

Overview of Big-O Notation

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Review of Big-O Notation

- The notation “ $g(n) = O(f(n))$ ” means:
 - “for all large n , $g(n)$ less than $c \cdot f(n)$ for some constant $c > 0$ ”.
- Examples:

$$20n + 50 = O(n).$$

$$5n^2 + 34n + 3 = O(n^2)$$

$$10 = O(1).$$

$$10 \cdot \log(n) + n = O(n).$$

$$n \cdot \log(n) + 20 \cdot n = O(n \log n).$$

$$2^n + 1000 \cdot n^{10} = O(2^n).$$

Review of Big-O Notation

- “Runtime of algorithm is $O(n)$ ” means that:
 - “In worst case, algorithm requires $O(n)$ operations.”
 - Typically, ‘ n ’ will measure size of input.
- Why is this important?
 - We can only apply $O(2^n)$ algorithms to tiny datasets.
 - We can apply $O(n^2)$ algorithms to medium-sized dataset.
 - We can apply an $O(nd)$ algorithm if one of ‘ n ’ or ‘ d ’ is medium-sized.
 - We can apply $O(n)$ algorithms to huge datasets.

Review of Big-O Notation

- Examples of algorithm runtimes:
 - Finding the maximum in a list of 'n' numbers: $O(n)$.
 - We do a constant number of operations for each of the 'n' numbers.
 - Finding item number 'i' in a list of 'n' numbers: $O(1)$.
 - Just return that number.
 - Printing each element times each other element: $O(n^2)$:
 - A constant number of operations for each combination.
 - Finding an element in a sorted list: $O(\log n)$.
 - Sorting a list of N numbers: $O(n \log n)$.
 - Well-known result, see algorithms class.
 - Classifying an example in depth-m decision tree: $O(m)$.
 - You do a constant number of operations at each depth.
 - N.B., no dependence on dimension 'd'.

Review of Big-O Notation

- Examples of algorithm runtimes depending on two parameters:
 - Finding the maximum of an 'n' by 'd' matrix: $O(nd)$
 - Fitting decision stump with 'n' objects and 'd' features:
 - $O(n^2 d)$ if score costs $O(n)$.
 - $O(n d \log n)$ if all scores cost $O(n \log n)$.
 - Fitting a decision tree of depth 'm':
 - Naïve analysis: 2^{m-1} trees, so $O(2^{m-1} n d \log(n))$.
 - But each object appears once at each depth: $O(m n d \log n)$.
 - Finding optimal decision tree:
 - There are $O(C(n) \cdot n!)$ tree structures, where $C(n)$ is huge.
 - Then you also have to find the thresholds for each structure.
 - You are never going to do this.