

# Announcements

- No homework this week
- Exam 1 on Friday

# Exam 1

- In class on Friday
- Random assigned seating
- 3Qs in 45 minutes
- 6 one-sided pages of notes
- No textbook or electronic aids
- On Chapters 3 & 4
- Review video online

# What constitutes a proof for this class?

- Needs to be a proof:
  - Establish result from known premises
  - Probably consists of English sentences
  - Might be quite short depending on problem
- How much detail to give?
  - Hard to spell out exactly
  - Rule of thumb: enough to convince the grader that you *could* fill in further details if pressed

# Last Time

- Divide and Conquer
- Karatsuba Multiplication
- Master Theorem

# Divide and Conquer

This is the first of our three major algorithmic techniques.

1. Break problem into pieces
2. Solve pieces recursively
3. Recombine pieces to get answer

# Karatsuba Multiplication

KaratsubaMult(N,M)

If  $N+M < 99$ , Return Product(N,M)

Let X be a power of  $2^{\lfloor \log(N+M)/2 \rfloor}$

Write  $N = AX + B$ ,  $M = CX + D$

$P_1 \leftarrow \text{KaratsubaMult}(A, C)$

$P_2 \leftarrow \text{KaratsubaMult}(B, D)$

$P_3 \leftarrow \text{KaratsubaMult}(A+B, C+D)$

Return  $P_1X^2 + [P_3 - P_1 - P_2]X + P_2$

# Runtime Recurrence

Karatsuba multiplication on inputs of size  $n$  spends  $O(n)$  time, and then makes three recursive calls to problems of (approximately) half the size.

If  $T(n)$  is the runtime for  $n$ -bit inputs, we have the recursion:

$$T(n) = \begin{cases} O(1) & \text{if } n = O(1) \\ 3T(n/2 + O(1)) + O(n) & \text{otherwise} \end{cases}$$

# Generalization

We will often get runtime recurrences with D&C looking something like this:

$$T(n) = O(1) \text{ for } n = O(1)$$

$$T(n) = a T(n/b + O(1)) + O(n^d) \text{ otherwise.}$$



# Master Theorem

**Theorem:** Let  $T(n)$  be given by the recurrence:

$$T(n) = \begin{cases} O(1) & \text{if } n = O(1) \\ aT(n/b + O(1)) + O(n^d) & \text{otherwise} \end{cases}$$

Then we have that

$$T(n) = \begin{cases} O(n^{\log_b(a)}) & \text{if } a > b^d \\ O(n^d \log(n)) & \text{if } a = b^d \\ O(n^d) & \text{if } a < b^d \end{cases}$$

# Karatsuba Runtime

$$T(n) = \begin{cases} O(1) & \text{if } n = O(1) \\ 3T(n/2 + O(1)) + O(n) & \text{otherwise} \end{cases}$$

- $a = 3, b = 2, d = 1$
- $a > b^d$

Runtime:  $O(n^{\log_2(3)}) = O(n^{1.585\dots})$ .

Better than easy  $O(n^2)$  algorithm!

# Question: Runtimes

Suppose that a divide and conquer algorithm needs to solve 4 recursive subproblems of half the size and do  $O(n^2)$  additional work. What is the runtime?

- A)  $O(n)$
- B)  $O(n \log(n))$
- C)  $O(n^2)$
- D)  $O(n^2 \log(n))$
- E)  $O(n^3)$

# Question: Runtimes

Suppose that a divide and conquer algorithm needs to solve 4 recursive subproblems of half the size and do  $O(n^2)$  additional work. What is the runtime?

A)  $O(n)$

$a = 4, b = 2, d = 2.$

B)  $O(n \log(n))$

$a = b^d.$

C)  $O(n^2)$

So runtime is  $O(n^d \log(n))$

D)  $O(n^2 \log(n))$

E)  $O(n^3)$

# Today

- Strassen's algorithm
- Mergesort
- Order statistics

# Note

In divide and conquer, it is important that the recursive subcalls are a constant *fraction* of the size of the original.

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For example, if we have

$$T(n) = 2T(n-1)$$

then  $T(n) = O(2^n)$ .

# Matrix Multiplication

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Can we do better?

# Block Matrix Multiplication

If you divide the matrix into blocks, you can get the product of the full matrix in terms of products of the blocks.

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} E & F \\ G & H \end{bmatrix} = \begin{bmatrix} AE + BG & AF + BH \\ CE + DG & CF + DH \end{bmatrix}.$$

Here, A,B,C,... are  $(n/2) \times (n/2)$  matrices.

# Easy D&C

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} E & F \\ G & H \end{bmatrix} = \begin{bmatrix} AE + BG & AF + BH \\ CE + DG & CF + DH \end{bmatrix}.$$

Compute 8 products of  $(n/2) \times (n/2)$  matrices, and do some addition to get answer.

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Compute 8 products of  $(n/2) \times (n/2)$  matrices, and do some addition to get answer.

$$T(n) = 8T(n/2) + O(n^2).$$

Master Theorem:  $a = 8$ ,  $b = 2$ ,  $d = 2$ .

$a > b^d$  so  $O(n^C)$  with  $C = \log_2 8 = 3$ .  $O(n^3)$ .

# Strassen's Algorithm

Instead compute:

- $M_1 = (A+D)(E+H)$
- $M_2 = (C+D)E$
- $M_3 = A(F-H)$
- $M_4 = D(G-E)$
- $M_5 = (A+B)H$
- $M_6 = (C-A)(E+F)$
- $M_7 = (B-D)(G+H)$

Entries of product matrix are:

- $M_1 + M_4 - M_5 + M_7$
- $M_3 + M_5$
- $M_2 + M_4$
- $M_1 - M_2 + M_3 + M_6$

How do you figure this out?

Don't ask me. Magic, probably.



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Improvement for very large matrices.

Best known algorithm:  $O(n^{2.376...})$  very impractical.

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**Base Case:** Your base case will be the non-recursive case of your algorithm (which your algorithm does need to have).

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**Use Induction:** Prove correctness by induction on problem size.

**Base Case:** Your base case will be the non-recursive case of your algorithm (which your algorithm does need to have).

**Inductive Step:** Assuming that the (smaller) recursive calls are correct, show that algorithm works.



# Sorting

**Problem:** Given a list of  $n$  numbers, return those numbers in ascending order.

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**Example:**

Input: {0, 5, 2, 7, 4, 6, 3, 1}

Output: {0, 1, 2, 3, 4, 5, 6, 7}

# Divide and Conquer

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1. Split into Subproblems      - Split list in two
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How do we make a divide and conquer algorithm?

1. Split into Subproblems
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  - Sort each sublist
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  - ???

# Merge

**Problem:** Given two sorted lists, combine them into a single sorted list.



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We want something that takes advantage of the individual lists being sorted.

# Merge Idea

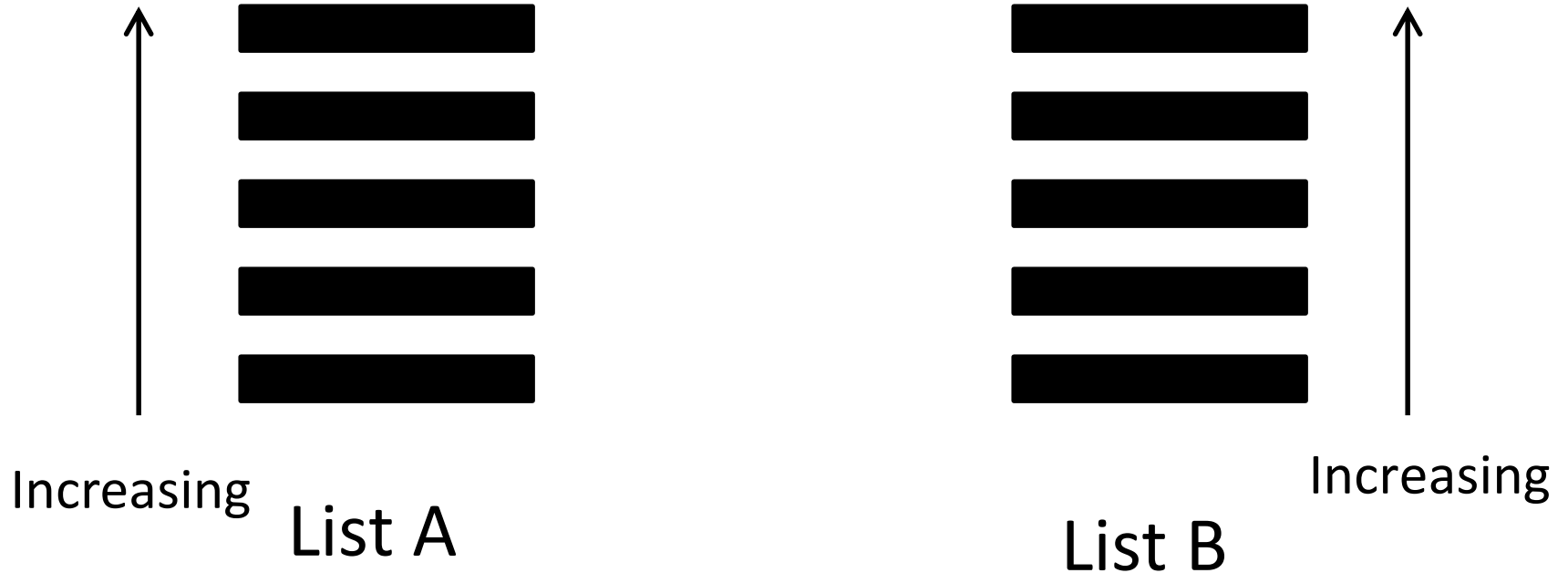


List A

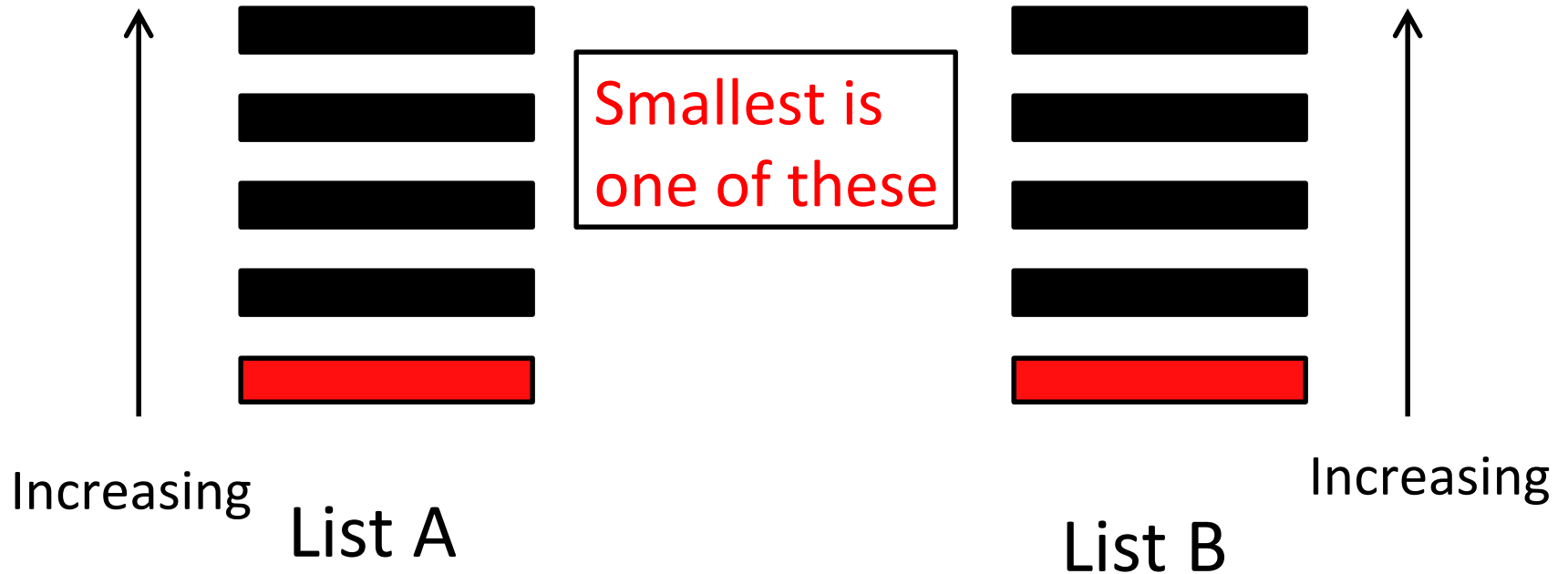


List B

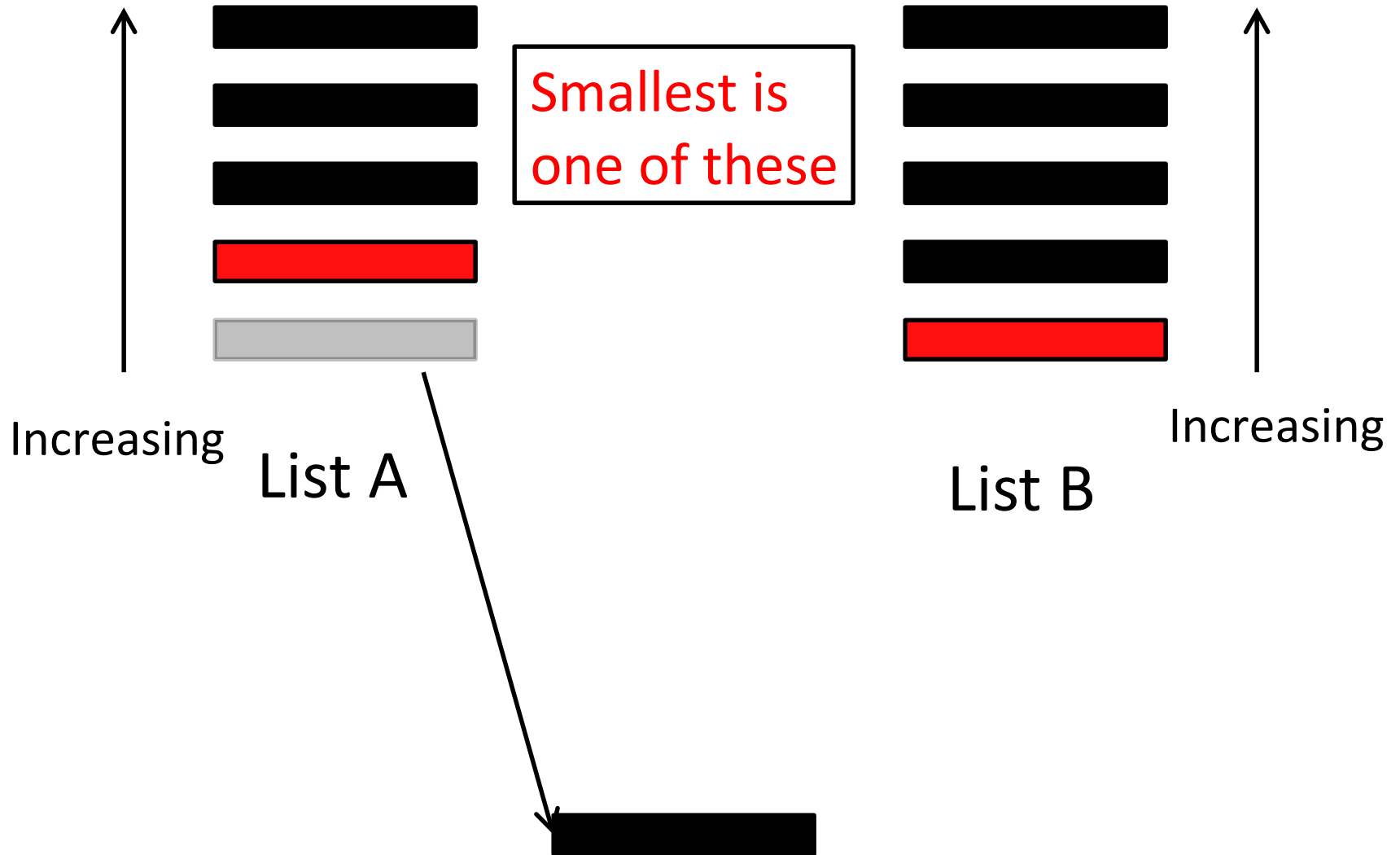
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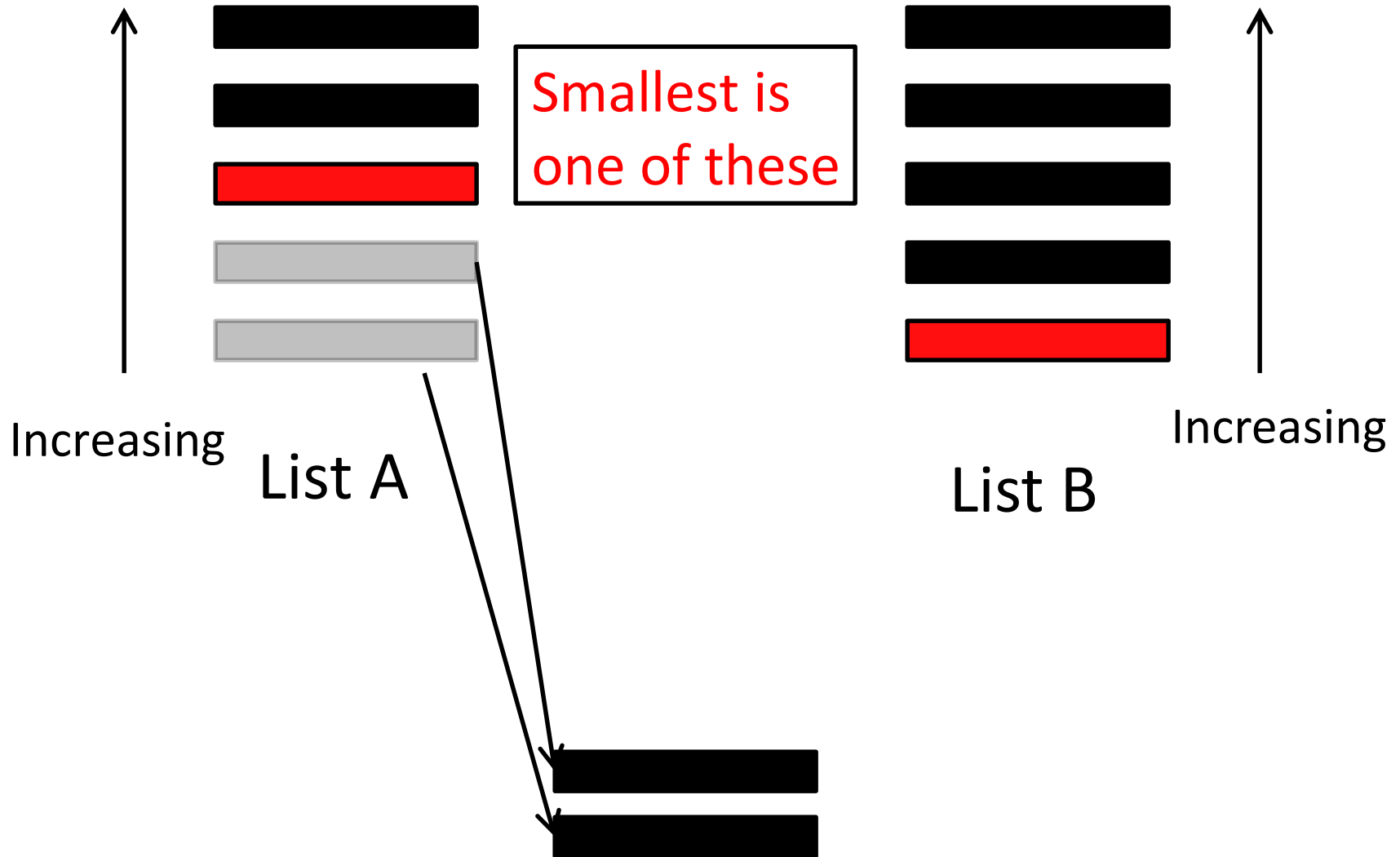
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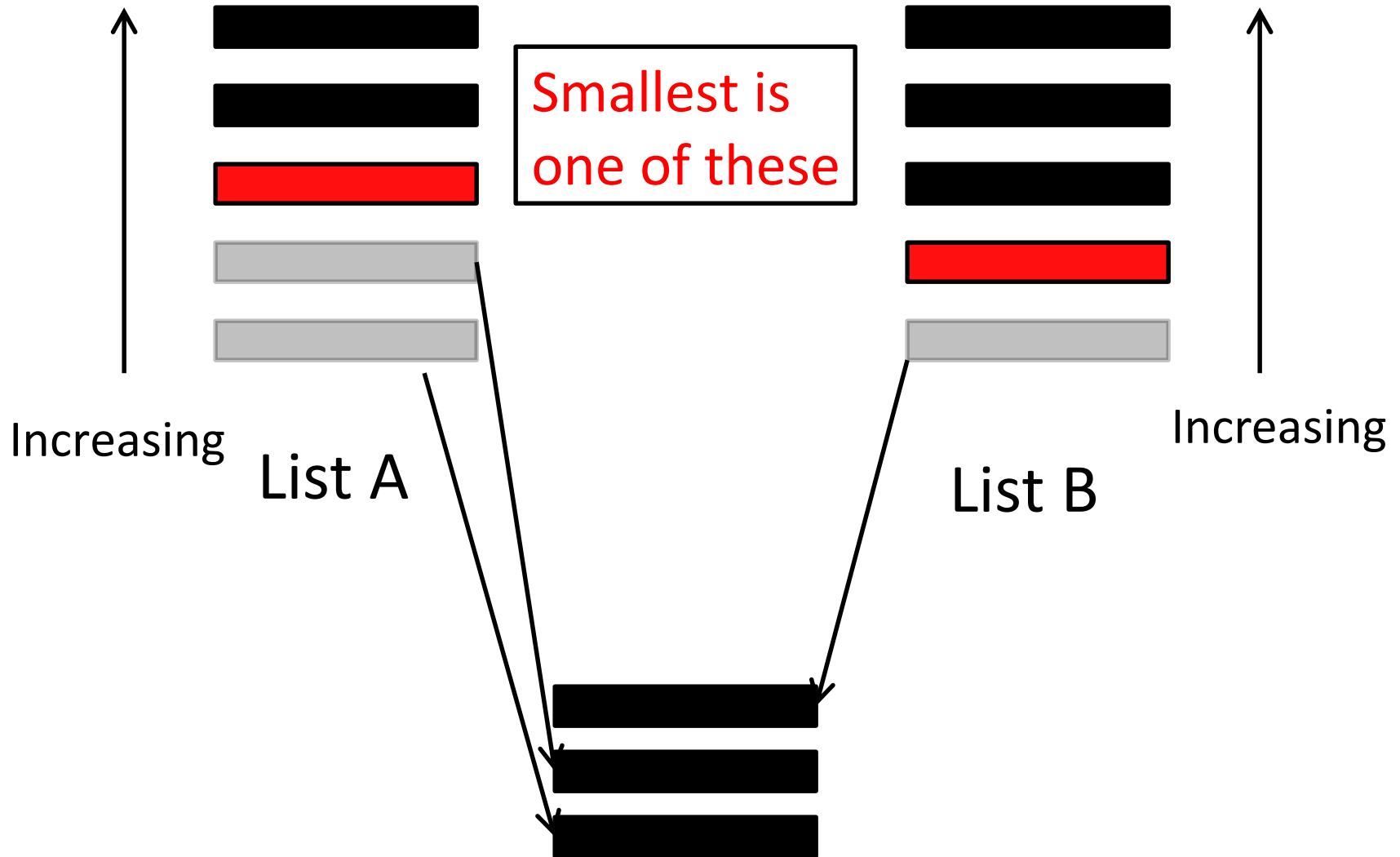
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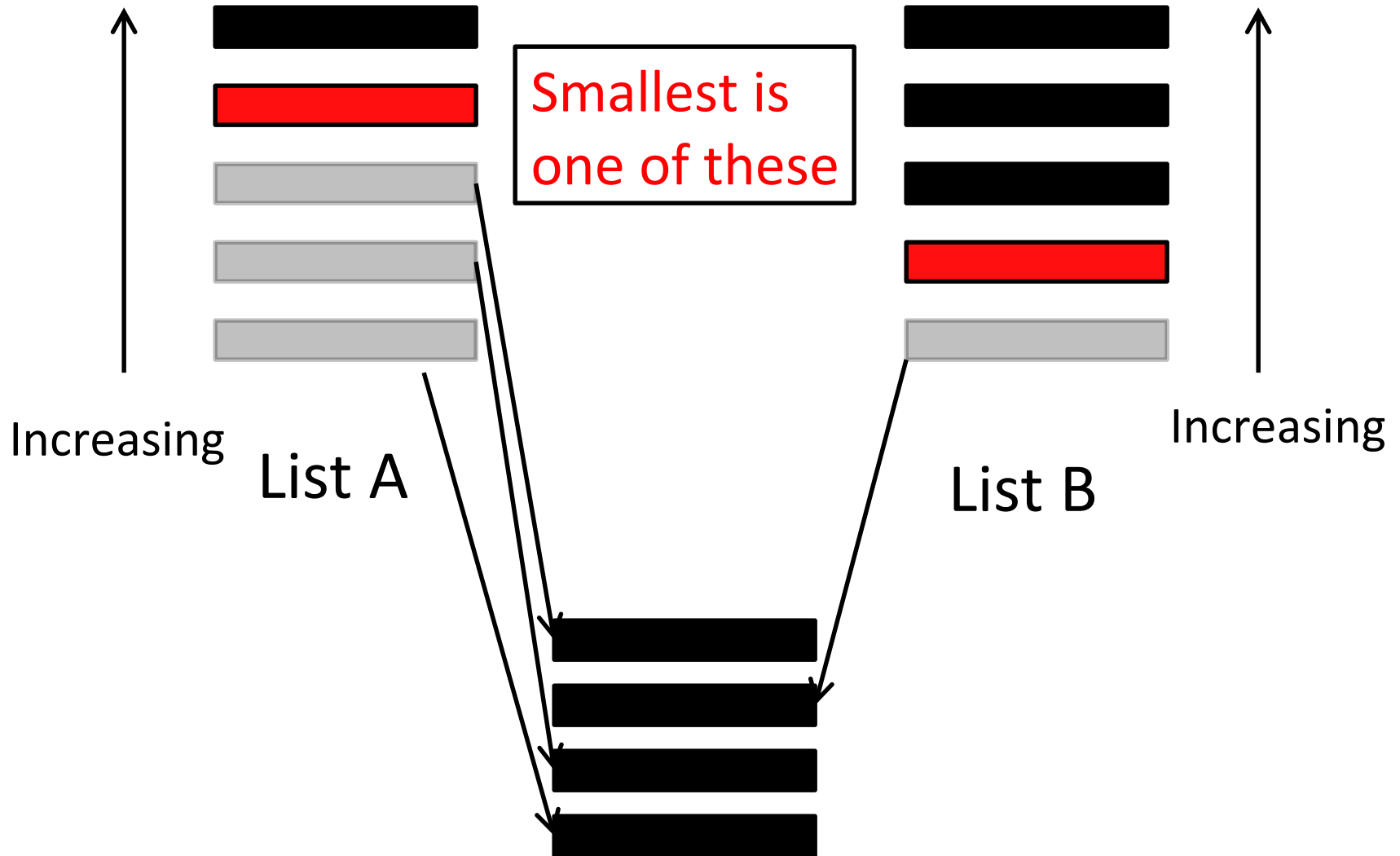
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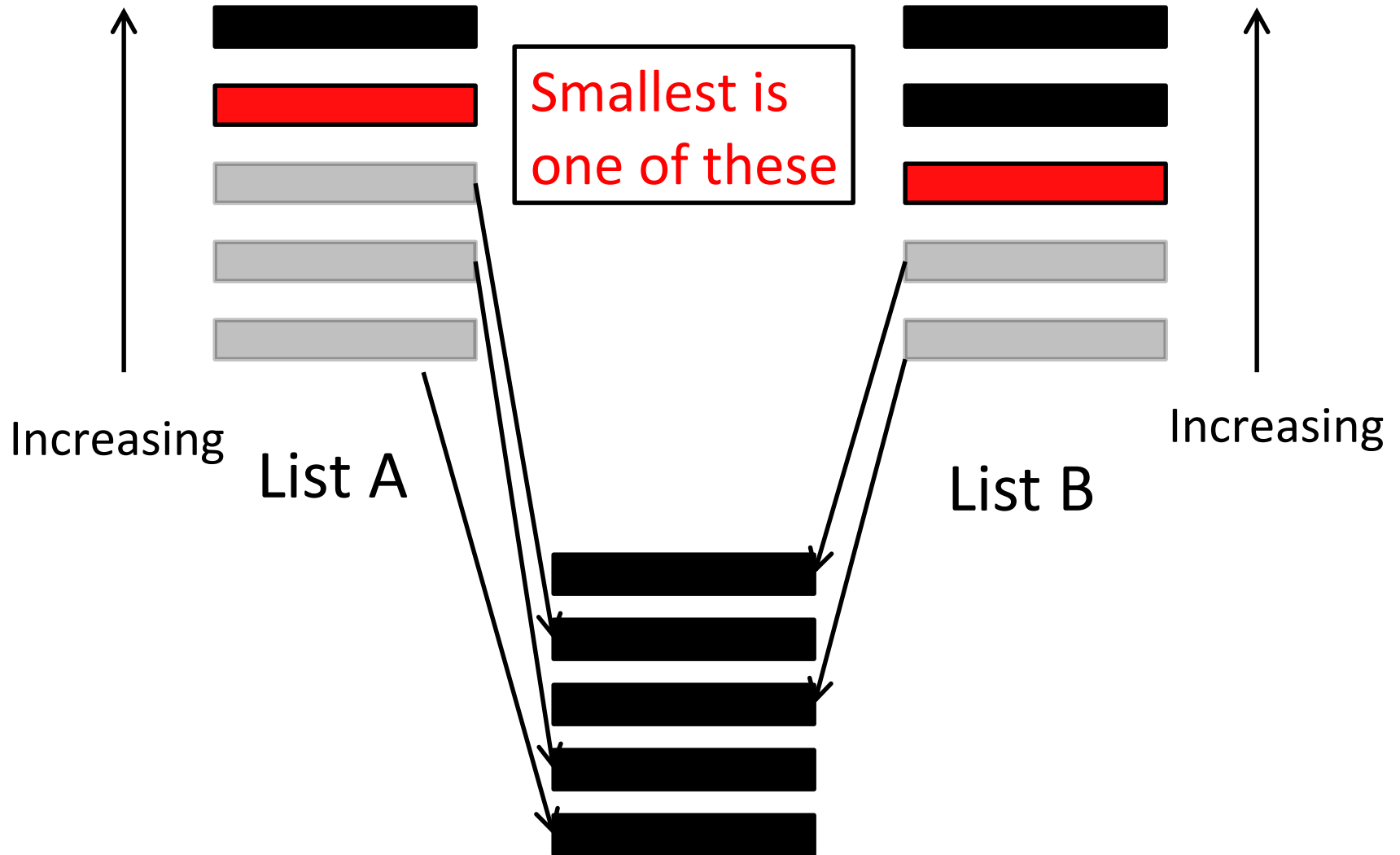


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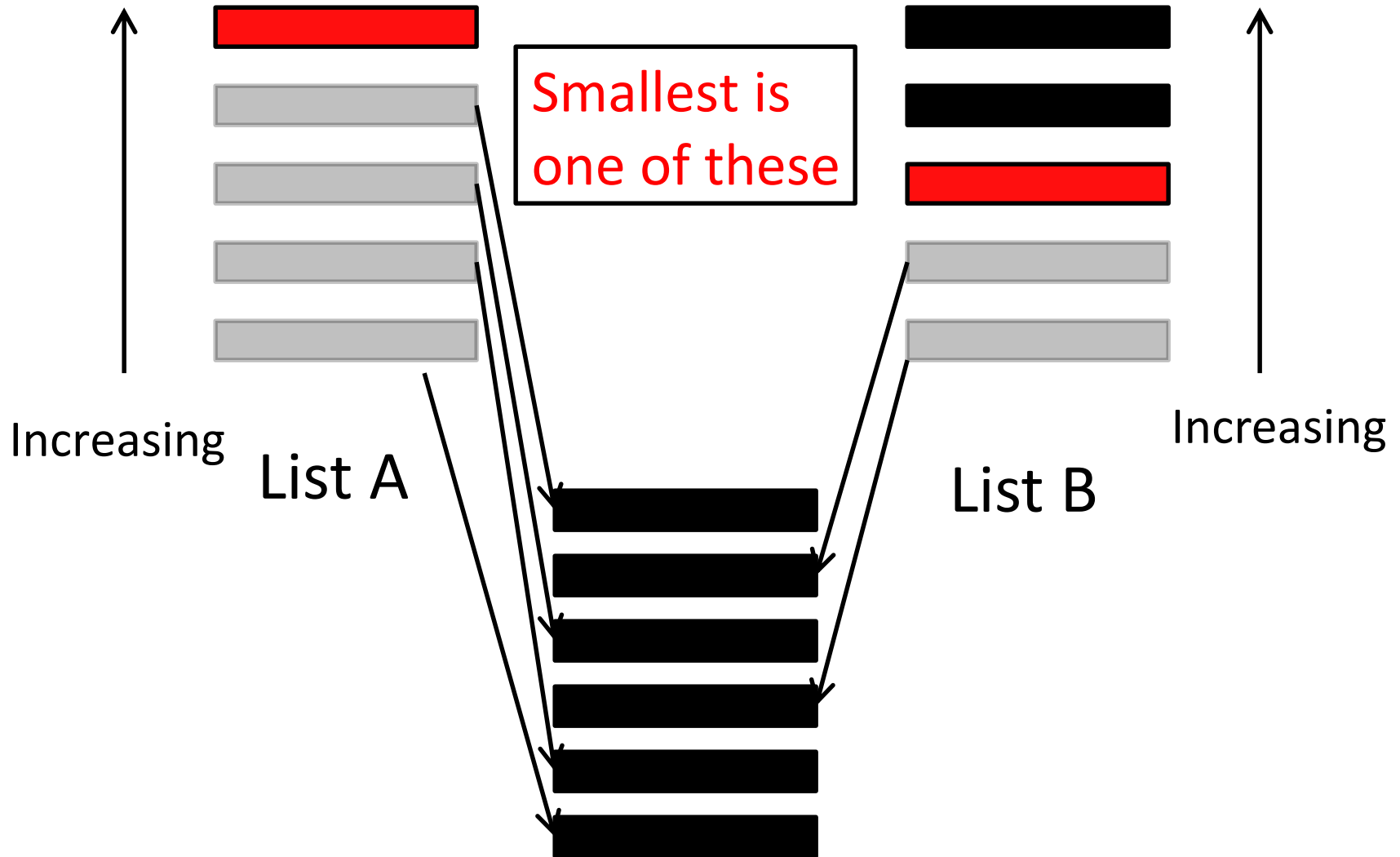




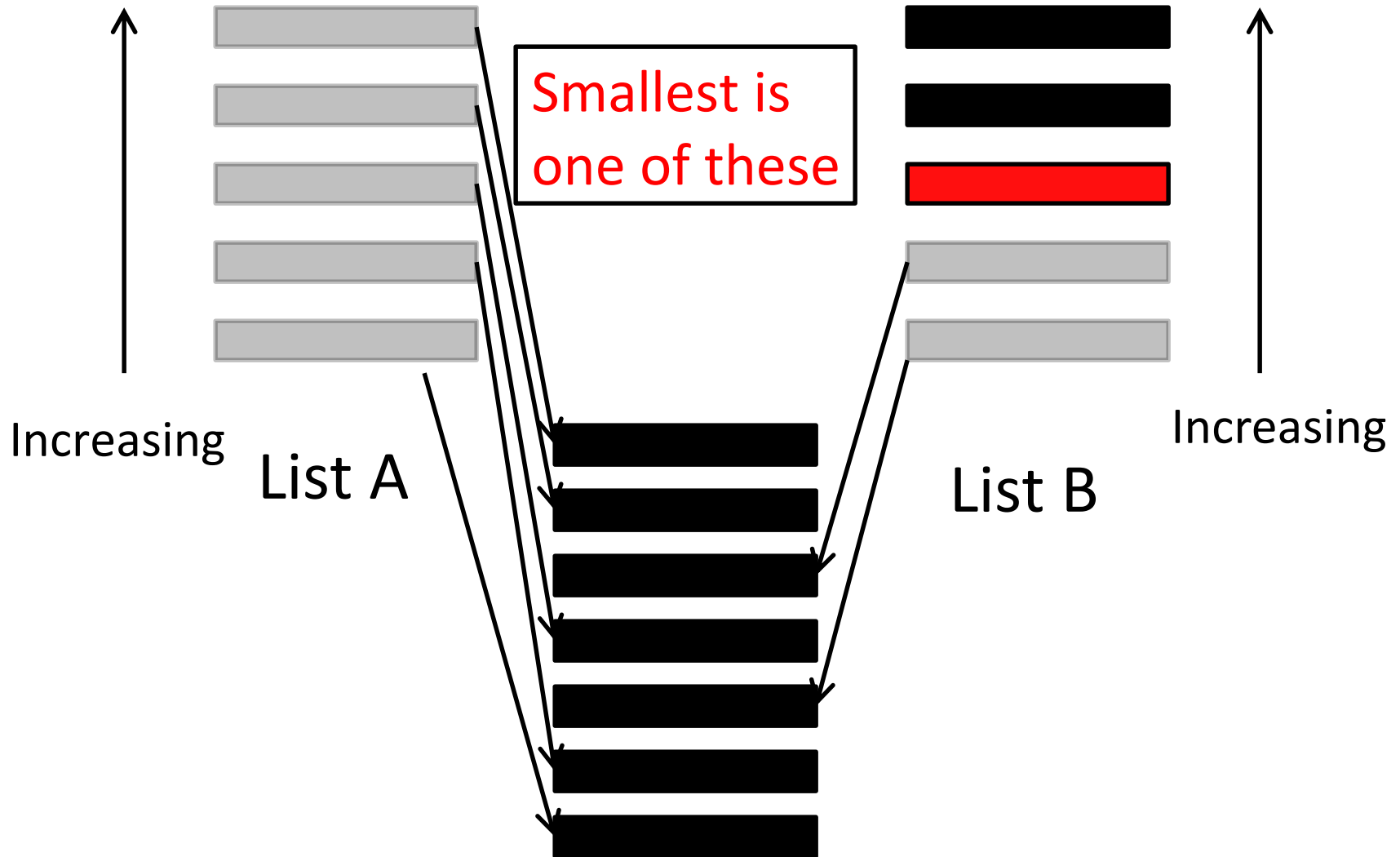
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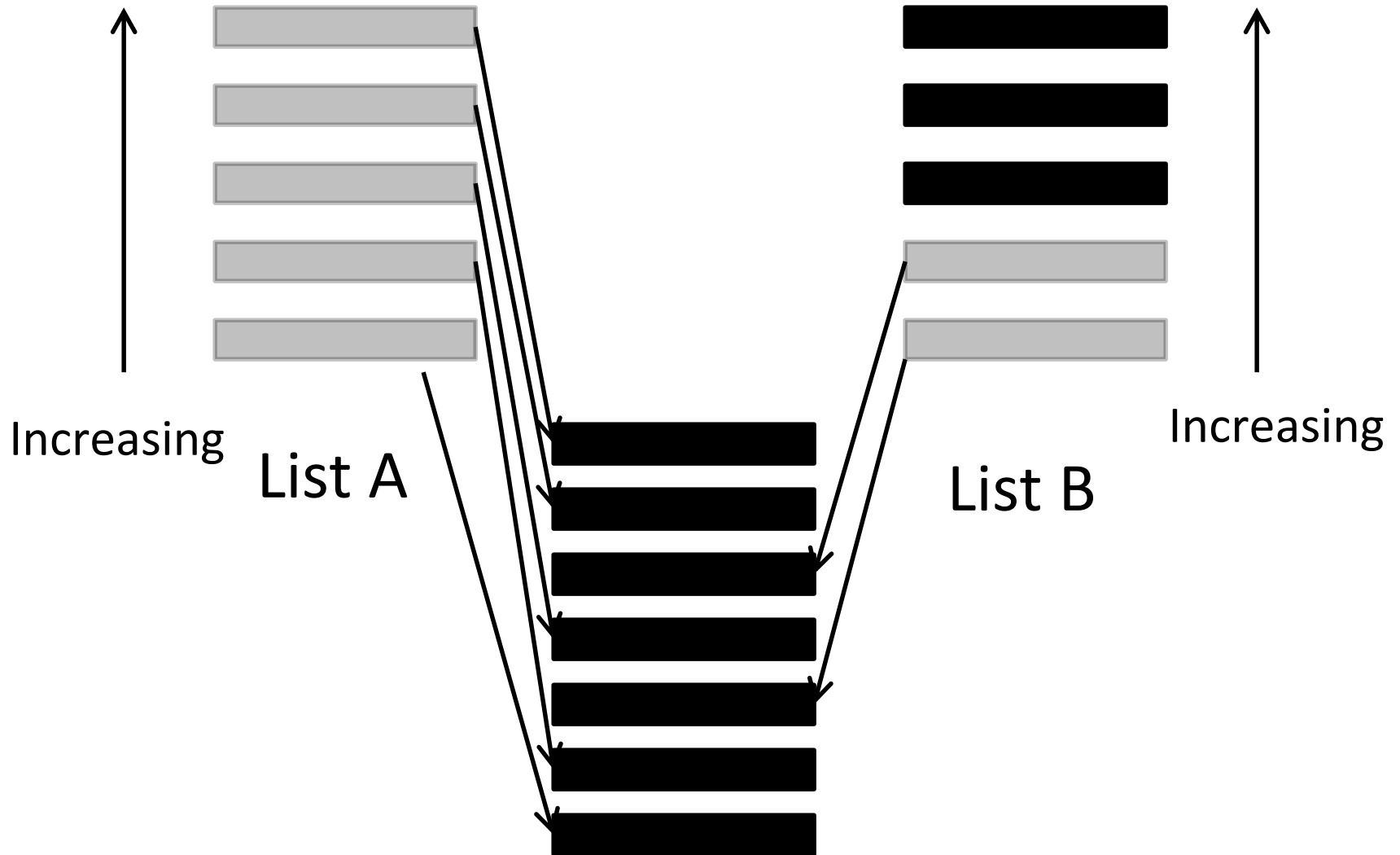
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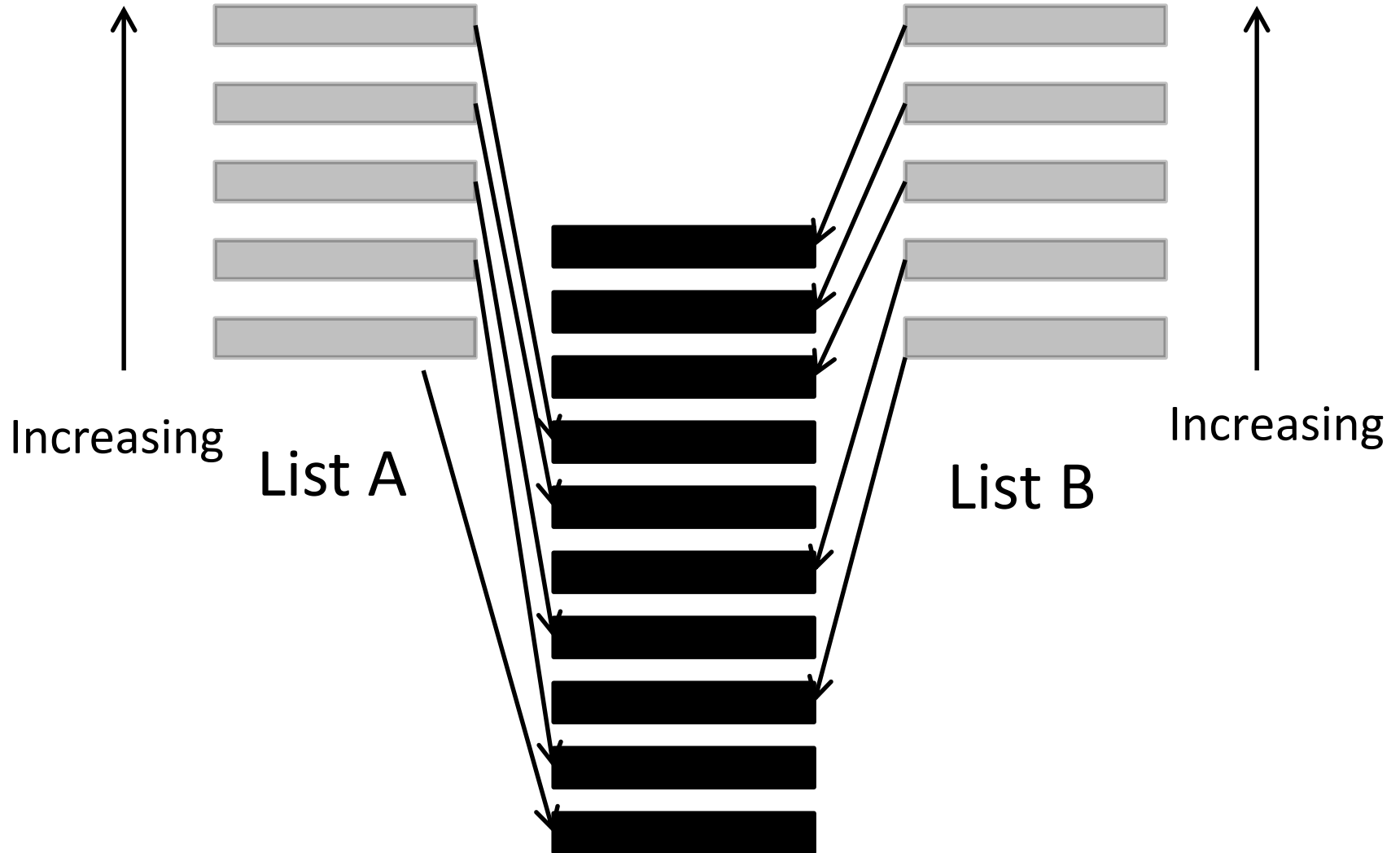
# Merge Idea



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# Merge

```
Merge(A,B)
```

```
  C ← List of length Len(A)+Len(B)
```

```
  a ← 1, b ← 1
```

```
  For c = 1 to Len(C)
```

```
    If (b > Len(B))
```

```
      C[c] ← A[a], a++
```

```
    Else if (a > Len(A))
```

```
      C[c] ← B[b], b++
```

```
    Else if A[a] < B[b]
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    Else
```

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      C[c] ← B[b], b++
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  Return C
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```

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      C[c] ← B[b], b++
```

```
  Return C
```

Runtime:  $O(|A| + |B|)$

# MergeSort

```
MergeSort(L)
```

```
  If Len(L) = 1          \ \ Base Case
```

```
    Return L
```

```
  Split L into equal  $L_1$  and  $L_2$ 
```

```
  A  $\leftarrow$  MergeSort( $L_1$ )
```

```
  B  $\leftarrow$  MergeSort( $L_2$ )
```

```
  Return Merge(A, B)
```



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$2T(n/2)$

$O(n)$

# Question: Runtime

What is the runtime of MergeSort?

- A)  $O(n)$
- B)  $O(n \log(n))$
- C)  $O(n^2)$
- D)  $O(n^{\log(3)})$
- E)  $O(n^2 \log(n))$

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Master Theorem:

$$a = 2, b = 2, d = 1$$

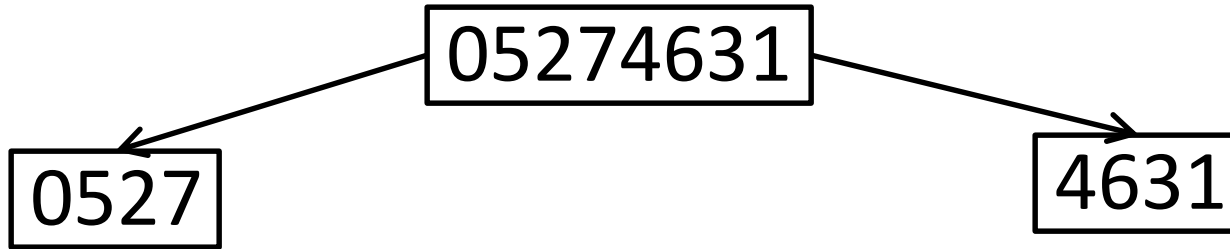
$$a = b^d$$

$$O(n^d \log(n)) = O(n \log(n))$$

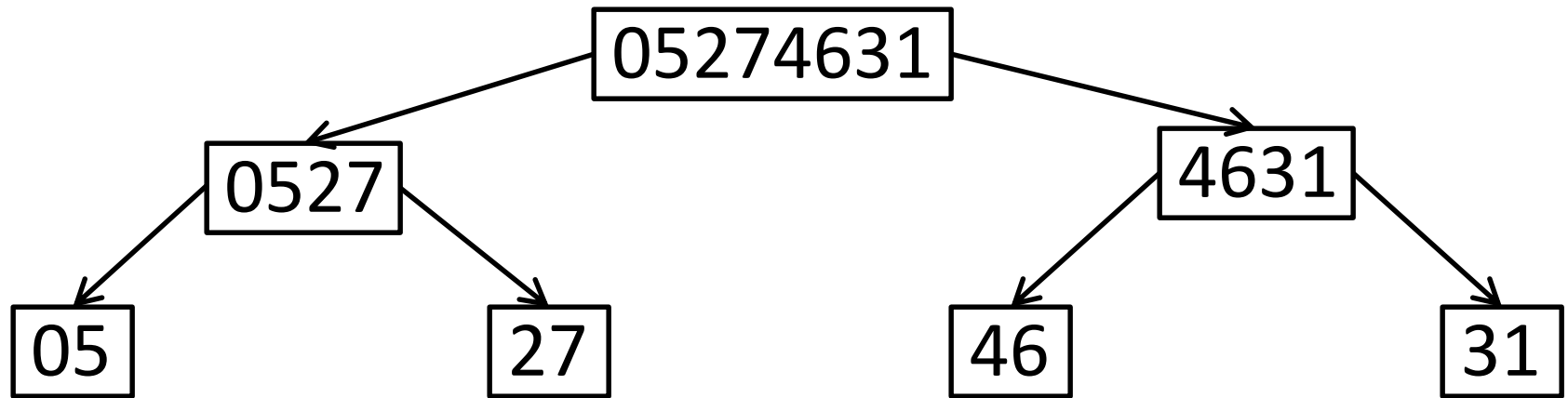
# Example

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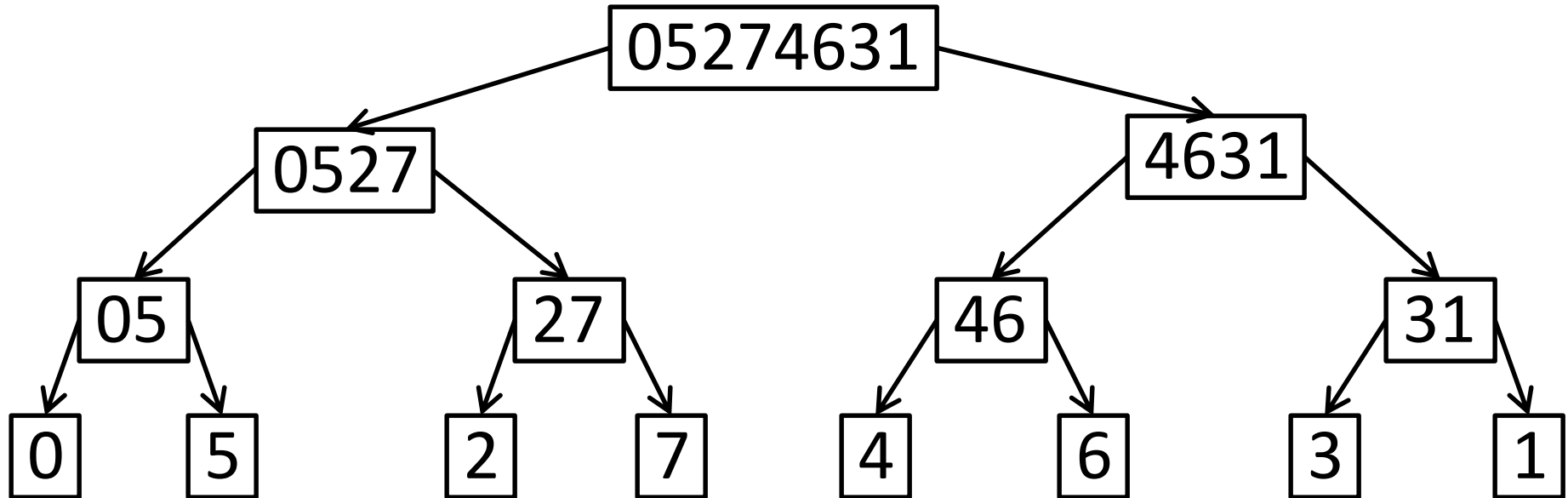
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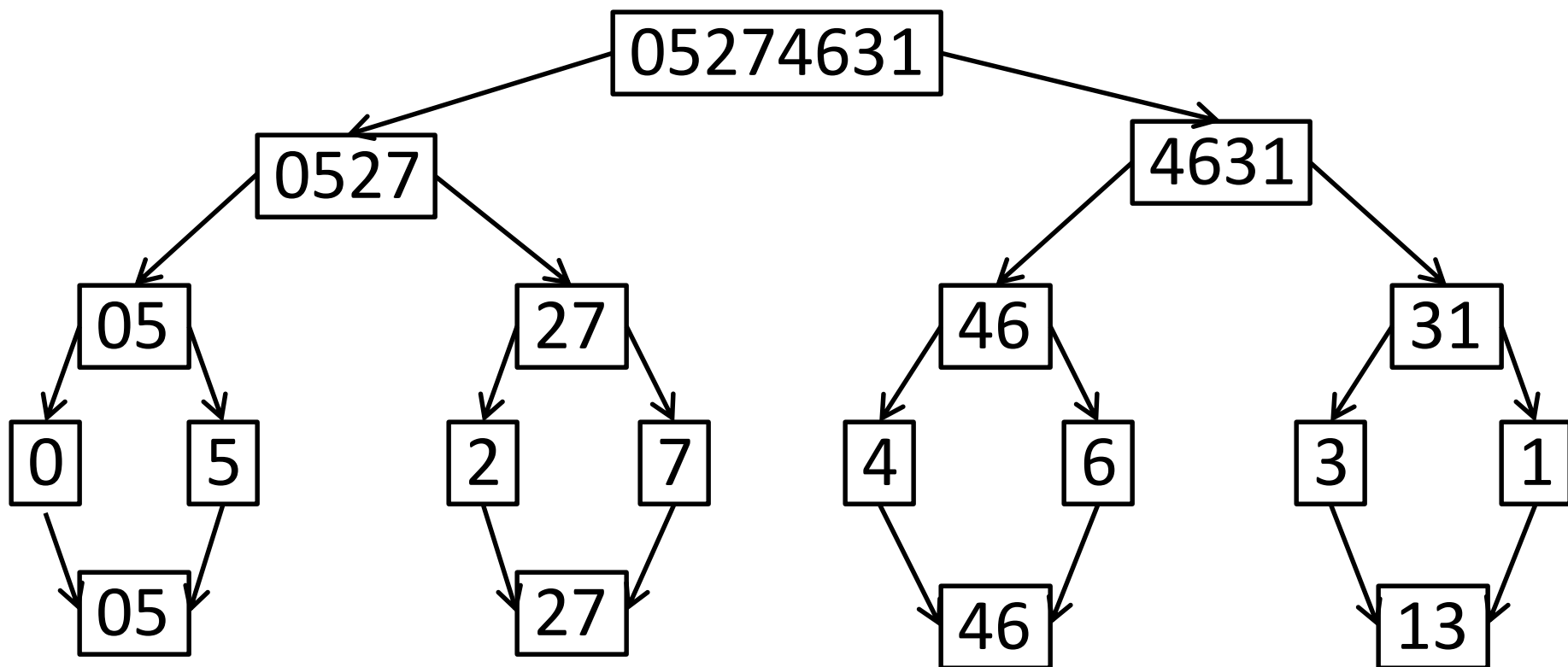


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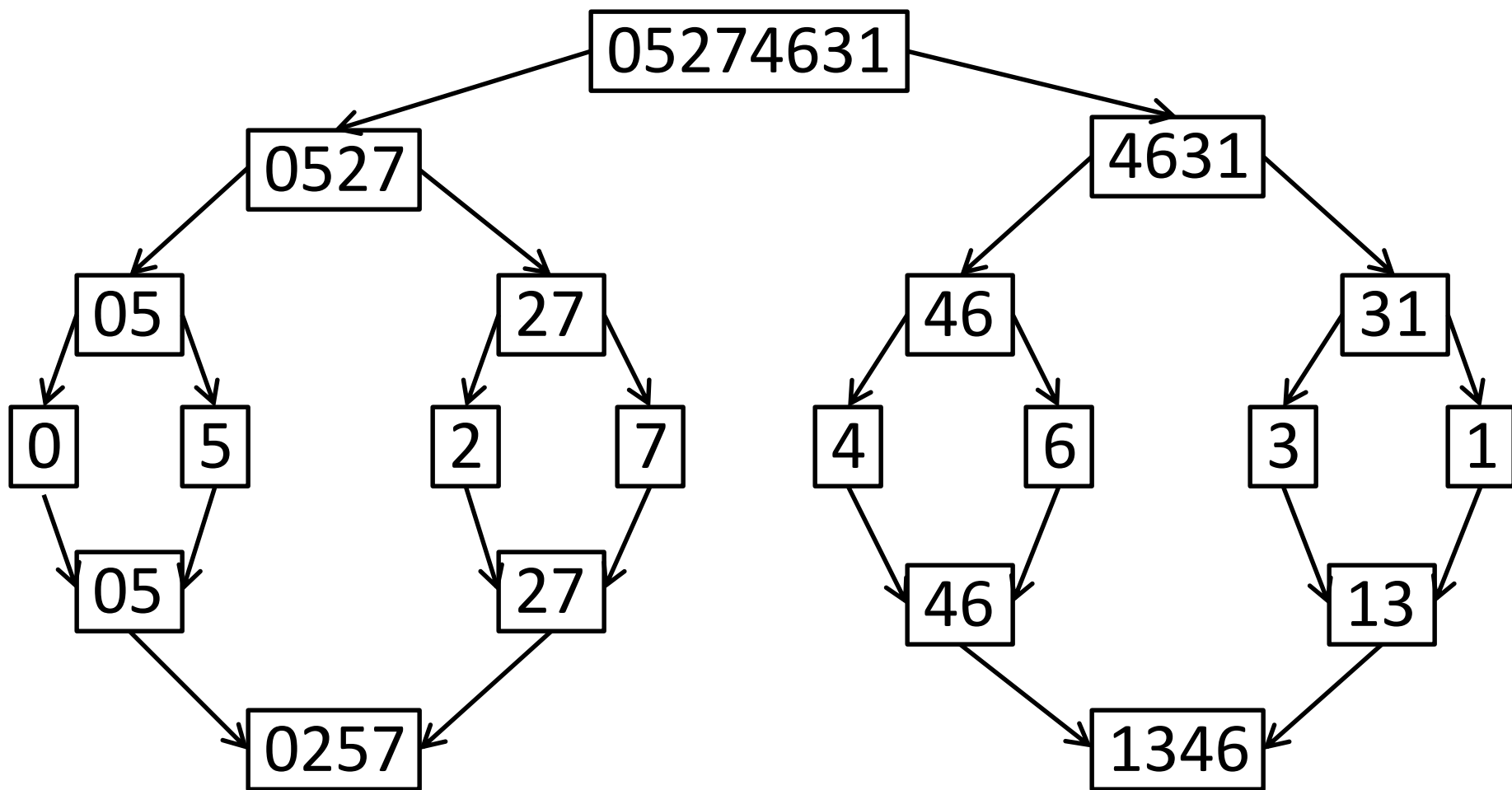




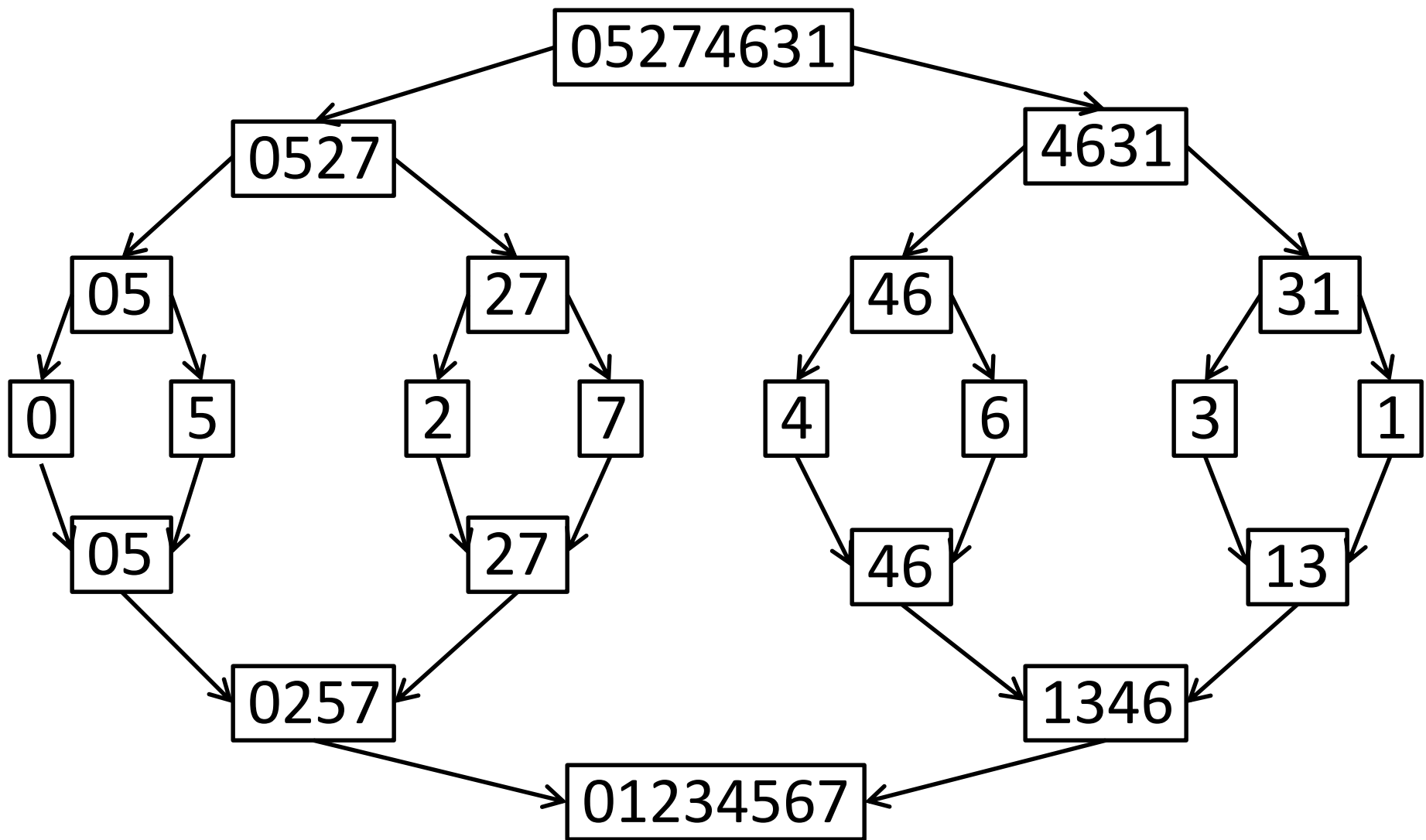
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# Optimaility

$O(n \log(n))$  is *optimal* for comparison based sorting algorithms.

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- Naïve implementation takes  $O(n^2)$  time.
- Use priority queues.

# Heap Sort

```
HeapSort (L)
```

```
    Priority Queue Q
```

```
    For x in L
```

```
        Q.Insert(x)
```

```
    List C
```

```
    While (Q not empty)
```

```
        C ← C.Append(Q.DeleteMin())
```

```
    Return C
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Runtime  $O(n \log(n))$

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**Problem:** Given a list  $L$  of numbers and an integer  $k$ , find the  $k$ th largest element of  $L$ .

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**Problem:** Given a list  $L$  of numbers and an integer  $k$ , find the  $k$ th largest element of  $L$ .

**Naïve Algorithm:** Sort  $L$  and return  $k$ th largest.  
 $O(n \log(n))$  time.