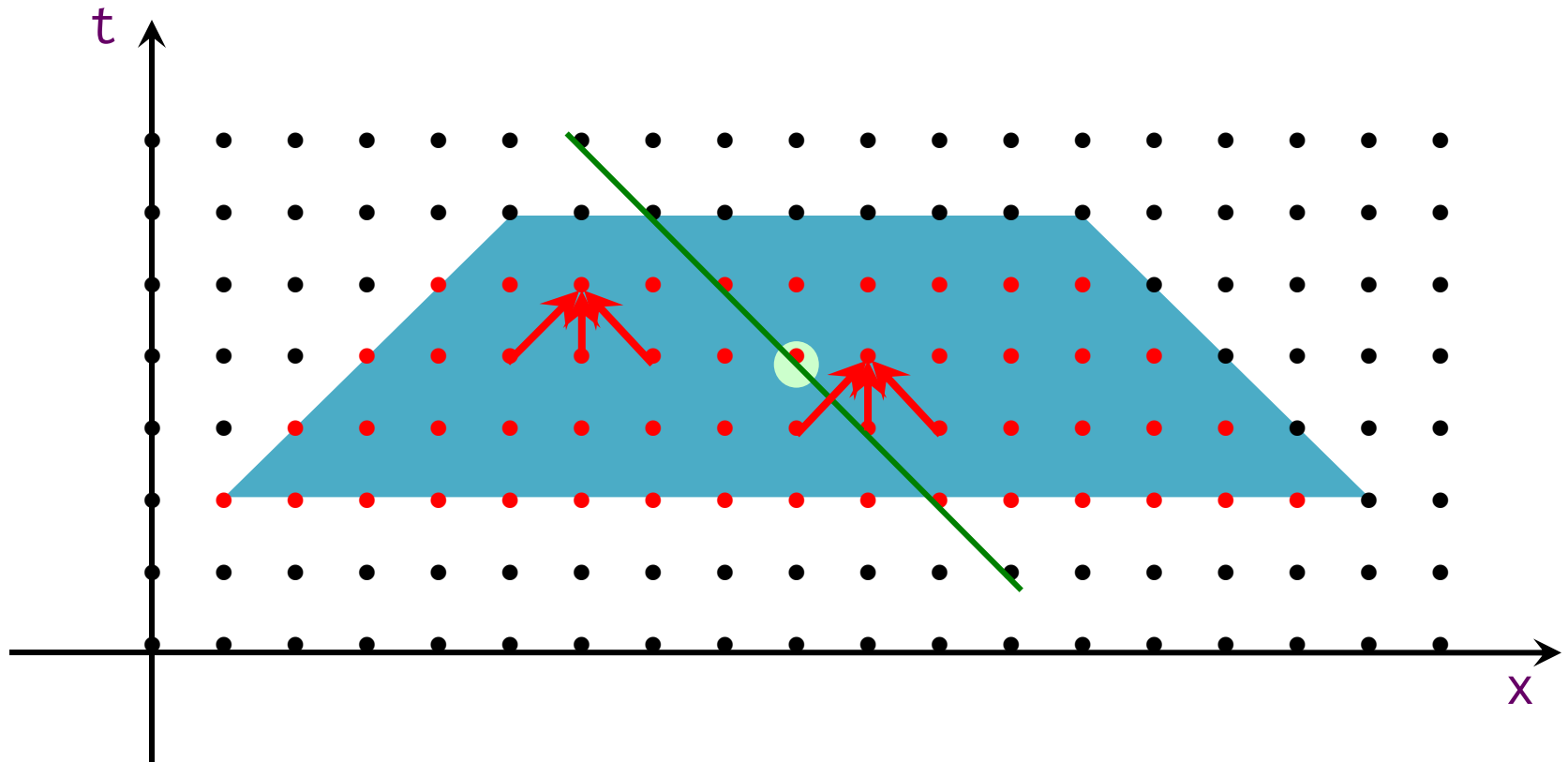
# Base Case

If  $\text{height} = 1$ , compute all space-time points in the trapezoid. Any order of computation is valid, since no point depends on another.



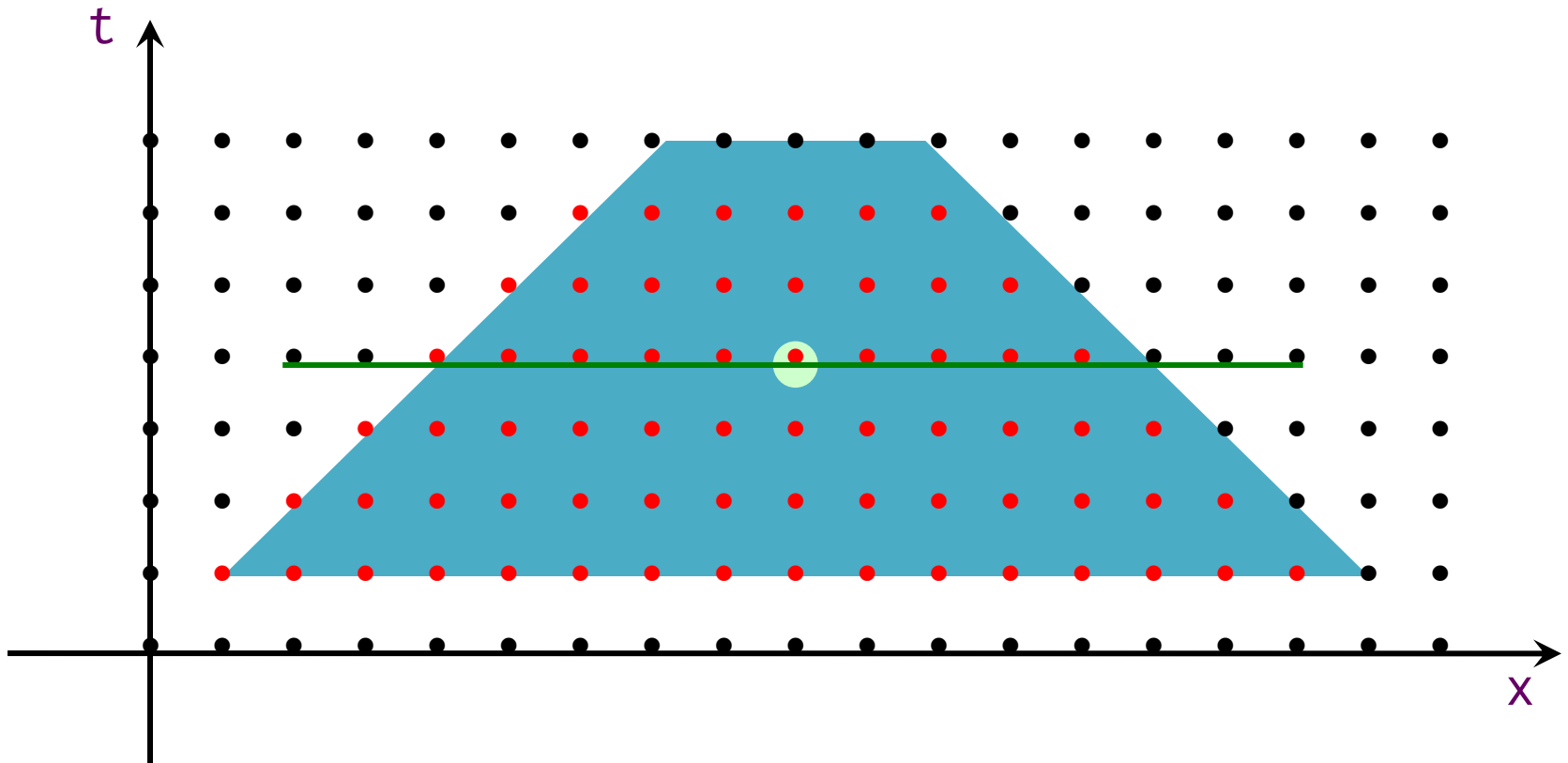
# Space Cut

If  $\text{width} \geq 2 \cdot \text{height}$ , cut the trapezoid with a line of slope  $-1$  through the center. Traverse the trapezoid on the left first, and then the one on the right.



# Time Cut

If  $\text{width} < 2 \cdot \text{height}$ , cut the trapezoid with a horizontal line through the center. Traverse the bottom trapezoid first, and then the top one.

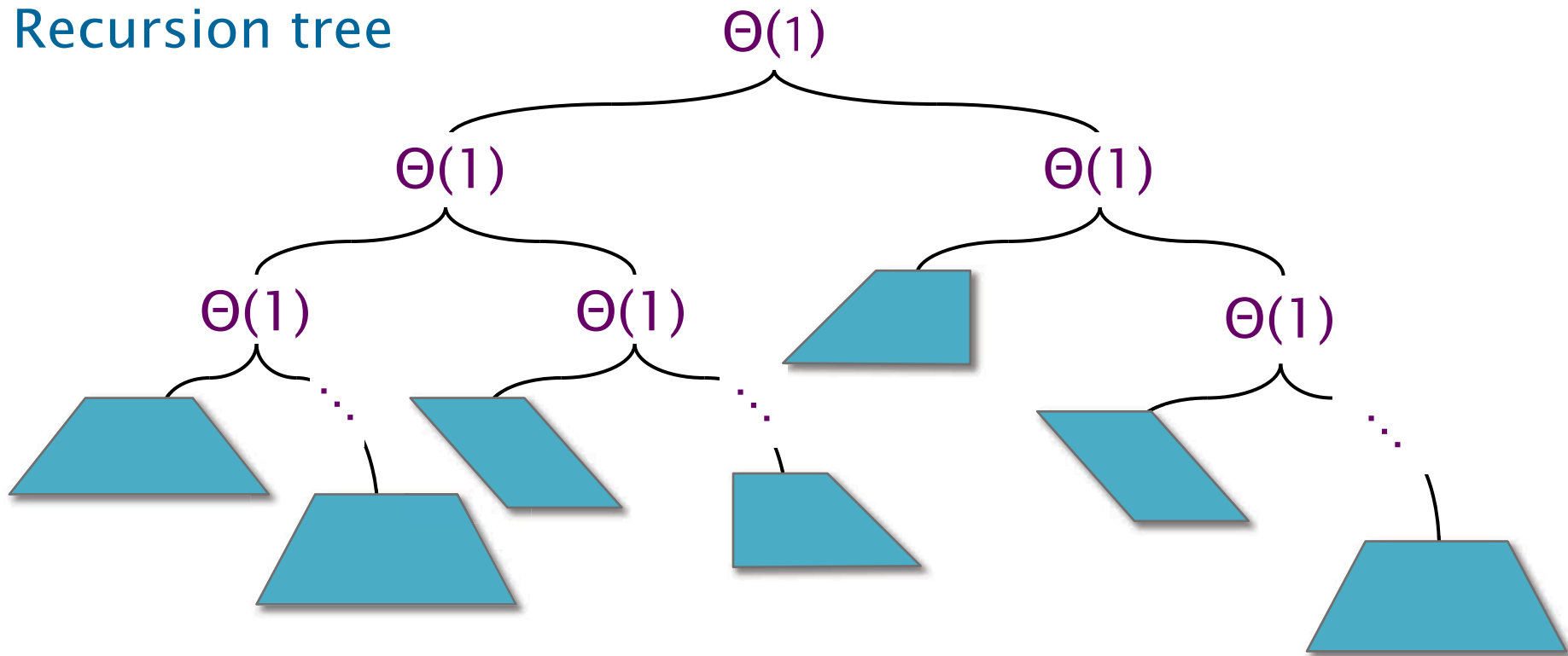


# C Implementation

```
void trapezoid(int64_t t0, int64_t t1, int64_t x0, int64_t dx0,
              int64_t x1, int64_t dx1)
{
    int64_t lt = t1 - t0;
    if (lt == 1) { //base case
        for (int64_t x = x0; x < x1; x++)
            u[t1%2][x] = kernel( &u[t0%2][x] );
    } else if (lt > 1) {
        if (2 * (x1 - x0) + (dx1 - dx0) * lt >= 4 * lt) { //space cut
            int64_t xm = (2 * (x0 + x1) + (2 + dx0 + dx1) * lt) / 4;
            trapezoid(t0, t1, x0, dx0, xm, -1);
            trapezoid(t0, t1, xm, -1, x1, dx1);
        } else { //time cut
            int64_t halflt = lt / 2;
            trapezoid(t0, t0 + halflt, x0, dx0, x1, dx1);
            trapezoid(t0 + halflt, t1, x0 + dx0 * halflt, dx0,
                    x1 + dx1 * halflt, dx1);
        }
    }
}
```

# Cache Analysis

## Recursion tree



- Each leaf represents  $\Theta(hw)$  points, where  $h = \Theta(w)$ .
- Each leaf incurs  $\Theta(w/\mathcal{B})$  misses, where  $w = \Theta(\mathcal{M})$ .
- $\Theta(NT/hw)$  leaves.
- #internal nodes = #leaves - 1 do not contribute substantially to  $Q$ .
- $Q = \Theta(NT/hw) \cdot \Theta(w/\mathcal{B}) = \Theta(NT/\mathcal{M}^2) \cdot \Theta(\mathcal{M}/\mathcal{B}) = \Theta(NT/\mathcal{M}\mathcal{B})$ .
- For  $d$  dimensions,  $Q = \Theta(NT/\mathcal{M}^{1/d}\mathcal{B})$

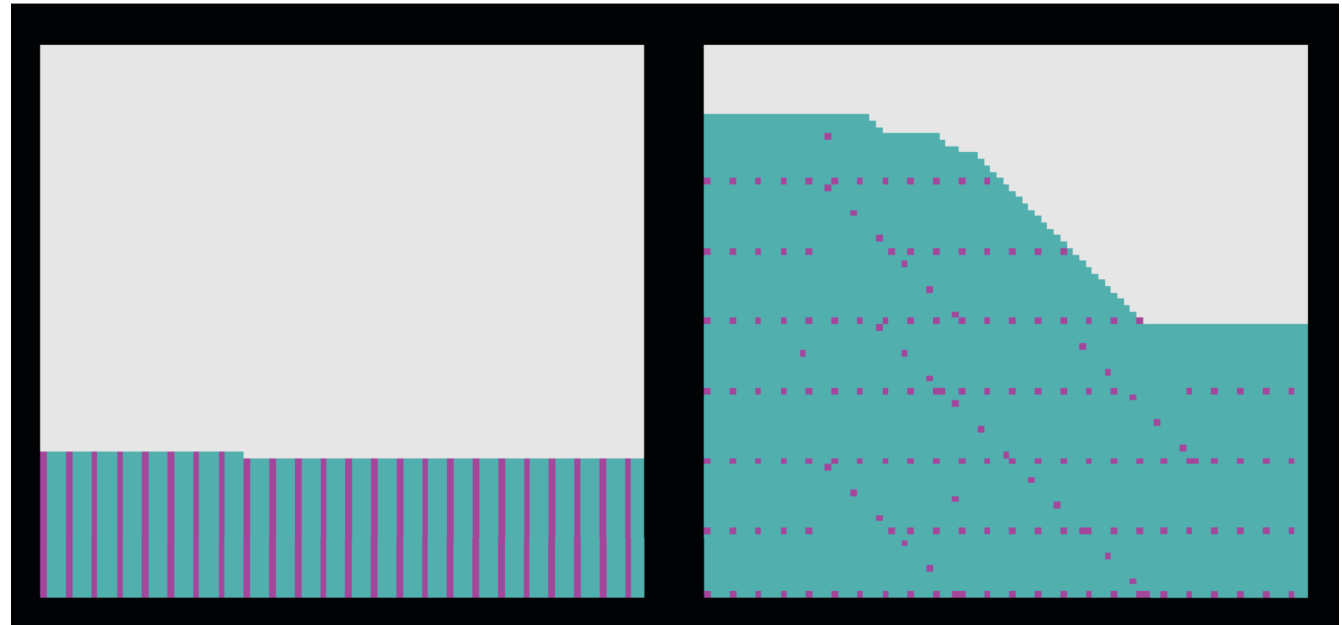


# Simulation: 3-Point Stencil

- Rectangular region

- $N = 95$

- $T = 87$

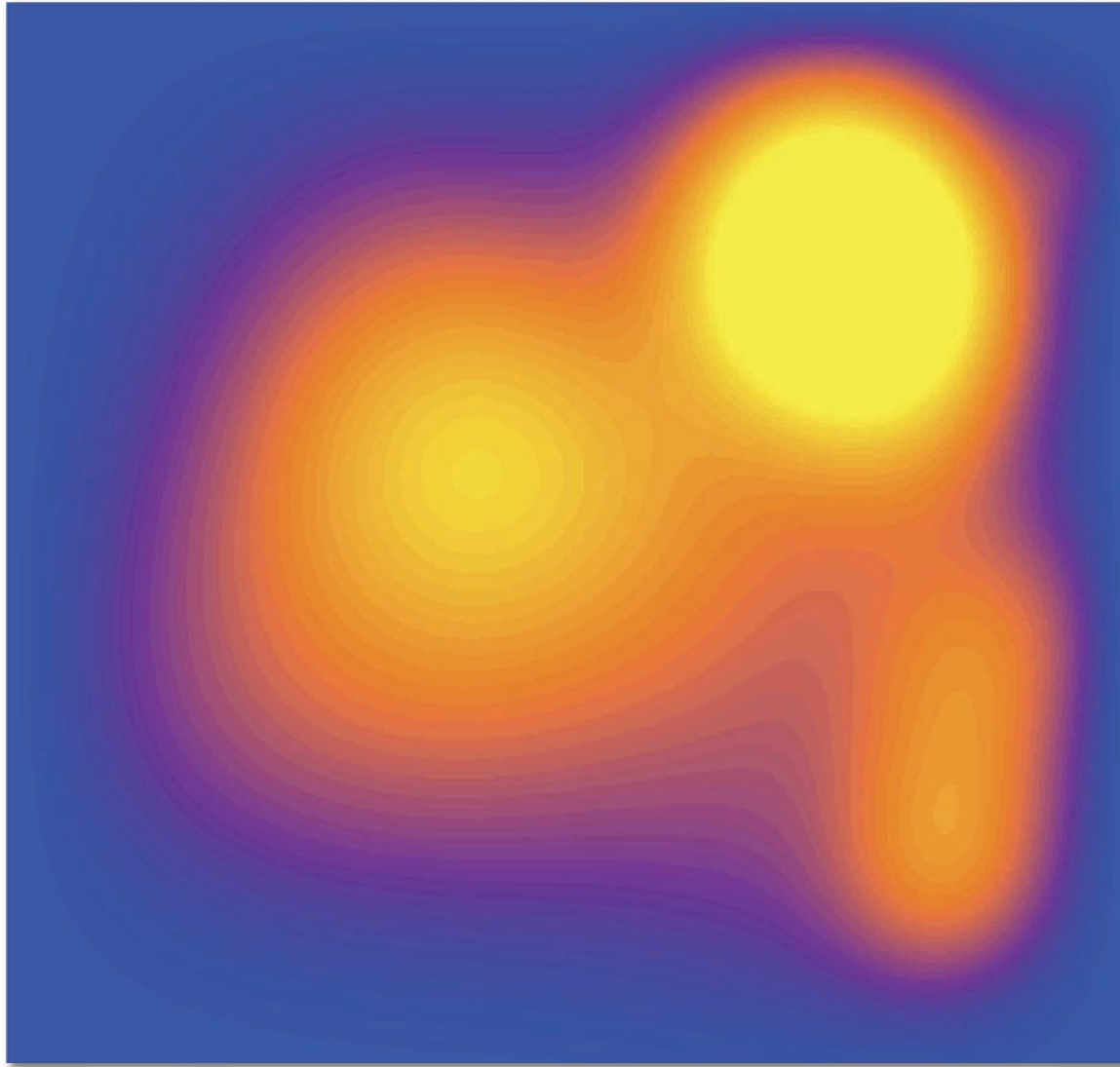


Looping

Trapezoid

- Fully associative LRU cache
  - $\mathcal{B} = 4$  points
  - $\mathcal{M} = 32$  points
- Cache-hit latency = 1 cycle
- Cache-miss latency = 10 cycles

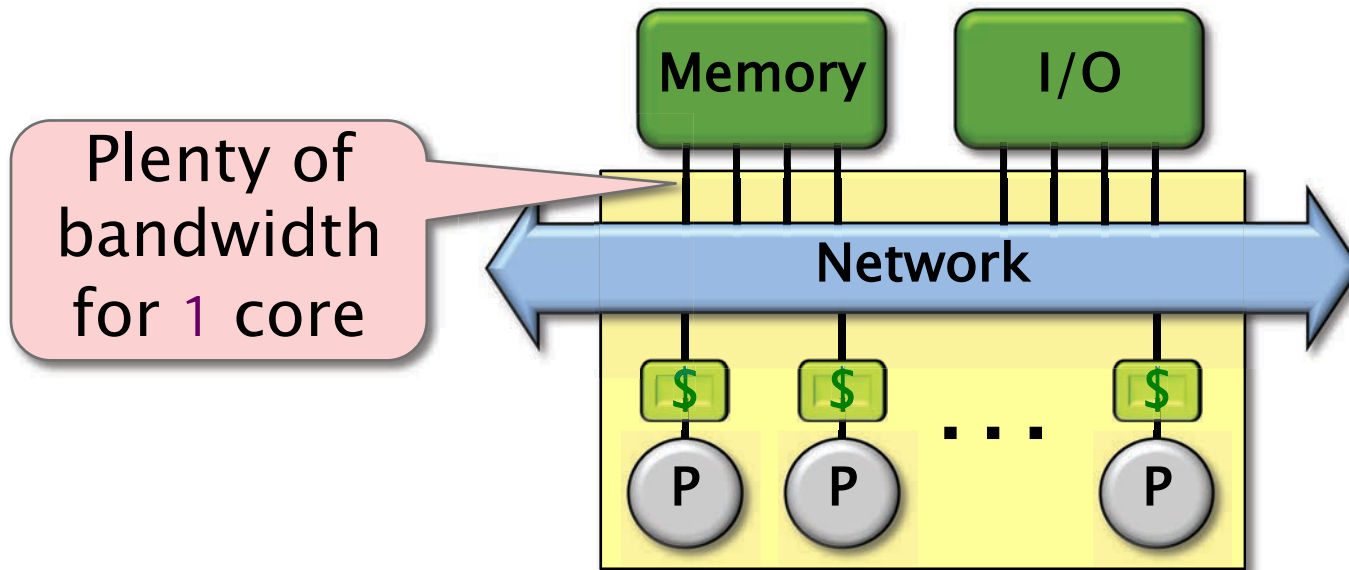
# Looping v. Trapezoid on Heat



# Impact on Performance

Q. How can the cache-oblivious trapezoidal decomposition have so many fewer cache misses, but the advantage gained over the looping version be so marginal?

A. **Prefetching** and a good memory architecture. One core cannot saturate the memory bandwidth.



# CACHING AND PARALLELISM



# Cilk and Caching

**Theorem.** Let  $Q_p$  be the number of cache misses in a deterministic Cilk computation when run on  $P$  processors, each with a private cache, and let  $S_p$  be the number of successful steals during the computation. In the ideal-cache model, we have

$$Q_p = Q_1 + O(S_p \mathcal{M}/\mathcal{B}) ,$$

where  $\mathcal{M}$  is the cache size and  $\mathcal{B}$  is the size of a cache block.

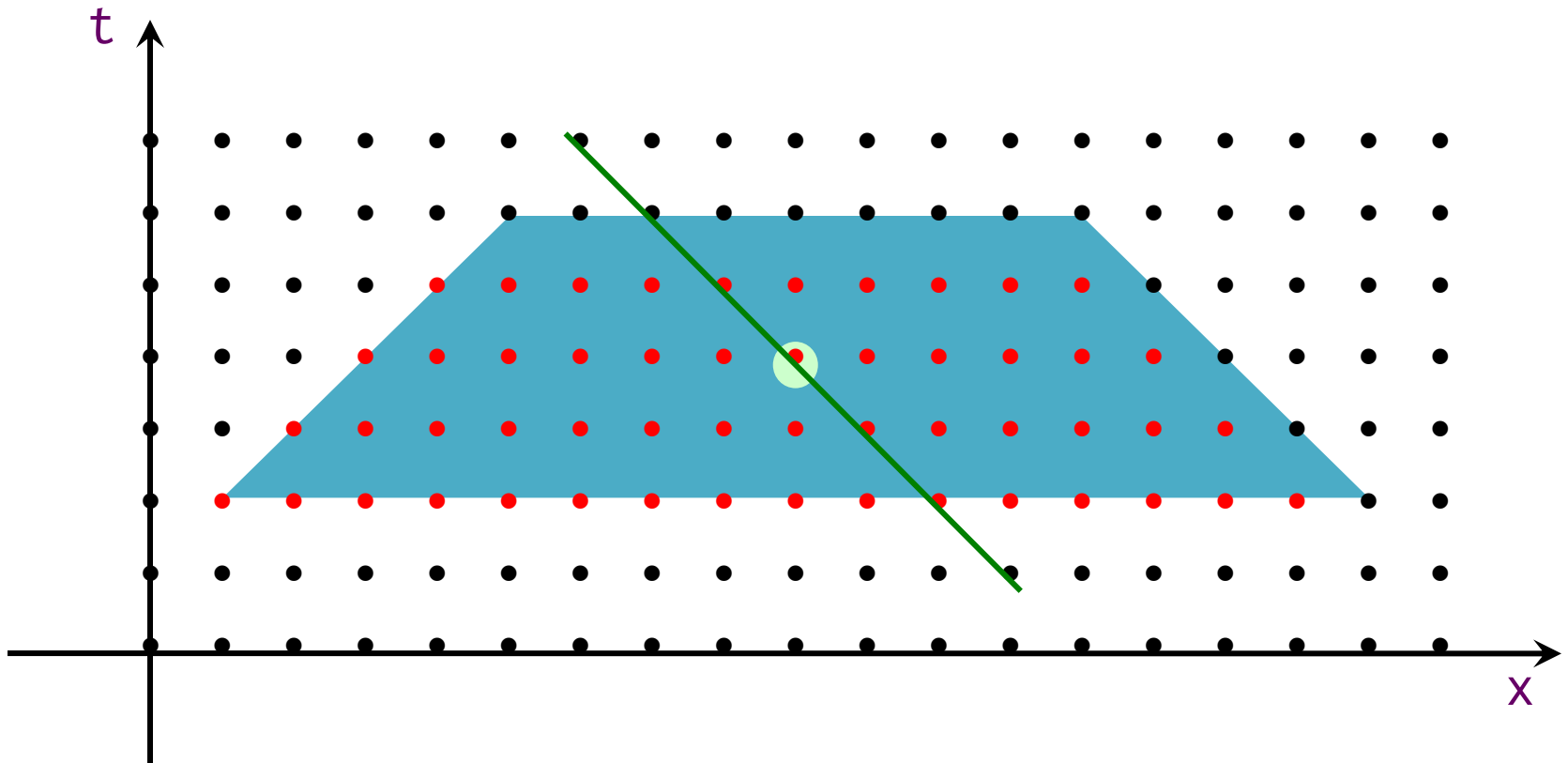
*Proof.* After a worker steals a continuation, its cache is completely cold in the worst case. But after  $\mathcal{M}/\mathcal{B}$  (cold) cache misses, its cache is identical to that in the serial execution. The same is true when a worker resumes a stolen subcomputation after a `cilk_sync`. The number of times these two situations can occur is at most  $2S_p$ . ■

**MORAL:** Minimizing cache misses in the serial elision essentially minimizes them in parallel executions.

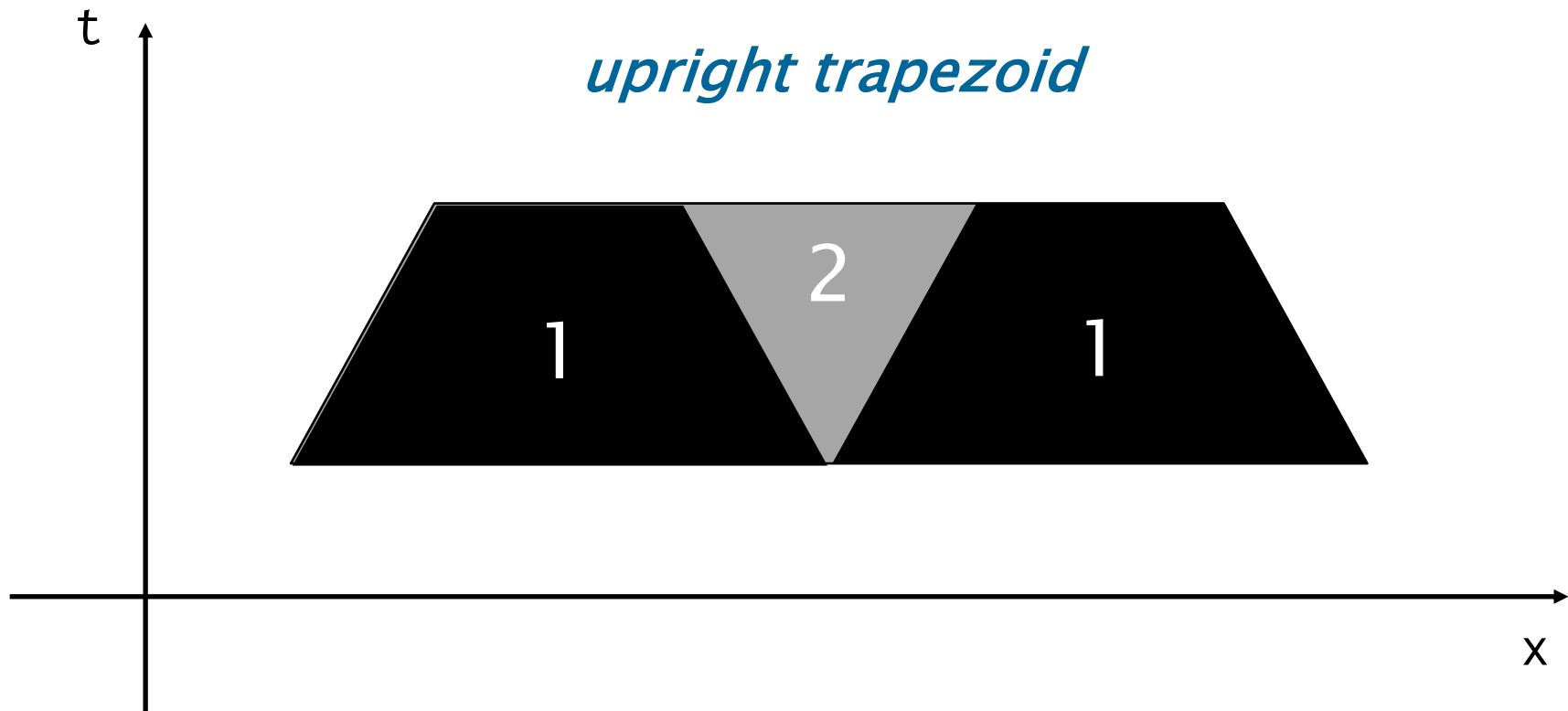
# Does this work in parallel?

Space cut: If  $\text{width} \geq 2 \cdot \text{height}$ , cut the trapezoid with a line of slope  $-1$  through the center.

Traverse the trapezoid on the left first, and then the one on the right.

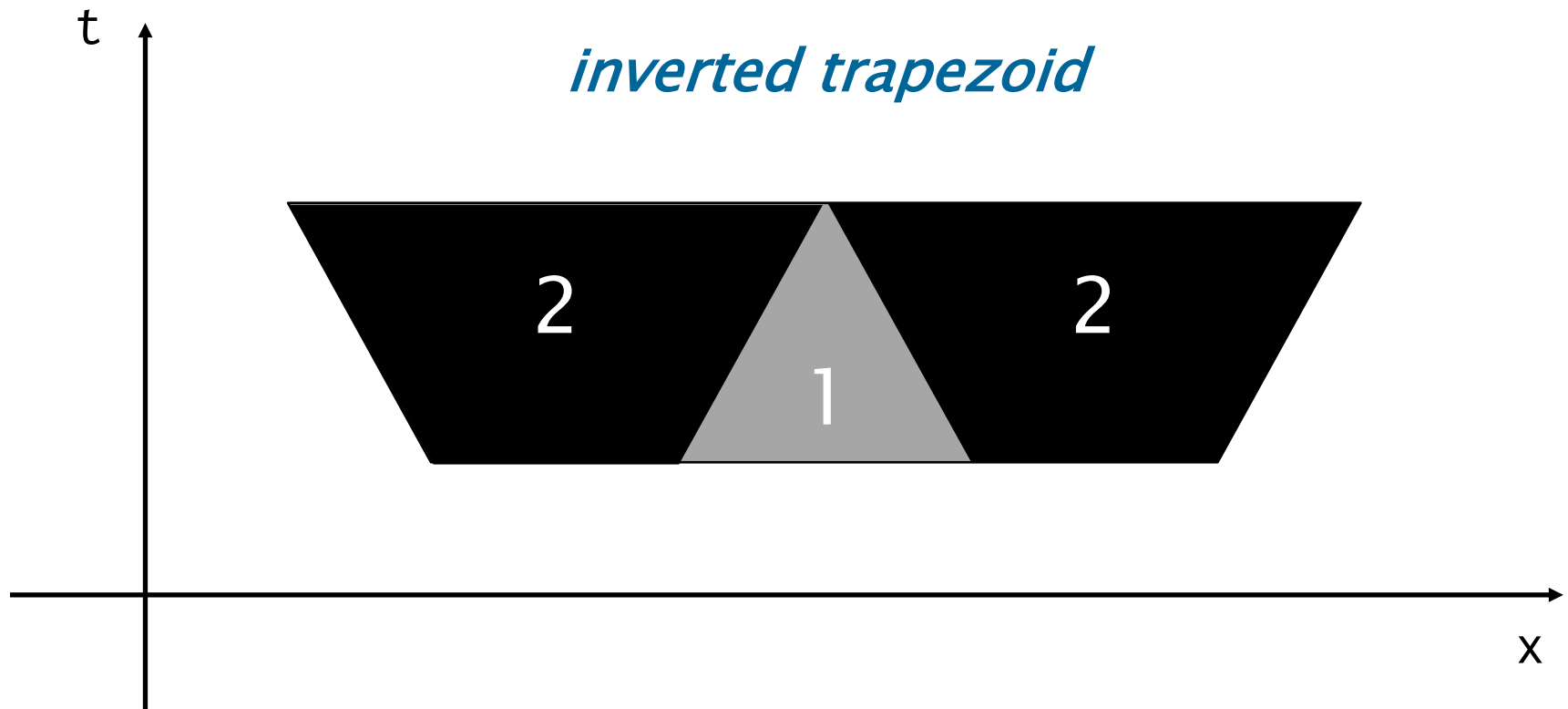


# Parallel Space Cuts



A *parallel space cut* produces two black trapezoids that can be executed in parallel and a third gray trapezoid that executes in series with the black trapezoids.

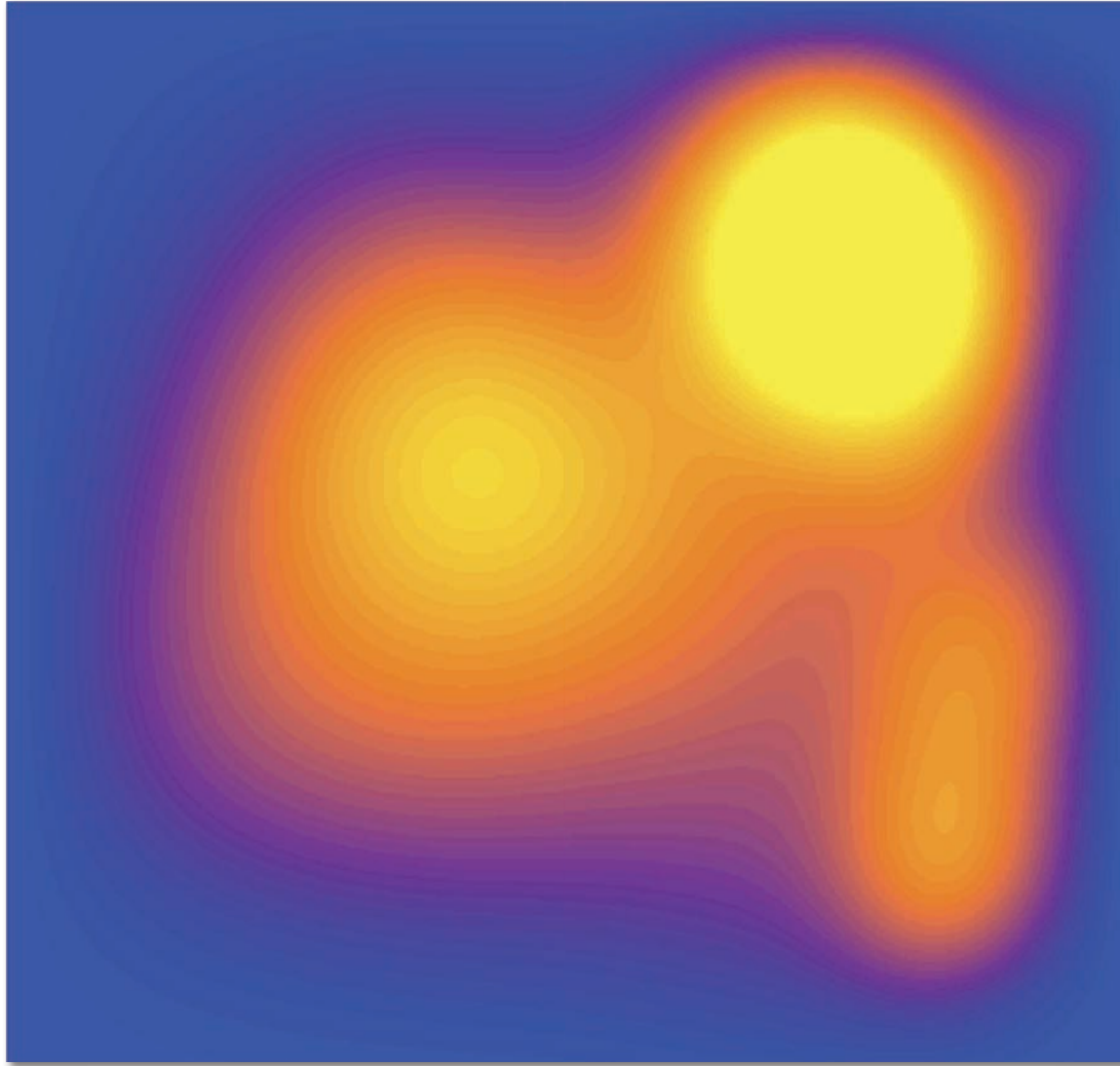
# Parallel Space Cuts



A *parallel space cut* produces two black trapezoids that can be executed in parallel and a third gray trapezoid that executes in series with the black trapezoids.



# Parallel Looping v. Parallel Trap.



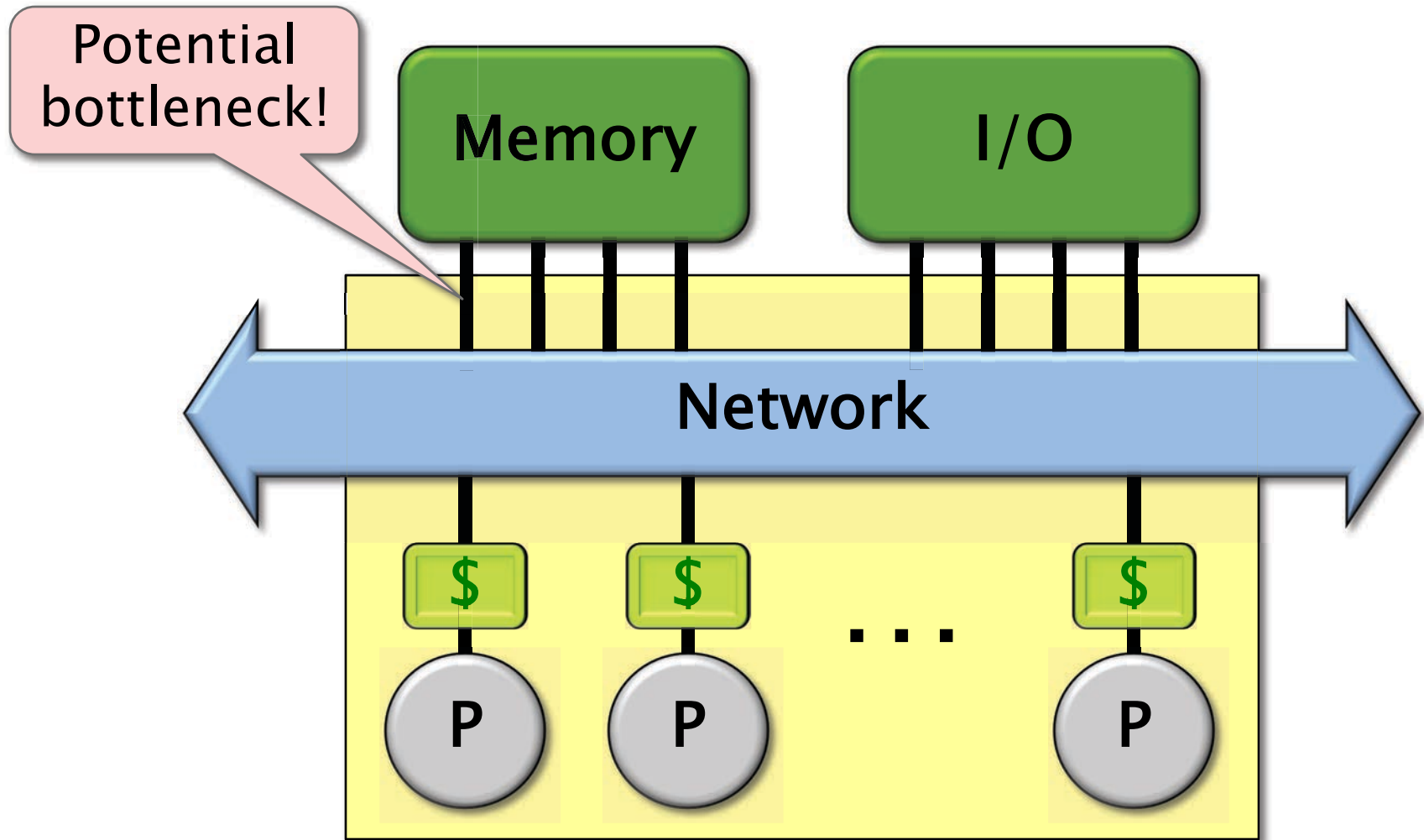
# Performance Comparison

Heat equation on a  $3000 \times 3000$  grid for 1000 time steps (4 processor cores with 8MB LLC)

Code	Time	
Serial looping	128.95s	} 1.93x
Parallel looping	66.97s	
Serial trapezoidal	66.76s	} 3.96x
Parallel trapezoidal	16.86s	

The parallel looping code achieves less than half the potential speedup, even though it has far more parallelism.

# Memory Bandwidth



# Impediments to Speedup

- ✓ Insufficient parallelism
- ✓ Scheduling overhead
- ✓ Lack of memory bandwidth
- ✓ Contention (locking and true/false sharing)

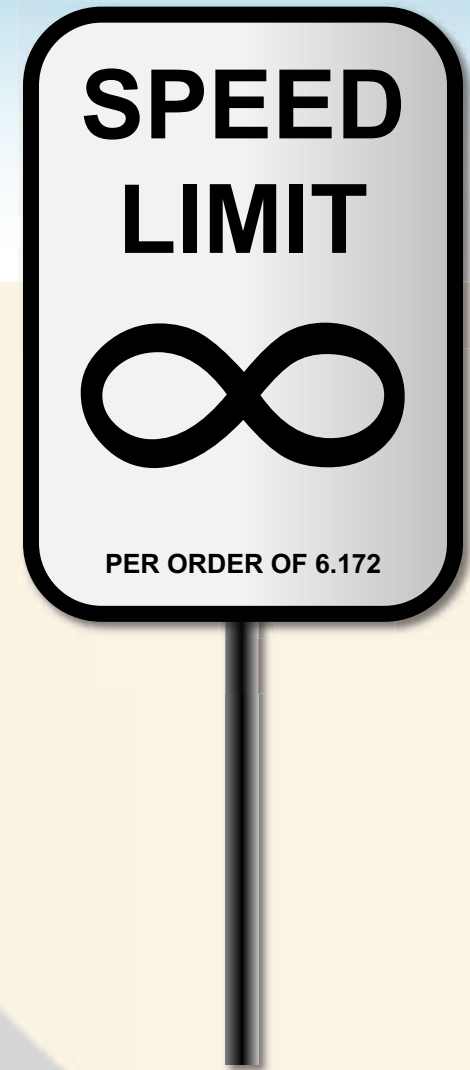
Cilkscale can diagnose the first two problems.

Q. How can we diagnose the third?

A. Run  $P$  identical copies of the serial code in parallel  
— if you have enough memory.

Tools exist to detect lock contention in an execution, but not the *potential* for lock contention. Potential for true and false sharing is even harder to detect.

# CACHE-OBLIVIOUS SORTING



# OUTLINE

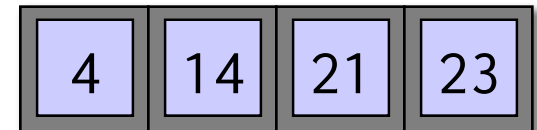
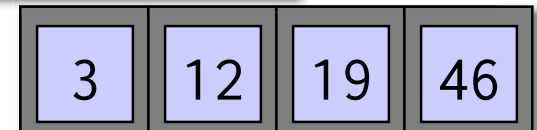
- Simulation of Heat Diffusion
- Cache-Oblivious Stencil Computations
- Caching and Parallelism
- Cache-Oblivious Sorting

# Merging Two Sorted Arrays

```
void merge(int64_t *C, int64_t *A, int64_t na,
           int64_t *B, int64_t nb) {
    while (na>0 && nb>0) {
        if (*A <= *B) {
            *C++ = *A++; na--;
        } else {
            *C++ = *B++; nb--;
        }
    }
    while (na>0) {
        *C++ = *A++; na--;
    }
    while (nb>0) {
        *C++ = *B++; nb--;
    }
}
```

Time to merge  $n$   
elements =  $\Theta(n)$ .

Number of cache  
misses =  $\Theta(n/B)$ .



# Merge Sort

```
void merge_sort(int64_t *B, int64_t *A, int64_t n) {  
    if (n==1) {  
        B[0] = A[0];  
    } else {  
        int64_t C[n];  
        cilk_spawn merge_sort(C, A, n/2);  
                   merge_sort(C+n/2, A+n/2, n-n/2);  
        cilk_sync;  
        merge(B, C, n/2, C+n/2, n-n/2);  
    }  
}
```

19	3	12	46	33	4	21	14
----	---	----	----	----	---	----	----



# Merge Sort

```
void merge_sort(int64_t *B, int64_t *A, int64_t n) {  
    if (n==1) {  
        B[0] = A[0];  
    } else {  
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        cilk_sync;  
        merge(B, C, n/2, C+n/2, n-n/2);  
    }  
}
```

19    3    12    46

33    4    21    14

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        cilk_sync;  
        merge(B, C, n/2, C+n/2, n-n/2);  
    }  
}
```

19

3

12

46

33

4

21

14

# Merge Sort

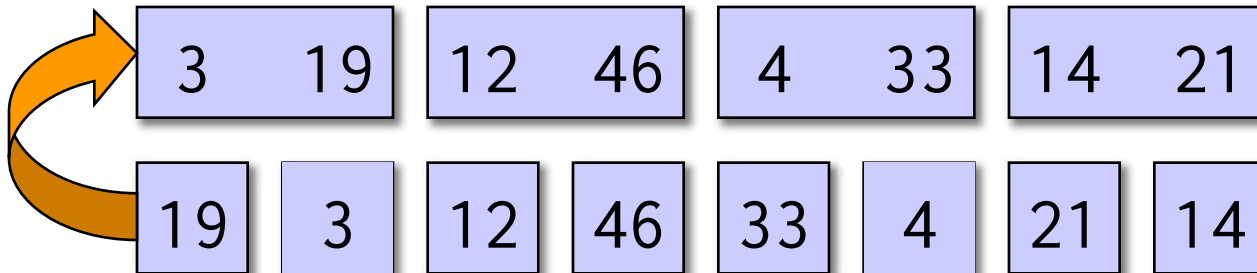
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        cilk_sync;  
        merge(B, C, n/2, C+n/2, n-n/2);  
    }  
}
```

19	3	12	46	33	4	21	14
----	---	----	----	----	---	----	----

# Merge Sort

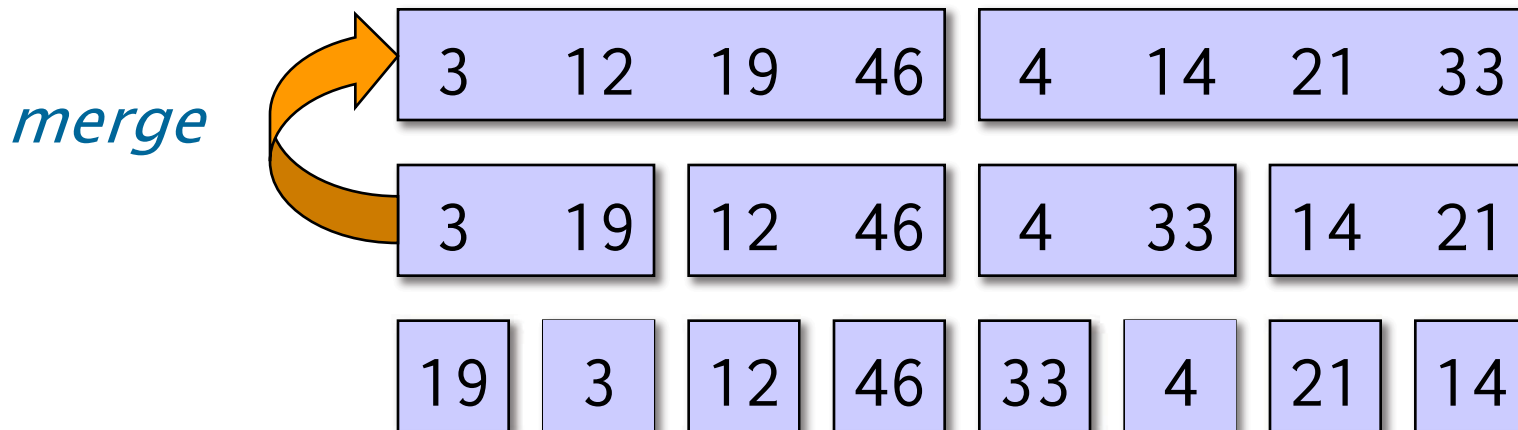
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        cilk_sync;  
        merge(B, C, n/2, C+n/2, n-n/2);  
    }  
}
```

*merge*



# Merge Sort

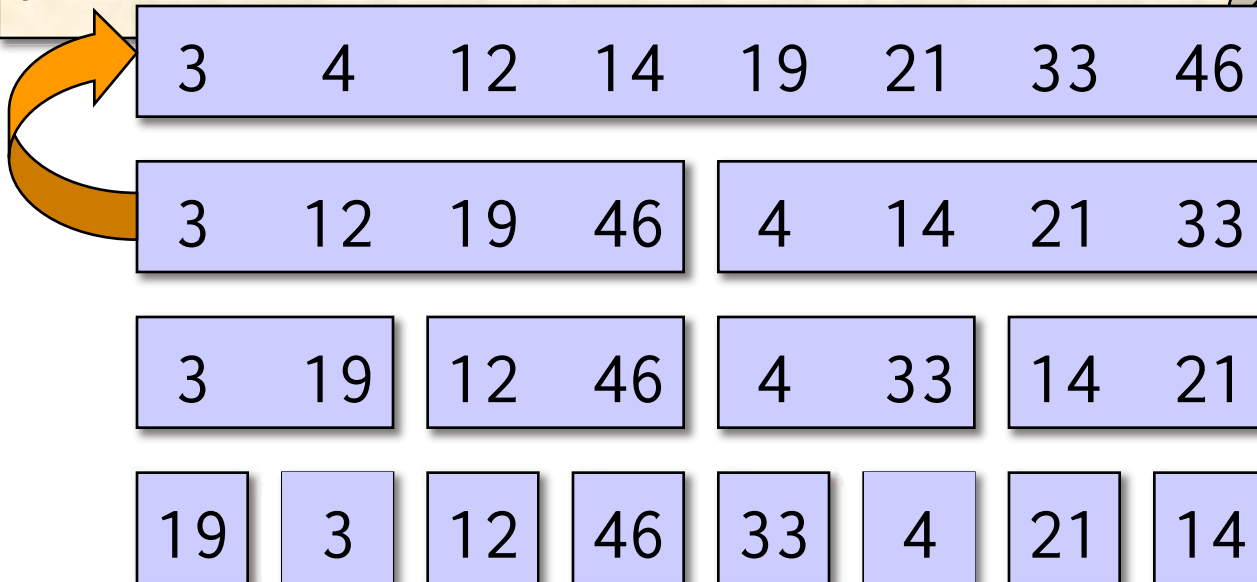
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    }  
}
```



# Merge Sort

```
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        cilk_sync;  
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    }  
}
```

*merge*



# Work of Merge Sort

```
void merge_sort(int64_t *B, int64_t *A, int64_t n) {  
    if (n==1) {  
        B[0] = A[0];  
    } else {  
        int64_t C[n];  
        merge_sort(C, A, n/2);  
        merge_sort(C+n/2, A+n/2, n-n/2);  
        merge(B, C, n/2, C+n/2, n-n/2);  
    }  
}
```

## CASE 2

$$n^{\log_b a} = n^{\log_2 2} = n$$

$$f(n) = \Theta(n^{\log_b a} \lg^0 n)$$

*Work:*

$$\begin{aligned} W(n) &= 2W(n/2) + \Theta(n) \\ &= \Theta(n \lg n) \end{aligned}$$

# Recursion Tree

Solve  $W(n) = 2W(n/2) + \Theta(n)$ .



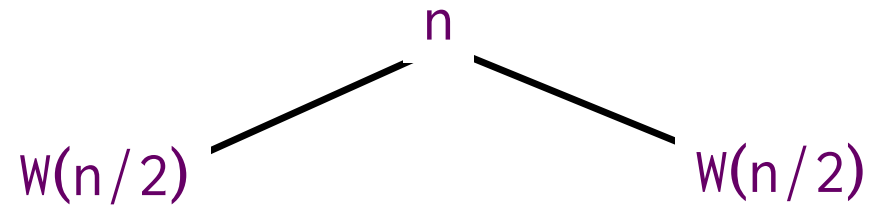
# Recursion Tree

Solve  $W(n) = 2W(n/2) + \Theta(n)$ .

$W(n)$

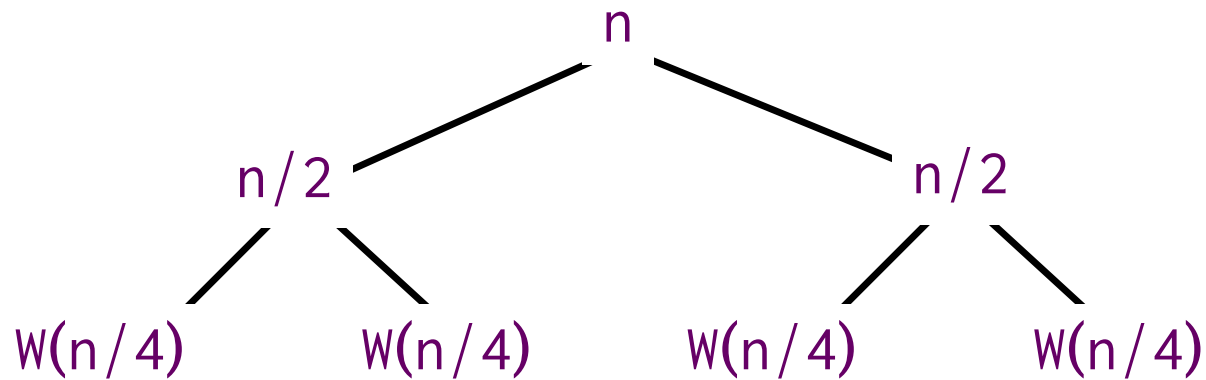
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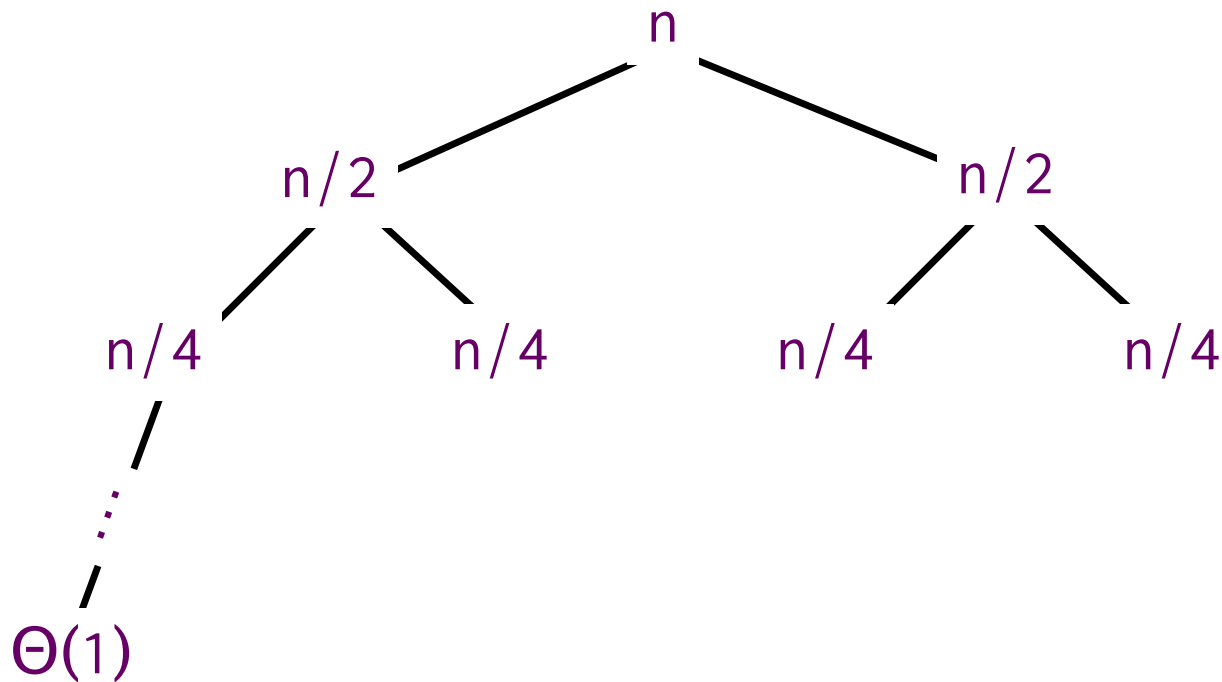
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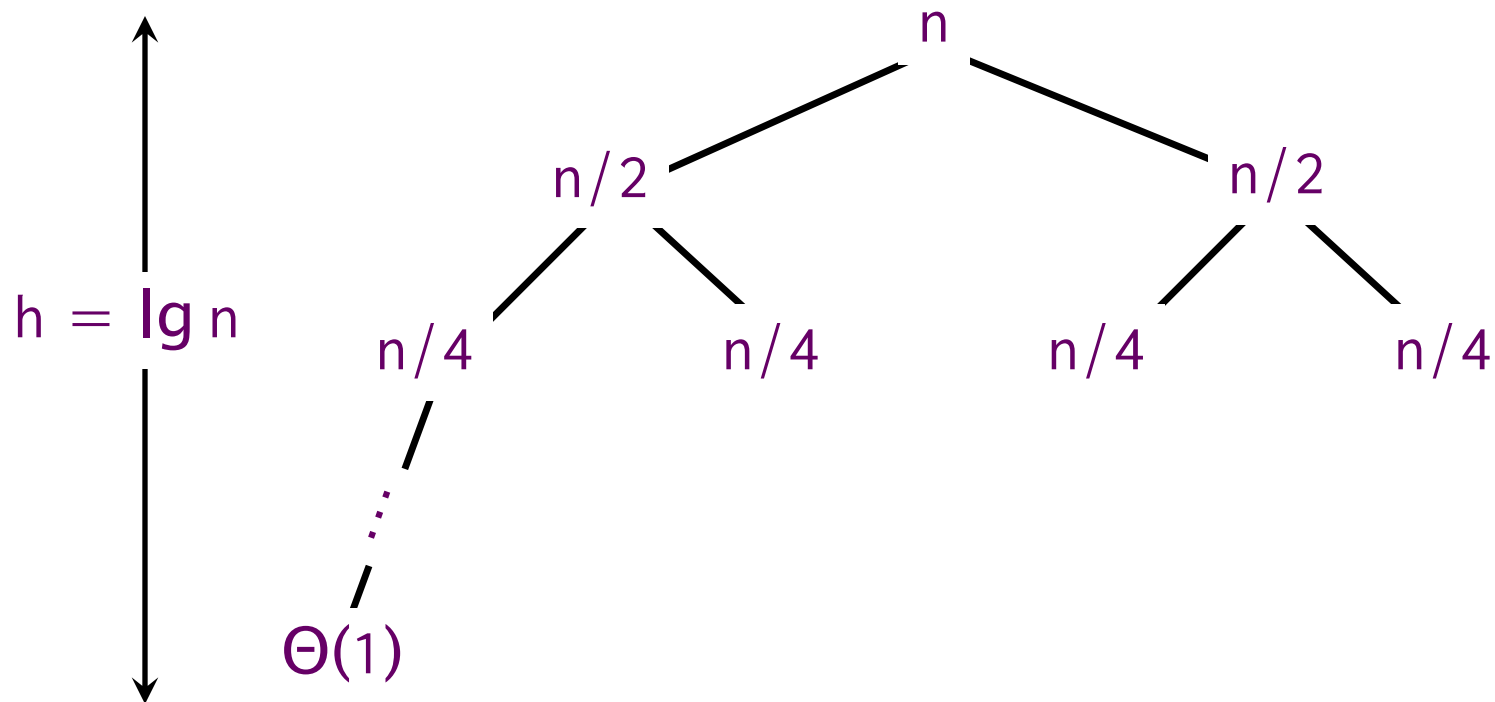
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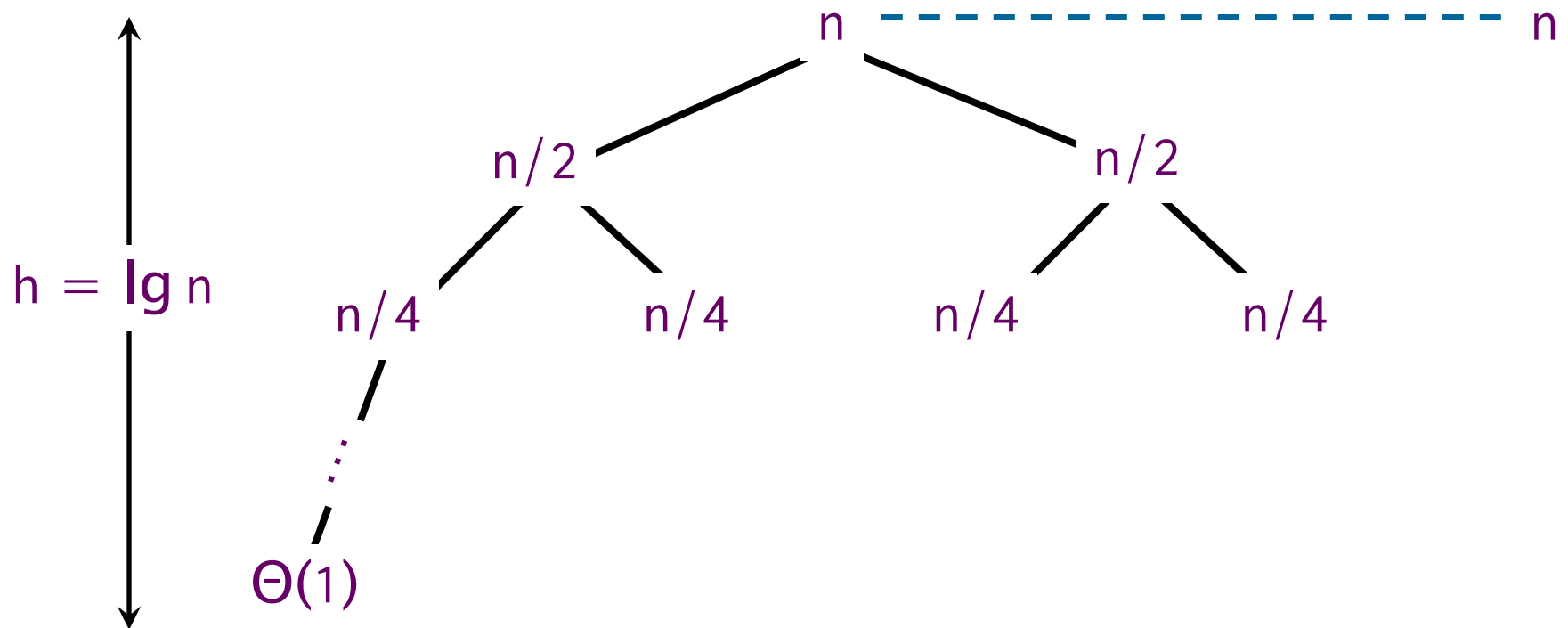
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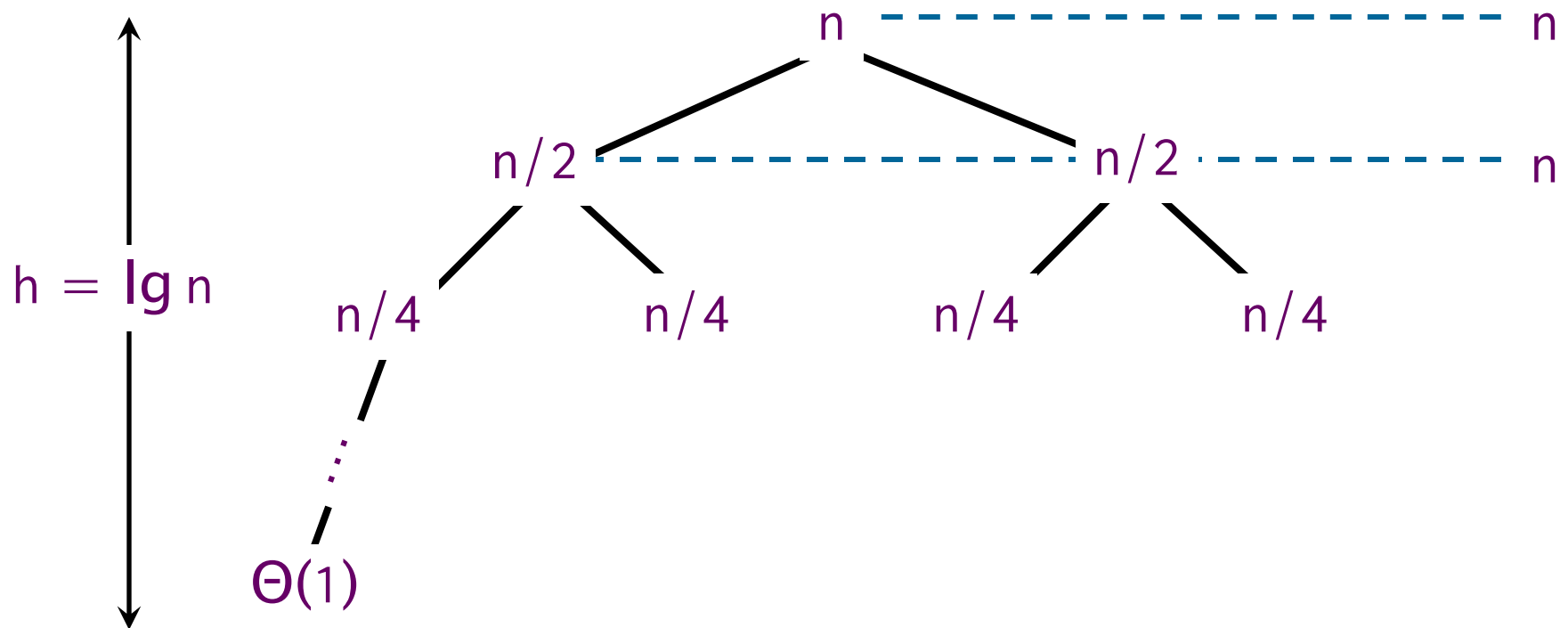
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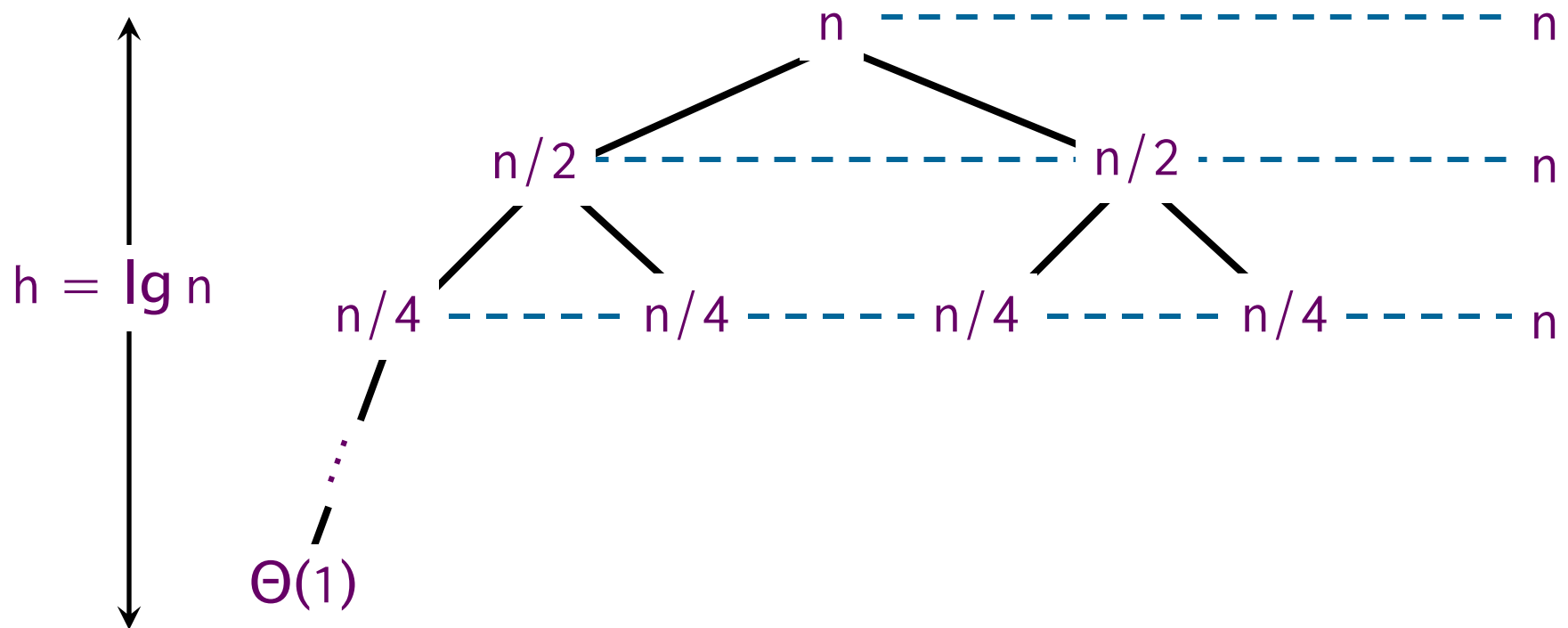
# Recursion Tree

Solve  $W(n) = 2W(n/2) + \Theta(n)$ .



# Recursion Tree

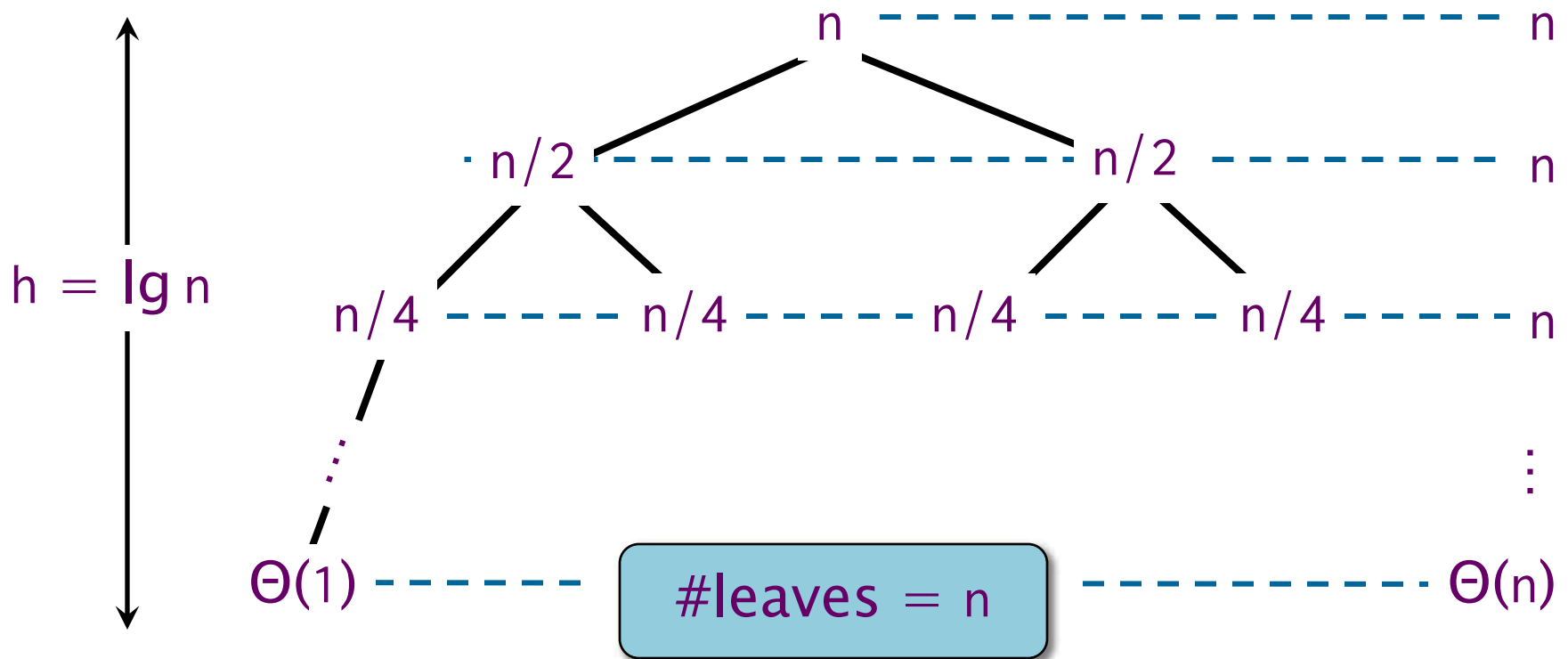
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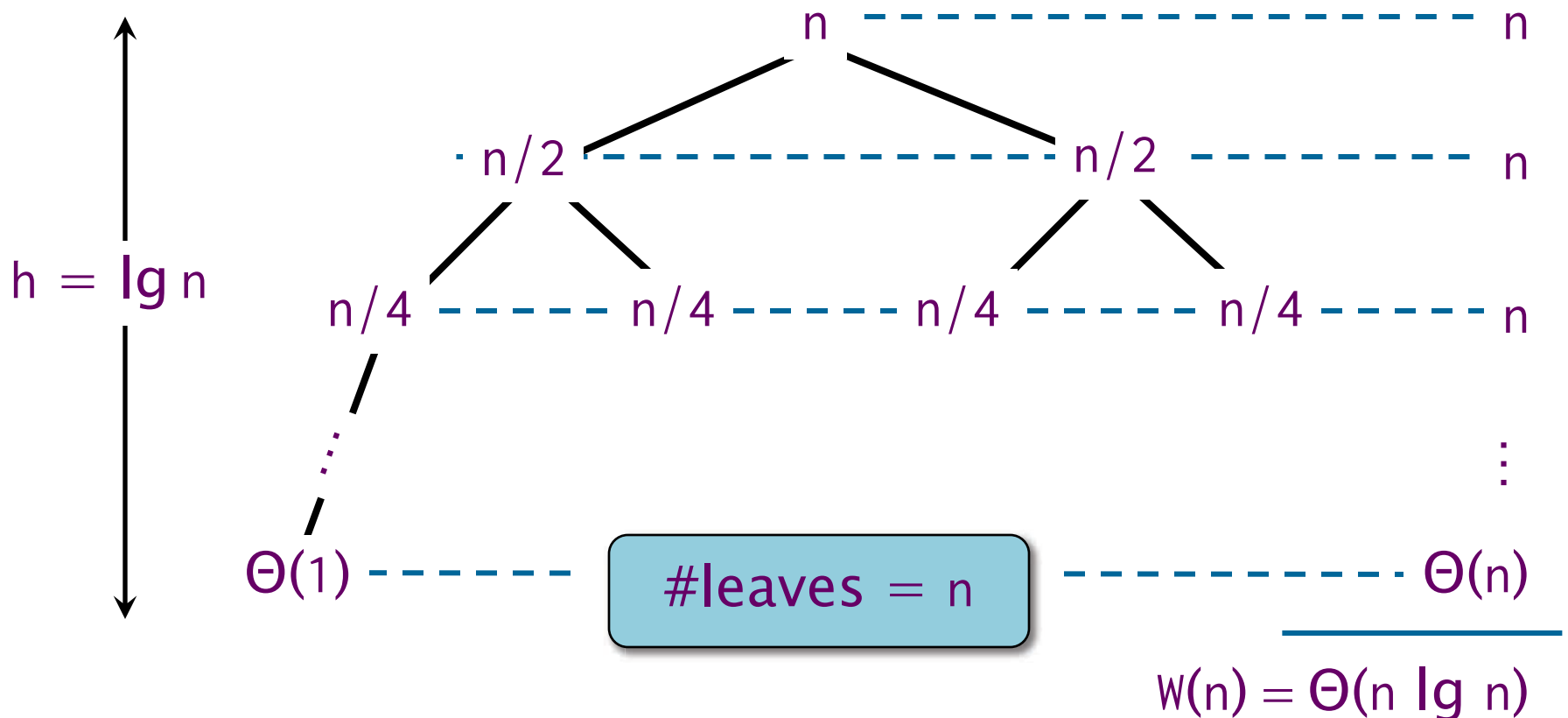
# Recursion Tree

Solve  $W(n) = 2W(n/2) + \Theta(n)$ .



# Recursion Tree

Solve  $W(n) = 2W(n/2) + \Theta(n)$ .



# Now with Caching

## Merge subroutine

$$Q(n) = \Theta(n/\mathcal{B}) .$$

## Merge sort

$$Q(n) = \begin{cases} \Theta(n/\mathcal{B}) & \text{if } n \leq c\mathcal{M}, \text{ constant } c \leq 1; \\ 2Q(n/2) + \Theta(n/\mathcal{B}) & \text{otherwise.} \end{cases}$$

# Cache Analysis of Merge Sort

$$Q(n) = \begin{cases} \Theta(n/\mathcal{B}) & \text{if } n \leq c\mathcal{M}, \text{ constant } c \leq 1; \\ 2Q(n/2) + \Theta(n/\mathcal{B}) & \text{otherwise.} \end{cases}$$

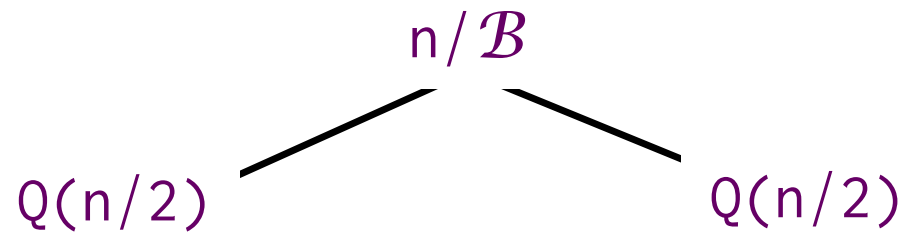
Recursion tree

$Q(n)$

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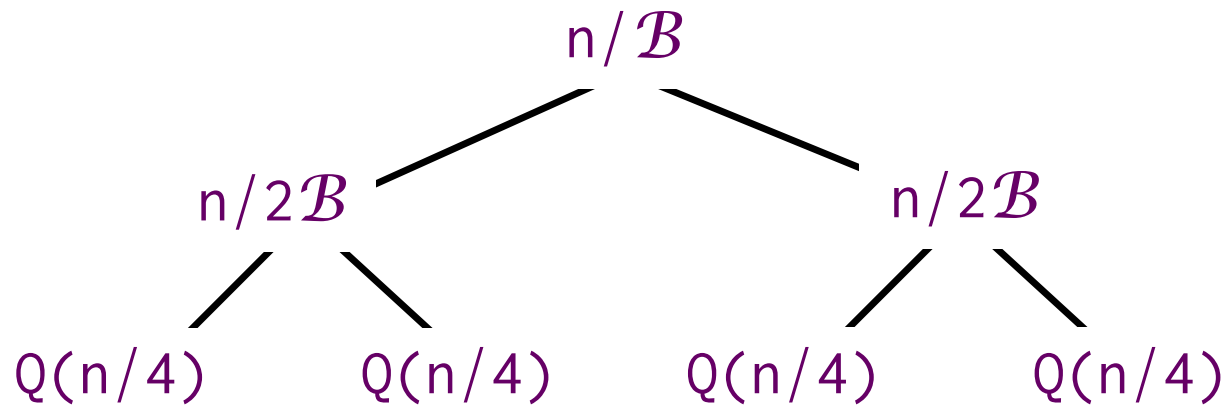
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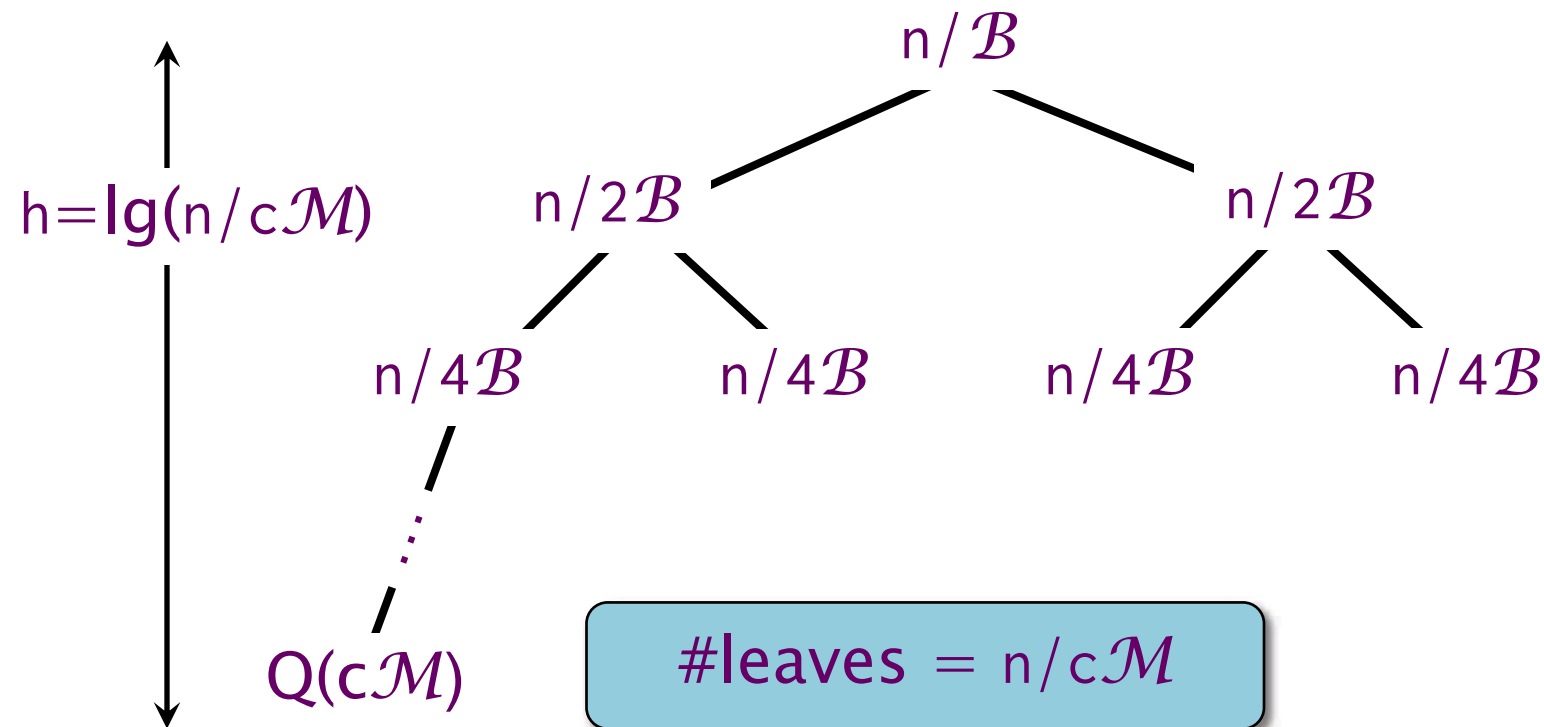
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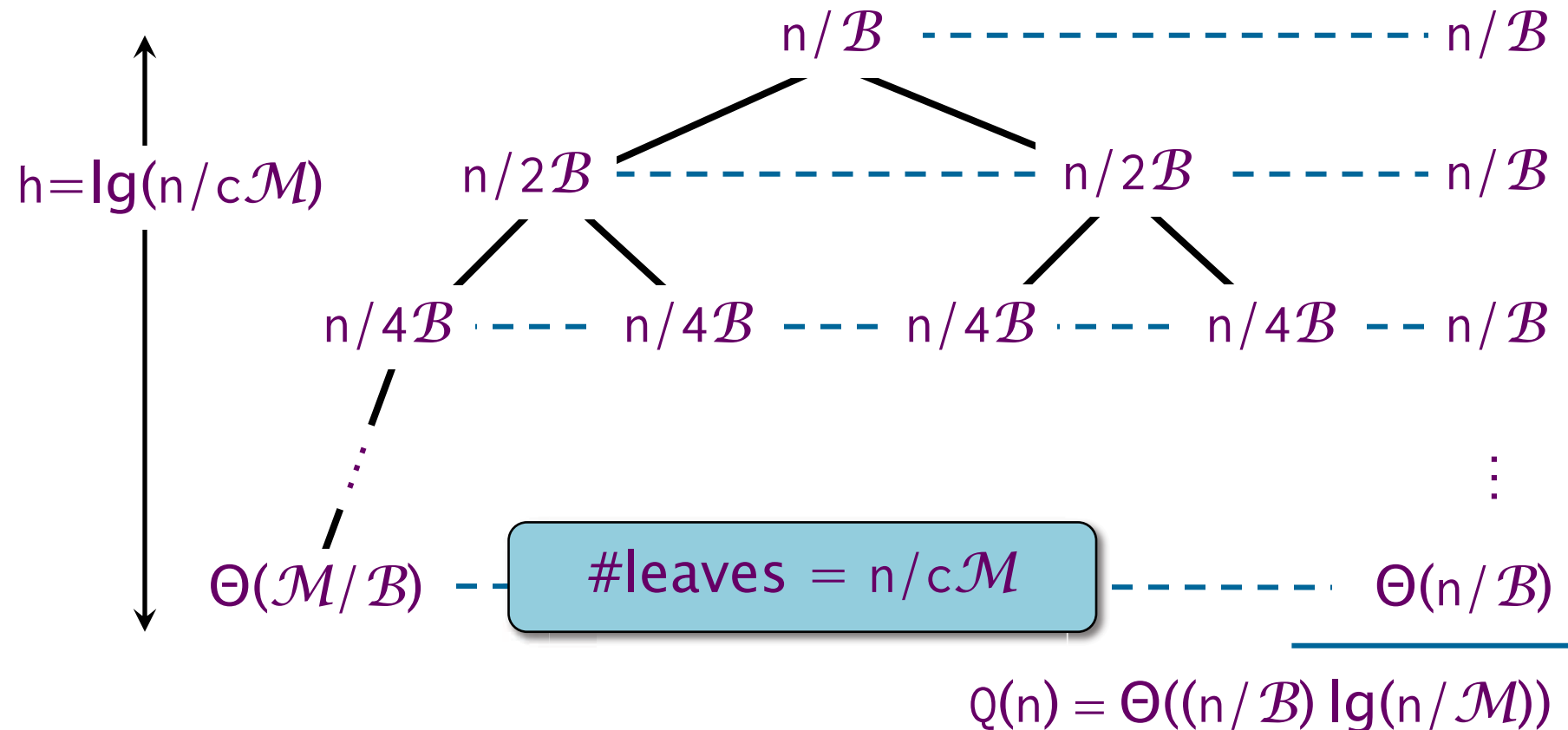
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# Bottom Line for Merge Sort

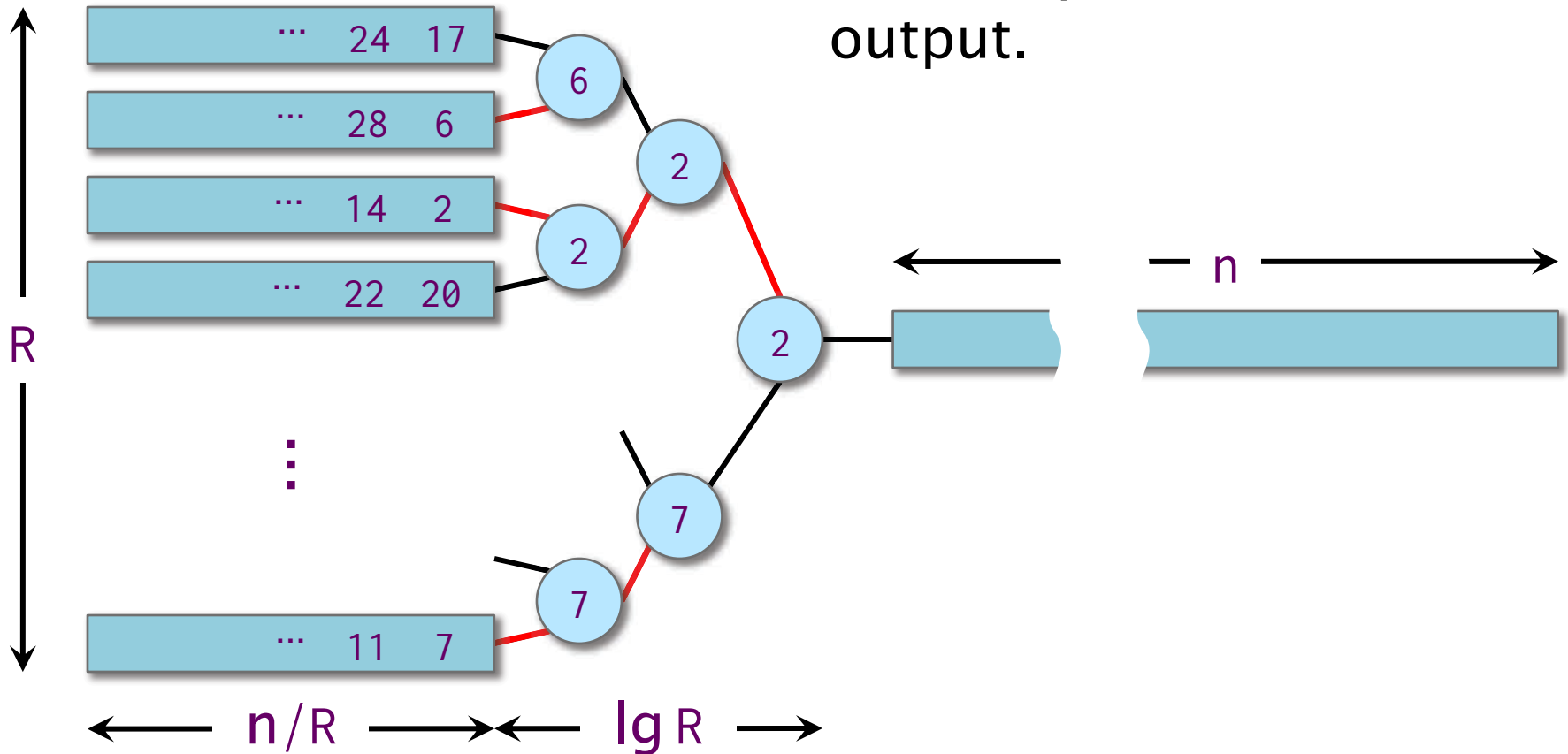
$$Q(n) = \begin{cases} \Theta(n/\mathcal{B}) & \text{if } n \leq c\mathcal{M}, \text{ constant } c \leq 1; \\ 2Q(n/2) + \Theta(n/\mathcal{B}) & \text{otherwise;} \end{cases}$$
$$= \Theta( (n/\mathcal{B}) \lg(n/\mathcal{M}) ).$$

- For  $n \gg \mathcal{M}$ , we have  $\lg(n/\mathcal{M}) \approx \lg n$ , and thus  $W(n)/Q(n) \approx \Theta(\mathcal{B})$ .
- For  $n \approx \mathcal{M}$ , we have  $\lg(n/\mathcal{M}) \approx \Theta(1)$ , and thus  $W(n)/Q(n) \approx \Theta(\mathcal{B} \lg n)$ .

# Multiway Merging

IDEA: Merge  $R < n$  subarrays with a tournament.

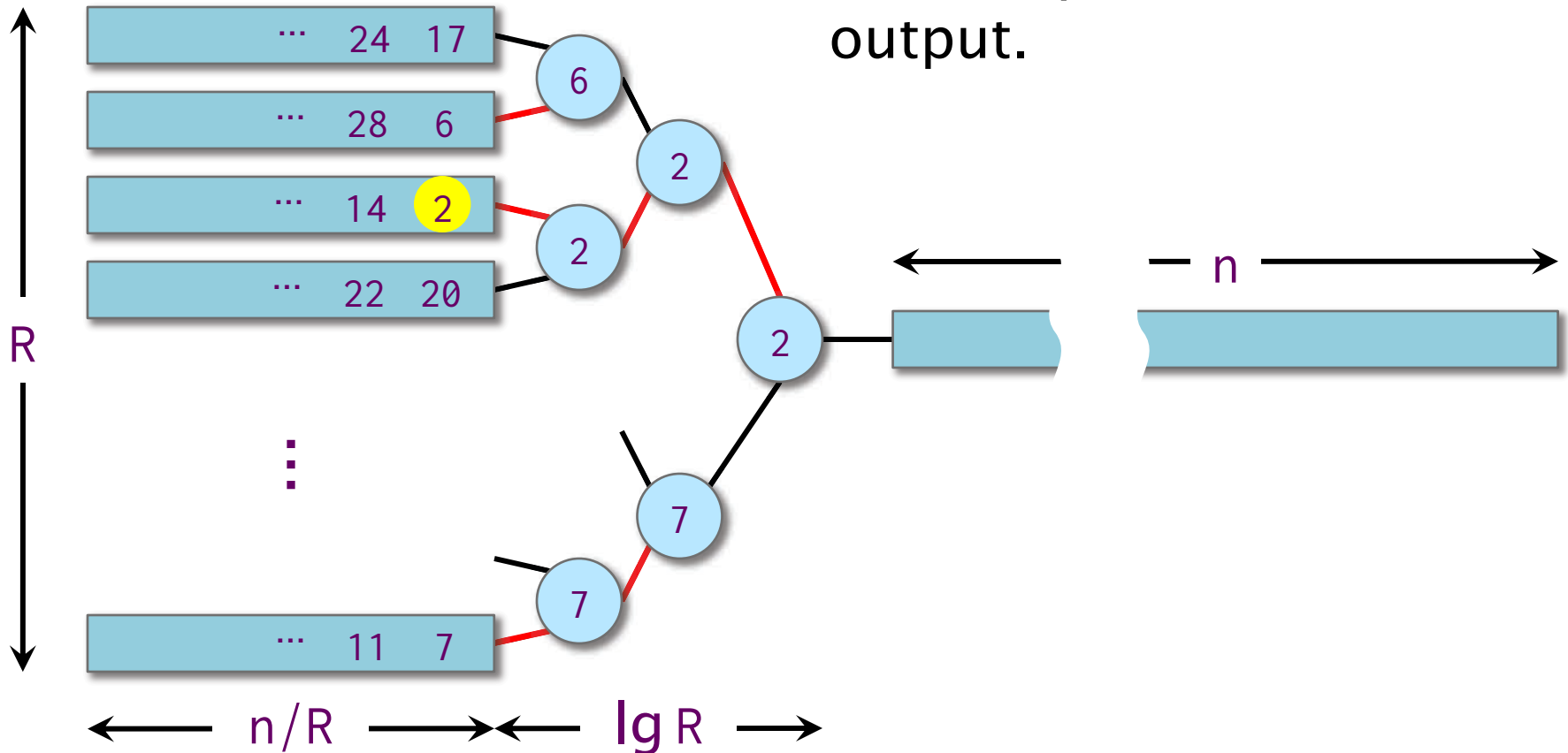
- Tournament takes  $\Theta(R)$  work to produce the first output.



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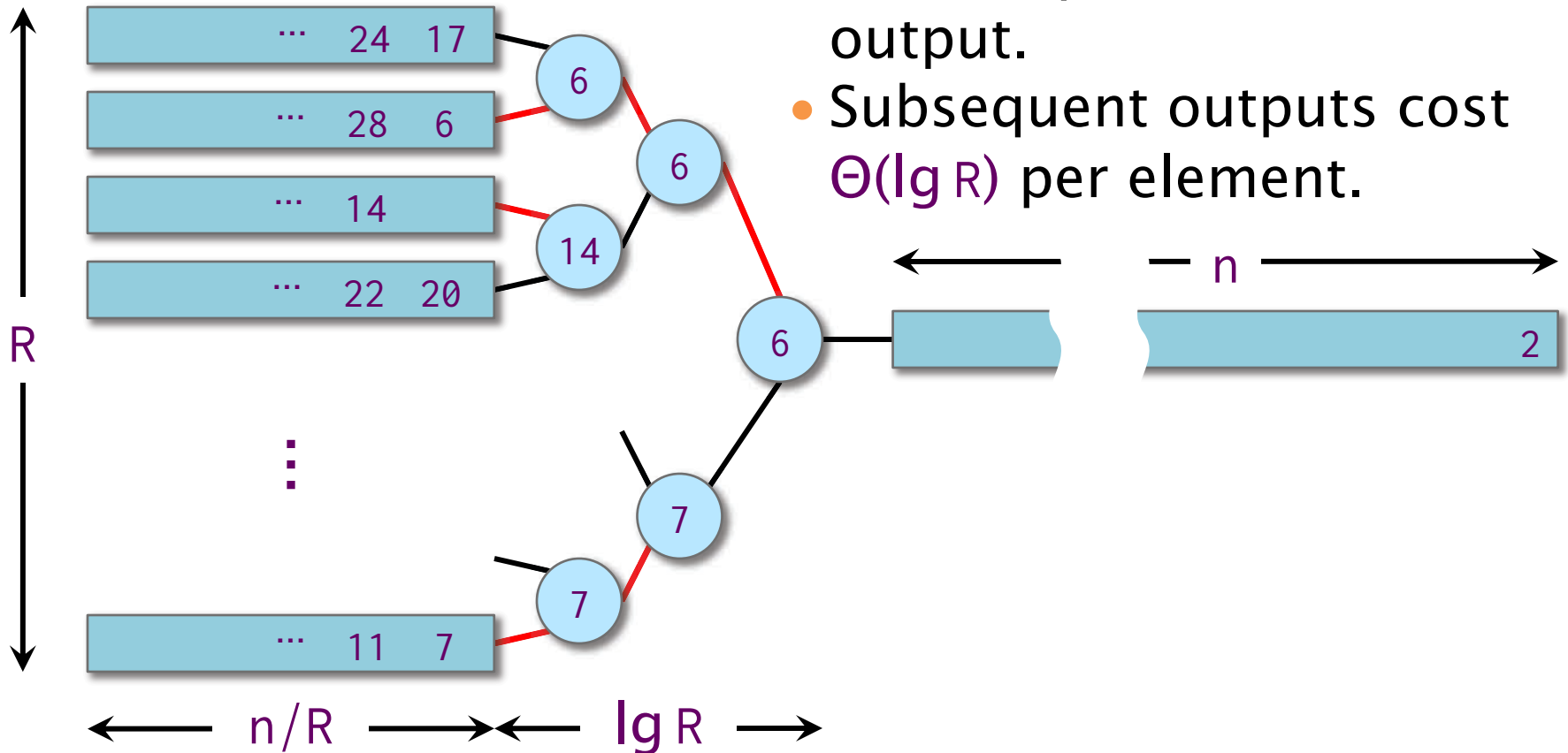
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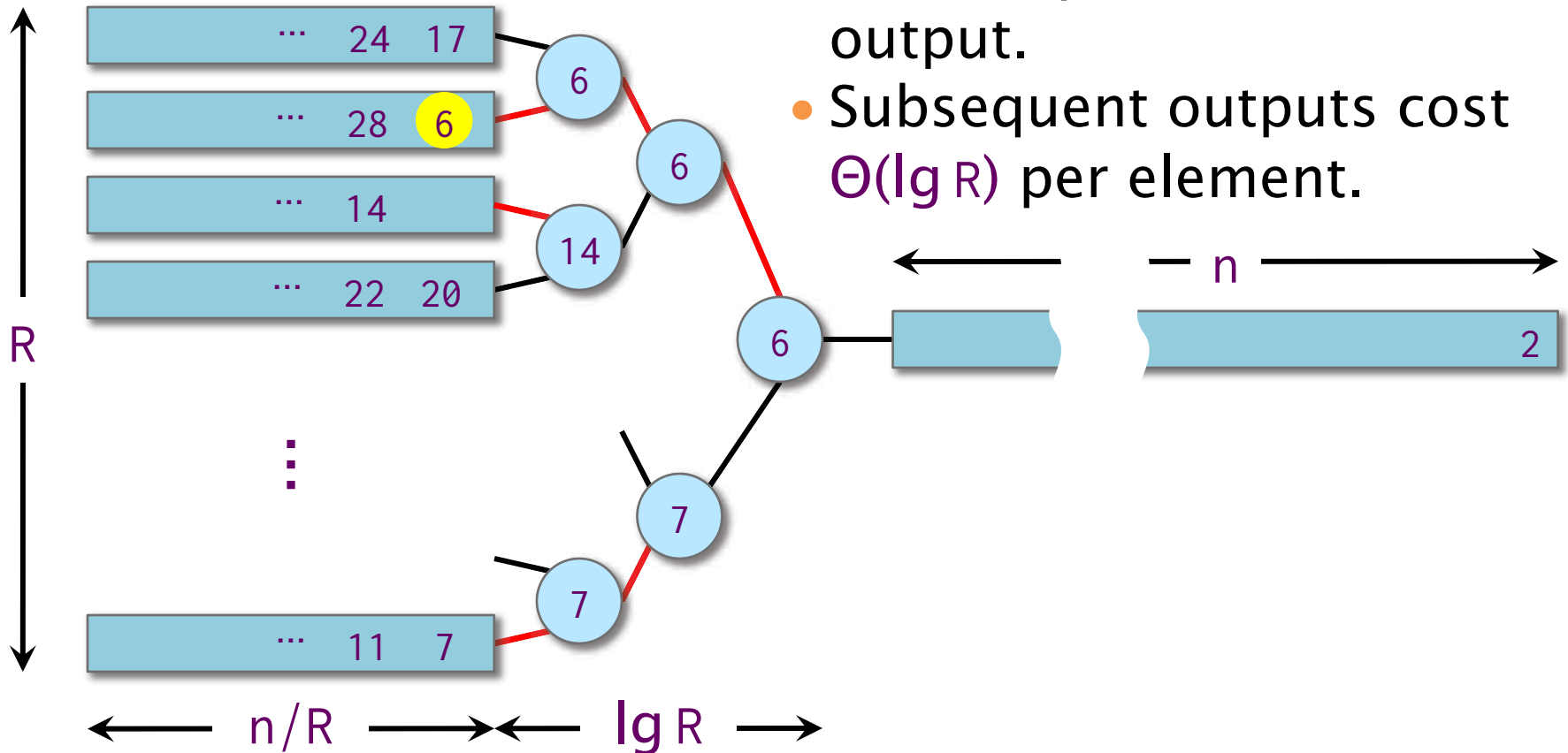
- Tournament takes  $\Theta(R)$  work to produce the first output.
- Subsequent outputs cost  $\Theta(\lg R)$  per element.



# Multiway Merging

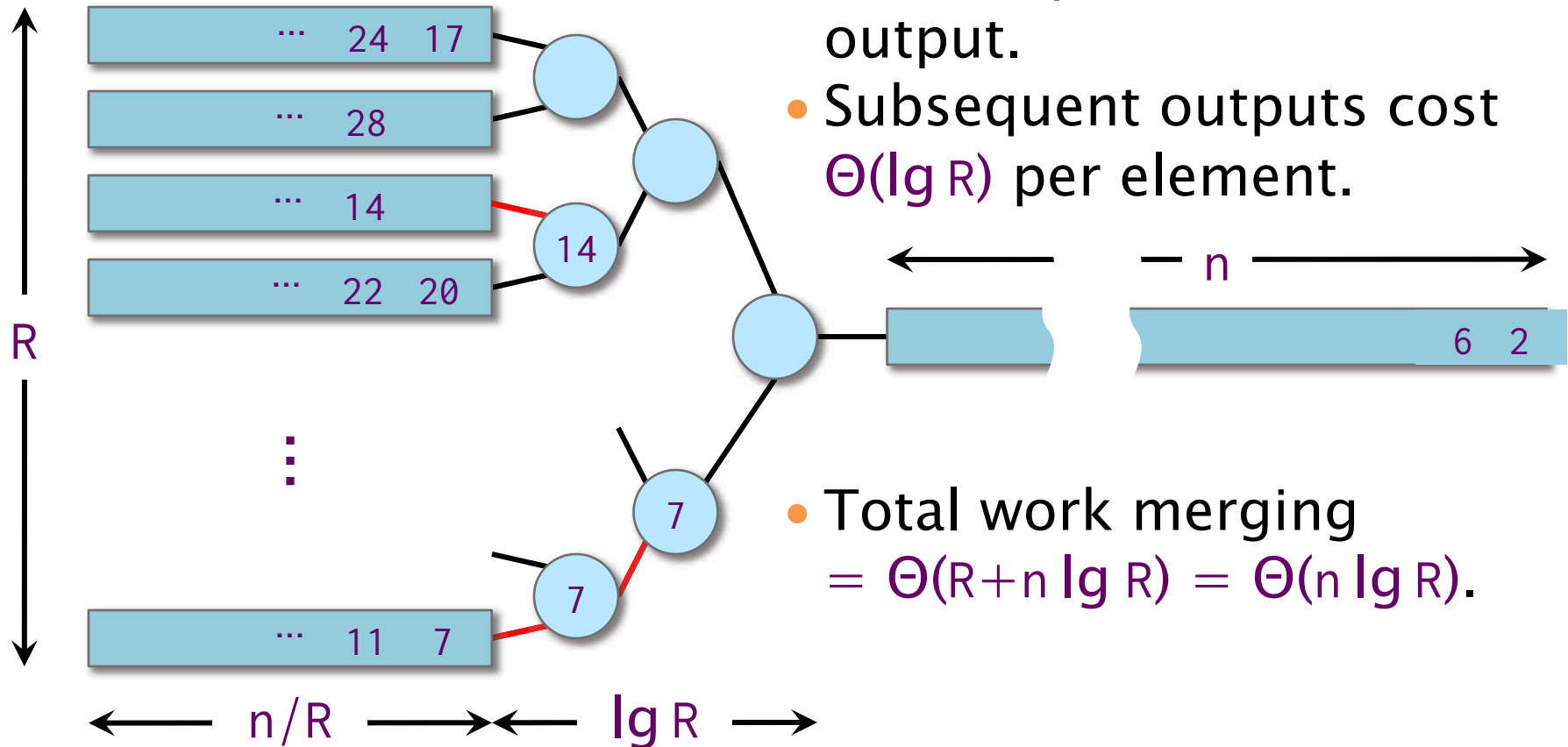
**IDEA:** Merge  $R < n$  subarrays with a tournament.

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# Multiway Merging

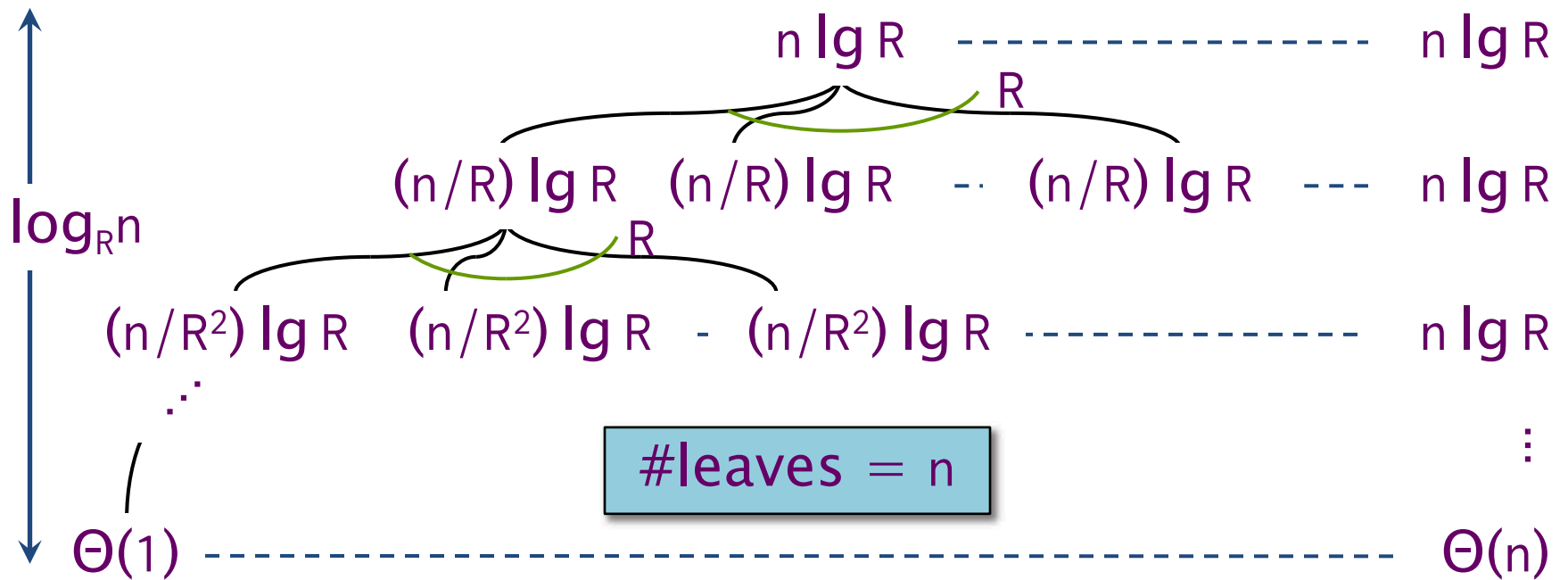
IDEA: Merge  $R < n$  subarrays with a tournament.



# Work of Multiway Merge Sort

$$W(n) = \begin{cases} \Theta(1) & \text{if } n = 1; \\ R \cdot W(n/R) + \Theta(n \lg R) & \text{otherwise.} \end{cases}$$

Recursion tree



*Same as binary merge sort.*

$$\begin{aligned} W(n) &= \Theta((n \lg R) \log_R n + n) \\ &= \Theta((n \lg R)(\lg n) / \lg R + n) \\ &= \Theta(n \lg n) \end{aligned}$$

# Caching Recurrence

Assume that we have  $R < c\mathcal{M}/\mathcal{B}$  for a sufficiently small constant  $c \leq 1$ .

Consider the  $R$ -way merging of contiguous arrays of total size  $n$ . If  $R < c\mathcal{M}/\mathcal{B}$ , the entire tournament plus 1 block from each array can fit in cache.

$\Rightarrow Q(n) \leq \Theta(n/\mathcal{B})$  for merging.

**$R$ -way merge sort**

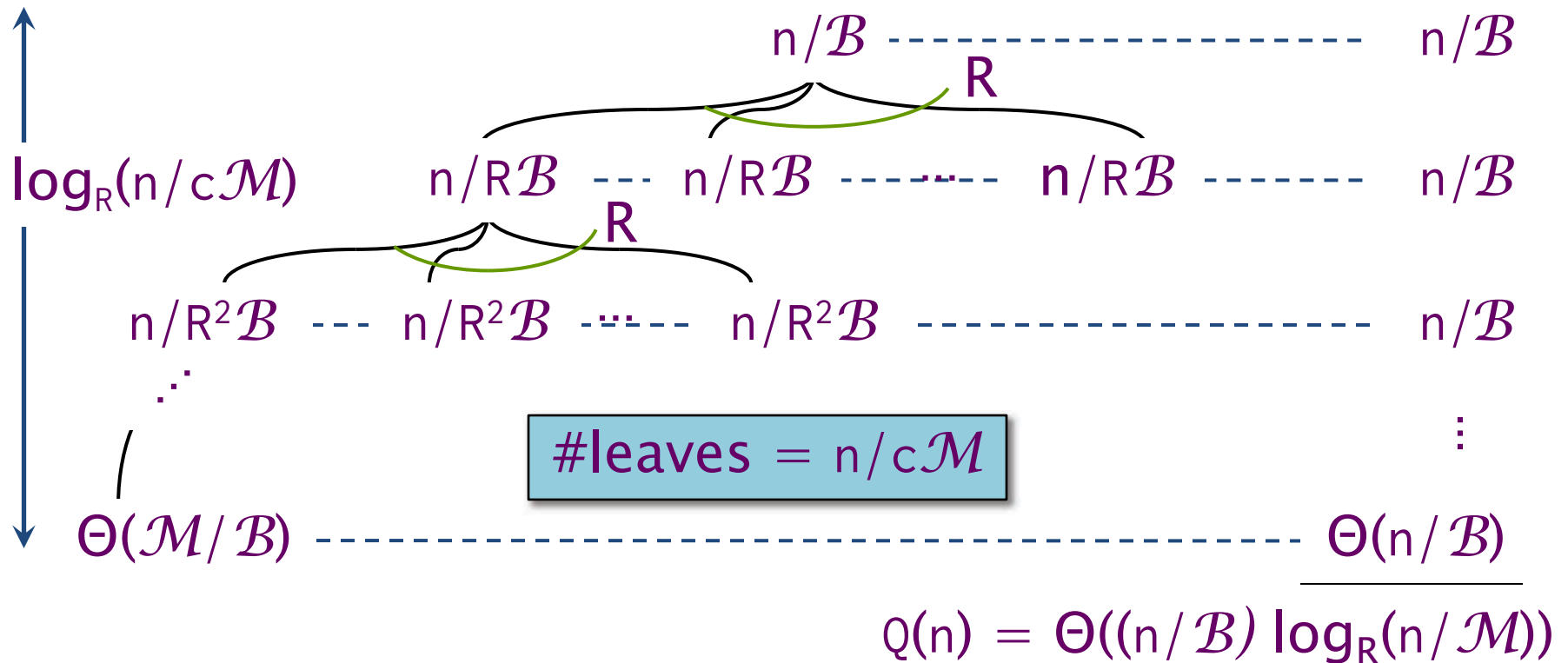
$$Q(n) \leq \begin{cases} \Theta(n/\mathcal{B}) & \text{if } n < c\mathcal{M}; \\ R \cdot Q(n/R) + \Theta(n/\mathcal{B}) & \text{otherwise.} \end{cases}$$



# Cache Analysis

$$Q(n) \leq \begin{cases} \Theta(n/\mathcal{B}) & \text{if } n < c\mathcal{M}; \\ R \cdot Q(n/R) + \Theta(n/\mathcal{B}) & \text{otherwise} \end{cases}$$

Recursion tree



# Tuning the Voodoo Parameter

We have

$$Q(n) = \Theta((n/\mathcal{B}) \log_R(n/\mathcal{M})) ,$$

which decreases as  $R \leq c\mathcal{M}/\mathcal{B}$  increases.

Choosing  $R$  as big as possible yields

$$R = \Theta(\mathcal{M}/\mathcal{B}) .$$

By the tall-cache assumption and the fact that  $\log_{\mathcal{M}}(n/\mathcal{M}) = \Theta((\lg n)/\lg \mathcal{M})$ , we have

$$\begin{aligned} Q(n) &= \Theta((n/\mathcal{B}) \log_{\mathcal{M}/\mathcal{B}}(n/\mathcal{M})) \\ &= \Theta((n/\mathcal{B}) \log_{\mathcal{M}}(n/\mathcal{M})) \\ &= \Theta((n \lg n)/\mathcal{B} \lg \mathcal{M}) . \end{aligned}$$

Hence, we have  $W(n)/Q(n) \approx \Theta(\mathcal{B} \lg \mathcal{M})$ .

# Multiway versus Binary Merge Sort

We have

$$Q_{\text{multiway}}(n) = \Theta((n \lg n) / \mathcal{B} \lg \mathcal{M})$$

versus

$$\begin{aligned} Q_{\text{binary}}(n) &= \Theta((n / \mathcal{B}) \lg(n / \mathcal{M})) \\ &= \Theta((n \lg n) / \mathcal{B}) , \end{aligned}$$

as long as  $n \gg \mathcal{M}$ , because then  $\lg(n / \mathcal{M}) \approx \lg n$ .  
Thus, multiway merge sort saves a factor of  $\Theta(\lg \mathcal{M})$  in cache misses.

**Example** (ignoring constants)

- L1-cache:  $\mathcal{M} = 2^{15} \Rightarrow 15\times$  savings.
- L2-cache:  $\mathcal{M} = 2^{18} \Rightarrow 18\times$  savings.
- L3-cache:  $\mathcal{M} = 2^{23} \Rightarrow 23\times$  savings.

# Optimal Cache-Oblivious Sorting

## Funnelsort [FLPR99]

1. Recursively sort  $n^{1/3}$  groups of  $n^{2/3}$  items.
2. Merge the sorted groups with an  $n^{1/3}$ -funnel.

A  **$k$ -funnel** merges  $k^3$  items in  $k$  sorted lists, incurring at most

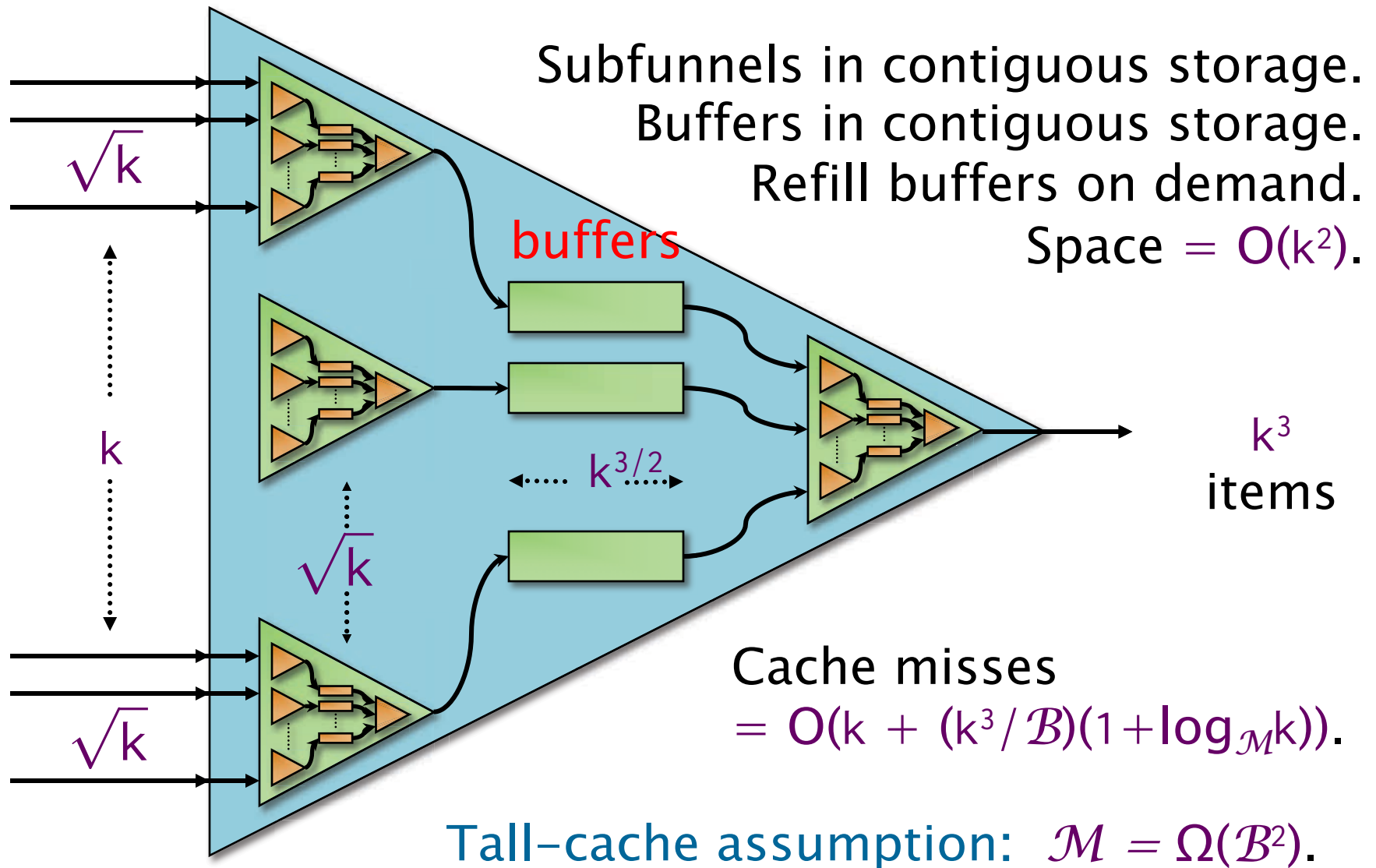
$$\Theta(k + (k^3/\mathcal{B})(1 + \log_{\mathcal{M}} k))$$

cache misses. Thus, funnelsort incurs

$$\begin{aligned} Q(n) &\leq n^{1/3}Q(n^{2/3}) + \Theta(n^{1/3} + (n/b)(1 + \log_{\mathcal{M}} n)) \\ &= \Theta(1 + (n/\mathcal{B})(1 + \log_{\mathcal{M}} n)), \end{aligned}$$

cache misses, which is asymptotically optimal [AV88].

# Construction of a $k$ -funnel



# Other C–O Algorithms

## Matrix Transposition/Addition

$$\Theta(1 + mn / \mathcal{B})$$

Straightforward recursive algorithm.

## Strassen's Algorithm

$$\Theta(n + n^2 / \mathcal{B} + n^{\lg 7} / \mathcal{B} \mathcal{M}^{(\lg 7)/2 - 1})$$

Straightforward recursive algorithm.

## Fast Fourier Transform

$$\Theta(1 + (n / \mathcal{B})(1 + \log_{\mathcal{M}} n))$$

Variant of Cooley–Tukey [CT65] using cache-oblivious matrix transpose.

## LUP–Decomposition

$$\Theta(1 + n^2 / \mathcal{B} + n^3 / \mathcal{B} \mathcal{M}^{1/2})$$

Recursive algorithm due to Sivan Toledo [T97].

# C-O Data Structures

## Ordered-File Maintenance

$$O(1 + (\lg^2 n) / \mathcal{B})$$

INSERT/DELETE or delete anywhere in file while maintaining  $O(1)$ -sized gaps. Amortized bound [BDFC00], later improved in [BCDFC02].

## B-Trees

INSERT/DELETE:	$O(1 + \log_{\mathcal{B}+1} n + (\lg^2 n) / \mathcal{B})$
SEARCH:	$O(1 + \log_{\mathcal{B}+1} n)$
TRAVERSE:	$O(1 + k / \mathcal{B})$

Solution [BDFC00] with later simplifications [BDIW02], [BFJ02].

## Priority Queues

$$O(1 + (1 / \mathcal{B}) \log_{\mathcal{M}/\mathcal{B}}(n / \mathcal{B}))$$

Funnel-based solution [BF02]. General scheme based on buffer trees [ABDHMM02] supports INSERT/DELETE.