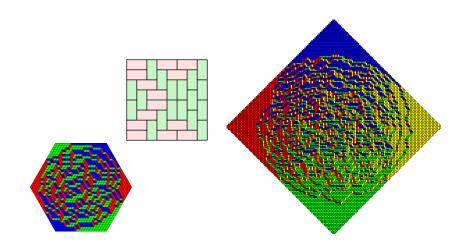
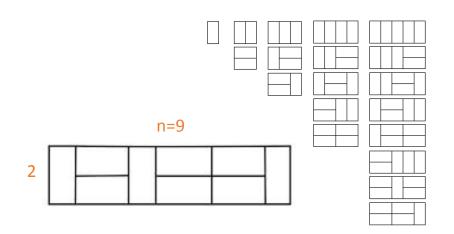
# Unit II: Dynamic Programming CISC 380 Algorithms

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## Domino Tilings



#### Domino Tilings



Problem: Find the number of domino tilings of a 2 by n region.

#### Example: Domino Tilings WITH recursion

Problem: Count the number of domino tilings of a  $2 \times n$  region.

```
function DominoTilings(n)

if n \le 0 then return 0

if n == 1 then return 1

if n == 2 then return 2

return DominoTilings(n - 1) + DominoTilings(n - 2)
```

The running time exponential!

#### A Bottom-Up Approach to Fibonacci

```
function Fib2(n)

if n=0 then

return 0

if n=1 then

return 1

Create an array F[0\dots n] F[0]=0, F[1]=1

for i=2\rightarrow n do

F[i]=F[i-1]+F[i-2]

return F[n]
```

#### Recursive Version of Longest Increasing Subsequence

#### WARNING: THIS CODE BLOWS UP!!!!

```
function LIS(a_1, \ldots, a_m)
   if m=1 then
       return 1
   maxSoFar = 1
   for i = 1 \rightarrow m-1 do
       if a_i < a_m then
           solnl = LIS(a_1, \ldots, a_i)
           if solnl + 1 > maxSoFar then
               maxSoFar = solnI + 1
   return maxSoFar
```

#### Longest Increasing Subsequence

```
Let L[j] = the length of the longest increasing subsequence in
a_1, \ldots, a_i which ends at a_i and includes a_i.
  function LIS(A = [a_1, \ldots, a_n])
      for i = 1 \rightarrow n do \triangleright solve each of the n subproblems
          L[i] = 1
         for i = 1 \rightarrow i - 1 do
             if (L[i] + 1 > L[j] \text{ AND } a_i < a_i) then
                 L[i] = L[i] + 1
      Let max = 1
      for k = 1 \rightarrow n do
                                    if (L[k] > L[max]) then
             max = k
      return (L[max])
```

#### Longest Increasing Subsequence: Finding the Sequence

This modified code keeps track of the next to last index i that gives the max in the Prev array. This allows us to find the actual subsequence and not just the length.

```
function LIS(A = [a_1, \ldots, a_n])
   for i = 1 \rightarrow n do \triangleright solve each of the n subproblems
       L[i] = 1, Prev[j] = NULL
       for i = 1 \rightarrow i - 1 do
           if (L[i] + 1 > L[j] \text{ AND } a_i < a_i) then
              L[i] = L[i] + 1
              Prev[i] = i
   Let max = 1
   for k = 1 \rightarrow n do
                                 if (L[k] > L[max]) then
           max = k
```

#### Longest Increasing Subsequence: Finding the Sequence

This code allows us to use the values stored in L and Prev to output the actual longest increasing subsequence.

```
Let max = 1 for k = 1 \rightarrow n do \qquad \qquad \triangleright find the max subproblem sol'n if (L[k] > L[max]) then max = k
Let i = max output(a_i) \qquad \triangleright output the LIS ending at (and including) max while (Prev[i] \neq \text{NULL}) do i = Prev[i] ouptut(a_i)
```

#### Longest Increasing Subsequence: Finding the Subsequence

This is an alternative (more general) approach for recovering the LIS that does not use the *Prev* array.

```
function RecoverLIS(i, L)
   if i = NUII then
       return ()
   output(a_i)

    b find the previous index - Prev[i]

   maxSoFar = 1
   maxSoFarINDEX = NULL
   for i = 1 \rightarrow m - 1 do
       if a_i < a_m then
          solnI = L[i]
          if solnl + 1 > maxSoFar then
              maxSoFar = solnI + 1
              maxSoFarIndex = i
   RecoverLIS(maxSoFarIndex, L)
```

## Longest Increasing Subsequence: Example

Α	100	9	10	200	300	12	13
L	?	?	?	?	?	?	?
Prev	?	?	?	?	?	?	?

## Longest Increasing Subsequence: Example

#### Solution:

Α	100	9	10	200	300	12	13
L	1	1	2	3	4	3	4
Prev	null	null	2	3	4	3	6

#### Longest Common Subsequence (LCS)

```
Let L[i][j] = \text{length of LCS of } x_1 x_2 \dots x_i \text{ with } y_1 y_2 \dots y_j
  function LCS(X = x_1 x_2 \dots x_n, Y = y_1 y_2 \dots y_m)
       for i = 0 \rightarrow n do

    initialize first column 0

           L[i][0] = 0
       for i = 0 \rightarrow m do
                                                         ▷ initialize first row 0
           L[0][i] = 0
       for i = 1 \rightarrow n do
                                                                      ▷ each row
           for i=1 \rightarrow m do

    each column

                if x_i == y_i then
                    L[i][j] = 1 + L[i-1][j-1]
                else
                     L[i][i] = \max\{L[i-1][i], L[i][i-1]\}
       return (L[n]|m|)
```

# Longest Common Subsequence (LCS): Recursive Sol'n (Attempt 1)

```
function \operatorname{LCSR}(X = x_1x_2 \dots x_i, \ Y = y_1y_2 \dots y_j)

if i == 0 OR j == 0 then \qquad \qquad \triangleright base case return 0

if x_i == y_i then \qquad \qquad \triangleright last characters match result = 1 + \operatorname{LCSR}([x_1, \dots, x_{i-1}], [y_1, \dots, y_{j-1}])

else \qquad \qquad \qquad \triangleright last characters are different result = \max\{\operatorname{LCSR}([x_1, \dots, x_{i-1}], \ Y), 

\operatorname{LCSR}(X, [y_1, \dots, y_{j-1}])

return result
```

The running time is exponential!

#### Longest Common Subsequence (LCS): Memoization

```
Initialize a n by m matrix A to unknown
  function LCSM(X = x_1x_2...x_i, Y = y_1y_2...y_i)
      if i == 0 OR j == 0 then
                                                           ▷ base case
          return 0
      if A[i][j] \neq \text{unknown then}
                                           ▷ Check if already solved.
         return A[i][j]
                                              ▷ last characters match
      if x_i == y_i then
          result = 1 + LCSM([x_1, ..., x_{i-1}], [y_1, ..., y_{i-1}])
                                        ▷ last characters are different
      else
          result = \max\{LCSM([x_1, ..., x_{i-1}], Y),
  LCSM(X, [y_1, \ldots, y_{i-1}])
      A[i][i] = result

    Store answer if not already solved.

      return result.
Running Time: O(mn)
```

#### Longest Increasing Subsequence Revisited

#### Finding the actual LIS

```
function RecoverLIS(i, L)
   if i = NULL then
      return ()
   output(a_i)
   maxSoFar = 1.
                          maxSoFarINDEX = NULL
   for i=1 \rightarrow m-1 do
      if a_i < a_m then
         solnI = L[i]
         if solnl + 1 > maxSoFar then
            maxSoFar = solnl + 1
            maxSoFarIndex = i
   RECOVERLIS(maxSoFarIndex, L)
```

#### LCS: Recovering the Subsequence

```
function RecoverLCS(i, j, L, X, Y)
     if i = 0 or j = 0 then
        return
     else if x_i = y_i then
        OUTPUT(x_i)
        RECOVERLCS(i-1, j-1, L, X, Y)
     else if L[i, j-1] > L[i-1, j] then
         RECOVERLCS(i, i-1, L, X, Y)
     else
         RECOVERLCS(i-1, i, L, X, Y)
What is the running time?
```

#### LCS: Recovering the Subsequence

```
function RecoverLCS(i, j, L, X, Y)
     if i = 0 or j = 0 then
         return
     else if x_i = y_i then
         OUTPUT(x_i)
         RECOVERLCS(i-1, j-1, L, X, Y)
     else if L[i, j-1] > L[i-1, j] then
         RECOVERLCS(i, j - 1, L, X, Y)
      else
         RECOVERLCS(i-1, j, L, X, Y)
Running Time: O(m+n)
(m \text{ and } n \text{ are the lengths of } X \text{ and } Y)
```

#### Knapsack: Without Repetition

Running Time: O(nB)

```
Let K[b][j] = \max value achievable using a subset of objects 1, \ldots, j and total capacity b
```

```
function KNAPSACKNOREPEAT(B, w_1, \ldots, w_n, v_1, \ldots, v_n)
                                                  ▷ initialize first row 0
    for j = 0 \rightarrow n do
        K[0][i] = 0
    for b = 0 \rightarrow B do

    initialize first column 0

        K[b][0] = 0
    for i = 1 \rightarrow n do
                                                              ▷ each row
        for b = 1 \rightarrow B do
                                                         ▷ each column
            if w_i > b then
                K[b][i] = K[b][i-1]
            else
                K[b][j] = \max\{v_i + K[b - w_i][j - 1], K[b][j - 1]\}
    return K[B][n]
```

#### Knapsack: Without Repetition Example

```
Object 1 2 3 4
Capacity: 5 weight 1 3 4 4
value 3 4 5 6
```

- 1. Fill in the dynamic programming table K.
- 2. What is the optimal value?
- 3. What subset does it correspond to?
- 4. How would you find the corresponding subset programmatically?

#### Knapsack: Without Repetition Example Soln

Object 1 2 3 4 weight 1 3 4 4 value 3 4 5 6

Columns are objects and rows are capacities.

	0	1	2	3	4
0	0	0	0	0	0
1	0	3	3	3	3
2	0	3	3	3	3
3	0	3	4	4	4
4	0	3	7	7	7
5	0	3	7	8	9

#### Knapsack: Without Repetition Advanced Example

```
Object 1 2 3 4 5
Capacity: 20 weight 2 3 4 5 9
value 3 4 5 8 10
```

- 1. Fill in the dynamic programming table K.
- 2. What is the optimal value?
- 3. What subset does it correspond to?
- 4. How would you find the corresponding subset programmatically?

## Knapsack: Without Repetition Advanced Example

 object
 1
 2
 3
 4
 5

 Capacity: 20
 weight 2
 3
 4
 5
 9

 value
 3
 4
 5
 8
 10

Columns are objects and rows are capacities.

	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	0	0	0	0
2	0	3	3	3	3	3
3	0	3	4	4	4	4
4	0	3	4	5	5	5
5	0	3	7	7	8	8
6	0	3	7	8	8	8
7	0	3	7	9	11	11
8	0	3	7	9	12	12
9	0	3	7	12	13	13
10	0	3	7	12	15	15

	0	1	2	3	4	5
11	0	3	7	12	16	16
12	0	3	7	12	17	17
13	0	3	7	12	17	17
14	0	3	7	12	20	20
15	0	3	7	12	20	20
16	0	3	7	12	20	21
17	0	3	7	12	20	22
18	0	3	7	12	20	23
19	0	3	7	12	20	25
20	0	3	7	12	20	26

#### Knapsack: Without Repetition

#### How to recover the subset of items? Work backwards!

- 1. run KnapsackNoRepeat to populate K
- 2. call the recursive function below with parameters B, n to output the actual subset of items

```
function RecoverItems(b,j)

if b = 0 OR j = 0 then 
ightharpoonup  Base Case return

if K[b][j] = K[b][j-1] then 
ightharpoonup  b We didn't use jth item RecoverItems(b, j - 1)

else

Output Item j 
ightharpoonup  We did use jth item RecoverItems(b - w_j, j - 1)
```

Running Time: O(n)

#### CISC 380: Today's Agenda

- Class Logistics
  - Welcome to Zoom!
  - Changes to Student Notetaker Assignment
  - Assignment 4 posted today (or tomorrow)
- Knapsack WITH repetition
  - Subproblem definition & recurrence
  - Filling the table iteration & memoization
- On your own: Knapsack WITHOUT Repetition Example (2 slides before this)

#### Knapsack: WITH Repetition

Recall the solution to Knapsack WITHOUT repetition:

 $K[b][j] = \max$  value achievable using a subset of objects  $1, \ldots, j$  and total capacity b If  $w_j \leq b$ , then

$$K[b,j] = \max\{v_j + K[b-w_j, j-1], K[b, j-1]\}$$

else

$$K[b,j] = K[b,j-1]$$

Base Cases:

$$K[b,0)] = 0, K[0,j] = 0$$

Now we are allowed to use objects multiple times (Knapsack WITH repetition). Come up with a subproblem & recurrence for this new problem.

#### Knapsack: With Repetition

```
Let K[b] = \max value achievable with capacity b, all objects 1, \ldots, n are allowed.
```

```
function KnapsackWithRepeat(B, w_1, \ldots, w_n, v_1, \ldots, v_n)
K[0] = 0
for b = 0 \rightarrow B do
K[b] = 0
for l = 1 \rightarrow n do \triangleright each possible "last" object
if w_l \leq b AND K[b] < K[b - w_l] + v_l then
K[b] = K[b - w_l] + v_l
return K[B]
```

Running Time: O(nB)

#### Knapsack: WITH Repetition Example

	object	1	2	3	4
Capacity: 10	weight	6	3	4	2
	value	30	14	16	9

- 1. Fill in the dynamic programming table K.
- 2. What is the optimal value?
- 3. What subset does it correspond to?
- 4. How would you find the corresponding subset programmatically?

#### Knapsack: WITH Repetition Example Soln

Object 1 2 3 4
Capacity: 10 weight 6 3 4 2
value 30 14 16 9

	0	1	2	3	4	5	6	7	8	9	10
B:	0	0	9	14	18	23	30	32	39	44	48

## Knapsack: WITH Repetition Example

	object	1	2
Capacity: 300	weight	100	200
	value	50	150

#### Chain Matrix Multiply

```
Subproblem: C[i][j] = \min \text{ cost of computing } A_i \times \ldots \times A_j
Let s = j - i be the "width" of the subproblem C[i][j]
  function CHAINMATRIXMULTIPLY (m_0, m_1, \ldots, m_n)
       for i=1 \rightarrow n do
                                                                ▶ Base Case
           C[i][i] = 0
       for s = 1 \rightarrow n - 1 do
                                                  ▷ each possible "width"
           for i = 1 \rightarrow n - s do
               Let i = i + s
               C[i][i] = \infty
               for l = i \rightarrow i - 1 do
                   if C[i][j] > m_{i-1}m_lm_i + C[i][l] + C[l+1][j] then
                       C[i][j] = m_{i-1}m_im_i + C[i][l] + C[l+1][j]
       return C[1][n]
```

#### Chain Matrix Multiply

```
Subproblem: C[i][j] = \min \text{ cost of computing } A_i \times \ldots \times A_j
Let s = j - i be the "width" of the subproblem C[i][j]
  function CHAINMATRIXMULTIPLY (m_0, m_1, \ldots, m_n)
       for i=1 \rightarrow n do
                                                              ▶ Base Case
           C[i][i] = 0
      for s = 1 \rightarrow n - 1 do
                                                 ▷ each possible "width"
          for i = 1 \rightarrow n - s do
               Let i = i + s
               C[i][j] = m_{i-1}m_im_i + C[i][i] + C[i+1][j]
               for l = i + 1 \to i - 1 do
                  if C[i][j] > m_{i-1}m_lm_i + C[i][l] + C[l+1][j] then
                       C[i][j] = m_{i-1}m_im_i + C[i][l] + C[l+1][i]
       return C[1][n]
```

## Chain Matrix Multiply: Example

$$A\times B\times C\times D$$
 
$$A=5\times 2,\ B=2\times 1,\ C=1\times 1,\ D=1\times 10$$

	1	2	3	4
1	?	?	?	?
2		?	?	?
3			?	?
4				?

#### Chain Matrix Multiply: Example Soln

$$A\times B\times C\times D$$
 
$$A=5\times 2,\ B=2\times 1,\ C=1\times 1,\ D=1\times 10$$

	1	2	3	4
1	0	10	12	62
2		0	2	22
3			0	10
4				0

#### CMM: Recovering the Solution

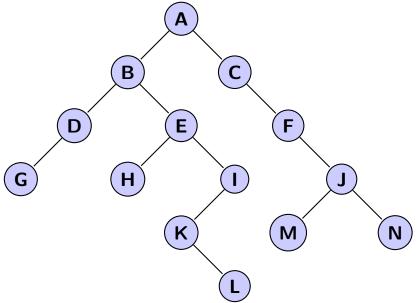
```
Recovery(i,j)
    if (i == j) print "A_" + i
    else
         1 = ''best 1"
         //to get 1 you could:
         //(1)store it in a ''prev" table as you fill C
         //(2) or use the recurrence to recompute it here
         print "("
         Recovery(i,1)
         print ") x ("
         Recovery(1+1, j)
         print ")"
```

What is the running time? Why is this not exponential? Why do we not need to use memoization for this recovery code?

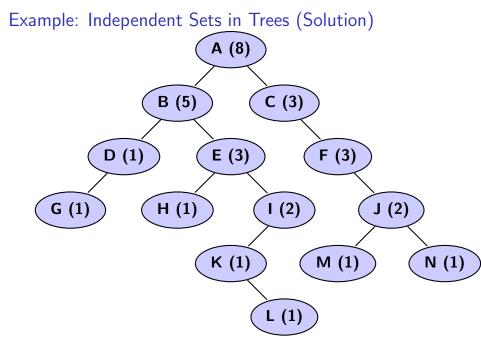
#### Independent Sets in Trees

```
Subproblem: I(j) = \text{size of the maximum IS in the subtree rooted}
 at i.
                    function MaxIS(node n)
                                                   if n is a leaf then
                                                                                                                                                                                                                                                                                                                                                                                                                                                                               ▷ Base Case
                                                                                 return 1
                                                    if n.maxIS \neq NULL then \triangleright Check if it's already solved.
                                                                                 return n maxIS
                                                    else
                                                                                 n.maxIS = \max\{\sum_{k \in n.children} MAXIS(k), 1 + \sum_{k \in n.children} 
                    \sum_{k \in n. \text{children}} \sum_{i \in k. \text{children}} \text{MAXIS}(j)
                                                    return n.maxIS
```

#### Example: Independent Sets in Trees



Find the size of the max IS in the subtree rooted at each node.



Find the size of the max IS in the subtree rooted at each node.

#### Independent Sets in Trees

How to recover the actual independent set.

```
function RECOVERSET(node n)
   if n is a leaf then
                                                     ▷ Base Case
       output n
   else
       if n.maxIS == \sum_{k \in n.children} k.maxIS then
           for all k \in n.children do
              RECOVERSET(k)
       else
                                             ▷ n is in the max IS
           output n
           for all k \in n.children do
              for all j \in k.children do
                  RecoverSet(i)
```