

CS7960 L2 : Review of Sequential Model

Turing Machines (Alan Turing 1936)

- single tape: moveL moveR, read, write
 - each constant time
 - constant pointer memory
 - tape infinite (extra memory)

Von Neumann Architecture (Von Neumann + Eckert + Mauchly 1945)

- based on ENIAC
 - CPU + Memory (RAM): read, write, op = constant time
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Scanning (max)

- TM : $O(n)$
- VNA: $O(n)$

Sorting

- TM : $O(n^2)$
- VNA: $O(n \log n)$

Searching

- TM : $O(n)$
- VNA: $O(\log n)$

how big is $\log n$, n , $n \log n$, n^2 :

10^x	1	2	3	4	5	6
7	8	9				
search	0.000001	0.000001	0.000001	0.000002	0.000001	
	0.000002	0.000002	0.000007	0.001871		
MAX	0.000003	0.000005	0.000006	0.000048	0.000387	

0.003988 0.040698 9.193987 >15 min
 QuiS | 0.000005 0.000030 0.000200 0.002698 0.029566
 0.484016 7.833908 137.9388
 BubS | 0.000003 0.000105 0.007848 0.812912 83.12960 ~2
 hour ~9 days ???

Gradations:

LOG	poly log (n)	:	$\log^c(n)$
P	poly (n)	:	n^c
-- NP --			
EXP	exp (n)	:	c^n

Theory:

- LOG not studied much since count loading of data
 - P is poly (n). Lots of neat algorithms.
Sometimes constant c (in n^c) important, sometimes not.
 - EXP usually hopeless, but 1.000001^n is ok.
 - NP : verify solution in P, find solution conjectured EXP.
if EXP number of (parallel) machines \rightarrow in P. (bits of solution argument)
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Matrices

Vector $v = [v_1 \ v_2 \ \dots \ v_n]^T$
 $u = [u_1 \ u_2 \ \dots \ u_m]^T$

Dot Products

$\langle v, u \rangle = v \text{ (dot) } u = v^T u$
= $\sum_i u_i * v_i$
(need $m = n$)

Theta(n)

$v^T u = R$ [$n \times m$] matrix.
 $R_{i,j} = v_i * u_j$

Theta(n^2)

Matrix Multiply:

$R = [n \times m]$ and $T = [m \times k]$ matrices

$R T = U$ a [$n \times k$] matrix

$U_{i,j} = \langle R_{i,*}, T_{*,j} \rangle = R_{i,*} T_{*,j}$

$O(n^3) \rightarrow O(n^{2.807})$ [Strassen 69] $\rightarrow \dots O(n^{2.376})$
[Coppersmith Winograd 90]
Omega(n^2)

Probability:

Let A, B be random variables.

$\Pr[A] * \Pr[B] = \Pr[A \text{ and } B]$ iff A and B are independent.

$\Pr[A \text{ and } B] < \Pr[A] + \Pr[B]$ "Union Bound"

Expected value $A = E[A] = \sum_{a \in U} a * \Pr[a = A]$

$E[A] + E[B] = E[A + B]$ "Linearity of Expectation"

Hash Functions:

$h : U \rightarrow [n]$

U := set of possible inputs, maybe $[m]$, maybe $[a-z, A-Z]^{\geq 28}$
 $[n]$:= output universe

H = family of hash functions.

If H *universal* for $x \neq y$ then $\Pr_{\{h \in H\}}[h(x) = h(y)] \leq 1/n$

Simple example

$h_{\{a,b\}}(x) = ((a x + b) \bmod p) \bmod n$
where a in $[1, p]$ and b in $[0, p]$, both at random, and $p > m$ and prime.

Multiply-Shift hashing (Dietzfelbinger 97)

high-order-bits($h_a(x)$) = $(a x \bmod 2^w), N$ // top M bits of first arg
where $a < 2^w$ (odd, at random), $w :=$ number of bits in machine word, $n = 2^N$