Design and Analysis of Algorithms CS218M

Amortized Complexity

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Autumn, 2022

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- Before you delete k elements you must do k push, each of $O(1)$. Hence Aggregate cost of k push and one Muplipop(S,k) is $O(k)$ and amortized cost per operation is $O(1)$.

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- Let, $CREDIT = \Sigma_{i=1}^{n} \hat{c}_i - \Sigma_{i=1}^{n} c_i$. We require that $CREDIT \geq 0$.
- Potential of a datatype configuration D is denoted $\Phi(D)$. It gives CREDIT retained by reaching D. It can define Credit

Potential

- Let op_i take datatype from config D_{i-1} to D_i.
  Hence $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1})$..
  Hence, $\Sigma_{i=1}^{n} \hat{c}_i = (\Sigma_{i=1}^{n} c_i) + \Phi(D_n) - \Phi(D_0)$
- If invariantly$\Phi(D_i) - \Phi(D_{i-1}) \geq 0$ then aggregate $\Sigma_{i=1}^{n} \hat{c}_i$ is upperbound on $Sigma_{i=1}^{n} c_i$.
- Potential function determines amortized cost $\hat{c}_i$ of each operation.
- We typically define $\Phi(D_0) = 0$ and require that $\Phi(D_i) \geq 0$.

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- Let $Op_i = Push(S, x)$: We have $\Delta\Phi = 1$.
  Hence, $\hat{c_i} = c_i + \Delta\Phi(D) = 1 + 1$.
- Let $Op_i = Pop(S$: We have $\Delta\Phi = -1$.
  Hence, $\hat{c_i} = c_i + \Delta\Phi(D) = 1 + (-1)$.
- Let $Op_i = Multipop(s, k)$: We have $\Delta\Phi = -min(s, k)$.
  Hence, $\hat{c_i} = c_i + \Delta\Phi = min(s, k) + (-min(s, k)) = 0$.

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- Let $Op_i = Multipop(s, k)$: We have $\Delta\Phi = -min(s, k)$.
  Hence, $\hat{c_i} = c_i + \Delta\Phi = min(s, k) + (-min(s, k)) = 0$.
- Hence amortized cost of each stack operation is $O(1)$.

Binary Counter

Counter value	A[7]	A[6]	A[5]	A[4]	A[3]	A[2]	A[1]	A[0]	Total cost
0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	1	1
2	0	0	0	0	0	0	1	0	3
3	0	0	0	0	0	0	1	1	4
4	0	0	0	0	0	1	0	0	7
5	0	0	0	0	0	1	0	1	8
6	0	0	0	0	0	1	1	0	10
7	0	0	0	0	0	1	1	1	11
8	0	0	0	0	1	0	0	0	15
9	0	0	0	0	1	0	0	1	16
10	0	0	0	0	1	0	1	0	18
11	0	0	0	0	1	0	1	1	19
12	0	0	0	0	1	1	0	0	22
13	0	0	0	0	1	1	0	1	23
14	0	0	0	0	1	1	1	0	25
15	0	0	0	0	1	1	1	1	26
16	0	0	0	1	0	0	0	0	31

Binary Counter of k bits

represented as bit-array $A[0..k-1]$.

```
INCREMENT(A)
1  i = 0
2  while i < A.length and A[i] == 1
3        A[i] = 0
4        i = i + 1
5  if i < A.length
6        A[i] = 1
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- Worst case complexity of *INC* is $\theta(k)$.
- Amortized worst case complexity of *INC* is ???.

Amortized Complexity: Counter

Let $\Phi(D)$ be number of trailing 1 in config D. Let $t_i = \Phi(D_{i-1}$.

- $c_i = t_i + 1$, Also $\Delta\Phi = -t_i$.
- Hence $\hat{c}_i = c_i + \Delta\Phi = 1$. Hence $O(1)$.

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  $T.size$ the size of table.
  $T.num$ Number of occupying elements in table.
- If $T.size = T.num$ then before insert, allocate double sized table and copy.

Dynamic Table Insert

TABLE-INSERT(T, x)

```
 1  if T.size == 0
 2      allocate T.table with 1 slot
 3      T.size = 1
 4  if T.num == T.size
 5      allocate new-table with 2 · T.size slots
 6      insert all items in T.table into new-table
 7      free T.table
 8      T.table = new-table
 9      T.size = 2 · T.size
10  insert x into T.table
11  T.num = T.num + 1
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- Initially, $T.num = T.size = 0$. Hence $\Phi_0 = 0$.
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- Case 1: $T.num < T.size$. Then,

$$\begin{aligned}
\hat{c}_i &= c_i + \Phi_i - \Phi_{i-1} \\
&= 1 + (2 \cdot num_i - size_i) - (2 \cdot num_{i-1} - size_{i-1}) \\
&= 1 + (2 \cdot num_i - size_i) - (2(num_i - 1) - size_i) \\
&= 3 .
\end{aligned}$$

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- Case 2: $T.num = T.size$. Then,

$$\begin{aligned}
\hat{c}_i &= c_i + \Phi_i - \Phi_{i-1} \\
&= num_i + (2 \cdot num_i - size_i) - (2 \cdot num_{i-1} - size_{i-1}) \\
&= num_i + (2 \cdot num_i - 2 \cdot (num_i - 1)) - (2(num_i - 1) - (num_i - 1)) \\
&= num_i + 2 - (num_i - 1) \\
&= 3 .
\end{aligned}$$

TABLE-DELETE

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- Double the table on INSERT when $\alpha_{i-1} = 1$.

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$$\Phi(T) = \begin{cases} 2 \cdot T.num - T.size & \text{if } \alpha(T) \geq 1/2\,, \\ T.size/2 - T.num & \text{if } \alpha(T) < 1/2\,. \end{cases}$$

Dynamic Programming

Problem Compute $C(n,k)$

$$\frac{n!}{k!\,(n-k)!}$$

{Rec: $n \geq 1$, $k \geq 0$, $n \geq k$}

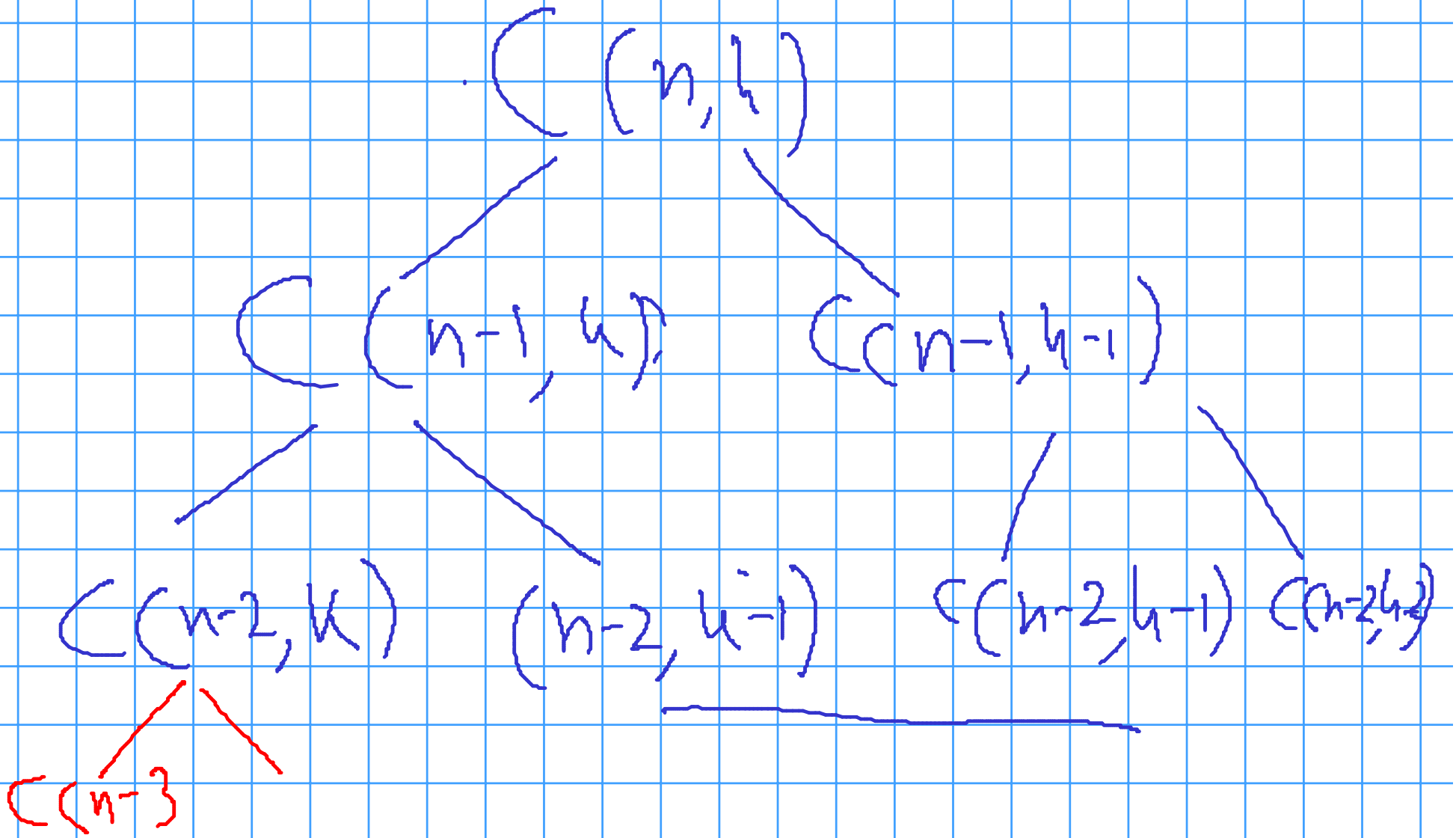
$C(n,k)$

if $k=0$ or $k=n$

return 1

else $C(n-1,k-1) + C(n-1,k)$

$$T(n,k) = 2 \cdot T(n,k-1) + \Theta(1)$$



Memorization

$T(i,j)$ $1 \le i \le n$, $0 \le j \le k$

1) Initialize $T(i,j) = -1$

C(n,k)

Allocate $T(i,j)$, n

Initialize

$T(i,j) = -1$ $1 \le i \le n$, $0 \le j \le k$

MEMOIZED-C(n,k)

```
MEMOIZED_C(n,k)
    if T[n,k] >= 0
        return T[n,k]
    else if k = 0 V k = n
        T[n,k] = 1
        return T[n,k]
    else
        T[n,k] = MEMOIZED(n-1, k) +
                 MEMOIZED(n-1, k-1)
        return T[n,k].
```

1. Give a name to procedure and describe input & output
2. Write a recursive procedure.
3. Determine num of distinct calls
4. Allocate table & initialize
5. Use memoized version of the recursive procedure

Input: A set S of n intervals
for interval i we have S[i] f[i]

Output subset T of S w(i) → weight positive int.
of maximum weight
s.t. disjoint(T).

1. MWNOSI(S) |S|=n

max (MWNOSI(S − S[1]),
MWNOSI(S') + W(1))