ساختمان داده و الگوریتم ها

مبحث پنجم: قضیه اصلی

سجاد شیرعلی شهرضا پائیز 1402 *دوشنبه، 17 مهر 1402*

اطلاع رساني

- بخش مرتبط کتاب برای این جلسه: 4.5، 4.4، 4.5
 - امتحانک اول
 - دوشنبه هفته آینده، 24 مهر 1402
 - ۰ به صورت حضوری در کلاس
 - در ساعت کلاس

زمان اجرای مرتب سازی ادغامی

چقدر سریع است؟

MERGESORT: IS IT FAST?

```
MERGESORT(A):
    n = len(A)
    if n <= 1:
        return A
    L = MERGESORT(A[0:n/2])
    R = MERGESORT(A[n/2:n])
    return MERGE(L,R)</pre>
```

CLAIM: MergeSort runs in time **O(n log n)**

AN ASIDE: $O(n log n) vs. O(n^2)$?

log(n) grows very slowly! (Much more slowly than n)

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ALL LOGARITHMS IN THIS COURSE ARE BASE 2

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log(2) = 1

log(4) = 2

...

log(64) = 6

log(128) = 7

...

log(4096) = 12
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log(# particles in the universe) < 280

AN ASIDE: $O(n log n) vs. O(n^2)$?

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log(# particles in the universe) < 280

n log n grows much more slowly than n²

Punchline: A running time of O(n log n) is a LOT better than O(n²)

MERGESORT: O(n log n) PROOF

Instead of counting every little operation and tracing all recursive calls, we can think about:

THE RECURSION TREE!

(and we'll add up all the work done across levels to compute the Big-O runtime)

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```

```
MERGE(L,R):
    result = length n array
    i = 0, j = 0
    for k in [0,...,n-1]:
        if L[i] < R[j]:
            result[k] = L[i]
            i += 1
        else:
            result[k] = R[j]
            j += 1
    return result</pre>
```

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We can see that MERGE is **O(n)**

MERGESORT: O(n log n) PROOF

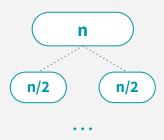
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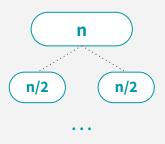
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```

```
MERGE(L,R):
  This means that within one
 recursive call that processes
an array/subarray of length n,
    the work done in that
    subproblem (creating
  subproblems & "merging"
    those results) is O(n).
   return result
```

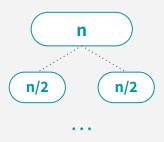


Level	Size of each Problem	# of Problems	Work done per Problem	Total work on this level			
0							
1							
	•••						
t							
•••							
?							





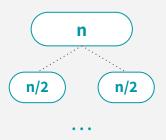
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0							
1							
t							
	•••						
log ₂ n							







Level	Size of each Problem	# of Problems	Work done per Problem	Total work on this level		
0	n					
1	n/2¹					
t	n/2 ^t					
•••						
log ₂ n	1					

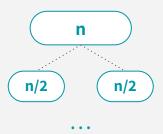






Level	Size of each Problem	# of Problems	Work done per Problem	Total work on this level		
0	n	1				
1	n/2 ¹	2 ¹				
t	n/2 ^t	2 ^t				
	••••					
log ₂ n	1	$2^{\log_2 n} = n$				

If a subproblem is of size \mathbf{n} , then the work done in that subproblem is $\mathbf{O}(\mathbf{n})$. \Rightarrow Work \leq $\mathbf{c} \cdot \mathbf{n}$ (c is a constant)

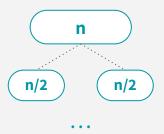






Level	Size of each Problem	# of Problems	Work done per Problem	Total work on this level			
0	n	1	c·n				
1	n/2 ¹	2 ¹	c · (n/2)				
	•••						
t	n/2 ^t	2 ^t	c·(n/2 ^t)				
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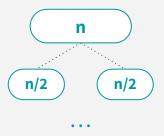






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t	n/2 ^t	2 ^t	c·(n/2 ^t)	$2^{t} \cdot c \cdot (n/2^{t}) = \mathbf{O(n)}$			
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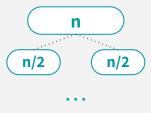


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t	n/2 ^t	2 ^t	c·(n/2 ^t)	$2^{t} \cdot c \cdot (n/2^{t}) = \mathbf{O(n)}$			
	•••						
log ₂ n	1	$2^{\log_2 n} = n$	c · (1)	$\mathbf{n} \cdot \mathbf{c} \cdot (1) = \mathbf{O(n)}$			

We have $(\log_2 n + 1)$ levels, each level has O(n) work total \Rightarrow $O(n \log n)$ work overall!

MERGESORT: O(n log n) RUNTIME

Using the "Recursion Tree Method" (i.e. drawing the tree & filling out the table), we showed that the runtime of MergeSort is **O(n log n)**







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t	n/2 ^t	2 ^t	c·(n/2 ^t)	$2^{t} \cdot c \cdot (n/2^{t}) = \mathbf{O(n)}$		
log ₂ n	1	$2^{\log_2 n} = n$	c · (1)	$\mathbf{n}\cdot\mathbf{c}\cdot(1)=\mathbf{O(n)}$		



رابطه بازگشتی

RUNTIMES FOR RECURSIVE ALGOS

Previously, we used the "Recursion Tree Method" (i.e. drawing the tree & filling out the table) to <u>manually add up all the work in the tree</u> and find that the runtime of MergeSort is **O(n log n)**.

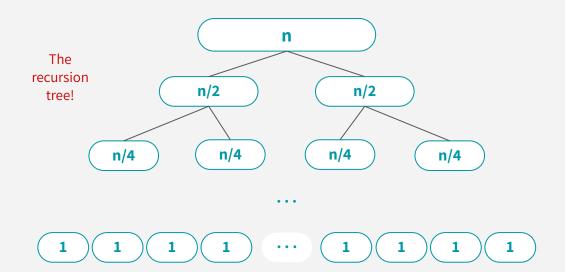
Drawing the tree & doing all that adding kind of takes a lot of work... Here's another way to reason about the runtime of a recursive algorithm like Mergesort:

INTRODUCING...

RECURRENCE RELATIONS

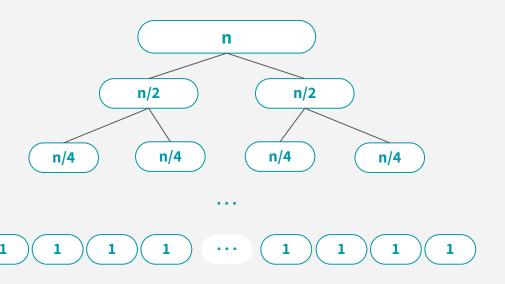
Basically, Recurrence Relations give us a *recursive* way to express runtimes for *recursive* algorithms!

We can then employ some math-ier approaches to analyze these recurrence relations.

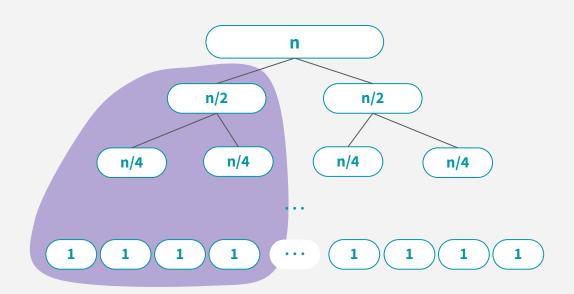


To build the recurrence relation for MergeSort, we can think of its runtime as follows:

Work in the whole tree =



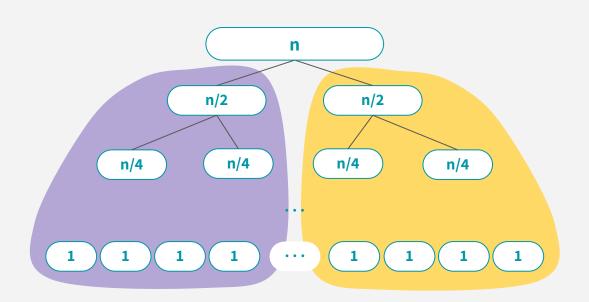
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total work in LEFT recursive call (left subtree)

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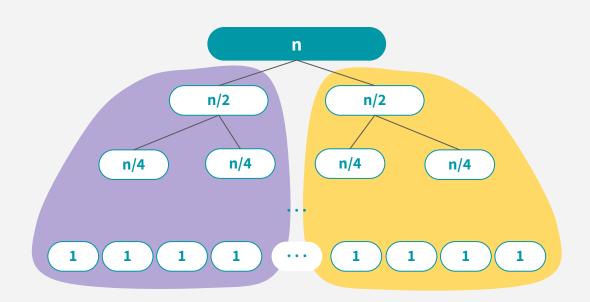
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total work in RIGHT recursive call (right subtree)

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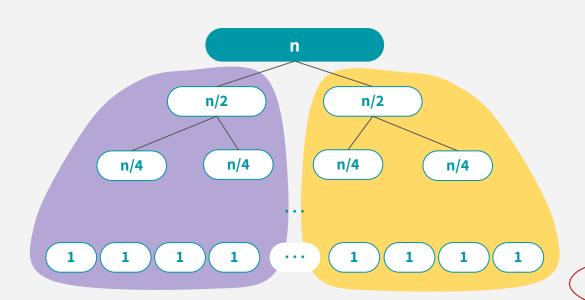


total work in RIGHT recursive call (right subtree)



work done within top problem

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total work in LEFT recursive call (left subtree)



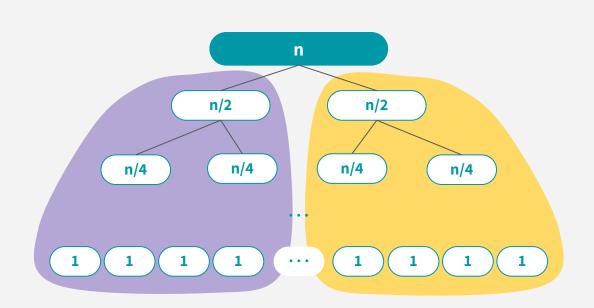
total work in RIGHT recursive call (right subtree)



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work to create suproblems & "merge" their solutions

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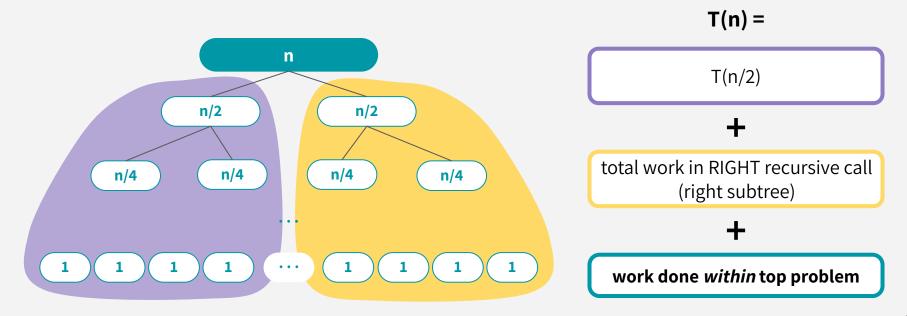
total work in LEFT recursive call (left subtree)

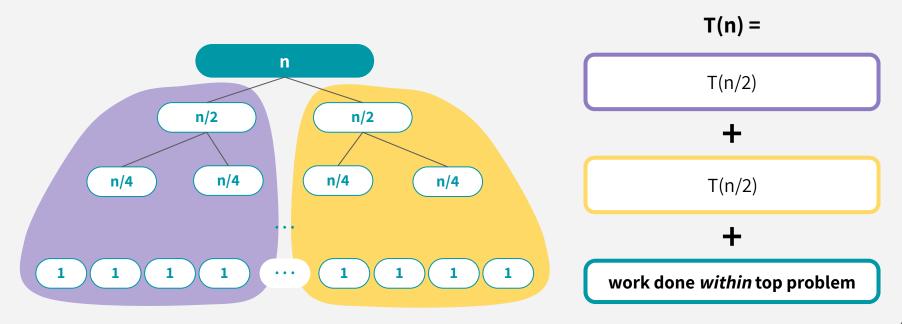


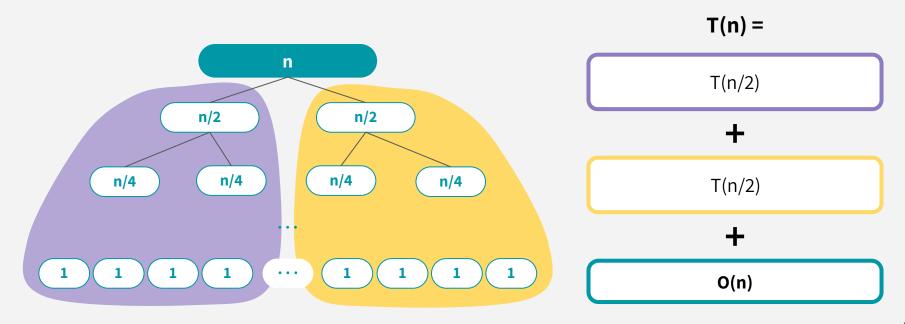
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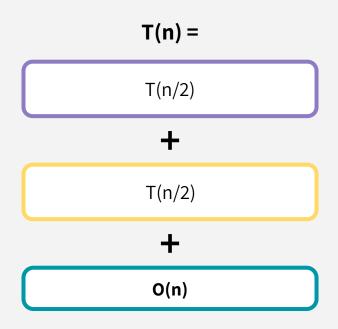


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A note:

We're making a simplifying assumption here that **n** is a perfect power of two (otherwise, we should use floors and ceilings).

Turns out that if we do incorporate floors and ceilings, we still get constant size subproblems at level Llog_bnJ, and generally, the stuff we'll do in this class with Recurrence Relations will still work if we forget about floors and ceilings here.



To build the recurrence relation for MergeSort, we can think of its runtime as follows:

$$T(n) = T(n/2) + T(n/2) + O(n)$$

since the subproblems are equal sizes, we can also write this as $2 \cdot T(n/2)$

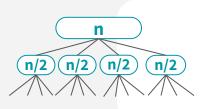
This is a *recursive* definition for T(n), so we also need a BASE CASE:

$$T(1) = O(1)$$

No matter what T is, T(1) = O(1). If it's greater than O(1), then the problem size wouldn't actually be 1.

Since we already used the Recursion Tree to compute the runtime of MergeSort, we know that $T(n) = O(n \log n)$.

EXAMPLE RECURRENCE RELATIONS

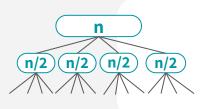


Useless Divide-and-Conquer Multiplication

$$T(n) = 4 \cdot T(n/2) + O(n)$$

 $T(n) = O(n^{\log_2 4}) = O(n^2)$

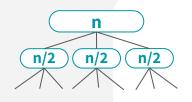
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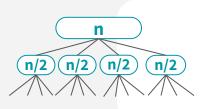


Karatsuba Integer Multiplication

$$T(n) = 3 \cdot T(n/2) + O(n)$$

 $T(n) = O(n^{\log_2 3}) \approx O(n^{1.6})$

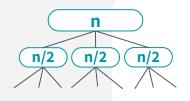
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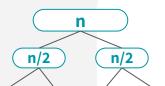
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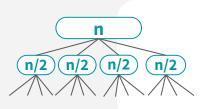
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MergeSort

$$T(n) = 2 \cdot T(n/2) + O(n)$$
$$T(n) = O(n \log n)$$

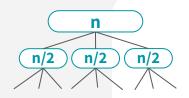
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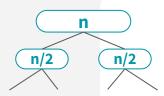
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$$T(n) = 2 \cdot T(n/2) + O(n)$$
$$T(n) = O(n \log n)$$





قضیه اصلی

فرمولی برای حل بسیاری از روابط بازگشتی (اما نه همه آنها!)

THE MASTER THEOREM

Suppose that $\mathbf{a} \ge \mathbf{1}$, $\mathbf{b} > \mathbf{1}$, and \mathbf{d} are constants (i.e. independent of \mathbf{n}).

Suppose $T(n) = a \cdot T(n/b) + O(n^d)$. The Master Theorem states:

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Suppose $T(n) = a \cdot T(n/b) + O(n^d)$. The Master Theorem states:

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

a: number of subproblems (branching factor)

b: factor by which input size shrinks (shrinking factor)

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$$a = 4$$

$$b = 2$$

$$d = 1$$

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USELESS DIVIDE & CONQUER MULTIPLICATION
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a: # of subproblems (branching factor)

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d: need to do O(n^d) work to create subproblems + "merge" solutions

$$T(n) = 4 \cdot T(n/2) + O(n)$$

 $T(n) = O(n^{\log_2 4}) = O(n^2)$

$$a = 4$$

 $b = 2$ $a > b^d$

$$d = 1$$

d = 1

$$T(n) = 3 \cdot T(n/2) + O(n)$$

 $T(n) = O(n^{\log_2 3}) \approx O(n^{1.6})$

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

if
$$a = b^{\alpha}$$

if $a < b^{d}$

a: # of subproblems (branching factor)

b: factor by which input size shrinks (shrinking factor)

$$T(n) = 4 \cdot T(n/2) + O(n)$$

$$a > b^d$$

$$T(n) = O(n^{\log_2 4}) = O(n^2)$$

$$d = 1$$

$$T(n) = 3 \cdot T(n/2) + O(n)$$

$$a = 3$$

$$a > b^d$$

$$T(n) = O(n^{\log_2 3}) \approx O(n^{1.6})$$

$$d = 1$$

$$T(n) = 2 \cdot T(n/2) + O(n)$$

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

if
$$a = b^d$$

a: # of subproblems (branching factor)

b: factor by which input size shrinks (shrinking factor)

$$T(n) = 4 \cdot T(n/2) + O(n)$$

$$a > b^d$$

$$T(n) = O(n^{\log_2 4}) = O(n^2)$$

$$d = 1$$

$$T(n) = 3 \cdot T(n/2) + O(n)$$

$$a > b^d$$

$$T(n) = O(n^{\log_2 3}) \approx O(n^{1.6})$$

$$d = 1$$

$$T(n) = 2 \cdot T(n/2) + O(n)$$

$$a = 2$$

$$d = 1$$

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

if
$$a = b^{d}$$

if $a < b^{d}$

a: # of subproblems (branching factor)

b: factor by which input size shrinks (shrinking factor)

$$T(n) = 4 \cdot T(n/2) + O(n)$$

$$a > b^d$$

$$T(n) = 0$$

$$T(n) = O(n^{\log_2 4}) = O(n^2)$$

$$d = 1$$

$$T(n) = 3 \cdot T(n/2) + O(n)$$

$$a > b^d$$

$$T(n) = O(n^{\log_2 3}) \approx O(n^{1.6})$$

$$d = 1$$

$$T(n) = 2 \cdot T(n/2) + O(n)$$

$$a = 2$$

$$a = b^d$$

$$d = 1$$

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

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$$T(n) = 4 \cdot T(n/2) + O(n)$$

 $T(n) = O(n^{\log_2 4}) = O(n^2)$

a = 4

$$a > b^d$$

$$T(n) = 3 \cdot T(n/2) + O(n)$$

$$a > h^0$$

$$T(n) = O(n^{\log_2 3}) \approx O(n^{1.6})$$

$$d = 1$$

$$T(n) = 2 \cdot T(n/2) + O(n)$$

$$a = 2$$

$$a = b^d$$

$$d = 1$$

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

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$$T(n) = 4 \cdot T(n/2) + O(n)$$

 $T(n) = O(n^{\log_2 4}) = O(n^2)$

$$a > b^d$$

$$T(n) = 3 \cdot T(n/2) + O(n)$$

 $T(n) = O(n^{\log_2 3}) \approx O(n^{1.6})$

$$a > b^d$$

$$= O(\mathsf{I}^{\mathsf{res}(2^{-1})}) \approx O(\mathsf{n}^{\mathsf{res}(1)})$$
 d:

$$a > b^{a}$$

$$T(n) = 2 \cdot T(n/2) + O(n)$$
$$T(n) = O(n \log n)$$

$$a = b^d$$

$$d = 1$$

