

CSC263

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1 Review

Abstract Data Types	Data Structures
specification	implementation
objects	data
operations	algorithms

Analysis - Runtime/complexity

- Worst-Case — Upper Bounds - O
- Best-Case — Lower Bounds - Ω
- ??? — Tight Bounds - Θ

When analyzing an algorithm, we are counting by steps. Steps are represented by any constant time operations, such that

$$t_A(x) = \text{Number of constant operations}$$

To be able to prove an upper bound, you need to compare two functions: usually, it's $t_A(x)$ compared to runtime. but runtime can have different values depending on the input size, and the order of the input, and that's where worst and best-case scenarios come into play. For a worst-case analysis, you take the largest possible value of runtime for a given input size, and the best case takes the smallest. can we use this fact in proving the tight bound?

Average-case running time

for each n , $S_n = \text{all inputs of size } n$ if we consider inputs to be random, S_n is a sample space.

we'll want a discrete probability distribution on S_n . For each $x \in S_n$, $\text{pr}(x)$

$$t(x) = \text{steps on input } x$$

Where x is a random variable.

$$\text{Average-case: } T(n) = E(t) = \sum_{x \in S_n} t(x) * Pr(x)$$

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EX: LinSearch(L, x):
number L is a linked list (precondition).
z = L.head (the first node)
while z != None and z.data != x:
    z = z.next
return z

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Average-case runtime? $S_n = ?$ with probability distribution? (aka $\Pr(x)$)
 $t(x)$?

Here, we need an exact expression for $t(x)$. There is a lot of different ways to count steps, but they would all be within a constant factor of each other. The trick we're going to do here is *choose some key operations s.t counting only these operation is within a constant factor of total time then set $t(x)$ = number of these key operations*

For LinSearch, the key operation is **`z.data != x`**