Running Times and Asymptotic Analysis

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CMSC 441 — Algorithms

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Outline

Running Times

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Properties of Θ

Reflexivity, Symmetry, Transitivity Polynomials Logarithms

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Random Access Machines

An important goal of analyzing an algorithm is to determine it's running time in a manner that is not tied to any specific computer architecture. To this end, we use an abstract computational model, the *Random Access Machine* (RAM).

- RAM can perform operations commonly found in real computers.
- Memory is flat: we do not concern ourselves with levels of cache or other details of real memory architectures.
- Data types are integers and floating point numbers.
- The model does not entail a fixed word size; however, for a particular algorithm, we typically assume some upper bound on word size to avoid unrealistic results.

Example: Binary Search

Input: a numeric array A of length n.

```
first = 1
 1
2 last = n
3
    while first < last
4
         middle = |(first + last)/2|
        if v \leq A[middle]
5
6
             last = middle
7
         else
8
              first = middle + 1
9
   if v == A[first]
         return first
10
11
    else
         return NIL
12
```

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Running Time of Binary Search

Simplified Running Time

We have that the running time of BINARY-SEARCH as a function of the length n of the input array is

 $T(n) = b + a \lg \lceil n \rceil$

where a and b are positive constants.

Further Simplification?

Intuitively, it seems that the important part of T(n) is the logarithmic term, which determines the "shape" of the curve, whereas the constant term just shifts the curve up by a constant amount. We need to justify this further simplification.

The Set Θ

Definition of Θ

Let g(n) be a function. The set $\Theta(g(n))$ consists of all funcitons f(n) for which there exists positive constants c_1 , c_2 , and n_0 such that, if $n \ge n_0$, then

$$c_1g(n) \leq f(n) \leq c_2g(n).$$

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If f(n) is in $\Theta(g(n))$, we may write

- "f(n) is $\Theta(g(n))$ "
- " $f(n) \in \Theta(g(n))$ "
- " $f(n) = \Theta(g(n))$ "

The Set Θ

Interpretation of Θ

For *n* sufficiently large, f(n) is sandwiched between $c_1g(n)$ and $c_2g(n)$. As *n* becomes large, f(n) has a similar "shape" as g(n).



Example: $b + a \lg \lceil n \rceil \in \Theta(\lg n)$

Proof of Upper Bound.

We will suppose that some $c_2 > 0$ exists such that

$$b + a \lg \lceil n \rceil \le c_2 \lg \lceil n \rceil$$

and try to derive a lower bound on n that makes this true. Rearranging the inequality gives the equivalent expression

$$\frac{b}{c_2-a} \leq \lg \lceil n \rceil,$$

assuming that $c_2 - a > 0$. Exponentiating each side results in

$$2^{b/(c_2-a)} \leq \lceil n \rceil.$$

This last inequality certainly holds if $n \ge 2^{b/(c_2-a)}$.

Example: $b + a \lg \lceil n \rceil \in \Theta(\lg n)$

Proof of Lower Bound. Similarly, suppose c_1 exists such that

$$b + a \lg \lceil n \rceil \ge c_1 \lg \lceil n \rceil$$

and, assuming that $c_1 - a < 0$, derive the equivalent expression

$$\lceil n \rceil \geq 2^{b/(c_1-a)}$$

This last inequality is certainly true if $n \ge 2^{b/(c_1-a)}$.

Example: $b + a \lg \lceil n \rceil \in \Theta(\lg n)$

Completing the Proof.

Let n_0 be the larger of $2^{b/(c_2-a)}$ and $2^{b/(c_1-a)}$. Then if $n \ge n_0$, we may follow our previous arguments "backwards" to conclude that

$$c_1 \lg \lceil n \rceil \leq b + a \lg \lceil n \rceil \leq c_2 \lg \lceil n \rceil.$$

and so $b + a \lg \lceil n \rceil \in \Theta(\lg \lceil n \rceil)$. The last step is to deal with the ceiling functions. We can show that, for $n \ge 2$,

$$\lg n \le \lg \lceil n \rceil < \lg (n+1) \le 2 \lg n,$$

so $\lg \lceil n \rceil \in \Theta(\lg n)$. It follows that

 $b + a \lg \lceil n \rceil \in \Theta(\lg n).$

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Some Properties of Θ

We will prove a few basic properties about Θ :

- Reflexivity. $f(n) \in \Theta(f(n))$
- ▶ Symmetry. $f(n) \in \Theta(g(n))$ if and only if $g(n) \in \Theta(f(n))$
- ▶ Transitivity. If $f(n) \in \Theta(g(n))$ and $g(n) \in \Theta(h(n))$, then $f(n) \in \Theta(h(n))$

Proof of Reflexivity.

The first property is really easy to prove. Since for all n,

$$1 \cdot f(n) \leq f(n) \leq 1 \cdot f(n),$$

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 $f(n) \in \Theta(f(n))$ (with $c_1 = c_2 = 1$ and n_0 any value).

Some Properties of Θ

Proof of Symmetry.

Suppose $f(n) \in \Theta(g(n))$. Then there exists positive constants c_1 , c_2 , and n_0 such that if $n \ge n_0$, then

$$c_1g(n) \leq f(n) \leq c_2g(n).$$

It follows that

$$\frac{1}{c_2}f(n)\leq g(n)\leq \frac{1}{c_1}f(n),$$

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so $g(n) \in \Theta(f(n))$. The converse is similar.

Some Properties of Θ

Proof of Transitivity.

Suppose $f(n) \in \Theta(g(n))$ and $g(n) \in \Theta(h(n))$. Then there exists positive constants c_1 , c_2 , n_0 and c'_1 , c'_2 , n'_0 such that if $n \ge n_0$, then

$$c_1g(n) \leq f(n) \leq c_2g(n).$$

and if $n \ge n'_0$, then

$$c_1'h(n) \leq g(n) \leq c_2'h(n).$$

So, if $n \ge \max(n_0, n'_0)$ we have that

 $c_1'c_1h(n) \leq f(n) \leq c_2'c_2h(n)$

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and so $f(n) \in \Theta(h(n))$.

Polynomials and Θ

Theorem

Let p(n) be a degree k polynomial with positive leading coefficient; that is,

$$p(n) = a_k n^k + a_{k-1} n^{k-1} + \dots + a_1 n + a_0$$

where a_0, a_1, \ldots, a_k are real coefficients with $a_k > 0$. Then $p(n) \in \Theta(n^k)$.

Examples

The theorem makes it much easier to determine the Θ class of polynomial functions.

1.
$$n^2 + n + 1 \in \Theta(n^2)$$
.
2. $\frac{n}{2} + 1 = \frac{1}{2}n + 1 \in \Theta(n)$.

Logarithms and $\boldsymbol{\Theta}$

The principle behind the previous Theorem is that when we have a sum of terms, the asymptotically largest (fastest growing) dominates and determines the Θ class.

Theorem

Any positive power of $\lg n$ is asymptotically smaller than any positive power of n; that is, for any positive constants α and β , $(\lg n)^{\beta}$ is asymptotically smaller than n^{α} .

Example

This tells us that $\lg n$ is much smaller asymptotically than n.

- 1. $(\lg n)^{700}$ is asymptotically smaller than $n^{0.00001}$.
- 2. $n^2 + \lg n \in \Theta(n^2)$. Apply the principle that the larger term determines the Θ class.
- 3. $n^2 + n \lg n \in \Theta(n^2)$.

Logarithms and Θ

Sketch of Proof.

Consider the limit

$$\lim_{n\to\infty}\frac{(\lg n)^{\beta}}{n^{\alpha}}.$$

Applying L'Hospital's rule some finite number of times, we show that the limit is zero. Therefore, for any $\epsilon > 0$ and *n* sufficiently large,

$$(\lg n)^{\beta} < \epsilon n^{\alpha},$$

so $(\lg n)^{\beta}$ is dominated by n^{α} .