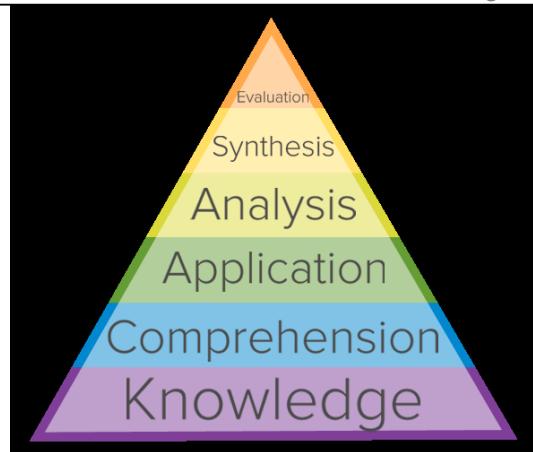


Bloom's Hierarchy of cognitive learning



- *What is thought is not said*
- *What is said is not heard*
- *What is heard is not understood*
- *What is understood is not believed*
- *What is believed is not yet advocated*
- *What is advocated is not yet acted on*
- *What is acted on is not yet completed*

Konrad
Lorenz
(Nobel
Prize
1973)

Syllabus

- "Virtues" of Computer Science
- Algorithms vs. Heuristic, Program
- Asymptotic Efficiency
- Example: Powering
- Example: Fibonacci Numbers
- Example: Polynomial Multiplication

"*Virtues*":

- problem specification
- formal semantics
- **algorithm design** #program
 #heuristic
- and analysis
(correctness, efficiency)
- proof of optimality
(complexity, CS422)



Algorithm \neq Heuristic, Program

An algorithm is a

- finite sequence of
- primitive instructions that,
executed according to their
- well-specified semantics,
provide a *mechanical*
solution to the *infinitely*
many instances
of a possibly *complex*
mathematical problem.

1. fully specified
(input/output)
2. guaranteed correct
(no heuristic/recipe)
3. analysis of cost
(runtime, memory, ...)
4. optimality proof (wrt
a model of computation)

Algorithm ≠ Heuristic, Program

```
    mov esi, offset list  
top:   mov edi, esi  
inner:  mov eax, [edi]  
        mov edx, [edi+4]  
        cmp eax, edx  
        jle no_swap  
        mov [edi+4], eax  
        mov [edi], edx  
no_swap: add edi, 4  
        cmp edi, list_end - 4  
        jb inner  
        add esi, 4  
        cmp esi, list_end - 4  
        jb top
```

- primitive operations
- their semantics
- their costs

```
# vowels list  
vowels = ['e', 'a', 'u', 'o', 'i']  
# sort the vowels  
vowels.sort()  
# print vowels  
print('Sorted list:', vowels)
```

Asymptotic Efficiency

n	$\log_2 n \cdot 10s$	$n \cdot \log n$ sec	n^2 msec	n^3 μ sec	2^n nsec
10	33sec	33sec	0.1sec	1msec	1msec
100	≈1min	11min	10sec	1sec	40 Mrd. Y
1000	≈1.5min	≈3h	17min	17min	
10 000	≈2min	1.5 days	≈1 day	11 days	
100 000	≈2.5min	19 days	4 months	32 years	

- Running times of some sorting algorithms
 - **BubbleSort:** $O(n^2)$ comparisons and copy instructions
 - **QuickSort:** typically $O(n \cdot \log n)$ steps
but $O(n^2)$ in the worst-case
 - **HeapSort:** always at most $O(n \cdot \log n)$ operations
 - **BucketSort:** $O(n)$ operations
 - SORT primitive: $O(1)$
- **Worst-case vs. average-case vs. best case**
w.r.t. input size =: $n \rightarrow \infty$

Example: Powering

Optimality?

Powering Problem: Given X and $n \in \mathbb{N}$.

Compute X^n with few(est number of) multiplications

- $X^n = X \cdot X \cdot \dots \cdot X$: $n-1$ multiplications
- Let $k := \lfloor n/2 \rfloor$ and recursively compute X^k ,
then compute $X^n = (X^k)^2$ or $X^n = (X^k)^2 \cdot X$
- #multiplications $T(n) \leq T(n/2) + 2$, $T(n) \leq 2 \cdot \log_2(n)$
- Asympt. optimality: Each multiplication at most doubles the degree of the intermediate results;
so computing x^n requires at least $\log_2 n$ of them.

Example: Fibonacci Numbers

```
FibIter(n)
if n=0 return 0;
fib := 1; fibL := 0;
while n>1 do
    tmp:=fibL;
    fibL := fib;
    fib := fibL + tmp;
    n := n-1 ; end
return fib;
```

$$F_0=0, \quad F_1=1, \quad F_n = F_{n-1} + F_{n-2}$$

FibRek(n)

```
if n=0 return 0; if n=1 return 0;
return FibRek(n-1) + FibRek(n-2);
```

$$F_n = (\varphi^n - (-1/\varphi)^n) / \sqrt{5}$$
$$\varphi := (1+\sqrt{5})/2$$

$$\begin{vmatrix} F_n \\ F_{n-1} \end{vmatrix} = \begin{vmatrix} 1 & 1 \\ 1 & 0 \end{vmatrix} \cdot \begin{vmatrix} F_{n-1} \\ F_{n-2} \end{vmatrix} = \begin{vmatrix} 1 & 1 \\ 1 & 0 \end{vmatrix}^{\textcircled{k}} \cdot \begin{vmatrix} F_{n-k} \\ F_{n-k-1} \end{vmatrix} \quad k := n-1$$

Example: Polynomial Multiplication

Long Multiplication

Input: coefficients of polynomials

$A(x) = a_0 + a_1x + a_2x^2 + \dots + a_{N-1}x^{N-1}$ and $B(x)$ of degree $< N$.

Output: coefficients of polynomial

w.l.o.g. $2|N$

$C(x) := A(x) \cdot B(x)$ of degree $\leq K := 2N-2$.

e.g. $\times(-5)$ or $+$

$T(N) := \#$ arithmetic operations (multiplications, linear combinat.s)

Recursive algorithm, Distributive law: $T(N) = 4 \cdot T(N/2) + O(N)$

"Naïve" $c_k = \sum_j a_j \cdot b_{k-j}$

$$(A_0(x) + A_1(x) \cdot x^{N/2}) \cdot (B_0(x) + B_1(x) \cdot x^{N/2}) = C_0(x) + C_1(x) \cdot x^{N/2} + C_2(x) \cdot x^N$$

$$C_0(x) = A_0(x) \cdot B_0(x) \quad C_1(x) = A_0(x) \cdot B_1(x) + A_1(x) \cdot B_0(x) \quad C_2(x) = A_1(x) \cdot B_1(x)$$

$$C(x) = [c_0, \dots, c_{N/2-1}, c_{N/2}, \dots, c_{N-1}, c_{N+1}, \dots, c_{3/2 \cdot N-1}, \dots, c_{2N-2}]$$

Example: Polynomial Multiplication

Karatsuba

Input: coefficients of polynomials

$A(x) = a_0 + a_1x + a_2x^2 + \dots + a_{N-1}x^{N-1}$ and $B(x)$ of degree $< N$.

Output: coefficients of polynomial

w.l.o.g. $2|N$

$C(x) := A(x) \cdot B(x)$ of degree $\leq K := 2N-2$.

e.g. $\times(-5)$ or $+$

$T(N) := \#$ arithmetic operations (multiplications, linear combinat.s)

Recursive algorithm, Distributive law: $T(N) = 4 \cdot T(N/2) + O(N)$

based on: „*Karatsuba law*“: $T(N) = 3 \cdot T(N/2) + O(N)$

$$(A_0(x) + A_1(x) \cdot x^{N/2}) \cdot (B_0(x) + B_1(x) \cdot x^{N/2}) = C_0(x) + C_1(x) \cdot x^{N/2} + C_2(x) \cdot x^N$$

$$C_0(x) := A_0(x) \cdot B_0(x), \quad C_2(x) := A_1(x) \cdot B_1(x)$$

$$C_1(x) := (A_0(x) + A_1(x)) \cdot (B_0(x) + B_1(x)) - C_0(x) - C_2(x)$$

Example: Polynomial Multiplication

$$\begin{aligned} T_1 &:= (A_0 + 2A_1 + 4A_2) \bullet (B_0 + 2B_1 + 4B_2) \\ T_2 &:= (A_0 + A_1 + A_2) \bullet (B_0 + B_1 + B_2) \\ T_3 &:= (4A_0 + 2A_1 + A_2) \bullet (4B_0 + 2B_1 + B_2) \end{aligned}$$

Toom

$$C_3 = -C_0 + \frac{1}{3}T_1 - 2T_2 + \frac{1}{6}T_3 - 3\frac{1}{2}C_4$$

w.l.o.g. $3|N$

$$C_2 = 3\frac{1}{2}C_0 - \frac{1}{2}T_1 + 5T_2 - \frac{1}{2}T_3 + 3\frac{1}{2}C_4$$

$$C_1 = -3\frac{1}{2}C_0 + \frac{1}{6}T_1 - 2T_2 + \frac{1}{3}T_3 - C_4$$

$$C_0 = A_0 \bullet B_0, \quad \text{Distributive law: } T(N) = 4 \cdot T(N/2) + O(N)$$

$$C_4 = A_2 \bullet B_2, \quad \text{„Karatsuba law“: } T(N) = 3 \cdot T(N/2) + O(N)$$

$$\text{„Toom’s law“: } T(N) = 5 \cdot T(N/3) + O(N)$$

$$\begin{aligned} &(A_0(x) + A_1(x) \cdot x^{N/3} + A_2(x) \cdot x^{2N/3}) \times (B_0(x) + B_1(x) \cdot x^{N/3} + B_2(x) \cdot x^{2N/3}) \\ &= C_0(x) + C_1(x) \cdot x^{N/3} + C_2(x) \cdot x^{2N/3} + C_3(x) \cdot x^{3N/3} + C_4(x) \cdot x^{4N/3} \end{aligned}$$

Example: Polynomial Multiplication

Toom-Cook

Input: coeff. of $A(x) = a_0 + a_1x + a_2x^2 + \dots$ of $\deg < N$, $B(x)$ of $\deg < M$

Output: coefficients c_0, \dots, c_K of $C(x) := A(x) \cdot B(x)$, $\deg(C) \leq K := N+M-2$

Long Multiplication: $N \cdot M$ products & linear combinations

$$\begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^K \\ 1 & x_1 & x_1^2 & \dots & x_1^K \\ \vdots & & & & \\ 1 & x_K & x_K^2 & \dots & x_K^K \end{pmatrix}^{-1} \cdot \begin{pmatrix} A(x_0) \cdot B(x_0) \\ A(x_1) \cdot B(x_1) \\ \vdots \\ A(x_K) \cdot B(x_K) \end{pmatrix} = \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_K \end{pmatrix}$$

Vandermonde Matrix
invertible for any
distinct fixed x_0, \dots, x_K

Evaluation/Interpolation: $K+1$ products, $O(N^2+M^2)$ linear comb.s

$$T(N, M) = (n+m-1) \cdot T(N/n, M/m) + O((N+M) \cdot (n+m))$$

$$\begin{aligned} &(A_0(x) + A_1(x) \cdot x^{N/n} + A_2(x) \cdot x^{2N/n} + \dots + A_{n-1}(x) \cdot x^{(n-1) \cdot N/n}) \\ &\times (B_0(x) + B_1(x) \cdot x^{M/m} + B_2(x) \cdot x^{2M/m} + \dots + B_{m-1}(x) \cdot x^{(m-1) \cdot M/m}) \end{aligned}$$