

## Complexity Workshop

Here is a small example of how to think about and work with algorithm complexity.

### Data Structure Efficiencies

The elementary unit of analysis is “*memory accesses*”, which may include storing an item (constructors), locating an item (recognizers), and retrieving an item (accessors).

<i>Arrays:</i>	A[i]	given index i, go straight to A[i]	O[1]
<i>Trees:</i>	((a b)(c d))	a series of branching decisions locates item i	O[ln n]
<i>Lists:</i>	(a b c)	look through all items for i	O[n]

### The Memory Hierarchy (1999 technology)

<i>Type</i>	<i>typical access time (<math>\wedge 10</math> ns)</i>		<i>typical capacity (bytes) (<math>\wedge 2</math> bits)</i>	
<i>registers</i>	2-5 ns	0	64-512	9-12
<i>primary cache</i>	4-10 ns	1	8K-256K	16-21
<i>secondary cache</i>	20-100 ns	2	512K-4M	22-25
<i>main memory</i>	50-1000 ns	3	8M-4G	26-35
<i>disk</i>	5-15 ms	7	500M-1T	32-41
<i>tape</i>	1-50 s	10	unlimited	

1 ns =  $10^{-9}$  s      1 ms =  $10^{-3}$  s

### Pragmatics of Nested Loops

Nested loops effectively do a brute-force search over all items. Consider searching a three dimensional matrix, indexed by (i, j, k):

```

for each i do
  for each j do
    for each k do
      <process>

```

If every item must be processed (e.g.: a pixel-based graphics display process), then the loops are unavoidable, and the best case is also the worst case.

However, if one item must be found out of  $n$ , then we can avoid some effort. The two effort avoidance techniques are

***data structure organization*** and ***smart, knowledgeable search***

**When is Organization better than Knowledge?**

Search is an exponential process. Ignoring effort other than direct search, and assuming the data can be structured as a binary decision process, then

Search-effort =  $2^n$                        $n$  is number of items in search pool

How deeply can loops be nested to have the same efficiency as search? Ignoring polynomial coefficients:

Loop-effort =  $n^k$                        $k$  is the number of nested loops

Point of equal effort:

$2^n = n^k$

Solving for  $k$ :

$n / \ln n = k$

Table of  $n / \ln n$ :

$\ln n$	$n$	$n / \ln n$
1	2	1
2	4	2
3	8	2.7
4	16	4
5	32	6.4
6	64	10
7	128	18
8	256	32
9	512	57
10	1024	102
11	2048	186
12	4096	341

Conclusion: worst case search through even small sets is worse than many nested loops.

**Pragmatics**

In fact, worst case search requires every decision to be incorrect. This is very difficult to achieve, since it requires perfect anti-knowledge.

Let  $p$  be the fraction of times that an incorrect decision is made. For example, in a sorted binary tree, the correct decision is made every time, since the structure of sorting gives the needed contextual information. In the perfectly sorted case, with no search errors:

$n$  decisions requires  $\ln n$  steps,  $2^n$  decisions requires  $n$  steps.

Conclusion: Sorting turns search into looping

In general, making  $P$  correct decisions:

$$2^{(nP)} = n^k$$

$$Pn / \ln n = k$$

Here is how knowledge effects search effort:

$\ln n$	$n$	<i>all wrong</i> $P=1$ $n / \ln n$	<i>half/half</i> $P=2^{-1}$	<i>1 in 10</i> $P=2^{-3}$	<i>1 in 100 wrong</i> $P=2^{-7}$
1	2	1			
2	4	2	1		
3	8	2.7	1.4		
4	16	4	2		
5	32	6.4	3.2		
6	64	10	5	1.2	
7	128	18	9	2	
8	256	32	16	4	
9	512	57	29	7	
10	1024	102	51	13	
11	2048	186	93	23	1.5
12	4096	341	170	42	2.7

When we guess correctly half of the time, the cross-over point between search and brute-force looping is at about 30 items for  $k=3$ . As errors decrease to 1 in 100, the number of items increases to over 4000. Thus, partial knowledge about the location of an item greatly increases the number of items that can be searched before a search strategy becomes less desirable than brute-force looping.