

Dynamic Programming

CS16: Introduction to Data Structures & Algorithms

Spring 2020

Outline

- ▶ Dynamic Programming
- ▶ Examples
 - ▶ Fibonacci
 - ▶ Seamcarve



What is Dynamic Programming?

- ▶ Algorithm design paradigm/framework
 - ▶ Design efficient algorithms for optimization problems
- ▶ Optimization problems
 - ▶ “find the *best* solution to problem \mathbf{x} ”
 - ▶ “what is the *shortest* path between \mathbf{u} and \mathbf{v} in \mathbf{G} ”
 - ▶ “what is the *minimum* spanning tree in \mathbf{G} ”
- ▶ Can also be used for non-optimization problems

When is Dynamic Programming Applicable?

- ▶ Condition **#1**: sub-problems
 - ▶ The problem can be solved recursively
 - ▶ Can be solved by solving sub-problems
- ▶ Condition **#2**: *overlapping* sub-problems
 - ▶ Same sub-problems need to be solved many times
- ▶ Core idea
 - ▶ solve each sub-problem once and store the solution
 - ▶ use stored solution when you need to solve sub-problem again

Steps to Solving a Problem w/ DP

- ▶ What are the **sub-problems**?
- ▶ What is the “**magic**” step?
 - ▶ Given solutions to sub-problems...
 - ▶ ...how do I combine them to get solution to the problem?
- ▶ In which **order** should I solve sub-problems?
 - ▶ so that solutions to sub-problems are available when I need them
- ▶ Design iterative **algorithm**
 - ▶ that solves sub-problems in right order and stores their solution

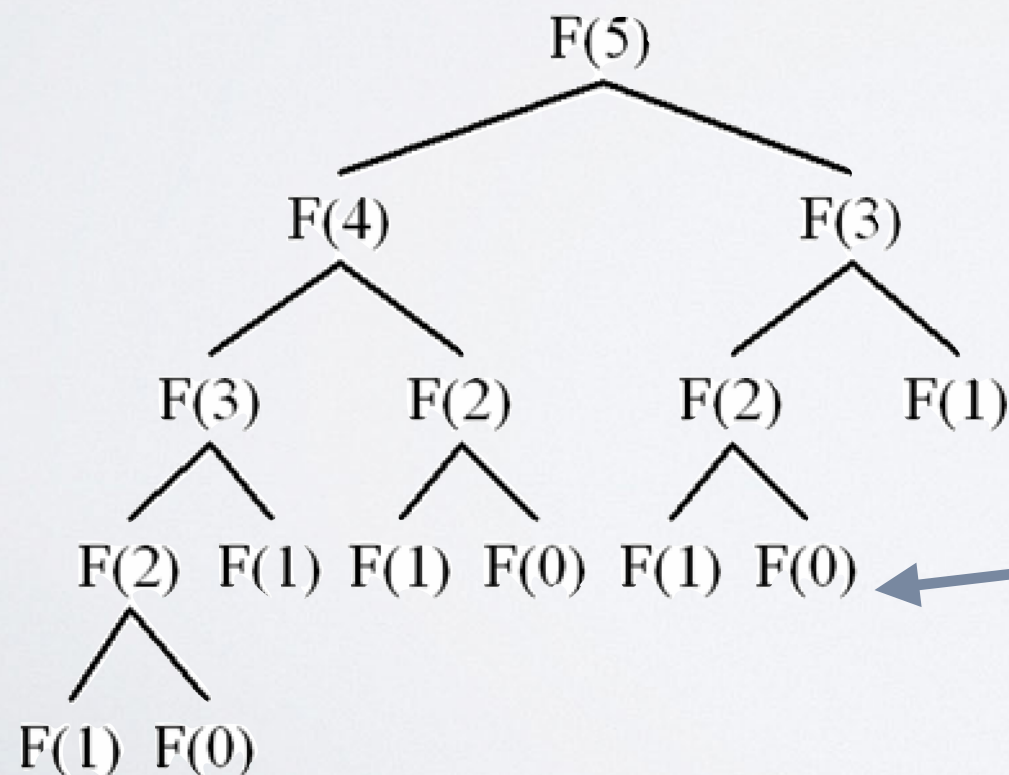
Fibonacci

Fibonacci



0, 1, 1, 2, 3, 5, 8, 13, 21, 34, ...

$$F(n) = F(n - 1) + F(n - 2)$$

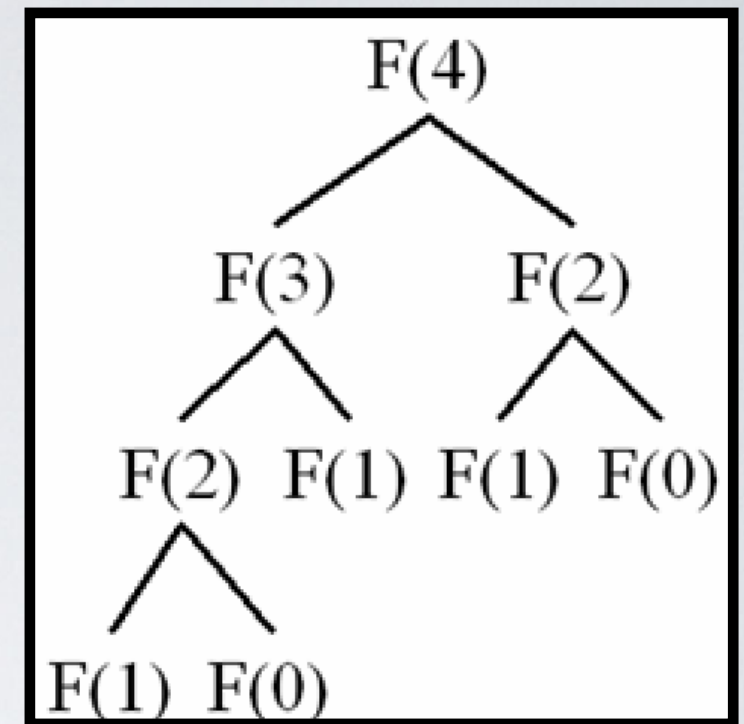


base cases:

$$F(0) = 0 \text{ \& } F(1) = 1$$

Fibonacci (Recursive)

- ▶ Defined by the recursive relation
 - ▶ $F_0 = 0, F_1 = 1$
 - ▶ $F_n = F_{n-1} + F_{n-2}$
- ▶ We can implement this recursively



```
function fib(n):  
    if n == 0:  
        return 0  
    if n == 1:  
        return 1  
    return fib(n-1) + fib(n-2)
```


Fibonacci (Recursive)

Big-O runtime of recursive **fib** function?

Activity #1

1 min

Fibonacci (Recursive)

Big-O runtime of recursive **fib** function?

Activity #1

1 min

Fibonacci (Recursive)

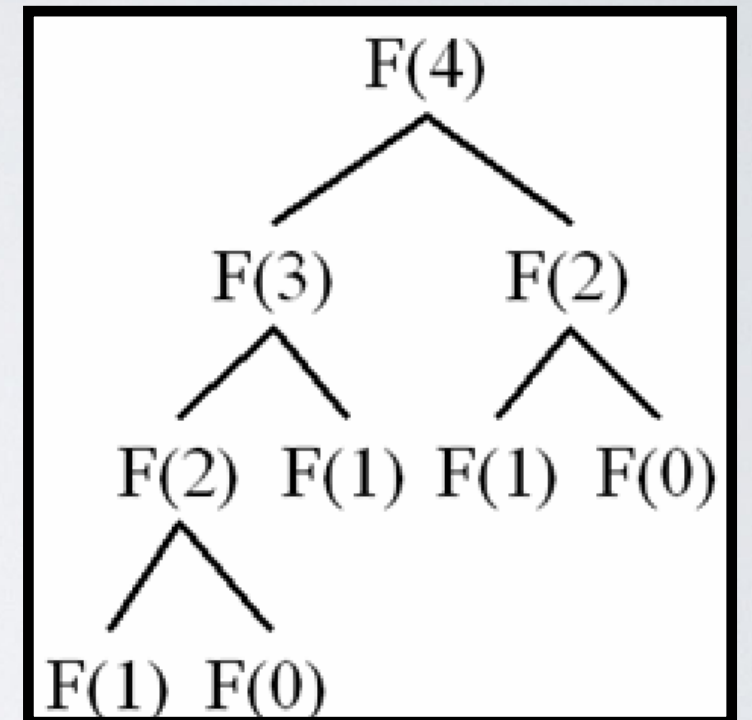
Big-O runtime of recursive **fib** function?

• **Activity #1**

O min

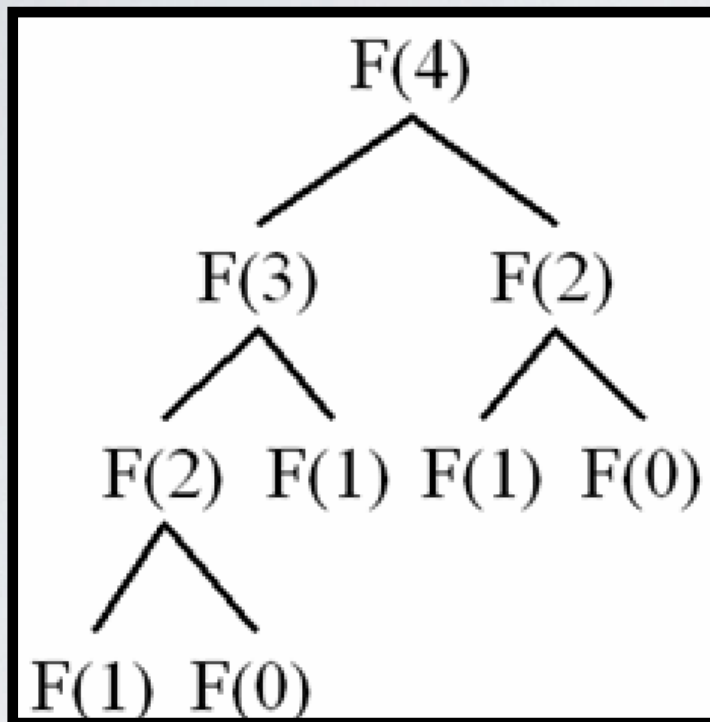
Fibonacci (Recursive)

```
function fib(n):  
    if n = 0:  
        return 0  
    if n = 1:  
        return 1  
    return fib(n-1) + fib(n-2)
```



- ▶ How many times does `fib` get called for `fib(4)`?
 - ▶ 8 times
- ▶ At each level it makes twice as many recursive calls as last
 - ▶ For `fib(n)` it makes approximately 2^n recursive calls
 - ▶ Algorithm is $O(2^n)$

Fibonacci (Recursive)



```
function fib(n):  
    if n == 0:  
        return 0  
    if n == 1:  
        return 1  
    return fib(n-1) + fib(n-2)
```

- ▶ How many times does `fib(1)` get computed?
- ▶ Instead of recomputing Fibonacci numbers over and over again
- ▶ Compute them *once* and store them for later

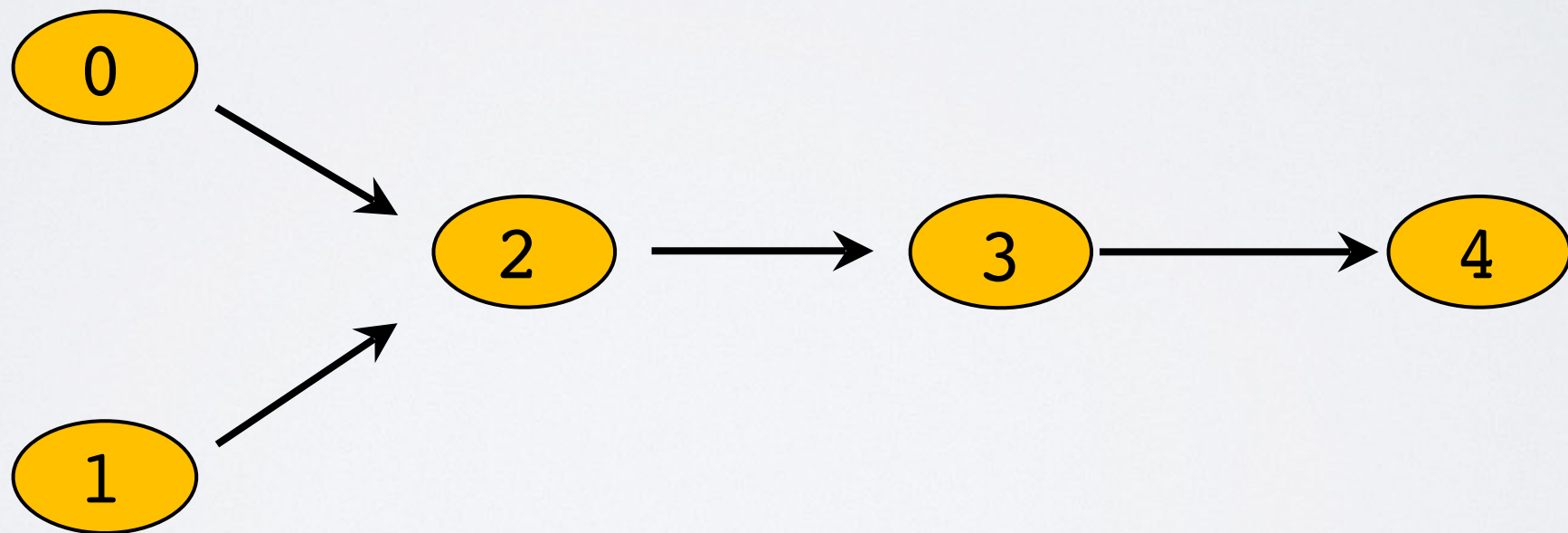
Fibonacci (Dynamic Programming)

- ▶ Given **n** compute
 - ▶ $\text{Fib}(\mathbf{n}) = \text{Fib}(\mathbf{n-1}) + \text{Fib}(\mathbf{n-2})$
 - ▶ with base cases $\text{Fib}(\mathbf{0}) = \mathbf{0}$ and $\text{Fib}(\mathbf{1}) = \mathbf{1}$
- ▶ What are the **sub-problems**?
 - ▶ $\text{Fib}(\mathbf{n-1}), \text{Fib}(\mathbf{n-2}), \dots, \text{Fib}(\mathbf{1}), \text{Fib}(\mathbf{0})$
- ▶ What is the **magic** step?
 - ▶ $\text{Fib}(\mathbf{n}) = \text{Fib}(\mathbf{n-1}) + \text{Fib}(\mathbf{n-2})$

Magic step is
usually not
provided!!

Fibonacci (Dynamic Programming)

- ▶ In which order should I solve sub-problems?
 - ▶ $\text{Fib}(0), \text{Fib}(1), \dots, \text{Fib}(n-1), \text{Fib}(n)$



Fibonacci (Dynamic Programming)

- ▶ Design iterative **algorithm**

```
function Fib(n):  
    fibs = []  
    fibs[0] = 0  
    fibs[1] = 1  
  
    for i from 2 to n:  
        fibs[i] = fibs[i-1] + fibs[i-2]  
  
    return fibs[n]
```

Fibonacci (Dynamic Programming)

- ▶ What's the runtime of **dynamicFib()**?
 - ▶ Calculates Fibonacci numbers from **0** to **n**
 - ▶ Performs **$O(1)$** ops for each one
 - ▶ Runtime is **$O(n)$**
- ▶ We reduced runtime of algorithm
 - ▶ From exponential to linear
 - ▶ with dynamic programming!

Seams

Finding Low Importance Seams



- ▶ **Idea:** remove **seams** not columns
 - ▶ (vertical) seam is a path from top to bottom
 - ▶ that moves left or right by at most one pixel per row

Finding Low Importance Seams

- ▶ How many seams in a $\mathbf{c} \times \mathbf{r}$ image?
 - ▶ At each row the seam can go Left, Right or Down
 - ▶ It chooses **1** out of **3** dirs at all but last row \mathbf{r}
 - ▶ So about $3^{\mathbf{r}-1}$ seams from some starting pixel
 - ▶ There are \mathbf{c} starting pixels so total number of seams is
 - ▶ about $\mathbf{c} \times 3^{\mathbf{r}-1}$
- ▶ For square $\mathbf{n} \times \mathbf{n}$ image
 - ▶ there are about $\mathbf{n} 3^{\mathbf{n}-1}$ possible seams

Finding Low Importance Seams

- ▶ Brute force algorithm
 - ▶ Try every possible seam & find least important one
- ▶ What is running time of brute force algorithm?
 - ▶ If image is **$n \times n$** brute force takes about **$n3^{n-1}$**
 - ▶ So brute force is **$\Omega(2^n)$** (i.e., exponential)

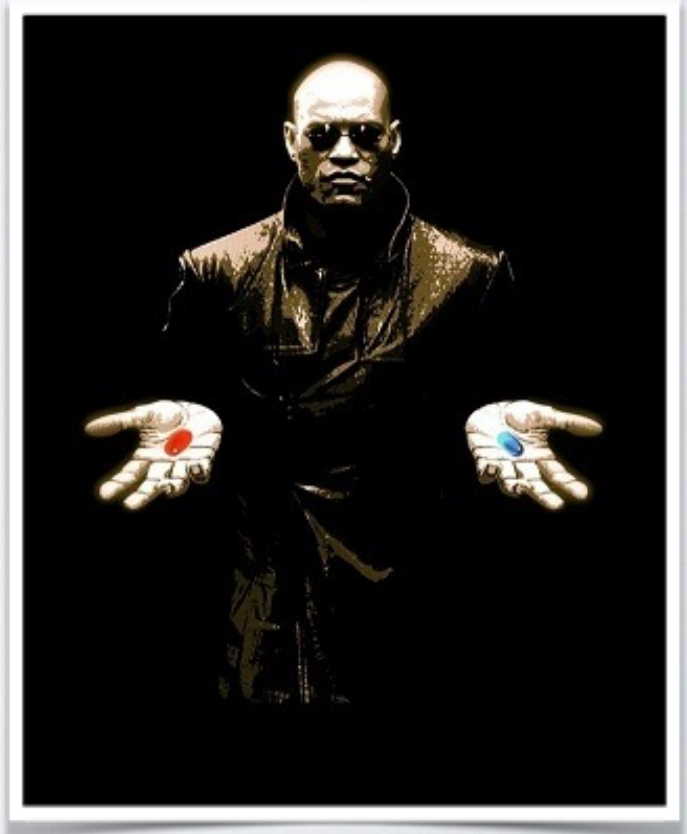
Seamcarve

- ▶ What is the runtime of Seamcarve?
- ▶ The algorithm
 - ▶ Iterate over all pixels from bottom to top
 - ▶ Populate **costs** and **dirs** arrays
 - ▶ Create seam by choosing minimum value in top row and tracing downward
- ▶ How many operations per pixel?
 - ▶ A constant number of operations per pixel (**4**)
- ▶ Constant number of operations per pixel means algorithm is linear
 - ▶ $O(n)$ where **n** is number of pixels
- ▶ Also could have counted # of nested loops in pseudocode...

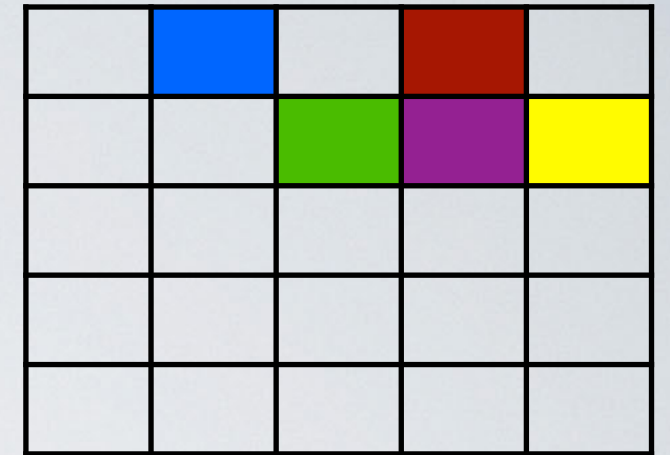
Seamcarve

- ▶ How can we possibly go from
 - ▶ exponential running time with brute force
 - ▶ to linear running time with Seamcarve?
 - ▶ What is the secret to this magic trick?

**Dynamic
Programming!**



Designing Seamcarve



- ▶ What are the subproblems?

- ▶ lowest cost seam (LCS) starting at  is

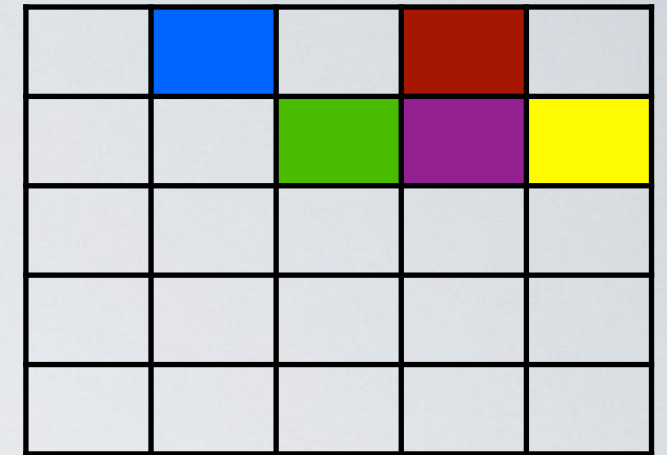
$$\text{red square} \parallel \min(\text{LCS}(\text{green square}), \text{LCS}(\text{purple square}), \text{LCS}(\text{yellow square}))$$

- ▶ Are they overlapping?

- ▶ Yes!

- ▶ ex: $\text{LCS}(\text{green square})$ is subproblem of $\text{LCS}(\text{blue square})$ and $\text{LCS}(\text{red square})$

Designing Seamcarve



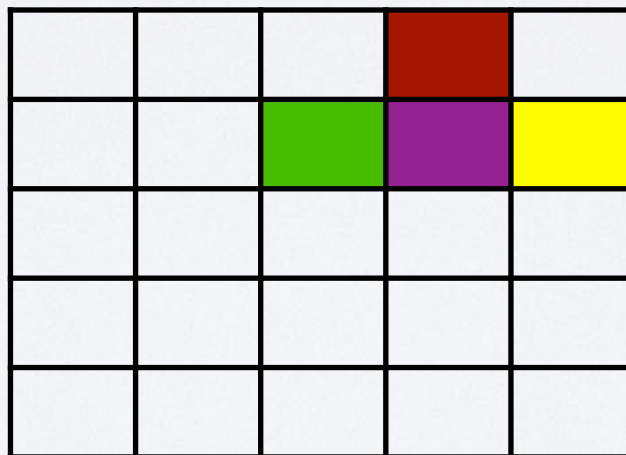
- ▶ What is the magic step?

$$\text{red} \parallel \min(\text{LCS}(\text{green}), \text{LCS}(\text{purple}), \text{LCS}(\text{yellow}))$$

- ▶ Which topological order should I use?
 - ▶ to solve LCS problem at cell (i, j)
 - ▶ we need to have solved problem at cells below

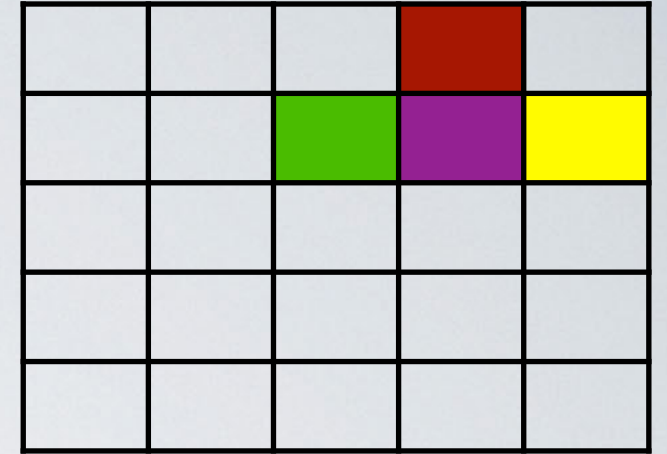
Designing Seamcarve

- ▶ Algorithm
 - ▶ compute cost of LCS for each cell going bottom up
 - ▶ store cost of LCS in an auxiliary 2D array...
 - ▶ ...so we can reuse them



$$\text{Cost}(\text{red}) = \text{Val}(\text{red}) + \min(\text{Cost}(\text{green}), \text{Cost}(\text{purple}), \text{Cost}(\text{yellow}))$$

Designing Seamcarve



- ▶ Problem

- ▶ Costs array only gives us *cost* of LCS at cell
- ▶ We need the seam. What happened?
- ▶ We used

$$\text{Cost}(\text{red}) = \text{Val}(\text{red}) + \min(\text{Cost}(\text{green}), \text{Cost}(\text{purple}), \text{Cost}(\text{yellow}))$$

- ▶ But recall that at “seam level” we had

$$\text{LCS}(\text{red}) = \text{red} \parallel \min(\text{LCS}(\text{green}), \text{LCS}(\text{purple}), \text{LCS}(\text{yellow}))$$

Designing Seamcarve

- ▶ It's OK!
 - ▶ We can keep track of minimum LCS
 - ▶ at each step in auxiliary structure Dirs

