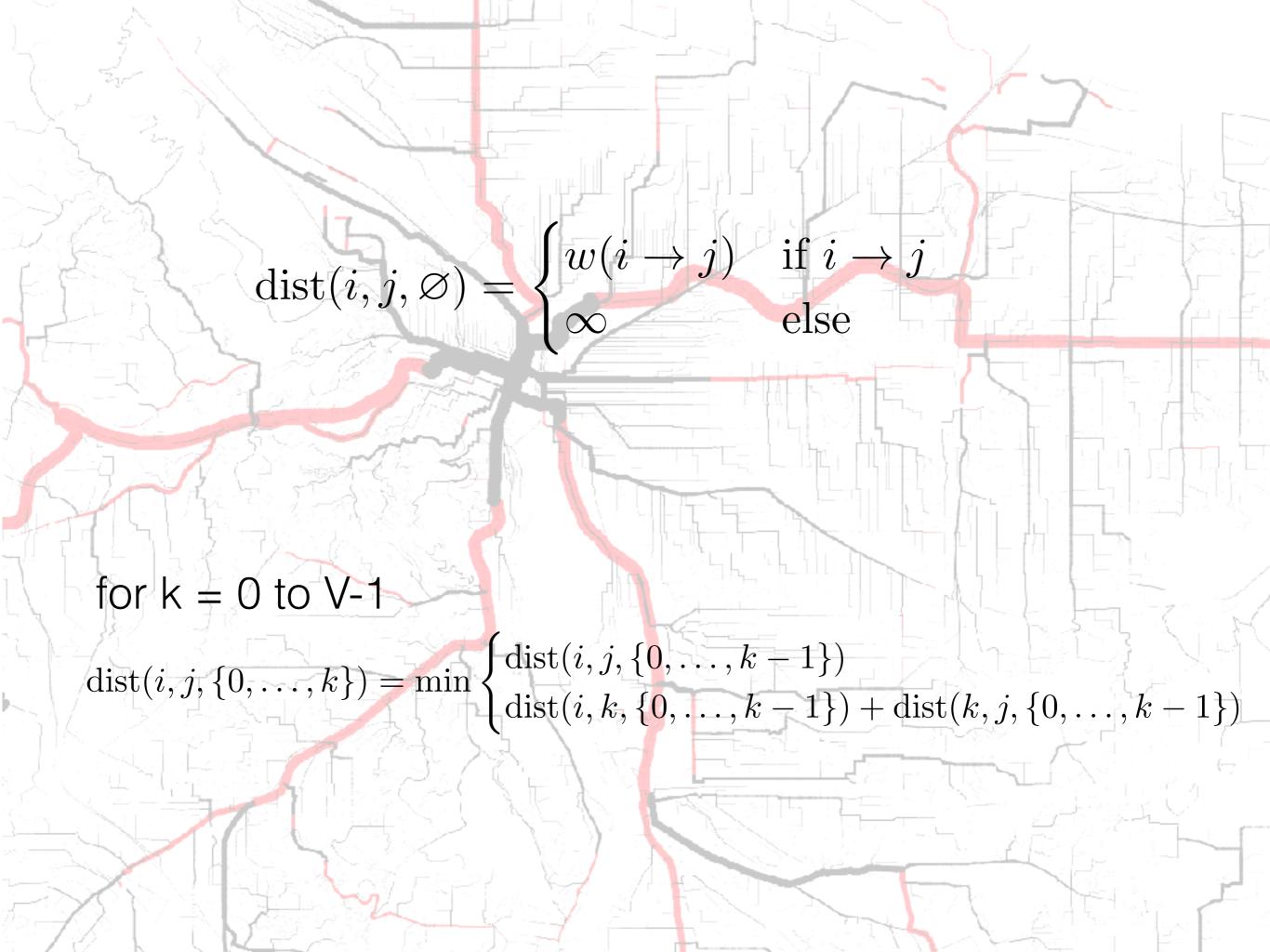
Dynamic programming (part 2)

CS 146 - Spring 2017



$$\begin{aligned} \operatorname{dist}(i,j,\varnothing) &= \begin{cases} w(i\to j) & \text{if } i\to j \\ \infty & \text{else} \end{cases} \\ \operatorname{dist}(i,j,\{0,\ldots,k\}) &= \min \begin{cases} \operatorname{dist}(i,j,\{0,\ldots,k-1\}) \\ \operatorname{dist}(i,k,\{0,\ldots,k-1\}) + \operatorname{dist}(k,j,\{0,\ldots,k-1\}) \end{cases} \end{aligned}$$

represented in 3d array

dist[i][j][k] = distance from i to j
using vertices 0 up to k-1, where k > 0

(shift k by 1 to have room for empty set)

Making recursion more time efficient

(by using more space efficient)

	memoization (top-down)	dynamic programming (bottom up)
store subproblems in	dictionary	array (2D if subpb has 2 params)
subproblem input	dictionary key	array index
subproblem output	dictionary value	entry at given index
code structure	recursive	for-loops

Today

- edit distance problem
- pretty printing

Edit distance problem

• edit: insert, delete or replace a character

ALGORITHM

ALGARITHM replace O with A

_LGARITHM delete A

LOGARITHM insert O

edit(ALGORITHM, LOGARITHM) = 3

Edit distance problem

• edit: insert, delete or replace a character

What is the edit distance between ...

POLYNOMIAL

EXPONENTIAL

Edit distance problem

• edit: insert, delete or replace a character

What is the edit distance between ...

POLYNOMIAL

EXPONENTIAL

efficient algorithm?

Step 1: understand the problem

Work out some examples



Understand the general input and output

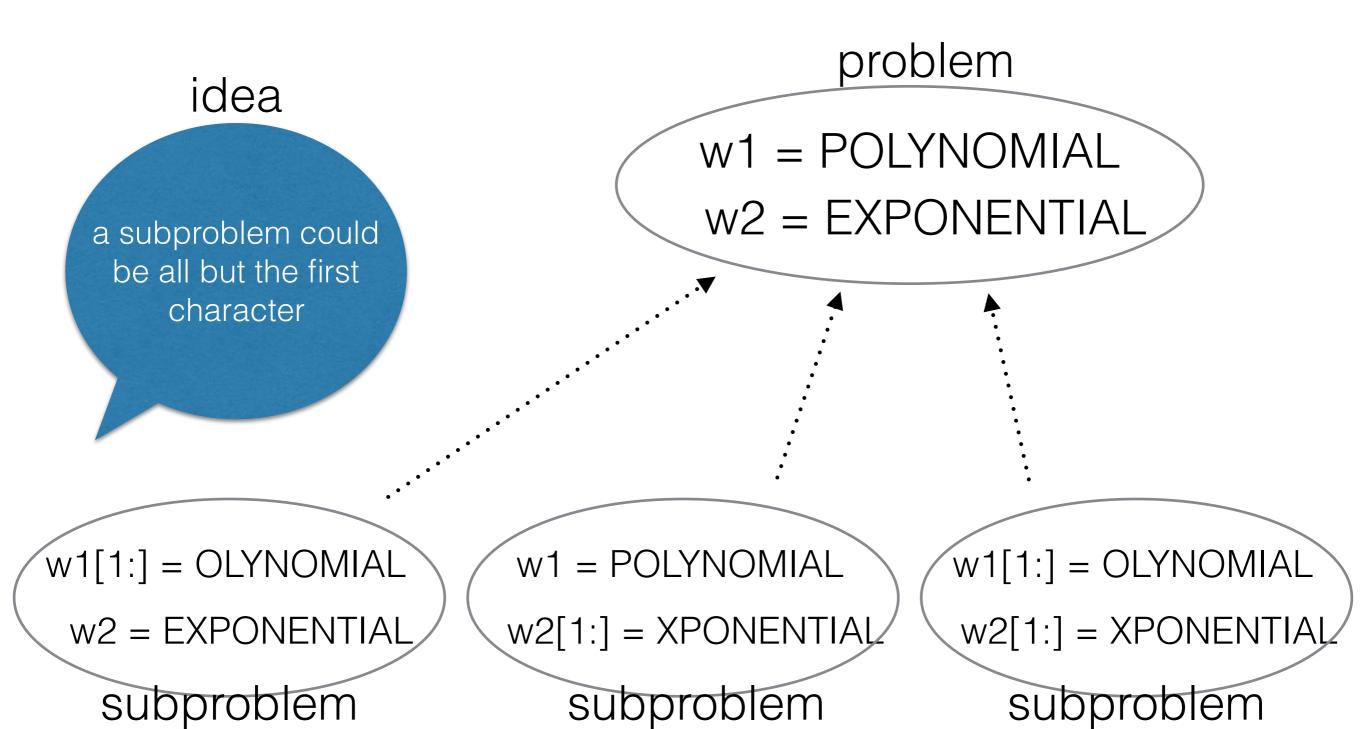
input: 2 strings w1 and w2

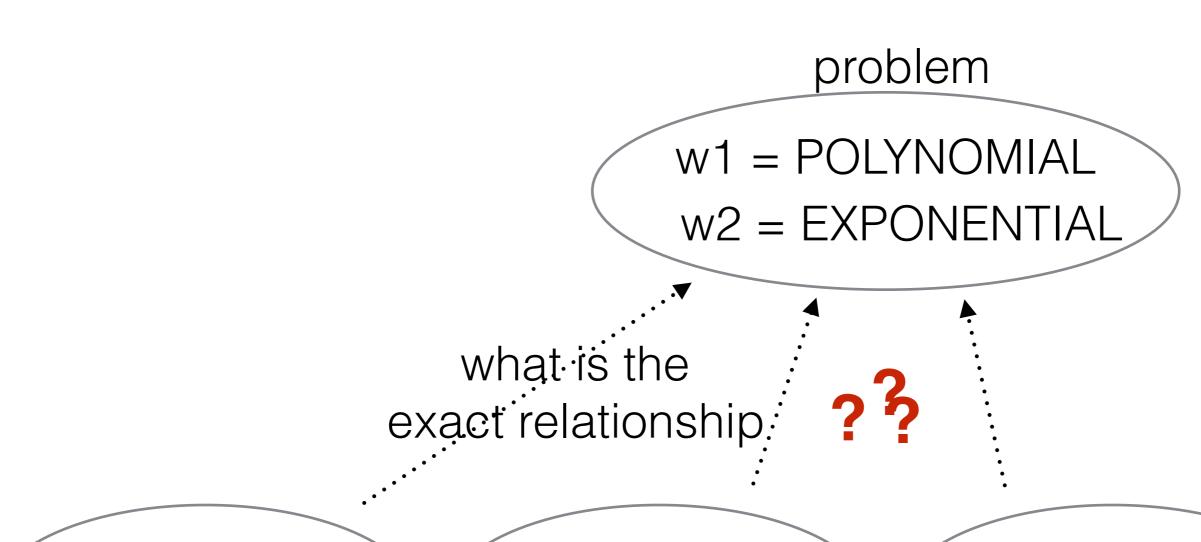
output: edit distance between w1 and w2

Decide on notation

edit(w1, w2) = an integer



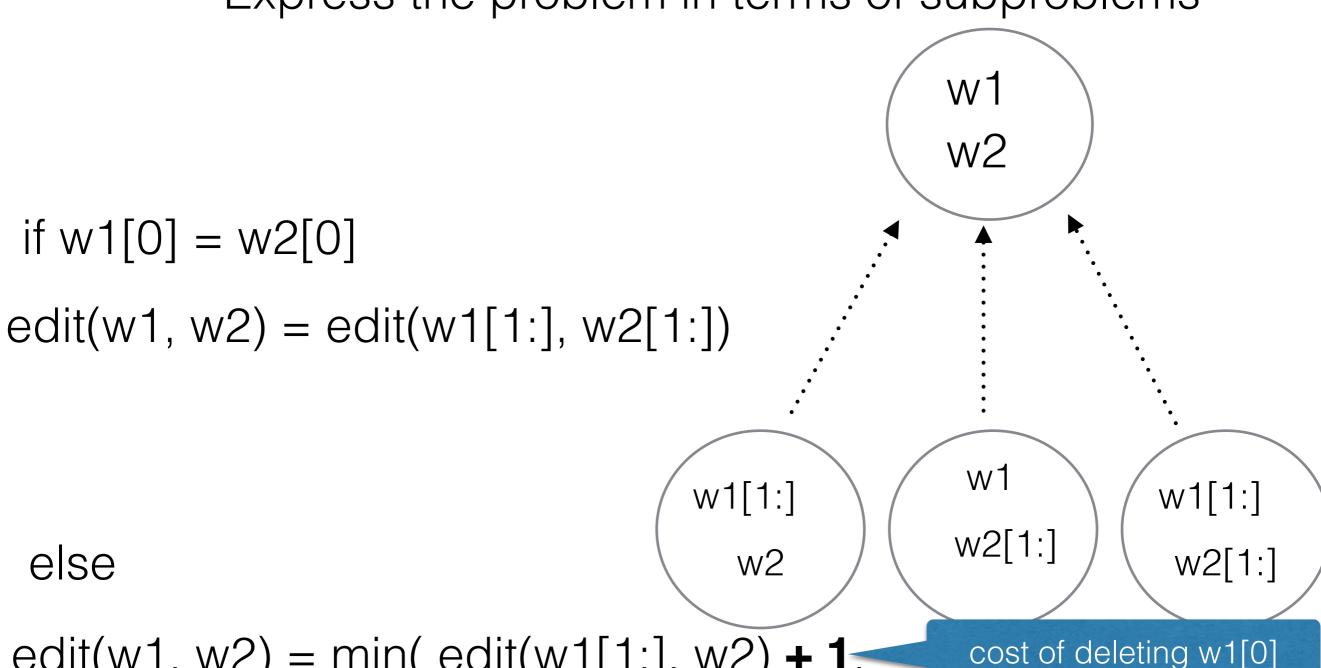




```
w1[1:] = OLYNOMIAL
w2 = EXPONENTIAL
subproblem
```

w1 = POLYNOMIAL w2[1:] = XPONENTIAL subproblem w1[1:] = OLYNOMIAL w2[1:] = XPONENTIAL subproblem

problem if first characters are a match w1 = POLYNOMIALno editing on 1st chars w2 = EXPONENTIAL else there must be an edit to make the 1st characters match w1 = POLYNOMIALw1[1:] = OLYNOMIALw1[1:] = OLYNOMIALw2[1:] = XPONENTIALw2[1:] = XPONENTIALw2 = EXPONENTIAL subproblem subproblem subproblem



edit(w1, w2) = min(edit(w1[1:], w2) + 1, edit(w1, w2[1:]) + 1, edit(w1[1:], w2[1:]) + 1)

cost of deleting w1[0] cost of inserting w1[0] cost of replacing w1[0]

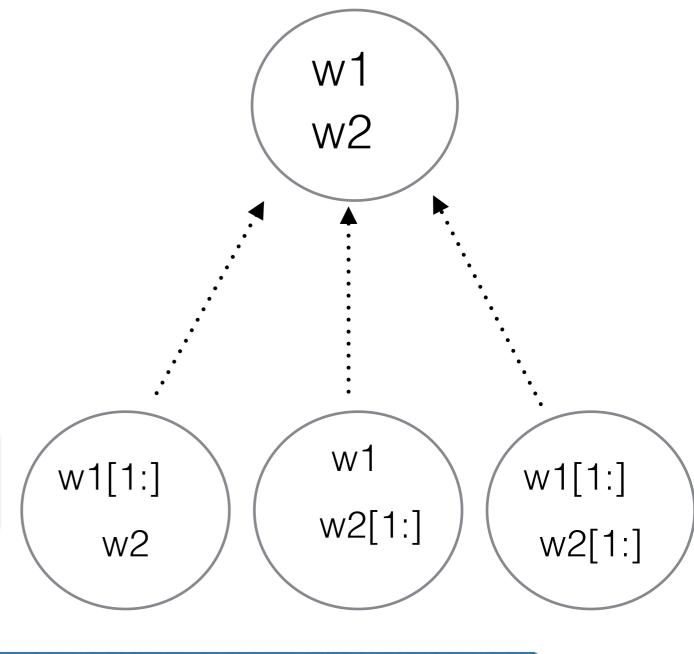
base case: w1 = empty string

edit(w1, w2) = length(w2)

cost of inserting all the letters of w2 into empty string

base case: w2 = empty string

edit(w1, w2) = length(w1)



cost of deleting all the letters of w1 until it becomes an empty string

base cases

if w1[0] = w2[0] edit(w1, w2) = edit(w1[1:], w2[1:])

else

```
edit(w1, w2) = min( edit(w1[1:], w2) + 1,
edit(w1, w2[1:]) + 1,
edit(w1[1:], w2[1:]) + 1)
```

Step 3: optimize!

- option 1: memoization
- option 2: dynamic programming

https://github.com/jnylam/SJSU-cs146-s17

your turn to code!

-> Dynamic programming

-> Edit distance

Time and space complexity

- brute force: time? space?
- dynamic programming (easy to analyze):
 - time: ?
 - space: ?
 - Note: time and and space not necessarily the same
- memoized (messier to analyze directly)

Time and space complexity

- memoized (messier to analyze directly b/c recursive)
 - every call either does actual work or retrieves from cache
 - # calls that do work = # subproblems = size of DP table, each doing (branching factor) work
 - # calls that retrieve from cache <= #subproblems * branching factor, each doing O(1)
 - memoize time = table size * branching factor = DP time
 - memoize space = map size = #subproblems = table
 size = DP space