


# Quick Sort

2	6	5	3	8	7	1	0
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1. Correct position in final, sorted array
2. Items to the left are smaller
3. Items to the right are larger

2	6	5	3	8	7	1	0
---	---	---	---	---	---	---	---




1. **itemFromLeft** that is larger than pivot

2	6	5	0	8	7	1	3
---	---	---	---	---	---	---	---

1. **itemFromLeft** that is larger than pivot
2. **itemFromRight** that is smaller than pivot


2	6	5	0	8	7	1	3
---	---	---	---	---	---	---	---



↑  
itemFromLeft

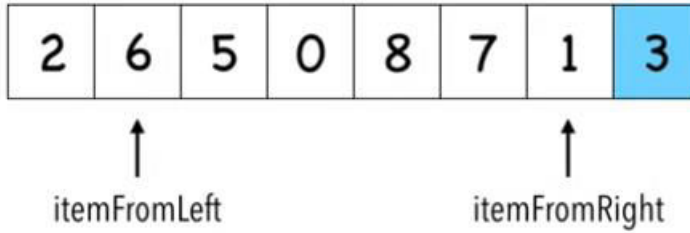
1. **itemFromLeft** that is larger than pivot
2. **itemFromRight** that is smaller than pivot

2	6	5	0	8	7	1	3
---	---	---	---	---	---	---	---

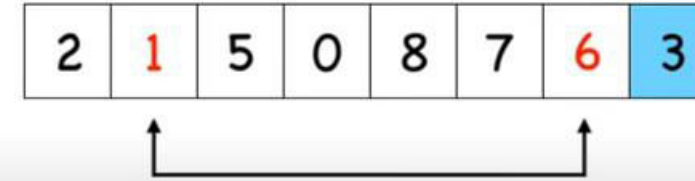


↑                      ↑  
itemFromLeft      itemFromRight

1. **itemFromLeft** that is larger than pivot
2. **itemFromRight** that is smaller than pivot

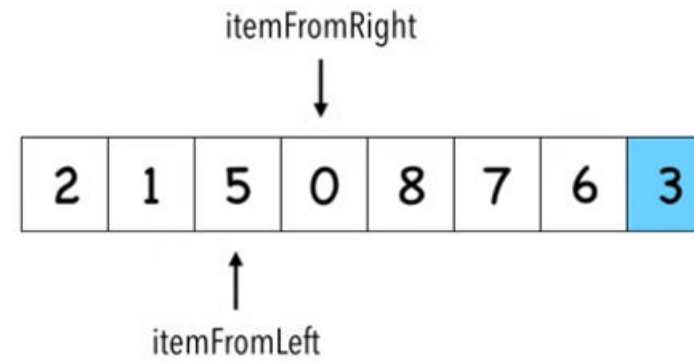
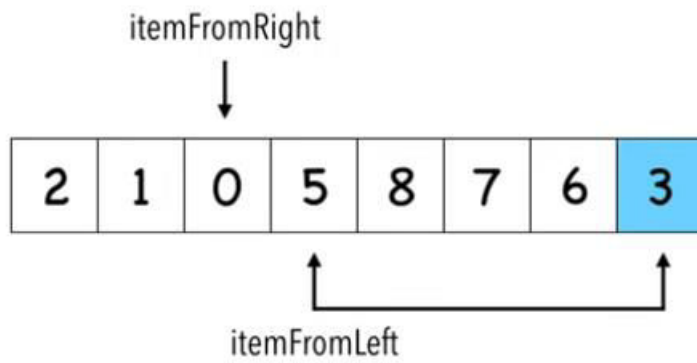


1. **itemFromLeft** that is larger than pivot
2. **itemFromRight** that is smaller than pivot

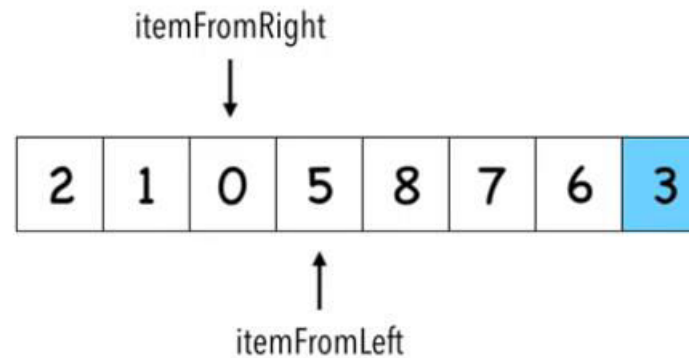
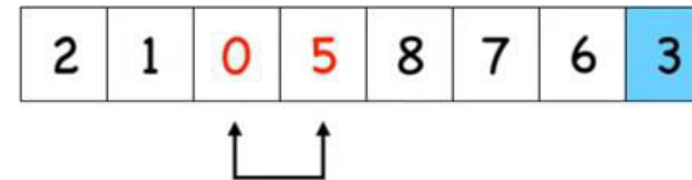


1. **itemFromLeft** that is larger than pivot
2. **itemFromRight** that is smaller than pivot





Swap **itemFromLeft** and **pivot**



Stop when index of **itemFromLeft** > index of **itemFromRight**

1. Correct position in final, sorted array
2. Items to the left are smaller
3. Items to the right are larger

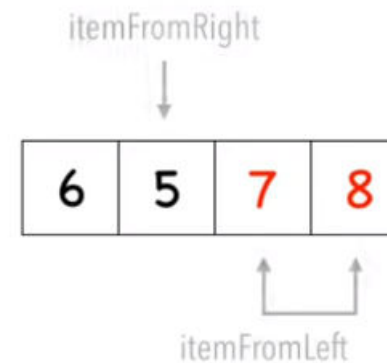
2	1	0	3	8	7	6	5
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8	7	6	5
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How to choose pivot:  
Median position element:

1. **itemFromLeft** that is larger than pivot
2. **itemFromRight** that is smaller than pivot

8	5	6	7
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Worst Case Complexity:  $O(n^2)$   
Average Case Complexity:  $\Theta(n \log n)$

Master's method is a quite useful method for solving recurrence equations because it directly gives us the cost of an algorithm with the help of the type of a recurrence equation and it is applied when the recurrence equation is in the form of:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

where,  $a \geq 1$ ,  $b > 1$  and  $f(n) > 0$ .

For example,

$$c + T\left(\frac{n}{2}\right) \rightarrow a = 1, b = 2 \text{ and } f(n) = c,$$

$$n + 2T\left(\frac{n}{2}\right) \rightarrow a = 2, b = 2 \text{ and } f(n) = n, \text{ etc.}$$

$T(n) = aT(n/b) + f(n)$ , where,  $n$  = size of input  $a$  = number of subproblems in the recursion  $n/b$  = size of each subproblem. All subproblems are assumed to have the same size. Recurrence relation helps in finding the subsequent term (next term) dependent upon the preceding term (previous term).

# Master's Theorem

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Taking an equation of the form:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

where,  $a \geq 1$ ,  $b > 1$  and  $f(n) > 0$

The Master's Theorem states:

- **CASE 1** - if  $f(n) = O(n^{\log_b a - \epsilon})$  for some  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$
- **CASE 2** - if  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$
- **CASE 3** - if  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some  $\epsilon > 0$ , and if  $af(n/b) \leq cf(n)$  for some  $c < 1$  and all sufficiently large  $n$ , then  $T(n) = \Theta(f(n))$ .

By the use of these three cases, we can easily get the solution of a recurrence equation of the form  $T(n) = aT\left(\frac{n}{b}\right) + f(n)$ .

### Example 1

$$T(n) = 2T\left(\frac{n}{2}\right) + n$$

Here,  $a = 2$ ,  $b = 2$ ,  $\log_b a = \log_2 2 = 1$

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

Also,  $f(n) = n$

So,  $n^{\log_b a} = n = f(n)$

(comparing  $n^{\log_b a}$  with  $f(n)$ )  $\Rightarrow f(n) = \Theta(n^{\log_b a})$

So, case 2 can be applied and thus  $T(n) = \Theta(n^{\lg_b a} \lg n) = \Theta(n \lg n)$ .

## Example 2

$$T(n) = 2T\left(\frac{n}{2}\right) + n^2$$

Here,  $a = 2$ ,  $b = 2$ ,  $\log_2 2 = 1$

$$\Rightarrow n^{\lg_b a} = n^1 = n$$

$$\text{Also, } f(n) = n^2$$

$$\Rightarrow f(n) = \Omega(n^{1+\epsilon}) \ (\epsilon = 1) \text{ (comparing } n^{\lg_b a} \text{ with } f(n))$$

Case 3 can be applied if rest of the conditions of case 3 gets satisfied for  $f(n)$ .

The condition is  $af(n/b) \leq cf(n)$  for some  $c < 1$  and all sufficiently large  $n$ .

For a sufficiently large  $n$ , we have,

$$af\left(\frac{n}{b}\right) = 2f\left(\frac{n}{2}\right) = 2\frac{n^2}{4} = \frac{n^2}{2} \leq \frac{1}{2}(n^2) \text{ (for } c = \frac{1}{2})$$

So, the condition is satisfied for  $c = \frac{1}{2}$ . Thus,  $T(n) = \Theta(f(n)) = \Theta(n^2)$



### Example 3

$$T(n) = 2T\left(\frac{n}{2}\right) + \sqrt{n}$$

Here,  $a = 2$   $b = 2$   $\log_2 2 = 1$

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

$$f(n) = \sqrt{n}$$

$$f(n) = O(n^{1-\epsilon}) \text{ (Case I)}$$

$$T(n) = \Theta(n)$$

### Example 4

$$T(n) = 3T\left(\frac{n}{4}\right) + n \lg n$$

Here,  $a = 3$   $b = 4$   $\log_4 3 = 0.792$

$f(n) = \Omega(n^{\log_4 3 + \epsilon})$  (Case 3)

$$3 \left(\frac{n}{4}\right) \lg \left(\frac{n}{4}\right) \leq \frac{3}{4} n \lg n = c * f(n), \quad c = \frac{3}{4}$$

So,  $T(n) = \Theta(n \lg n)$

### Example 5

$$T(n) = 2T\left(\frac{n}{2}\right) + n \lg n$$

Here,  $a = 2$   $b = 2$   $\log_2 2 = 1$

$$n^{\log_2 2} = n^1$$

$$f(n) = n \lg n$$

$f(n)$  must be polynomially larger by a factor of  $n^\epsilon$  but it is only larger by a factor of  $\lg n$ .

So, Master's theorem can't be applied.