

[320] Complexity + Big O (Worksheet: Complexity Analysis)

Department of Computer Sciences
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1

```
def search(L, target):  
    for x in L:  
        if x == target: #line A  
            return True  
    return False
```

*assume this is asked unless
otherwise stated*

Let $f(N)$ be the number of times line A executes, with $N = \text{len}(L)$. What is $f(N)$ in each case?

Worst Case (target is at end of list): $f(N) = \underline{\hspace{2cm}}$

Best Case (target is at beginning of list): $f(N) = \underline{\hspace{2cm}}$

Average Case (target in middle of list): $f(N) = \underline{\hspace{2cm}}$

A **step** is any unit of work with bounded execution time (it doesn't keep getting slower with growing input size).

We classify algorithm complexity by classifying the **order of growth** of a function $f(N)$, where f gives the number of steps the algorithm must perform for a given input size.

Big O definition: if $f(N) \leq C * g(N)$ for large N values and some fixed constant C , then $f(N) \in O(g(N))$

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Average Case (target in middle of list):

$$f(N) = \frac{N}{2} \in O(N)$$

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2

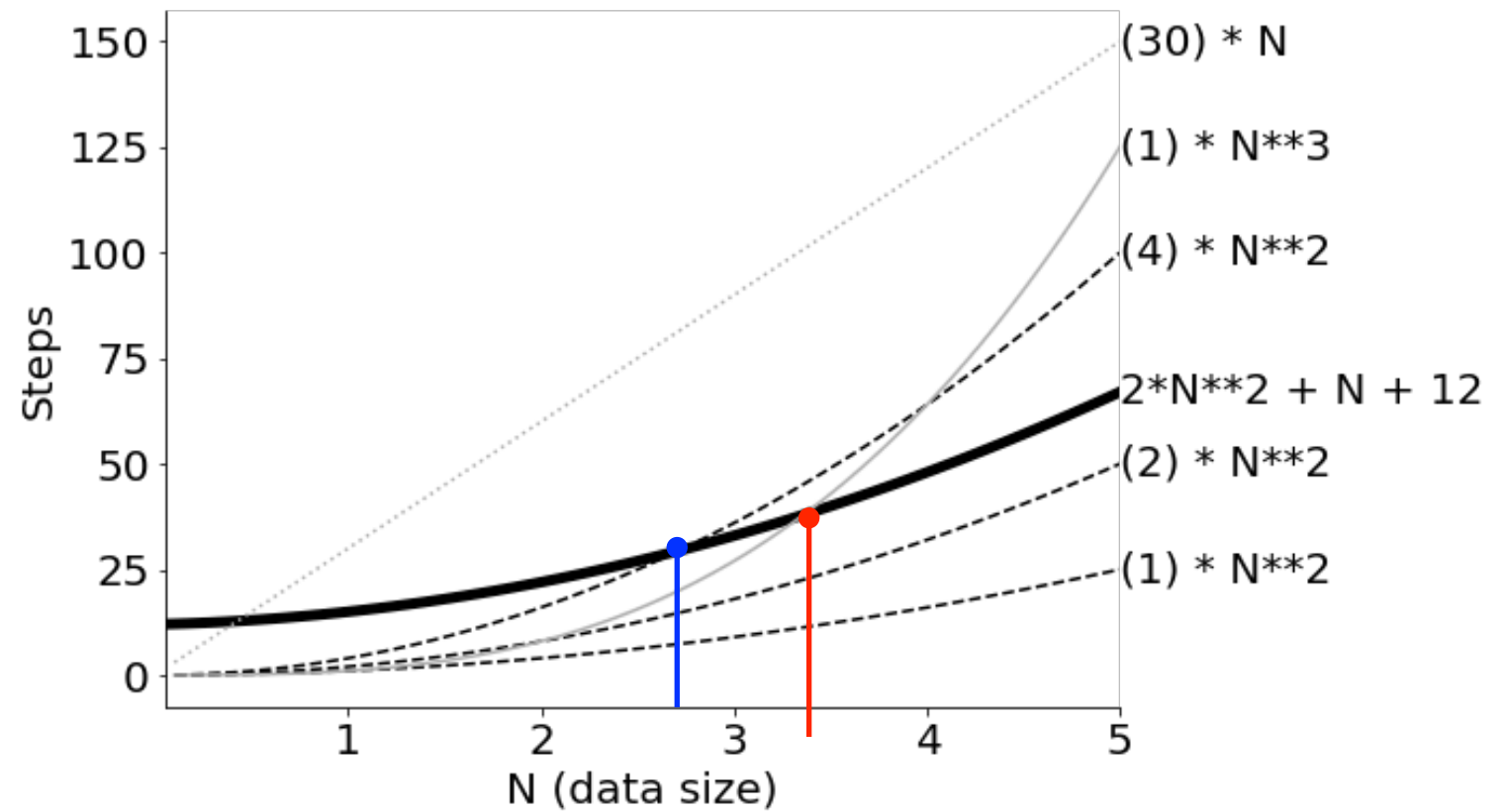
Let $f(N) = 2N^2 + N + 12$

If we want to show $f(N) \in O(N^3)$, what is a good lower bound on N ? Let's have $C=1$.

To show $f(N) \in O(N^2)$, do we pick 1, 2, or 4 for the C ? After picking C , what should we choose for N 's lower bound?

What is more informative to show?
 $f(N) \in O(N^3)$ or $f(N) \in O(N^2)$?

Somebody claims $f(N) \in O(N)$, offering $C=30$ and $N>0$. Suggest an N value to counter their claim.



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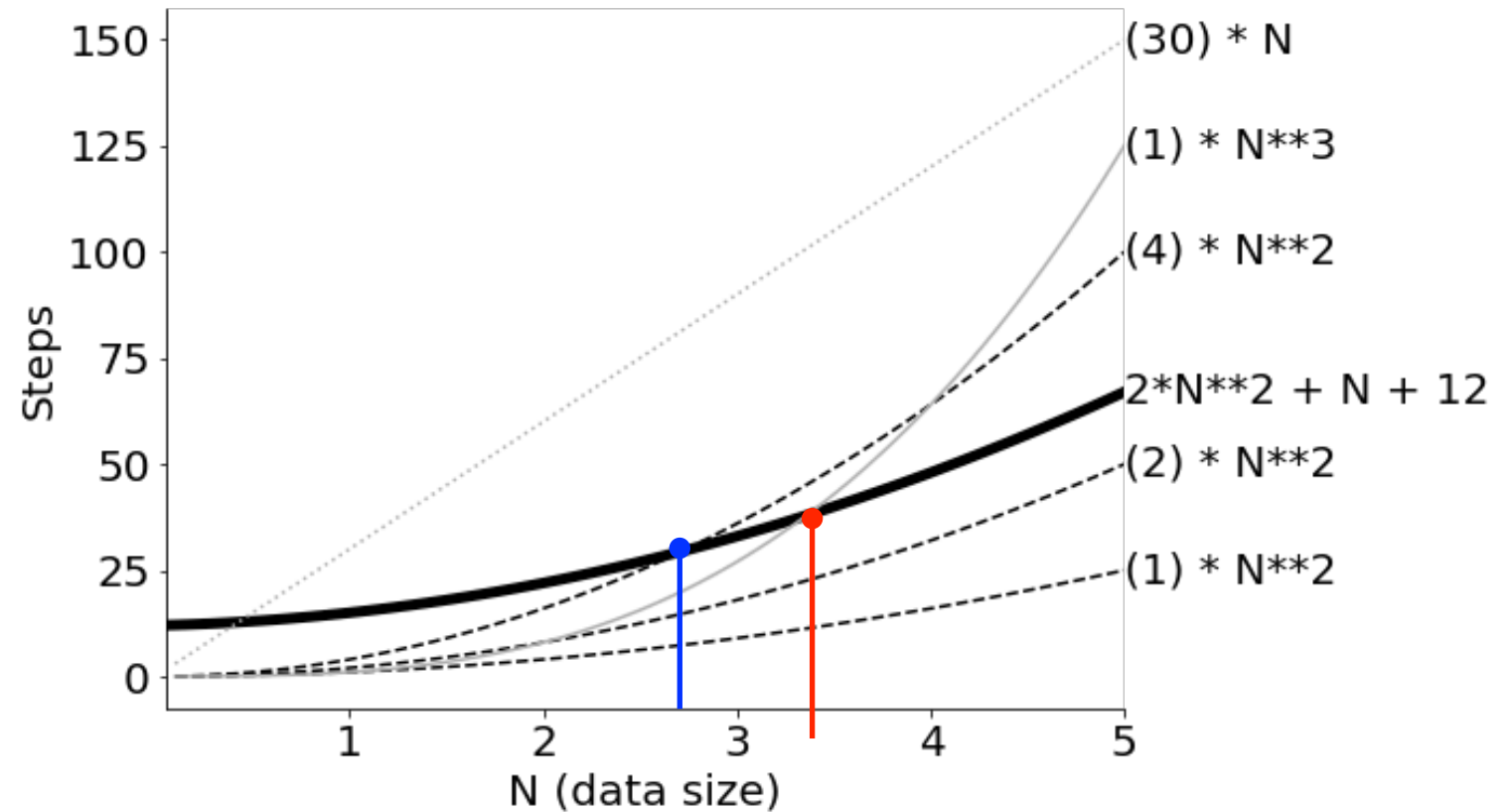
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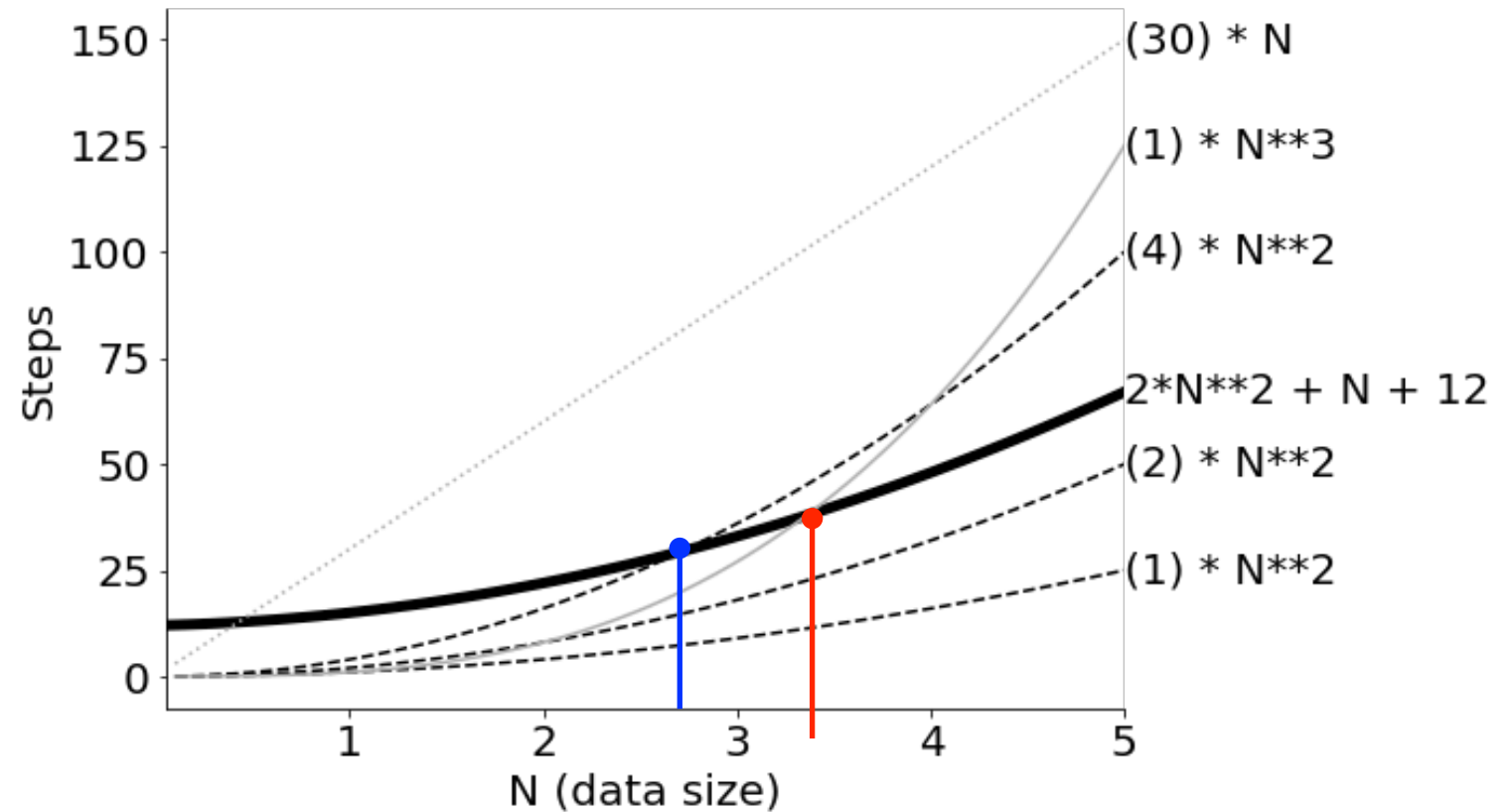
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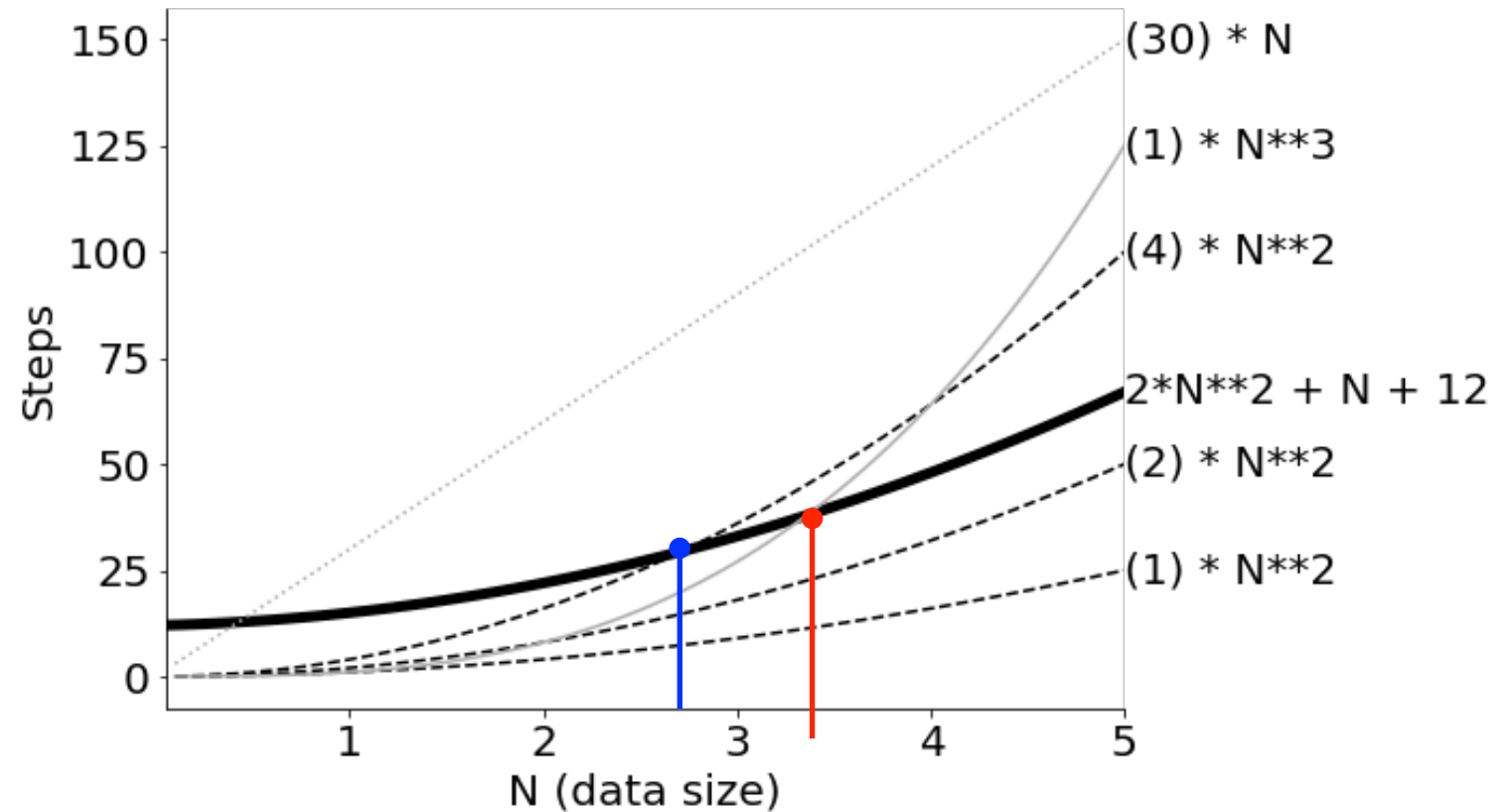
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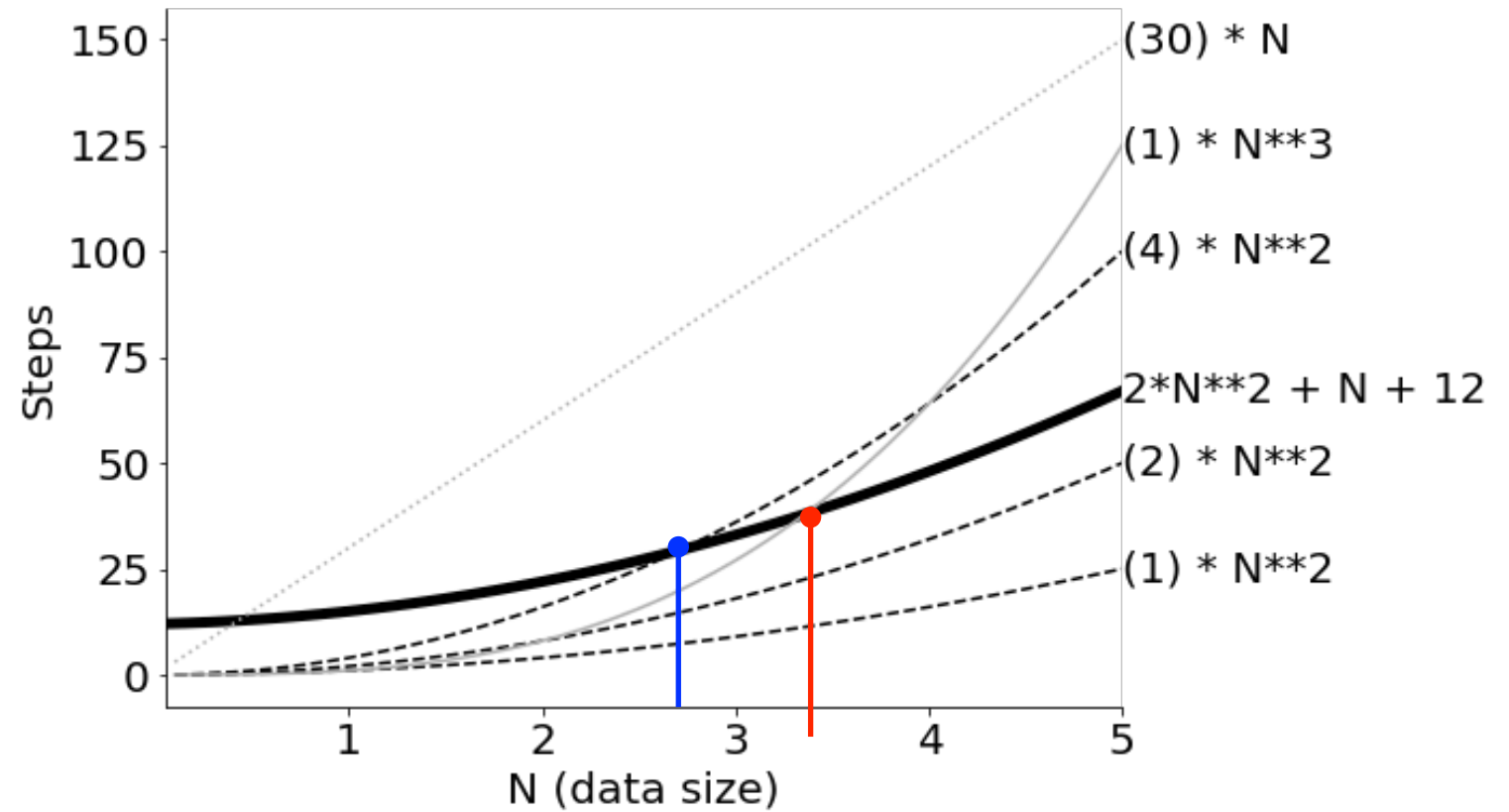
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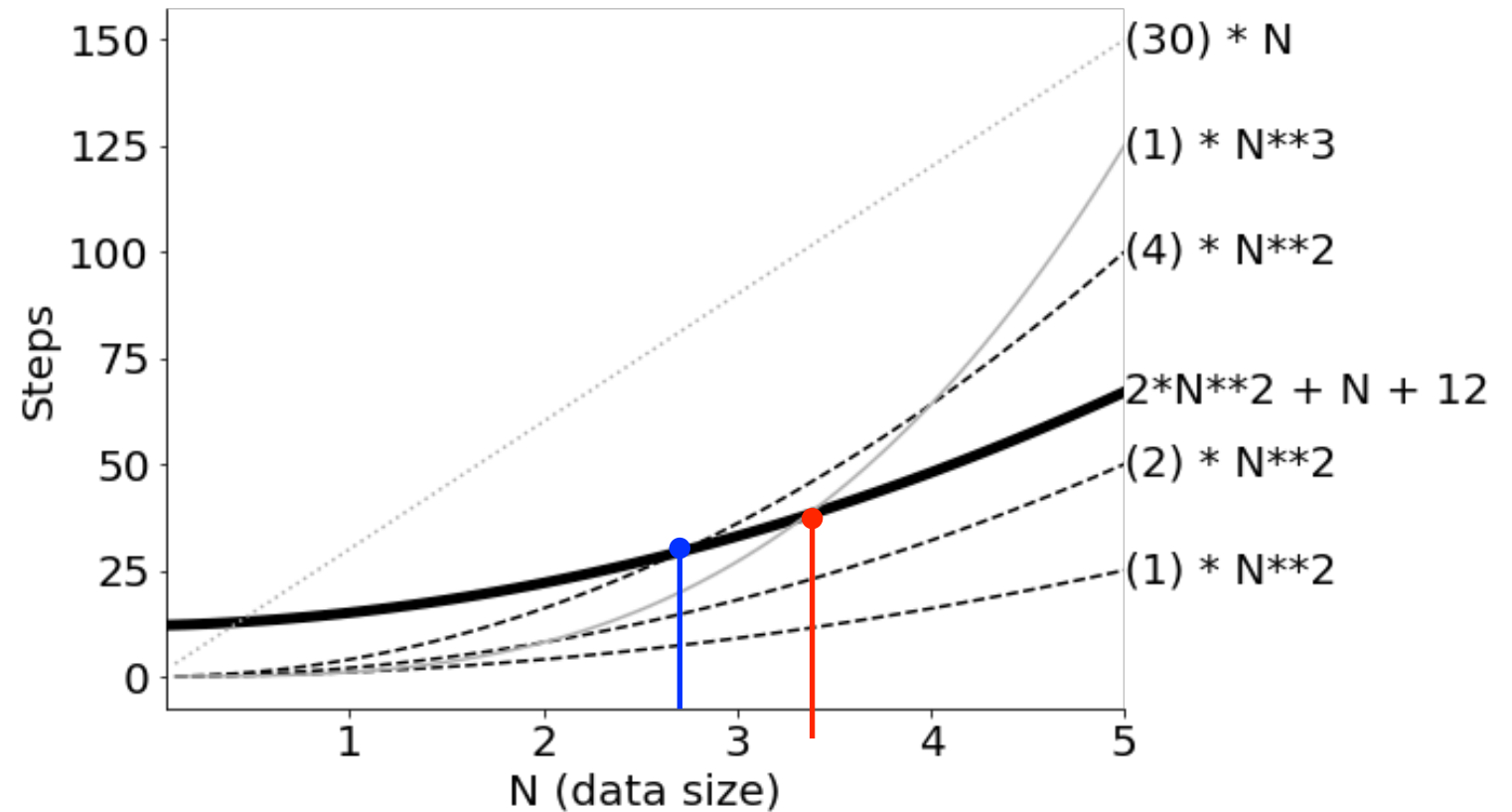
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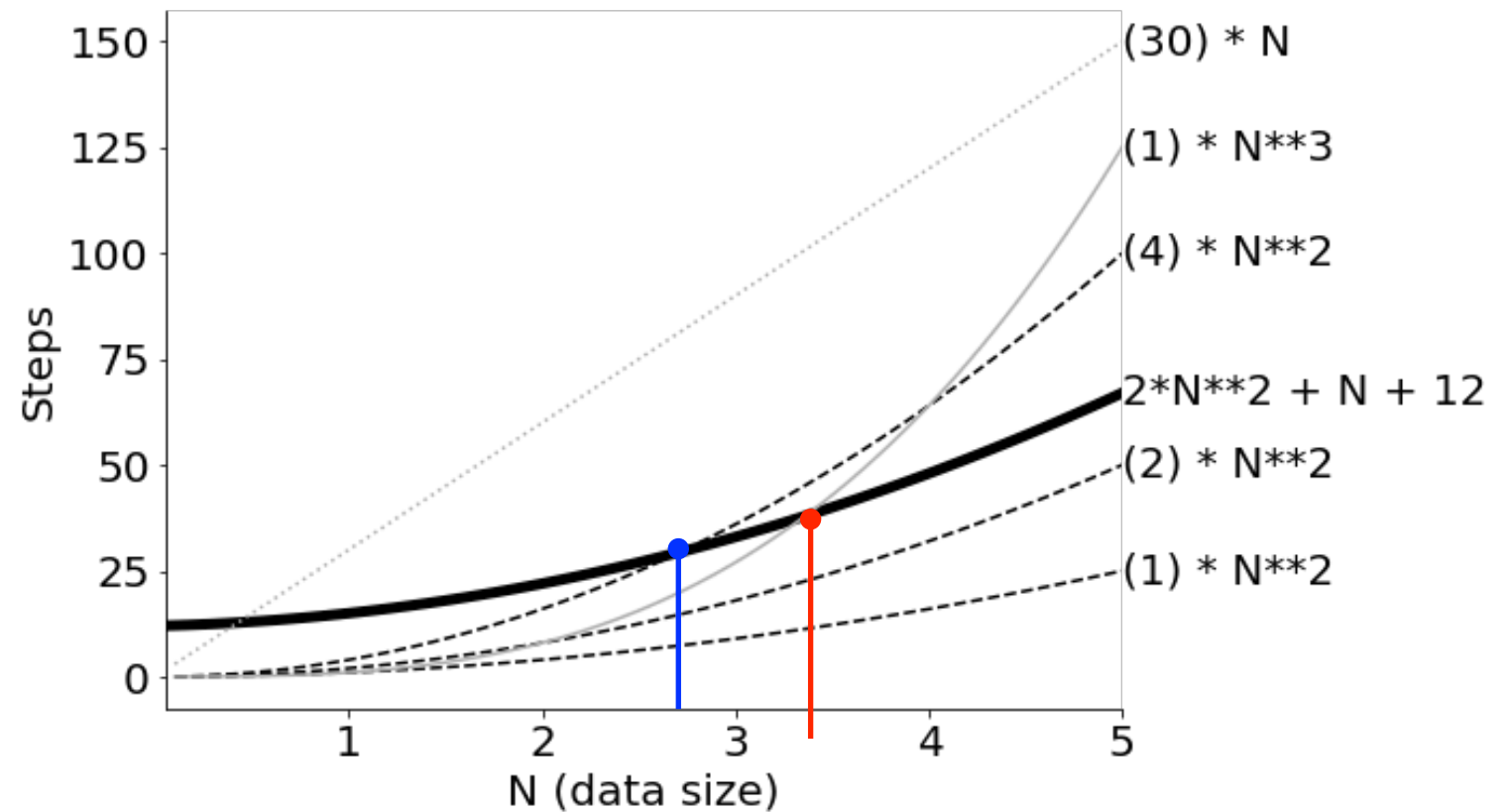
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Assume $N = 20$. and $2N^2 + N + 12 \leq 30N$.



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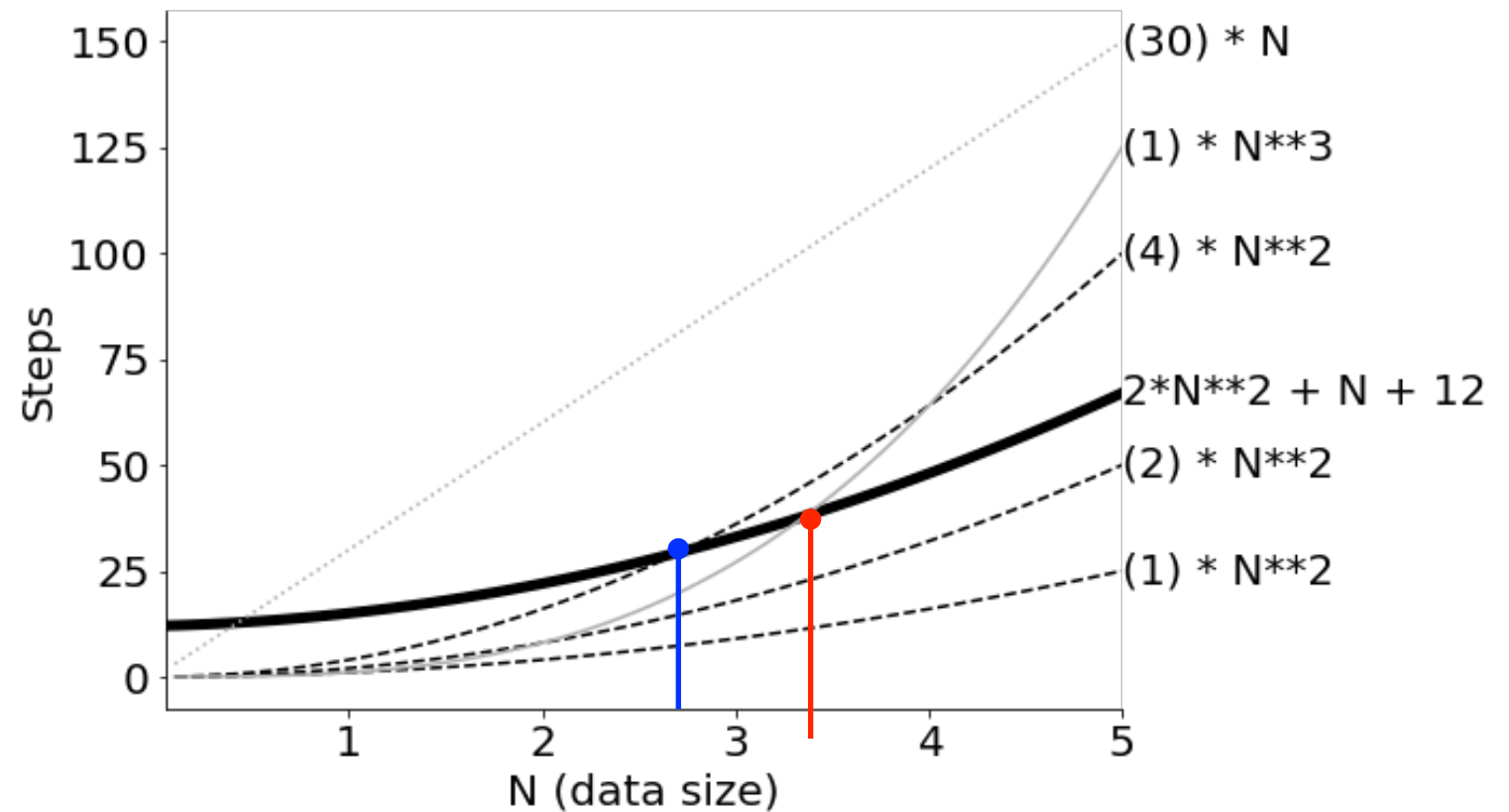
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Assume $N = 20$. and

$2N^2 + N + 12 \leq 30N$.

However, $800 + 20 + 12 \not\leq 600$.



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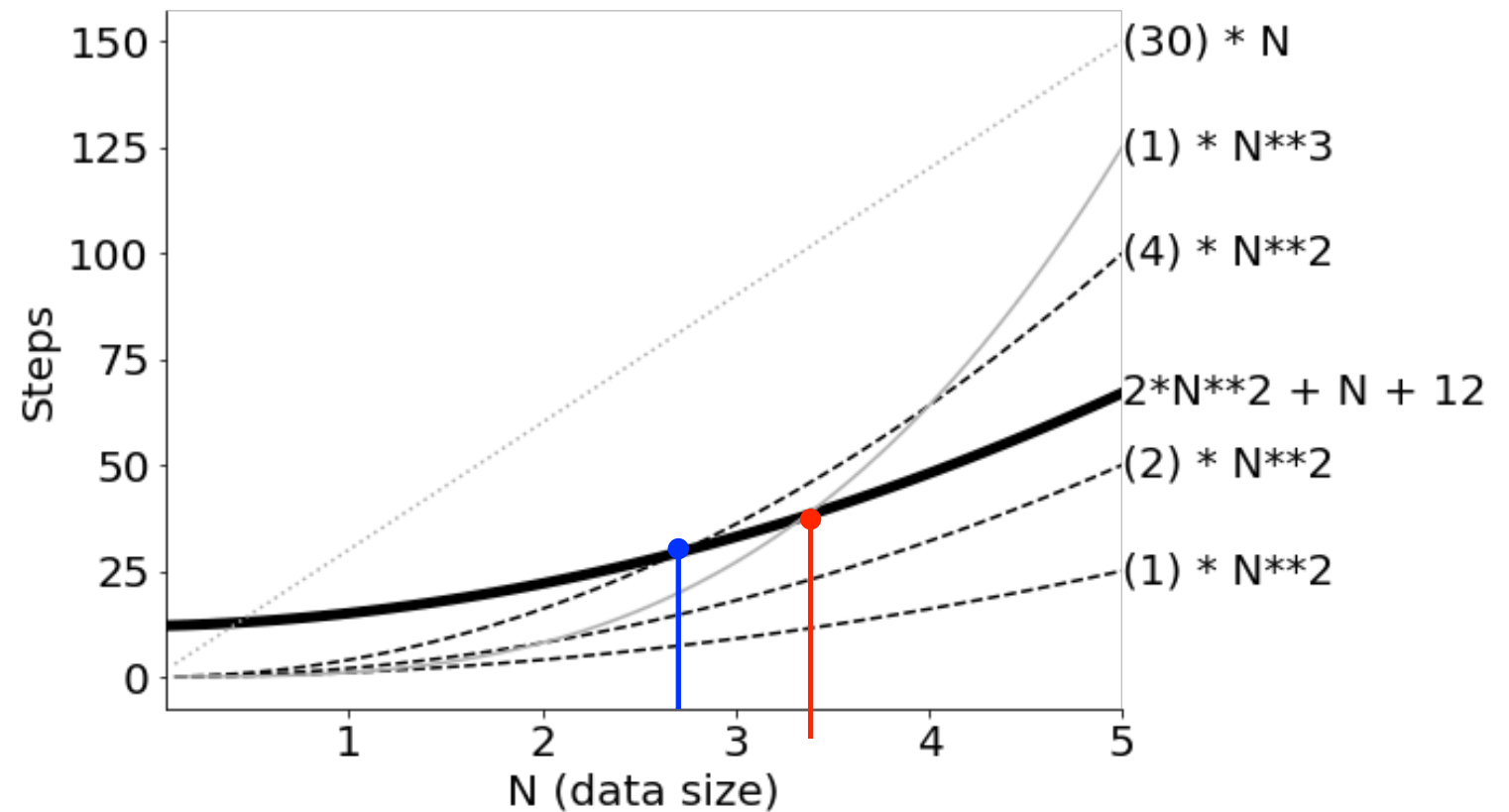
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Somebody claims $f(N) \in O(N)$, offering $C=30$ and $N>0$. Suggest an N value to counter their claim.

Assume $N = 20$. and
 $2N^2 + N + 12 \leq 30N$.

However, $800 + 20 + 12 \not\leq 600$.

Therefore, the suggest value of
 $N = 20$.



3

```
nums = [...]
```

```
first100sum = 0
```

```
for x in nums[:100]:  
    first100sum += x  
print(first100sum)
```

If we increase the size of nums from 20 items to 100 items, the code will probably take _____ times longer to run.

If we increase the size of nums from 100 to 1000, will the code take longer? Yes / No

The complexity of the code is $O(\text{_____})$, with $N=\text{len}(\text{nums})$.

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If we increase the size of nums from 20 items to 100 items, the code will probably take **5** times longer to run.

If we increase the size of nums from 100 to 1000, will the code take longer? Yes / No

No

The complexity of the code is $O(\text{_____})$, with $N=\text{len}(\text{nums})$.

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If we increase the size of nums from 100 to 1000, will the code take longer? Yes / No
No

The complexity of the code is $O(\mathbf{1})$, with $N=\text{len}(\text{nums})$.

4

Each of the following list operations are either $O(1)$ or $O(N)$, where N is $\text{len}(L)$. Circle those you think are $O(N)$.

`L.insert(0, x)`

`L.pop(0)`

`x = L[0]`

`x = max(L)`

`x = len(L)`

`L.append(x)`

`L.pop(-1)`

`L2.extend(L)`

`x = sum(L)`

`found = x in L`

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```
L = [...]  
for x in L:  
    avg = sum(L) / len(L)  
    if x > 2*avg:  
        print("outlier", x)
```

What is the big O complexity?

Is there a way to optimize the code?

```
L = [...]  
for x in L: N+1 steps  
    avg = sum(L) / len(L)  
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Is there a way to optimize the code?

Calculate **avg** outside the loop.

6

```
A = [...]
```

```
B = [...]
```

```
for x in A:  
    for y in B:  
        print(x*y)
```

how would you define the variable(s) to describe the size of the input data?

The complexity of code is

$O(\rule{1cm}{0.4pt})$

6

```
A = [...]    len(A) = M  
B = [...]
```

```
for x in A:  
    for y in B:  
        print(x*y)
```

how would you define the variable(s) to describe the size of the input data?

The complexity of code is

$O(\rule{1.5cm}{0.4pt})$

6

A = [...] $\text{len}(A) = M$
B = [...] $\text{len}(B) = N$

```
for x in A:  
    for y in B:  
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how would you define the variable(s) to describe the size of the input data?

$\text{len}(A) = M$ and $\text{len}(B) = N$

The complexity of code is

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len(A) = M and *len(B) = N*

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how would you define the variable(s) to describe the size of the input data?

$\text{len}(A) = M$ and $\text{len}(B) = N$

The complexity of code is

$$O((M+1)(N+1)) = O(MN + M + N + 1) = O(MN)$$

7

```
# assume L is already sorted, N=len(L)
def binary_search(L, target):
    left_idx = 0 # inclusive
    right_idx = len(L) # exclusive
    while right_idx - left_idx > 1:
        mid_idx = (right_idx + left_idx) // 2
        mid = L[mid_idx]
        if target >= mid:
            left_idx = mid_idx
        else:
            right_idx = mid_idx

    return right_idx > left_idx and L[left_idx] == target
```

Assume target = 20

how many times does this step run
when **N = 1**? **N = 2**? **N = 4**? **N = 8**?

If **f(N)** is the number of times this step runs,
then **f(N) =**

The complexity of binary search is
O(_____)

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The complexity of binary search is
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Assume target = 20

| | | | | | | | |
|---|---|---|---|----|----|----|----|
| 1 | 3 | 5 | 8 | 10 | 20 | 73 | 80 |
|---|---|---|---|----|----|----|----|

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        if target >= mid:
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        else:
            right_idx = mid_idx

    return right_idx > left_idx and L[left_idx] == target
```

how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) =$

The complexity of binary search is
 $O(\text{_____})$

Assume target = 20

| Idx | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 |
|-----|---|---|---|---|----|----|----|----|---|
| | 1 | 3 | 5 | 8 | 10 | 20 | 73 | 80 | |

7

```
mid = L[mid_idx]
if target >= mid:
    left_idx = mid_idx
else:
    right_idx = mid_idx
```

```
return right_idx > left_idx and L[left_idx] == target
```

how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) =$

The complexity of binary search is $O(\text{_____})$

Assume target = 20

[illegible]

7

```
mid = L[mid_idx]
if target >= mid:
    left_idx = mid_idx
else:
    right_idx = mid_idx
```

```
return right_idx > left_idx and L[left_idx] == target
```


how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) =$

The complexity of binary search is $O(\underline{\hspace{2cm}})$

Assume target = 20

[illegible]



```
mid = L[mid_idx]
if target >= mid:
    left_idx = mid_idx
else:
    right_idx = mid_idx
```

```
return right_idx > left_idx and L[left_idx] == target
```

how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) =$

The complexity of binary search is $O(\underline{\hspace{2cm}})$

Assume target = 20

[illegible]

7

```
mid = L[mid_idx]
if target >= mid:
    left_idx = mid_idx
else:
    right_idx = mid_idx
```

```
return right_idx > left_idx and L[left_idx] == target
```

how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) =$

The complexity of binary search is $O(\underline{\hspace{2cm}})$

Assume target = 20

[illegible]

7

```
mid = L[mid_idx]
if target >= mid:
    left_idx = mid_idx
else:
    right_idx = mid_idx
```

```
return right_idx > left_idx and L[left_idx] == target
```

how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) =$

The complexity of binary search is $O(\underline{\hspace{2cm}})$

Assume target = 20

[illegible]

7

```
mid = L[mid_idx]
if target >= mid:
    left_idx = mid_idx
else:
    right_idx = mid_idx
```

```
return right_idx > left_idx and L[left_idx] == target
```

how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) = \log_2 N$

The complexity of binary search is $O(\underline{\hspace{2cm}})$

Assume target = 20

[illegible]

7

```
mid = L[mid_idx]
if target >= mid:
    left_idx = mid_idx
else:
    right_idx = mid_idx
```

```
return right_idx > left_idx and L[left_idx] == target
```

how many times does this step run
when $N = 1$? $N = 2$? $N = 4$? $N = 8$?

If $f(N)$ is the number of times this step runs,
then $f(N) = \log_2 N$

The complexity of binary search is $O(\log N)$

Assume target = 20

[illegible]

```
s1 = tuple("...") # could be any string
s2 = tuple("...")
```

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):
```

```
    if p == s2:
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A?

what is the complexity of version B?

```
s1 = tuple("...") # could be any string    len(s1) = N  
s2 = tuple("...")    len(s2) = N
```

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):
```

```
    if p == s2:
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A?

what is the complexity of version B?

```
s1 = tuple("...") # could be any string    len(s1) = N
s2 = tuple("...")    len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):
```

```
    if p == s2:
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

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s1 = tuple("...") # could be any string    len(s1) = N
s2 = tuple("...")    len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):
```

```
    if p == s2:
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A?

what is the complexity of version B?

```
s1 = tuple("...") # could be any string    len(s1) = N
s2 = tuple("...")    len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

ABC BCA
ACB CAB
BAC CBA

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):
```

```
    if p == s2:
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A?

what is the complexity of version B?


```
s1 = tuple("...") # could be any string    len(s1) = N
s2 = tuple("...")    len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):
```

```
    if p == s2:
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A?

what is the complexity of version B?

```
s1 = tuple("...") # could be any string    len(s1) = N
s2 = tuple("...")    len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1): N! steps
```

```
    if p == s2:
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A?

what is the complexity of version B?

```
s1 = tuple("...") # could be any string  len(s1) = N
s2 = tuple("...")  len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1): N! steps
```

```
    if p == s2: N steps
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A?

what is the complexity of version B?

```
s1 = tuple("...") # could be any string     $len(s1) = N$ 
s2 = tuple("...")     $len(s2) = N$ 
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):  $N!$  steps
```

```
    if p == s2:  $N$  steps
```

```
        matches = True
```

version B

```
s1 = sorted(s1)
```

```
s2 = sorted(s2)
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A? $O(N * N!)$

what is the complexity of version B?

```
s1 = tuple("...") # could be any string     $\text{len}(s1) = N$ 
s2 = tuple("...")     $\text{len}(s2) = N$ 
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):  $N!$  steps
```

```
    if p == s2:  $N$  steps
```

```
        matches = True
```

version B

```
s1 = sorted(s1)     $N \log N$ 
```

```
s2 = sorted(s2)     $N \log N$ 
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A? $O(N * N!)$

what is the complexity of version B?

```
s1 = tuple("...") # could be any string     $len(s1) = N$ 
s2 = tuple("...")     $len(s2) = N$ 
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1):  $N!$  steps
```

```
    if p == s2:  $N$  steps
```

```
        matches = True
```

version B

```
s1 = sorted(s1)     $N \log N$ 
```

```
s2 = sorted(s2)     $N \log N$ 
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A? $O(N * N!)$

what is the complexity of version B? $O(N \log N + N \log N)$

```
s1 = tuple("...") # could be any string    len(s1) = N
s2 = tuple("...")                                len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1): N! steps
```

```
    if p == s2: N steps
```

```
        matches = True
```

version B

```
s1 = sorted(s1)    N logN
```

```
s2 = sorted(s2)    N logN
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A? $O(N * N!)$

what is the complexity of version B? $O(N \log N + N \log N) = O(2N \log N)$

```
s1 = tuple("...") # could be any string    len(s1) = N
s2 = tuple("...")                                len(s2) = N
```

For Example, $s1 = (A, B, C)$, then permutations of $s1$ are

```
ABC  BCA
ACB  CAB
BAC  CBA
```

| | | | | | |
|--------------|----------------|----------------|-----|--------------|-------------|
| N choices | N-1 choices | N-2 choices | ... | 2 choices | 1 choice |
|--------------|----------------|----------------|-----|--------------|-------------|

Total choices = $N * (N-1) * (N-2) * \dots * 2 * 1$

Therefore, total permutations for $(A, B, C) = 3 * 2 * 1$

version A

```
import itertools
```

```
matches = False
```

```
for p in itertools.permutations(s1): N! steps
```

```
    if p == s2: N steps
```

```
        matches = True
```

version B

```
s1 = sorted(s1)    N logN
```

```
s2 = sorted(s2)    N logN
```

```
matches = (s1 == s2)
```

Examples, merge sort, quick sort

assumed sorted is $O(N \log N)$

what is the complexity of version A? $O(N * N!)$

what is the complexity of version B? $O(N \log N + N \log N) = O(2N \log N) = O(N \log N)$

9

```
def selection_sort(L):  
    for i in range(len(L)):  
        idx_min = i  
        for j in range(i, len(L)):  
            if L[j] < L[idx_min]:  
                idx_min = j  
        # swap values at i and idx_min  
        L[idx_min], L[i] = L[i], L[idx_min]
```

if this runs $f(N)$ times, where $N=\text{len}(L)$,
then $f(N) =$

```
nums = [2, 4, 3, 1]  
selection_sort(nums)  
print(nums)
```

The complexity of selection sort is
 $O(\text{_____})$

9

```
def selection_sort(L):
    for i in range(len(L)):
        idx_min = i
        for j in range(i, len(L)):
            if L[j] < L[idx_min]:
                idx_min = j
        # swap values at i and idx_min
        L[idx_min], L[i] = L[i], L[idx_min]
```

if this runs $f(N)$ times, where $N=\text{len}(L)$,
then $f(N) =$

```
nums = [2, 4, 3, 1]
selection_sort(nums)
print(nums)
```

The complexity of selection sort is
 $O(\rule{1cm}{0.4pt})$

| i | # of items for the inner for loop |
|------------|--|
| 0 | N |
| 1 | N-1 |
| 2 | N-2 |
| ... | ... |
| N-1 | 1 |
| N | 0 |

9

```
def selection_sort(L):
    for i in range(len(L)):
        idx_min = i
        for j in range(i, len(L)):
            if L[j] < L[idx_min]:
                idx_min = j
        # swap values at i and idx_min
        L[idx_min], L[i] = L[i], L[idx_min]
```

```
nums = [2, 4, 3, 1]
selection_sort(nums)
print(nums)
```

if this runs $f(N)$ times, where $N=\text{len}(L)$,

then $f(N) = N + (N-1) + (N-2) + \dots + 2 + 1 + 0$

The complexity of selection sort is

$O(\underline{\hspace{2cm}})$

| i | # of items for the inner for loop |
|-----|-----------------------------------|
| 0 | N |
| 1 | N-1 |
| 2 | N-2 |
| ... | ... |
| N-1 | 1 |
| N | 0 |

9

```
def selection_sort(L):
    for i in range(len(L)):
        idx_min = i
        for j in range(i, len(L)):
            if L[j] < L[idx_min]:
                idx_min = j
        # swap values at i and idx_min
        L[idx_min], L[i] = L[i], L[idx_min]
```

```
nums = [2, 4, 3, 1]
selection_sort(nums)
print(nums)
```

if this runs $f(N)$ times, where $N=\text{len}(L)$,

then $f(N) = N + (N - 1) + (N - 2) + \dots + 2 + 1 + 0$

$$= \frac{N(N+1)}{2}$$

The complexity of selection sort is

$O(\underline{\hspace{2cm}})$

| i | # of items for the inner for loop |
|------------|--|
| 0 | N |
| 1 | N-1 |
| 2 | N-2 |
| ... | ... |
| N-1 | 1 |
| N | 0 |

9

```
def selection_sort(L):
    for i in range(len(L)):
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            if L[j] < L[idx_min]:
                idx_min = j
        # swap values at i and idx_min
        L[idx_min], L[i] = L[i], L[idx_min]
```

```
nums = [2, 4, 3, 1]
selection_sort(nums)
print(nums)
```

if this runs $f(N)$ times, where $N=\text{len}(L)$,

then $f(N) = N + (N-1) + (N-2) + \dots + 2 + 1 + 0$

$$= \frac{N(N+1)}{2} = \frac{N^2 + N}{2}$$

The complexity of selection sort is

$O(\underline{\hspace{2cm}})$

| i | # of items for the inner for loop |
|-----|-----------------------------------|
| 0 | N |
| 1 | N-1 |
| 2 | N-2 |
| ... | ... |
| N-1 | 1 |
| N | 0 |

9

```
def selection_sort(L):
    for i in range(len(L)):
        idx_min = i
        for j in range(i, len(L)):
            if L[j] < L[idx_min]:
                idx_min = j
        # swap values at i and idx_min
        L[idx_min], L[i] = L[i], L[idx_min]
```

```
nums = [2, 4, 3, 1]
selection_sort(nums)
print(nums)
```

if this runs $f(N)$ times, where $N=\text{len}(L)$,

then $f(N) = N + (N-1) + (N-2) + \dots + 2 + 1 + 0$

$$= \frac{N(N+1)}{2} = \frac{N^2 + N}{2} = \frac{N^2}{2} + \frac{N}{2}$$

The complexity of selection sort is

$O(\text{_____})$

| i | # of items for the inner for loop |
|-----|-----------------------------------|
| 0 | N |
| 1 | N-1 |
| 2 | N-2 |
| ... | ... |
| N-1 | 1 |
| N | 0 |

9

```

def selection_sort(L):
    for i in range(len(L)):
        idx_min = i
        for j in range(i, len(L)):
            if L[j] < L[idx_min]:
                idx_min = j
        # swap values at i and idx_min
        L[idx_min], L[i] = L[i], L[idx_min]

nums = [2, 4, 3, 1]
selection_sort(nums)
print(nums)

```

if this runs $f(N)$ times, where $N=\text{len}(L)$,

then $f(N) = N + (N-1) + (N-2) + \dots + 2 + 1 + 0$

$$= \frac{N(N+1)}{2} = \frac{N^2 + N}{2} = \frac{N^2}{2} + \frac{N}{2}$$

The complexity of selection sort is

$$O\left(\frac{N^2}{2} + \frac{N}{2}\right) = O\left(\frac{N^2}{2}\right)$$

| i | # of items for the inner for loop |
|-----|-----------------------------------|
| 0 | N |
| 1 | N-1 |
| 2 | N-2 |
| ... | ... |
| N-1 | 1 |
| N | 0 |

9

```
def selection_sort(L):
    for i in range(len(L)):
        idx_min = i
        for j in range(i, len(L)):
            if L[j] < L[idx_min]:
                idx_min = j
        # swap values at i and idx_min
        L[idx_min], L[i] = L[i], L[idx_min]
```

```
nums = [2, 4, 3, 1]
selection_sort(nums)
print(nums)
```

if this runs $f(N)$ times, where $N=\text{len}(L)$,

then $f(N) = N + (N-1) + (N-2) + \dots + 2 + 1 + 0$

$$= \frac{N(N+1)}{2} = \frac{N^2 + N}{2} = \frac{N^2}{2} + \frac{N}{2}$$

The complexity of selection sort is

$$O\left(\frac{N^2}{2} + \frac{N}{2}\right) = O\left(\frac{N^2}{2}\right) = O(N^2)$$

| i | # of items for the inner for loop |
|-----|-----------------------------------|
| 0 | N |
| 1 | N-1 |
| 2 | N-2 |
| ... | ... |
| N-1 | 1 |
| N | 0 |