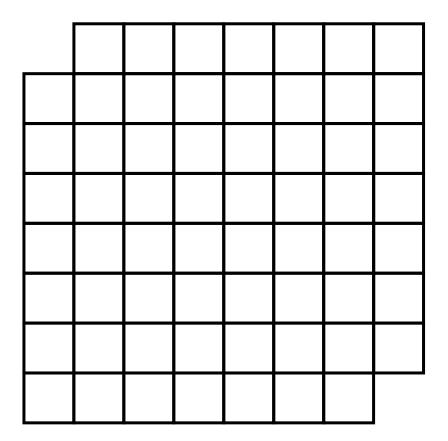
# CS 4102: Algorithms

### Lecture 12: Dynamic Programming

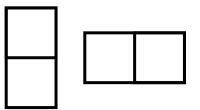
David Wu Fall 2019

### Warm-Up

# **Problem:** Can you fill a $8 \times 8$ board with two corners missing using $2 \times 1$ dominoes?

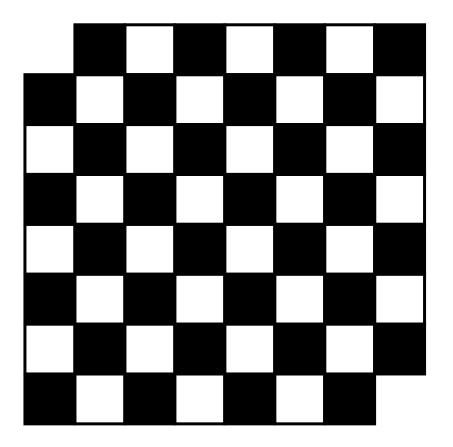


#### **Dominoes:**

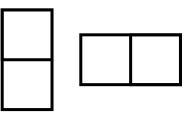


### Warm-Up

# **Problem:** Can you fill a $8 \times 8$ board with two corners missing using $2 \times 1$ dominoes?



#### Dominoes:



32 black squares30 white squares

# Today's Keywords

Dynamic Programming Matrix Chaining (Review) Longest Common Subsequence Seam Carving

**CLRS Readings:** Chapter 14

# Homework

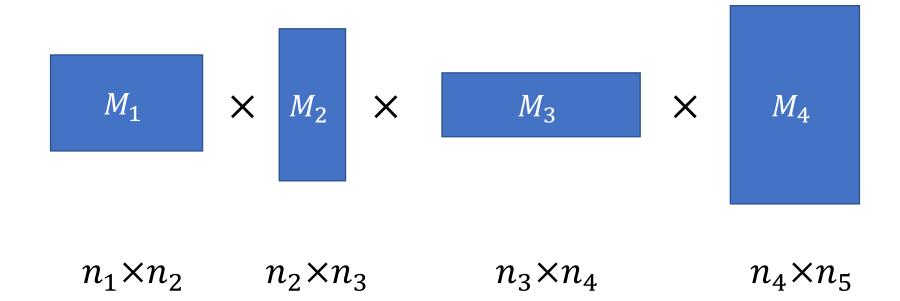
- HW4 due Saturday, October 12, 11pm
  - Divide and conquer, sorting, and dynamic programming
  - Written (use LaTeX!) Submit <u>both</u> **zip** and **pdf** (two <u>separate</u> attachments)!
- HW5 released next week (after exam)
  - Seam Carving
  - Dynamic Programming (implementation)
  - Java or Python

# Midterm

- Tuesday, October 15 (in class)
  - SDAC: Please schedule with SDAC for Tuesday
  - Mostly in-class with a take-home portion
- Practice Midterm (Last Semester's Midterm) available on Collab today
- Optional Review Session: Sunday, October 13 at 3pm, Olsson 120

# **Review: Matrix Chaining**

**Problem:** Given a sequence of matrices  $M_1, \ldots, M_n$ , what is the most efficient way to multiply them?



**Remember:** matrix multiplication is associative

#### More generally:

Best(i, j) = cheapest way to multiply together  $M_i$  through  $M_j$ Possible ways to compute  $M_i \times M_{i+1} \times \cdots \times M_j$ 

 $M_{i} \times M_{i+1,j} = M_{i} \times (M_{i+1} \times \dots \times M_{j})$   $M_{i,i+1} \times M_{i+2,j} = (M_{i} \times M_{i+1}) \times (M_{i+2} \times \dots \times M_{j})$   $M_{i,i+2} \times M_{i+3,j} = (M_{i} \times M_{i+1} \times M_{i+2}) \times (M_{i+3} \times \dots \times M_{j})$   $\vdots \qquad \vdots \qquad \vdots$   $M_{i,j-1} \times M_{j} = (M_{i} \times \dots \times M_{j-1}) \times M_{j}$ 

Position of the "split" changes

More generally:

Best(i, j) = cheapest way to multiply together M<sub>i</sub> through M<sub>j</sub>

Possible ways to compute  $M_i \times M_{i+1} \times \cdots \times M_i$ 

 $\langle M_{i+1} \times \dots \times M_{j}$   $Best(i,i) + Best(i+1,j) + n_{i}n_{i+1}n_{j+1}$   $Best(i,i+1) + Best(i+2,j) + n_{i}n_{i+2}n_{j+1}$   $Best(i,i+2) + Best(i+3,j) + n_{i}n_{i+3}n_{j+1}$  Best(i,i) = 0  $Best(i,j-1) + Best(j,j) + n_{i}n_{j}n_{j+1}$  $Best(i,j) = \min_{k=0,...,j-i-1} Best(i,i+k) + Best(i+k+1,j) + n_i n_{i+k+1} n_{j+1}$ 

Best
$$(i, i) = 0$$
 Best $(i, j) = \min_{k=0,...,j-i-1}$ Best $(i, i+k) + Best(i+k+1, j) + n_i n_{i+k+1} n_{j+1}$   
 $i = 1, j = 4$ 

$$1 \quad 2 \quad 3 \quad 4$$

$$n_1 = 5 \quad n_2 = 10 \quad M_1 \quad \times M_2 \quad \times M_3 \quad \times M_4$$
 $n_3 = 20 \quad n_4 = 8 \quad n_5 = 6$ 

$$M_1 \quad \times M_2 \quad \times M_3 \quad \times M_4$$

Best
$$(i, i) = 0$$
 Best $(i, j) = \min_{k=0,...,j-i-1}$ Best $(i, i+k) + Best(i+k+1, j) + n_i n_{i+k+1} n_{j+1}$   
 $i = 1, j = 4$ 
 $k = 0$ 
 $1$ 
 $2$ 
 $3$ 
 $4$ 
 $n_1 = 5$ 
 $n_2 = 10$ 
 $n_3 = 20$ 
 $n_4 = 8$ 
 $n_5 = 6$ 
 $M_1 \times M_2 \times M_3 \times M_4$ 

Best
$$(i, i) = 0$$
 Best $(i, j) = \min_{k=0,...,j-i-1}$ Best $(i, i+k) + Best(i+k+1, j) + n_i n_{i+k+1} n_{j+1}$   
 $i = 1, j = 4$ 
 $k = 1$ 
 $1$ 
 $2$ 
 $3$ 
 $4$ 
 $n_1 = 5$ 
 $n_2 = 10$ 
 $n_3 = 20$ 
 $n_4 = 8$ 
 $n_5 = 6$ 
 $M_1$ 
 $\times$ 
 $M_2$ 
 $\times$ 
 $M_3$ 
 $\times$ 
 $M_4$ 
 $12$ 

Best
$$(i, i) = 0$$
 Best $(i, j) = \min_{k=0,...,j-i-1}$ Best $(i, i+k) + Best(i+k+1, j) + n_i n_{i+k+1} n_{j+1}$   
 $i = 1, j = 4$ 
 $k = 2$ 
 $1$ 
 $2$ 
 $3$ 
 $4$ 
 $n_1 = 5$ 
 $n_2 = 10$ 
 $n_3 = 20$ 
 $n_4 = 8$ 
 $n_5 = 6$ 
 $M_1 \times M_2 \times M_3 \times M_4$ 
13

Best
$$(i,i) = 0$$
 Best $(i,j) = \min_{k=0,...,j-i-1}$ Best $(i,i+k) + Best(i+k+1,j) + n_i n_{i+k+1} n_{j+1}$   
 $i = 1, j = 4$ 
  
**Observation:** Value  
depends on values to its  
left and below
  
 $n_1 = 5$   $n_2 = 10$   
 $n_3 = 20$   $n_4 = 8$   $n_5 = 6$ 
  
 $M_1 \times M_2 \times M_3 \times M_4$ 
  
 $m_1 = 5$   $m_2 = 10$   
 $M_1 \times M_2 \times M_3 \times M_4$ 
  
 $m_1 = 5$   $m_2 = 10$ 

 $Best(i,i) = 0 \quad Best(i,j) = \min_{k=0,\dots,i-i-1} Best(i,i+k) + Best(i+k+1,j) + n_i n_{i+k+1} n_{j+1}$ *J* 1 2 3 4 1000 1800 2040 1 0 1600 2080 2 0 3 960 0 4 0  $n_1 = 5$   $n_2 = 10$  $n_3 = 20$   $n_4 = 8$   $n_5 = 6$  $M_1$  ×  $M_2$  ×  $M_3$  ×  $M_4$ 16

# **Run Time**

- 1. Initialize Best[i, i] to be all 0s
- 2. Starting at the main diagonal, working to the upper-right, fill in each cell using:  $\Theta(n^2)$  cells in the array
  - Best(i, i) = 0
  - $Best(i,j) = \min_{k=0,\dots,j-i-1} Best(i,i+k) + Best(i+k+1,j) + n_i n_{i+k+1} n_{j+1}$  $\Theta(n)$  options per cell

### $\Theta(n^3)$ overall run time

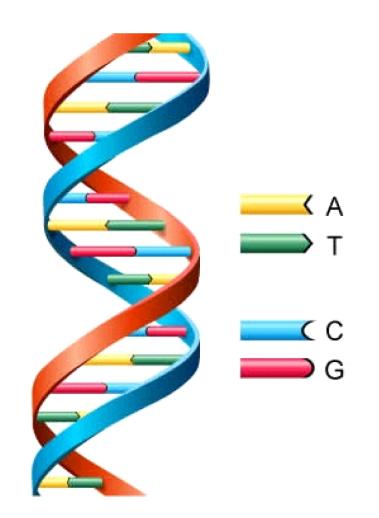
## Backtrack to Find the Best Order

# **Longest Common Subsequence**

Given two sequences X and Y, find the length of their longest common <u>subsequence</u>

#### Example:

- X = ATCTGAT
- Y =TGCATA



# Longest Common Subsequence

Given two sequences X and Y, find the length of their longest common <u>subsequence</u>

#### Example:

X = ATCTGATY = TGCATALCS = TCTA

**Brute force:** Compare every subsequence of *X* with *Y* 



<mark></mark> А Т
G

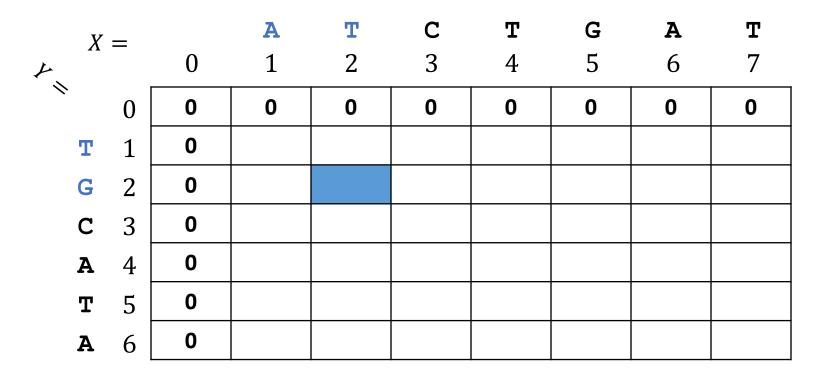
# **Dynamic Programming**

#### Requires optimal substructure

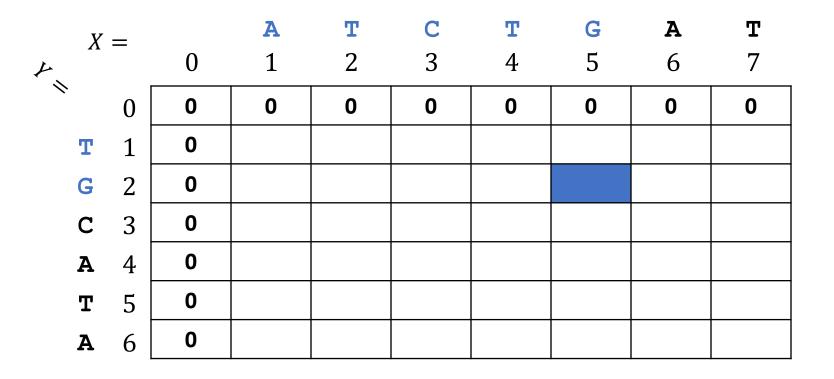
• Solution to larger problem contains the solutions to smaller ones

#### **General Blueprint:**

- 1. Identify recursive structure of the problem
  - What is the "last thing" done?
- 2. Select a good order for solving subproblems
  - "Top Down:" Solve each problem recursively
  - "Bottom Up:" Iteratively solve each problem from smallest to largest
- 3. Save solution to each subproblem in memory



$$i = 2$$
 and  $j = 2$ 



$$i = 2 \text{ and } j = 5$$

Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 

- Suppose X[i] = Y[j] i = 2 and j = 1
  - X = ATCTGAT**Observation:** We can <u>always</u> includeY = TGCATAthe last character (T) in the LCS

Why is this the case? (Argument for optimality)

- If last character in LCS is not  $\mathbb{T}$ , then can extend it to include  $\mathbb{T}$
- If the last character in LCS is  $\mathbb{T}$ , then it does not matter whether we use an earlier  $\mathbb{T}$  or the last  $\mathbb{T}$

Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 

Suppose X[i] = Y[j] i = 2 and j = 1

X = ATCTGAT**Observation:** We can <u>always</u> includeY = TGCATAthe last character (T) in the LCS

**Optimal choice:** always take the last character (add it to the LCS), and recursively solve LCS on remainder

$$LCS(i,j) = LCS(i - 1, j - 1) + 1$$

Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 

- Suppose  $X[i] \neq Y[j]$  i = 2 and j = 2
  - X =ATCTGAT**Observation:** At least one of theY =TGCATACharacters will not be in the LCS

Why is this the case? (Argument for optimality)

 Cannot take both, since otherwise, the last character of the two subsequences are different

Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 

Suppose  $X[i] \neq Y[j]$  i = 2 and j = 2

X =ATCTGAT**Observation:** At least one of theY =TGCATAcharacters will not be in the LCS

$$LCS(i,j) = LCS(i-1,j) \qquad Drop T$$
  
or  
$$LCS(i,j) = LCS(i,j-1) \qquad Drop G$$

Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 

Suppose  $X[i] \neq Y[j]$  i = 2 and j = 2

X =ATCTGAT**Observation:** At least one of theY =TGCATAcharacters will not be in the LCS

$$LCS(i,j) = max(LCS(i-1,j), LCS(i,j-1))$$

Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 

Suppose X[i] = Y[j]LCS(i, j) = LCS(i - 1, j - 1) + 1

Suppose  $X[i] \neq Y[j]$ LCS(i,j) = max(LCS(i-1,j), LCS(i,j-1))

Base case:

$$LCS(i,0) = 0 = LCS(0,j)$$

$$LCS(i,j) = \begin{cases} 0 & i = 0 \text{ or } j = 0 \\ LCS(i-1,j-1) + 1 & X[i] = Y[j] \\ max(LCS(i-1,j), LCS(i,j-1)) & X[i] \neq Y[j] \end{cases}$$

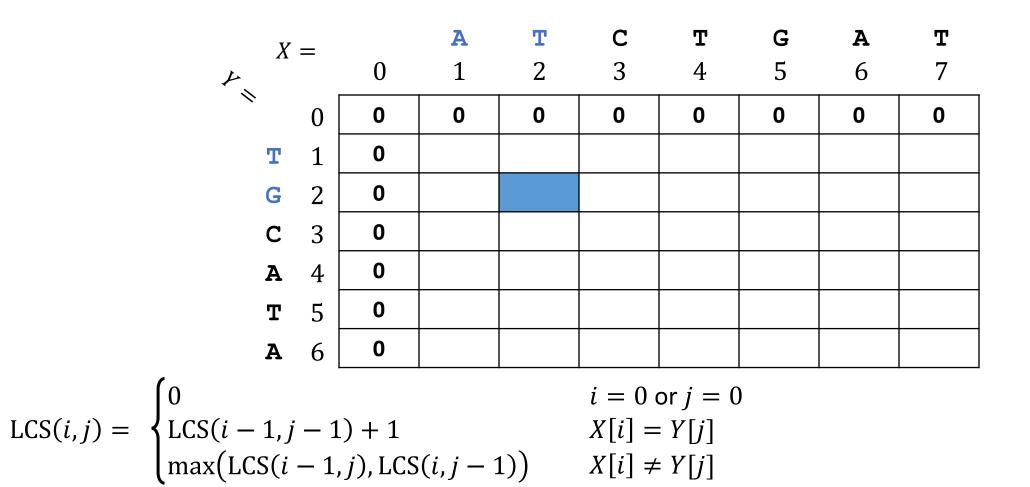
# **Dynamic Programming**

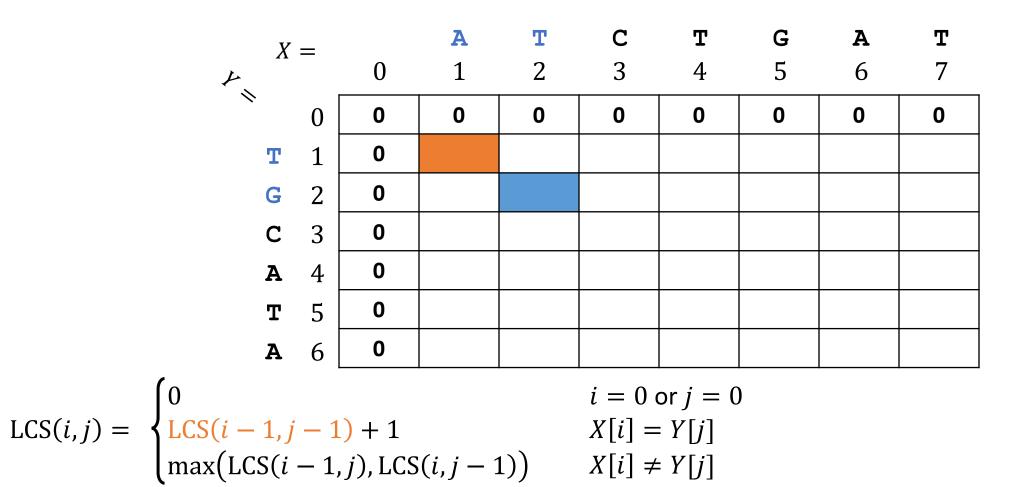
#### Requires optimal substructure

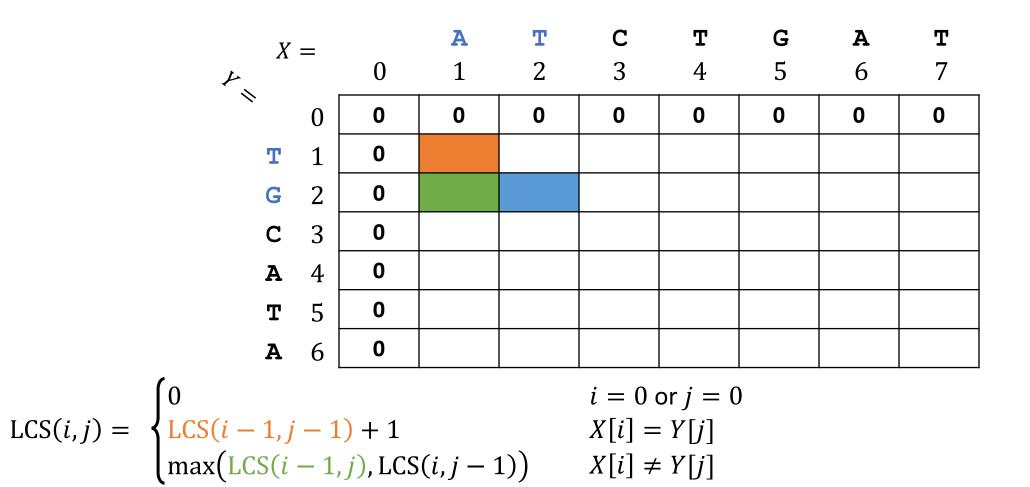
• Solution to larger problem contains the solutions to smaller ones

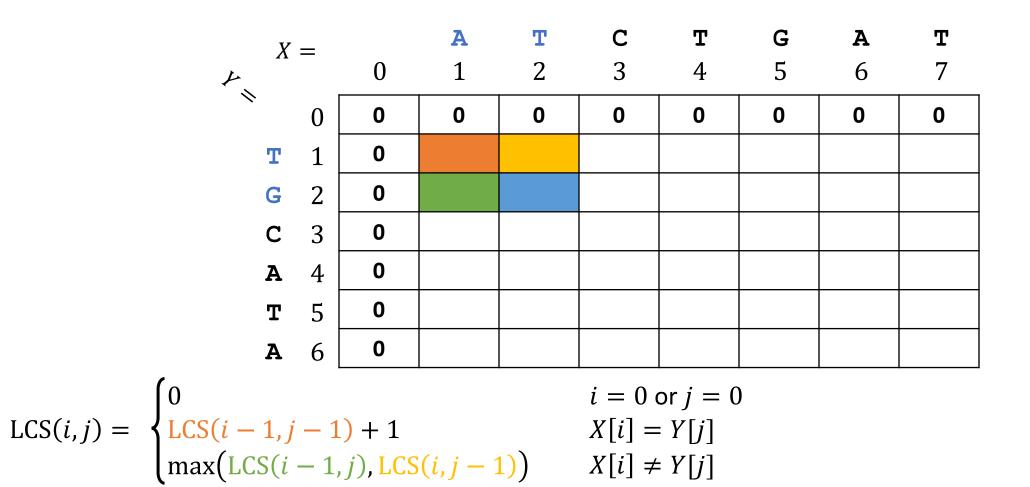
#### **General Blueprint:**

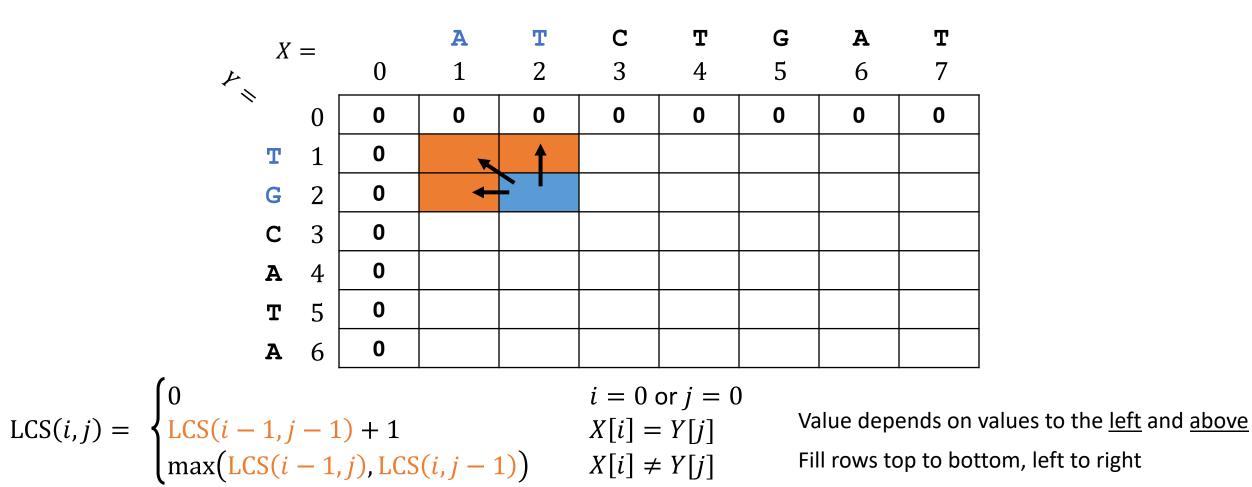
- 1. Identify recursive structure of the problem
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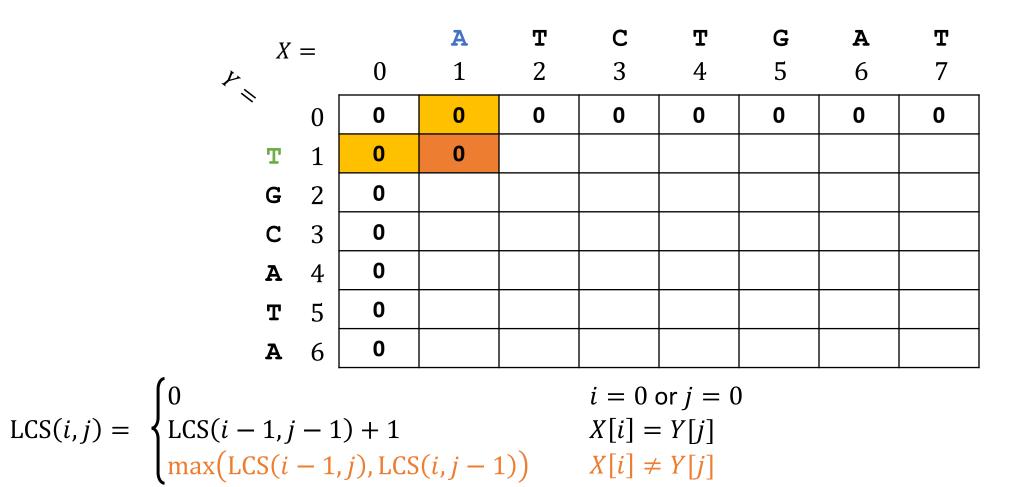


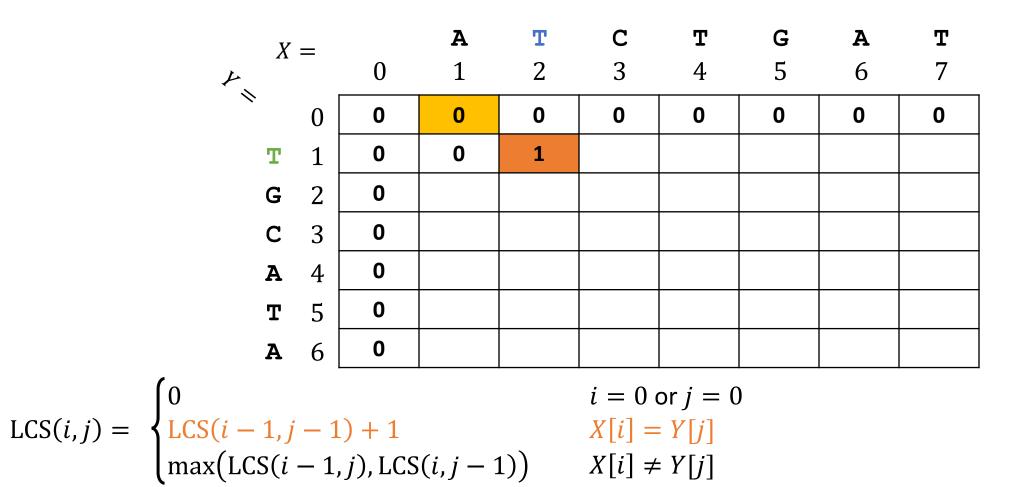


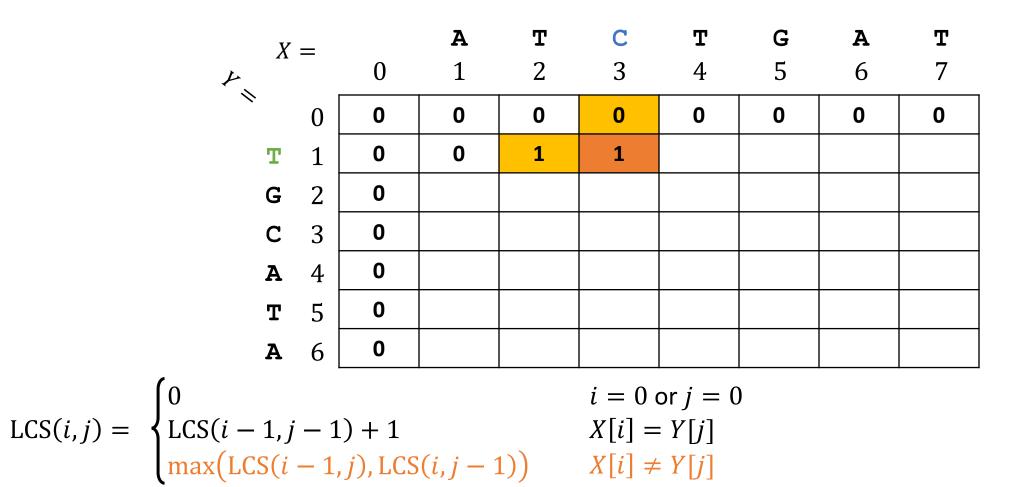


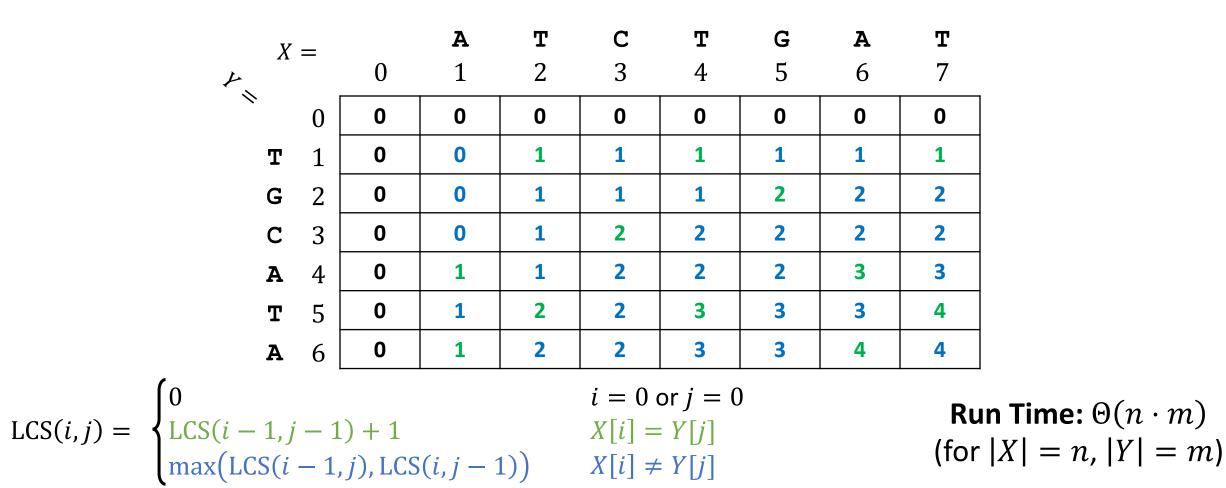


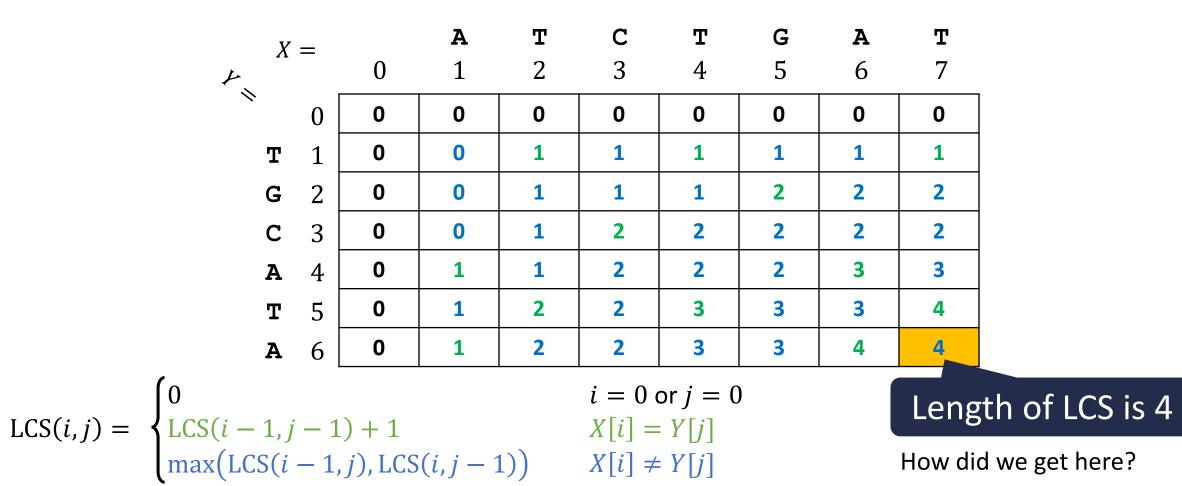




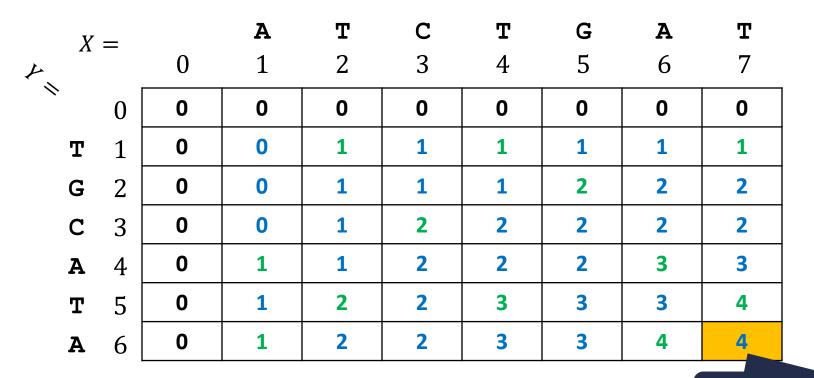








Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 

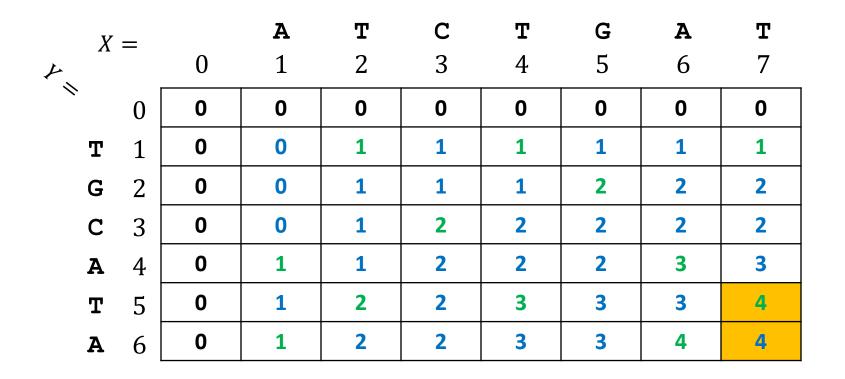


Take the character and move diagonally up if characters match Otherwise, move to the larger of the value above or to the left

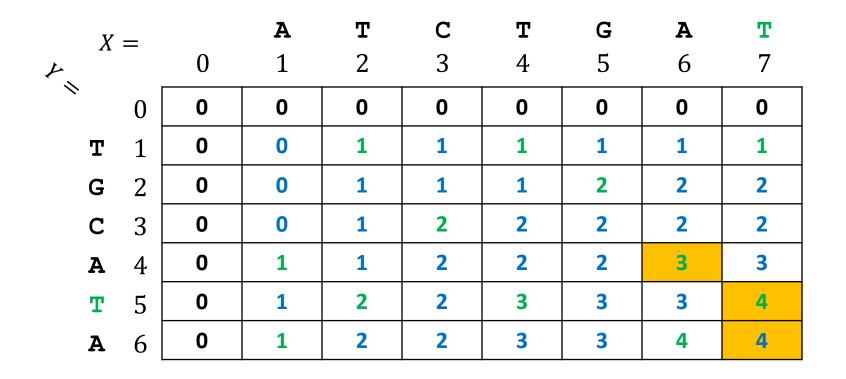
#### Length of LCS is 4

How did we get here?

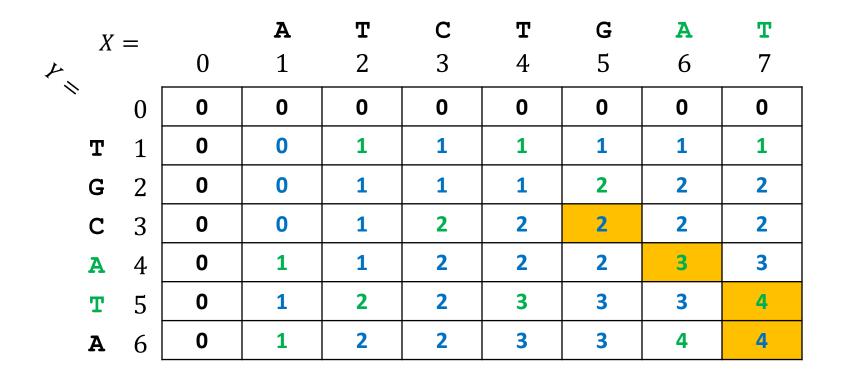
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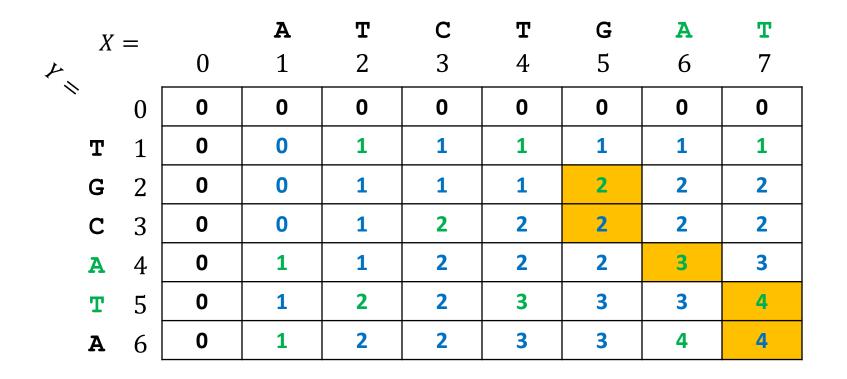
Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 



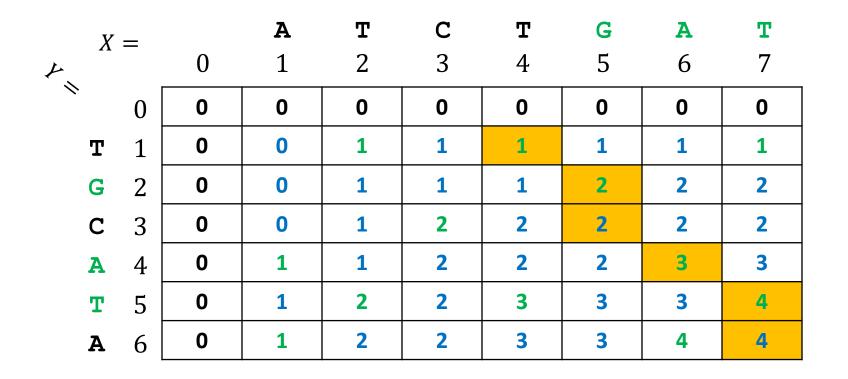
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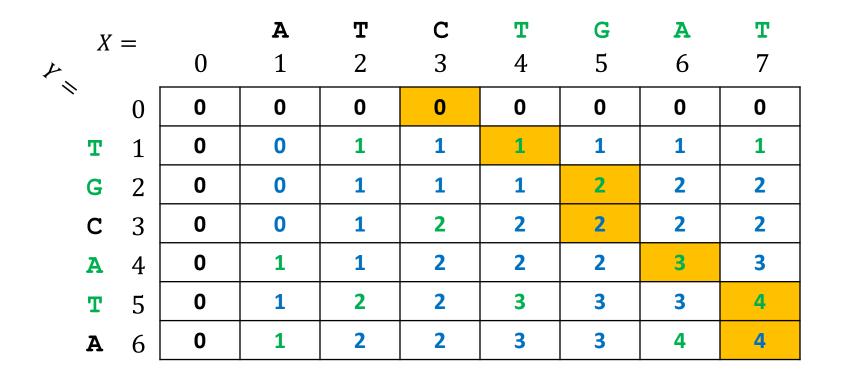
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Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 



Let LCS(i, j) denote the length of the longest common subsequence between the first *i* characters of *X* and first *j* character of *Y* 



Take the character and move diagonally up if characters match Otherwise, move to the larger of the value above or to the left

Not necessarily unique!

Method for image resizing that does not scale/crop the image

#### Method for image resizing that does not scale/crop the image

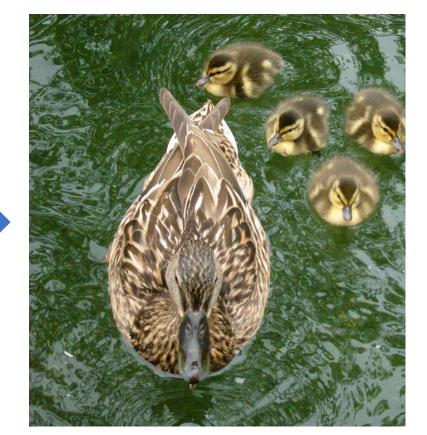


# Cropping

#### Removes a "block" of pixels

Cropped

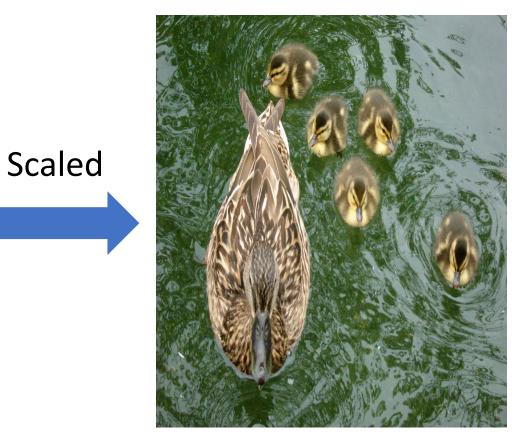




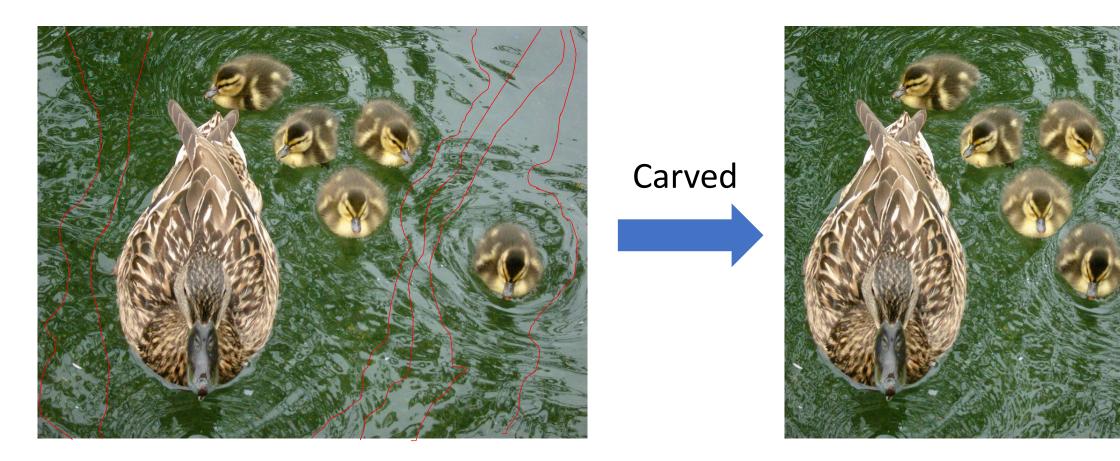
## Scaling

#### Removes "stripes" of pixels





#### Removes "least energy seam" of pixels Demo: http://nparashuram.com/seamcarving/



#### Method for image resizing that does not scale/crop the image

#### Cropped



#### Scaled



Carved



#### Seattle Skyline



Demo: http://nparashuram.com/seamcarving/

# **Energy of a Seam**

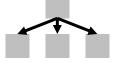
Sum of the energies of each pixel

• e(p) = energy of pixel p

Many choices

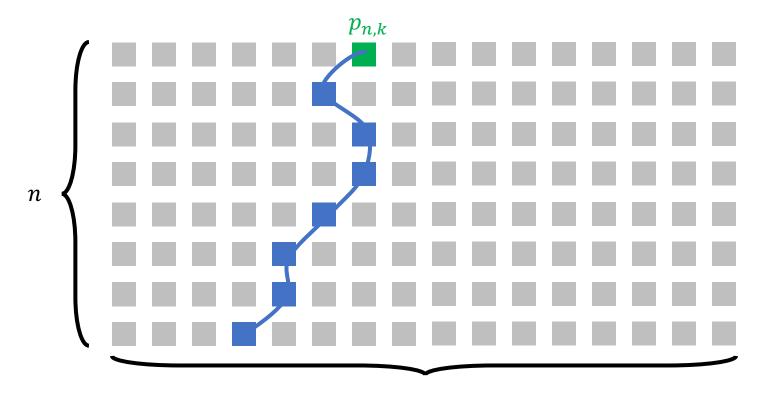
- **Example:** Gradient (how much the color of this pixel differs from its neighbors)
- Particular choice doesn't matter, we use it as a "black box"

S(i, j) = seam with minimal energy from the bottom of the image to pixel  $p_{i,j}$ 



Seam extends from one pixel to (diagonally) adjacent pixel on next row

**Goal:** find the least energy seam going from bottom to top, so delete:  $\min_{k=1,...,m} (S(n,k))$ 



# **Dynamic Programming**

#### Requires optimal substructure

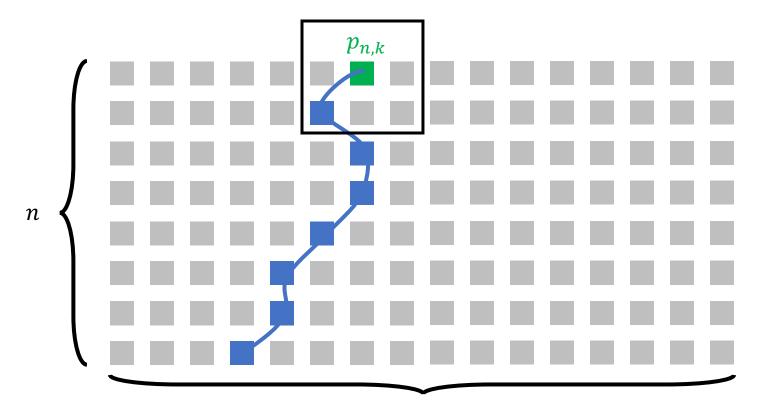
• Solution to larger problem contains the solutions to smaller ones

#### **General Blueprint:**

- 1. Identify recursive structure of the problem
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- 3. Save solution to each subproblem in memory

# **Computing** S(n, k)

Suppose we know the least energy seams for all rows up to n - 1(i.e., we know  $S(n - 1, \ell)$  for all  $\ell$ )



# Computing S(n, k)

Suppose we know the least energy seams for all rows up to n-1

(i.e., we know 
$$S(n - 1, \ell)$$
 for all  $\ell$ )  
 $p_{n,k}$   $S(n,k) = \min \begin{cases} S(n - 1, k - 1) + e(p_{n,k}) \\ S(n - 1, k) + e(p_{n,k}) \\ S(n - 1, k + 1) + e(p_{n,k}) \end{cases}$   
 $S(n - 1, k - 1)$   $S(n - 1, k)$   $S(n - 1, k + 1)$ 

# **Repeated Seam Removal**

Only need to update pixels that depend on the removed seam

At most 2n pixels change

 $\Theta(n)$  time to update pixels

 $\Theta(n+m)$  time to find minimum + backtrack

