

Big O Notation

Linear time:

- $O(n)$
- As n increases time to compute increases at same rate
- Ex. Looping through an array once (for loop)

Constant time:

- $O(1)$
- As n increases time to compute stays the same

Quadratic time:

- $O(n^2)$
- As n increases time increases much more
- Ex. Two for loops, one nested in another

Logarithmic time:

- $O(\log(n))$
- As n increases the smaller proportion of actual input the program has to go through so time is reduced compared to $O(n)$.

Different steps get added:

- Take the complexity for each step in the algorithm and add them together.
- Ex. Two steps, first = $O(n)$, second = $O(a)$, overall complexity is $O(n + a)$

Drop constants:

- $O(2n) = O(n)$
- Interested in linear vs quadratic, constants are dropped.

Different inputs leads to different variables:

- Ex. Two arrays are input to algorithm (a and b).
- First array is looped through, and second array is looped through inside first loop (nested).
- $O(a^2)$ is wrong.
- $O(a*b)$ is right.

Drop non-dominant terms:

- Ex. Two steps in algorithm
- First is $O(n)$
- Second is $O(n^2)$
- Overall complexity is $O(n + n^2)$ but n is linear and therefore non-dominant so is dropped leading to complexity of $O(n^2)$.
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