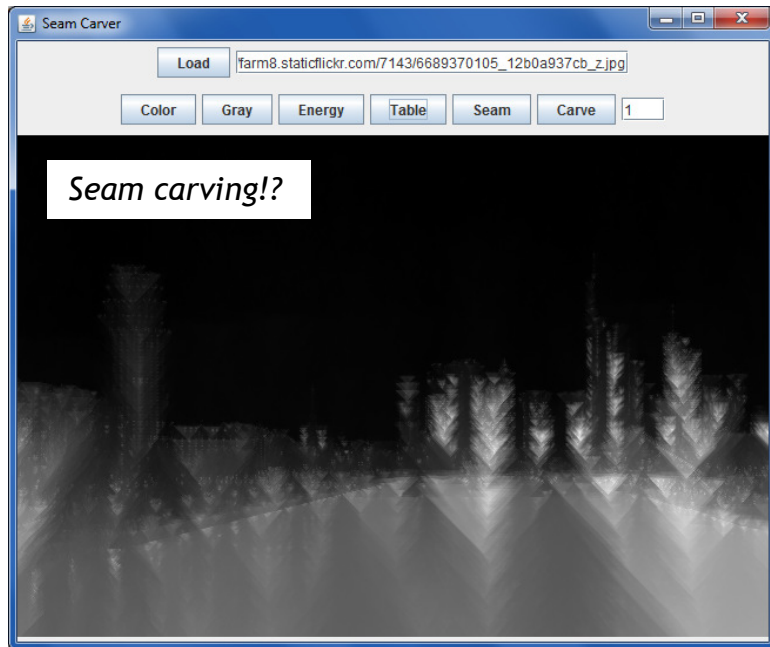


CS 60 today!



hw8 due Tues., 4/8

(1) BSTNode

(2) min-change (dyn prog)

(3) Floyd-Warshall!

hw9 ~ due Mon. 4/14

hw10 ~ due Mon. 4/21

hw11 ~ due Fri. 5/2

Ex. Cr. project...

Final exam ~ due Thurs. 5/15

Schedule...

Thanks, Meghana!

H5CKATHON

where ideas grow

REGISTER AT 5CHACKATHON.COM

april 11 - april 12 // 7pm - 10am
fmi: kmerrill27@gmail.com

Dynamic Programming

(1) Express a problem recursively

(2) Do as little as possible

Who is this talking about?



Dynamic Programming

(1) Express a problem recursively

(2) Do as little as possible

for the computer!

If you can make a small set of recursive calls
***one-time each*, storing the results in a table –**
you can gain huge big-O and real-time speedups!



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Dynamic programming

From Wikipedia, the free encyclopedia

For the programming paradigm, see [Dynamic programming language](#).

In [mathematics](#), [computer science](#), [economics](#), and [bioinformatics](#), **dynamic programming** is a method for solving complex problems by breaking them down into simpler subproblems. It is applicable to problems exhibiting the properties of [overlapping subproblems](#)^[1] and [optimal substructure](#) (described below). When applicable, the method takes far less time than naive methods that don't take advantage of the subproblem overlap (like [depth-first search](#)).

The idea behind dynamic programming is quite simple. In general, to solve a given problem, we need to solve different parts of the problem (subproblems), then combine the solutions of the subproblems to reach an overall solution. Often when using a more naive method, many of the subproblems are generated and solved many times. The dynamic programming approach seeks to solve each subproblem only once, thus reducing the number of computations: once the solution to a given subproblem has been computed, it is stored or "[memo-ized](#)": the next time the same solution is needed, it is simply looked up. This approach is especially useful when the number of repeating subproblems [grows exponentially](#) as a function of the size of the input.



The screenshot shows a web browser window with the URL `en.wikipedia.org/wiki/Dynamic_programming`. The browser's address bar and tabs are visible at the top. The Wikipedia logo, a globe made of puzzle pieces, is on the left. The page title is "Dynamic programming" and the subtitle is "From Wikipedia, the free encyclopedia". Navigation tabs for "Article", "Talk", "Read", "Edit", and "View history" are present. A search bar is on the right. The browser's bookmark bar contains various items like "home", "cs5", "cs60", "ACM", "RGBtoHSV", "RPSLS", "Summer2013", "MyCS", "RecDraw", "ArrowEvader", "PyVis", "DataNitro", and "importIO".

What title, what name, could I choose? In the first place I was interested in planning, in decision making, in thinking. But planning, is not a good word for various reasons. I decided therefore to use the word "programming". I wanted to get across the idea that this was dynamic, this was multistage, this was time-varying I thought, lets kill two birds with one stone. Lets take a word that has an absolutely precise meaning, namely dynamic, in the classical physical sense. It also has a very interesting property as an adjective, and that is its impossible to use the word dynamic in a pejorative sense. Try thinking of some combination that will possibly give it a pejorative meaning. It's impossible. Thus, I thought dynamic programming was a good name. It was something not even a Congressman could object to. So I used it as an umbrella for my activities."

The word *dynamic* was chosen by Bellman to capture the time-varying aspect of the problems, and because it sounded impressive.^[3] The word *programming* referred to the use of the method to find an optimal *program*, in the sense of a military schedule for training or logistics. This usage is the same as that in the phrases *linear programming* and *mathematical programming*, a synonym for *mathematical optimization*.^[4]

DP ~ *Fibonacci* 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

Use-it-or-lose-it is a *conceptual* guide:

`fib(0) = 1`

`fib(1) = 1`

`fib(N) = fib(N-1) + fib(N-2)`

`fib(10)`

DP ~ *Fibonacci* 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

Use-it-or-lose-it is a *conceptual* guide:

$$\text{fib}(0) = 1$$

$$\text{fib}(1) = 1$$

$$\text{fib}(N) = \text{fib}(N-1) + \text{fib}(N-2)$$

`fib(10)`

`fib(9)`

`fib(8)`

DP ~ Fibonacci

1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

Use-it-or-lose-it is a *conceptual* guide:

$$\text{fib}(0) = 1$$

$$\text{fib}(1) = 1$$

$$\text{fib}(N) = \text{fib}(N-1) + \text{fib}(N-2)$$

big-O?

fib(10)

fib(9)

fib(8)

fib(8)

fib(7)

fib(7)

fib(6)

fib(7)

fib(6)

fib(6)

fib(5)

fib(6)

fib(5)

fib(5)

fib(4)

6

5

5

4

5

4

4

3

5

4

4

3

4

3

3

2

but it's way too inefficient for large problems!

DP ~ Fibonacci

1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

Use-it-or-lose-it is a **conceptual** guide:

$$\text{fib}(0) = 1$$

$$\text{fib}(1) = 1$$

$$\text{fib}(N) = \text{fib}(N-1) + \text{fib}(N-2)$$

$O(2^N)$

fib(10)

fib(9)

fib(8)

fib(8)

fib(7)

fib(7)

fib(6)

fib(7)

fib(6)

fib(6)

fib(5)

fib(6)

fib(5)

fib(5)

fib(4)

6

5

5

4

5

4

4

3

5

4

4

3

4

3

3

2

Is there an identifiable source of the **inefficiency** here?

DP ~ *Fibonacci* 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

Use-it-or-lose-it is a **conceptual** guide:

What's the big-O if we make each call to fib **only once!**?

$$\text{fib}(0) = 1$$

$$\text{fib}(1) = 1$$

$$\text{fib}(N) = \text{fib}(N-1) + \text{fib}(N-2)$$

fib(10)

fib(9)

fib(8)

fib(8)

fib(7)

fib(7)

fib(6)

fib(7)

fib(6)

fib(6)

fib(5)

fib(6)

fib(5)

fib(5)

fib(4)

6

5

5

4

5

4

4

3

5

4

4

3

4

3

3

2

inefficiency here == **too many repeated calls!**

DP ~ *Fibonacci* 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

Use-it-or-lose-it is a **conceptual** guide:

What's the big-O if we make each call to fib **only once!**?

$$\text{fib}(0) = 1$$

$$\text{fib}(1) = 1$$

$$\text{fib}(N) = \text{fib}(N-1) + \text{fib}(N-2)$$

fib(10)

fib(9)

fib(8)

fib(8)

fib(7)

fib(7)

fib(6)

fib(7)

fib(6)

fib(6)

fib(5)

fib(6)

fib(5)

fib(5)

fib(4)

6

5

5

4

5

4

4

3

5

4

4

3

4

3

3

2

inefficiency here == **too many repeated calls!**

DP ~ *Fibonacci* 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

Use-it-or-lose-it is a *conceptual* guide:

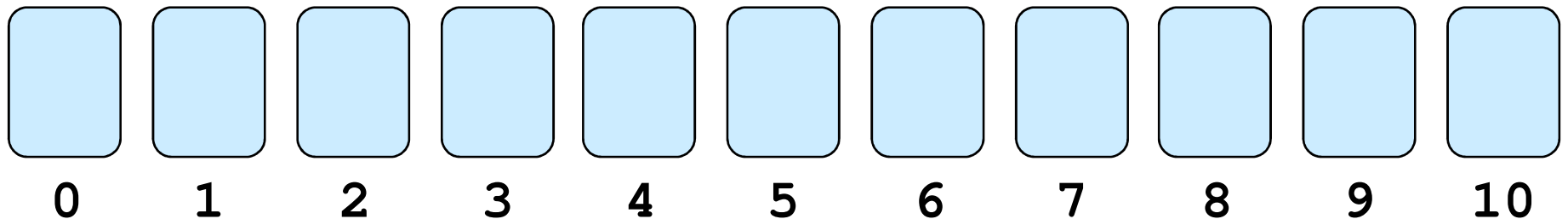
$$\text{fib}(0) = 1$$

$$\text{fib}(1) = 1$$

$$\text{fib}(N) = \text{fib}(N-1) + \text{fib}(N-2)$$

`fib(10)`

and *dynamic programming* makes a table of all possible calls:



...can make *some* exponential run-times into polynomial ones

Knapsack...

MY HOBBY:
EMBEDDING NP-COMPLETE PROBLEMS IN RESTAURANT ORDERS

CHOTCHKIES RESTAURANT	
~ APPETIZERS ~	
MIXED FRUIT	2.15
FRENCH FRIES	2.75
SIDE SALAD	3.35
HOT WINGS	3.55
MOZZARELLA STICKS	4.20
SAMPLER PLATE	5.80
~ SANDWICHES ~	
BARBECUE	6.55



Knapsack contents?

Knapsack...

Suppose *a friend* can consume a candy weight/cost of **13**.

Each candy has different "values"

How can you maximize their *candy-value* experience?

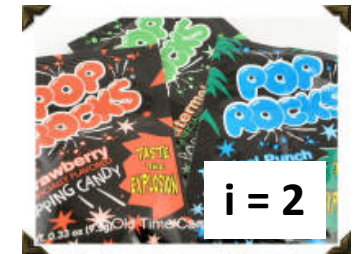
$$v_0 = 100$$
$$w_0 = 2$$



$$v_1 = 120$$
$$w_1 = 3$$



$$v_2 = 230$$
$$w_2 = 5$$



$$v_3 = 560$$
$$w_3 = 7$$



$$v_4 = 675$$
$$w_4 = 9$$



Knapsack...

First, think recursively...

`max_val (13) =`

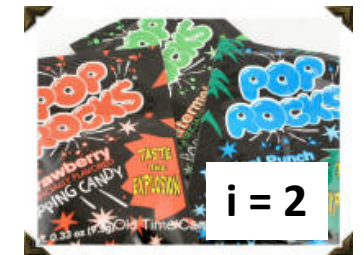
$$v_0 = 100$$
$$w_0 = 2$$



$$v_1 = 120$$
$$w_1 = 3$$



$$v_2 = 230$$
$$w_2 = 5$$



$$v_3 = 560$$
$$w_3 = 7$$



$$v_4 = 675$$
$$w_4 = 9$$



Try this: use it or use it or use it or use it or use it or lose it

Let's see it in action, too...

Knapsack...

First, think recursively...

`max_val(13) =`

$$\max \begin{cases} \max_val(\text{capacity}-1) \\ \max_val(\text{capacity}-w_i) + v_i \end{cases} \text{ over all possible } i$$

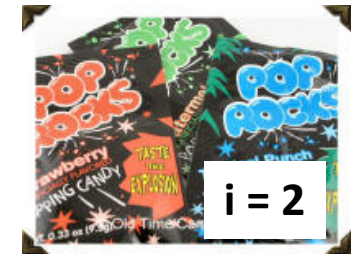
$$v_0 = 100 \\ w_0 = 2$$



$$v_1 = 120 \\ w_1 = 3$$



$$v_2 = 230 \\ w_2 = 5$$



$$v_3 = 560 \\ w_3 = 7$$



$$v_4 = 675 \\ w_4 = 9$$



What's the big-O runtime of this algorithm, as stated?

Knapsack with DP

Minimize work!

Create a table of possible recursive calls.

Compute the results of each one *bottom-up*.

No need for the overhead of the function calls!

$v_0 = 100$
 $w_0 = 2$



$v_1 = 120$
 $w_1 = 3$



$v_2 = 230$
 $w_2 = 5$



$v_3 = 560$
 $w_3 = 7$



$v_4 = 675$
 $w_4 = 9$



CAPACITY

Fill in this table from 0 up to the actual capacity of interest...

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

Is this table indexed by value or weight?

Does this table hold values or weights?

Which values are "easy" to fill in?

How do we fill in each cell?

$T[\text{cap}]$ = the maximum value obtainable for the capacity cap

(1) Fill in this table of best-results for the knapsack problem:

Name(s) _____

		CAPACITY													
		0	1	2	3	4	5	6	7	8	9	10	11	12	13
VALUE	0	0	0	100											

Fill in this table from 0 up to the actual capacity of interest...
 Each cell has $T[\text{cap}]$, the maximum value obtainable for the capacity **cap**

$v_0 = 100$
 $w_0 = 2$
 $i = 0$

$v_1 = 120$
 $w_1 = 3$
 $i = 1$

$v_2 = 230$
 $w_2 = 5$
 $i = 2$

$v_3 = 560$
 $w_3 = 7$
 $i = 3$

$v_4 = 675$
 $w_4 = 9$
 $i = 4$

(2) Estimate the big-O runtime of this table-filling algorithm, in terms of N, the number of items available.

What else does the big-O runtime depend on?

Quiz

(3) How could you *also* keep track of the items to take using a 2nd table with *one item per cell*?

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

One Item

Knapsack!



Why are these important for *knapsacking* at HMC ?

hw8pr2: Change

we want the *fewest* coins that will produce this **total**

1c is always available

these are the denominations (can use each as many times as we want)

Minimizing change...

fewest coins for a desired total

change(**27**, [1,5,10,25])

3

(1, 1, 25)

You'll write this two times...

recursively: in Racket
dynamic programming: in Java

we want the *fewest* coins that will produce this *total*

1c is always available

these are the denominations (can use each as many times as we want)

change(27, [1,5,10,25])

3

(1, 1, 25)

Minimizing change...

fewest coins for a desired total

change(27, [1,5,10])



change(18, [1,5,10,12])



change(18, [1,5,9,10,12])



we want the **fewest** coins that will produce this **total**

out of these denominations (we can use as many times as we want)

change(**18**, [1,5,9,10,12])

in Racket: *recursion use-it-or...-or-lose-it*

in Java: *dynamic programming!*

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

Fill in this table from 0 up to the actual amount of interest...

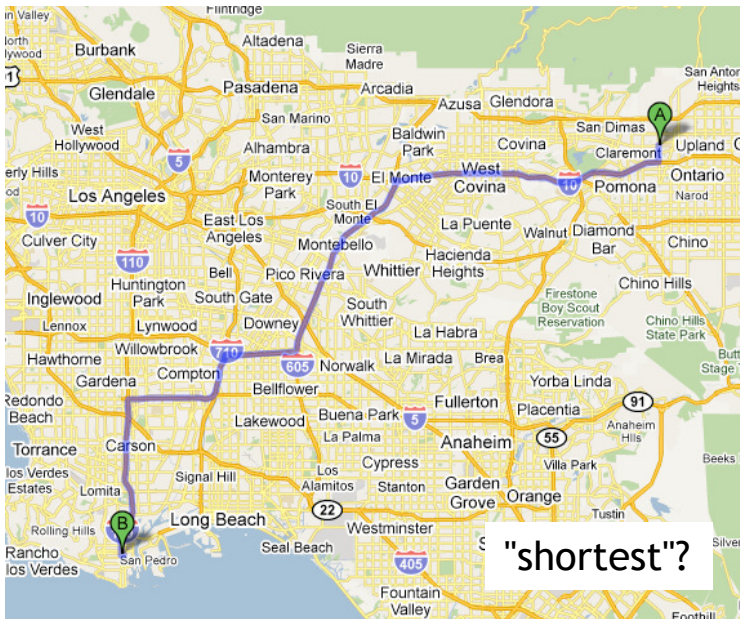
Each top cell has **Min[amt]**, the minimum # of coins that can create **amt**

Try filling out this table by hand – *this is the DP algorithm!*

16	17	18

Shortest paths!

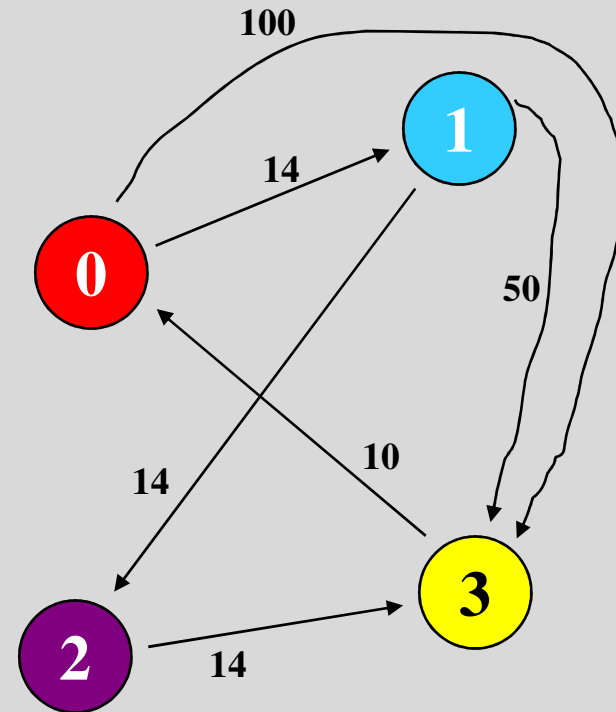
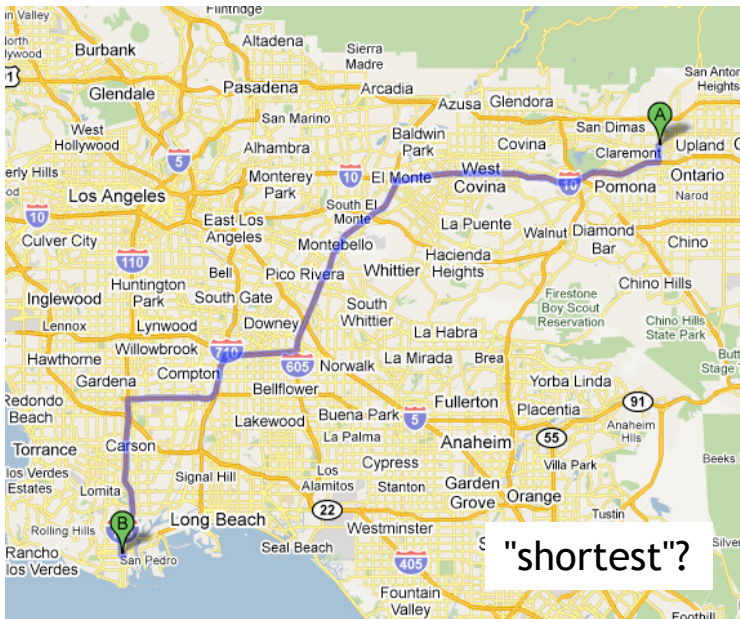
Suppose we'd like the shortest route from any city to any other.



didn't we solve this already?

Shortest paths!

Suppose we'd like the shortest route from any city to any other.



Computationally, many path-planners represent their maps as **GRAPHS**

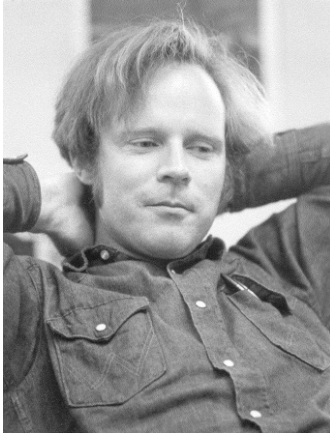
Racket: min-path

What's the big-O running time?

hw8pr3: Floyd-Warshall

Robert Floyd

From Wikipedia, the free encyclopedia



For other persons named Robert Floyd, see [Robert Floyd \(disambiguation\)](#).

Robert W Floyd (June 8, 1936 – September 25, 2001) was an eminent computer scientist.

Born in [New York](#), Floyd finished school at age 14. At the [University of Chicago](#), he received a [Bachelor's degree in liberal arts](#) in 1953 (when still only 17) and a second Bachelor's degree in [physics](#) in 1958.

Becoming a computer operator in the early 1960s, he began publishing many noteworthy papers and was appointed an associate professor at [Carnegie Mellon University](#) by the time he was 27 and became a full professor at [Stanford University](#) six years later. He obtained this position without a [Ph.D.](#)

His contributions include the design of [Floyd's algorithm](#), which effici

All-pairs shortest paths!

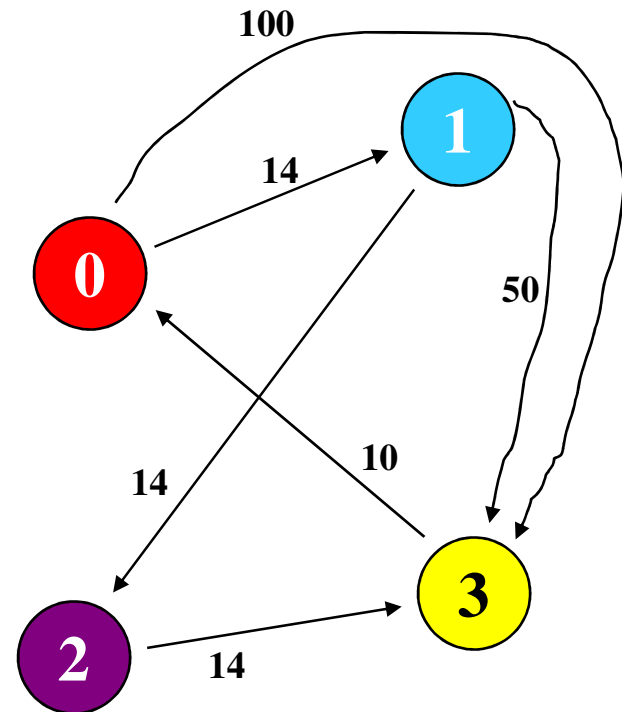
Warshall's algorithm

There is an interesting anecdote about the [Warshall's algorithm](#), now known as [Warshall's algorithm](#). A colleague at Technical Operations bet a bottle of [rum](#) on who first could determine if this [algorithm](#) always works. Warshall came up with his [proof](#) overnight, winning the bet and the [rum](#), which he shared with the loser of the bet. Because Warshall did not like sitting at a desk, he did much of his creative work in unconventional places such as on a [sailboat](#) in the [Indian Ocean](#) or in a [Greek lemon orchard](#).

Idea: start with an adjacency matrix

What's going on here?

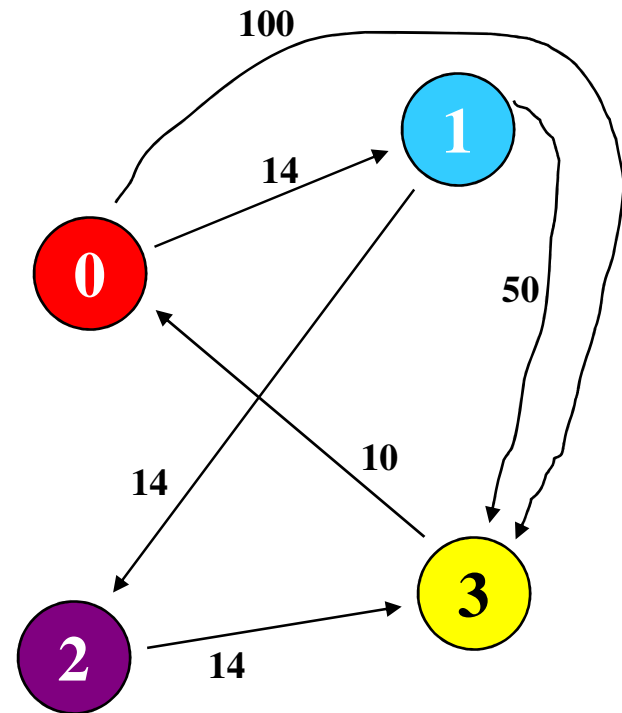
0	14	inf	100
inf	0	14	50
inf	inf	0	14
10	inf	inf	0



Insight: consider waypoints 1 at a time

		dst "to"			
		0	1	2	3
SRC "from"	0	0	14	inf	100
	1	inf	0	14	50
	2	inf	inf	0	14
	3	10	inf	inf	0

intermediate nodes:



Step 1 check each *src* to each *dst* THROUGH 0

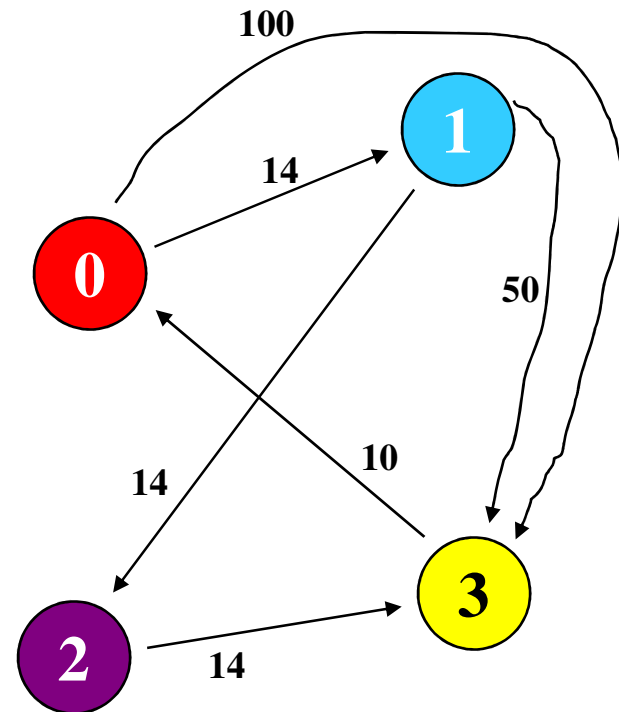
1 entry will change – which?

dst
"to"

	0	1	2	3
0	0	14	inf	100
1	inf	0	14	50
2	inf	inf	0	14
3	10	inf	inf	0

src
"from"

intermediate nodes: 0



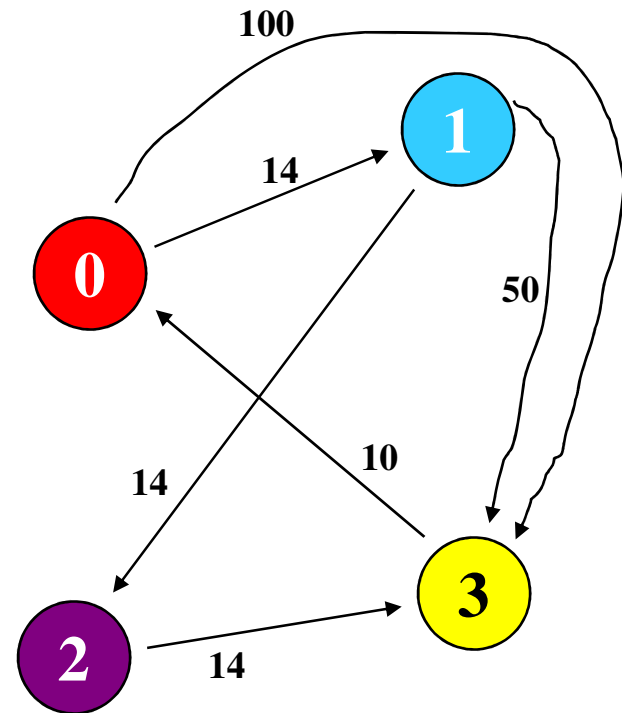
Step 1 check each *src* to each *dst* THROUGH 0

dst
"to"

	0	1	2	3
0	0	14	inf	100
1	inf	0	14	50
2	inf	inf	0	14
3	10	24	inf	0

src
"from"

intermediate nodes: 0

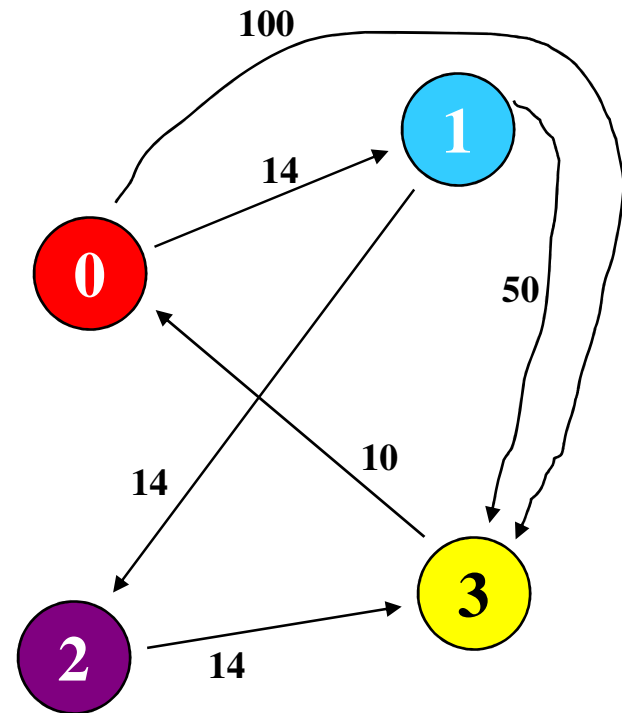


Step 2 check each *src* to each *dst* THROUGH ①

3 entries will change – which?

		dst "to"			
		①	②	③	④
SRC "from"	①	0	14	inf	100
	②	inf	0	14	50
	③	inf	inf	0	14
	④	10	24	inf	0

intermediate nodes: ① ②



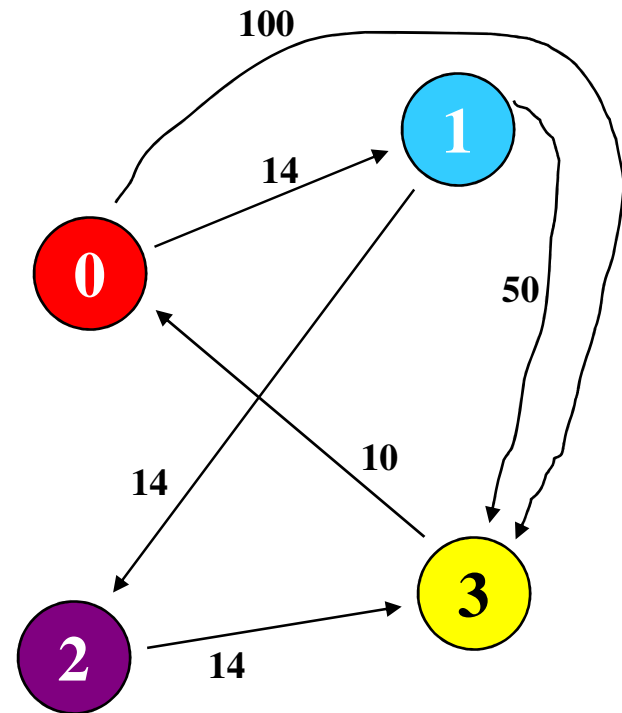
Step 2 check each *src* to each *dst* THROUGH ①

dst
"to"

	①	②	③	
①	0	14	28	64
②	inf	0	14	50
③	inf	inf	0	14
④	10	24	38	0

src
"from"

intermediate nodes: ① ②

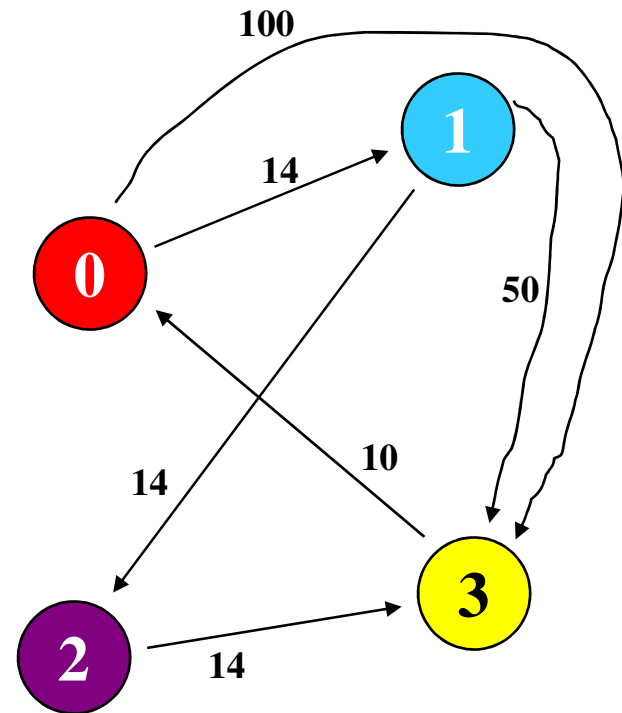


Step 3 check each *src* to each *dst* THROUGH 2

2 entries will change – which?

		dst "to"			
		0	1	2	3
SRC "from"	0	0	14	28	64
	1	inf	0	14	50
	2	inf	inf	0	14
	3	10	24	38	0

intermediate nodes: 0 1 2

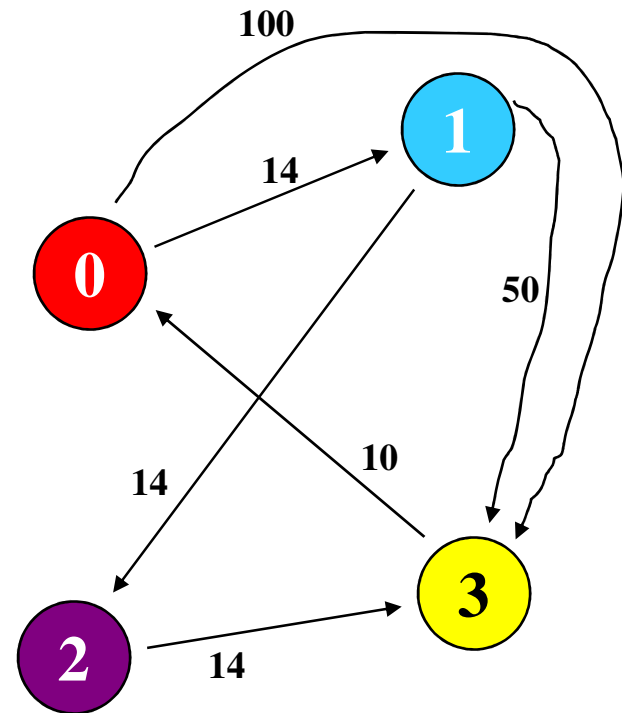


Step 3 check each *src* to each *dst* THROUGH 2

2 entries will change – which?

		dst "to"			
		0	1	2	3
SRC "from"	0	0	14	28	42
	1	inf	0	14	28
	2	inf	inf	0	14
	3	10	24	38	0

intermediate nodes: 0 1 2



Done!

We now have all of the shortest-distances
between all (src, dst) pairs!

What's the big-O runtime?

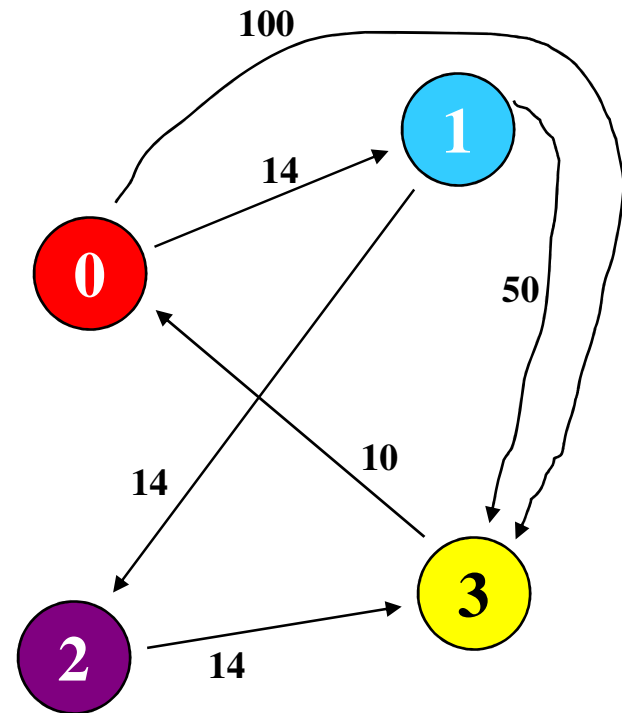
And what about the *paths*?

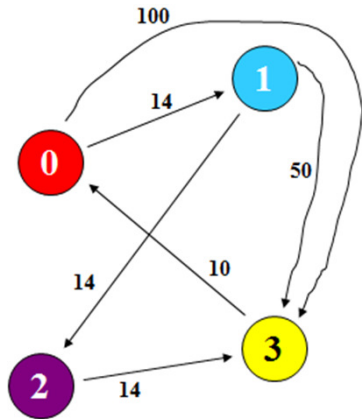
dst
"to"

	0	1	2	3
0	0	14	28	42
1	38	0	14	28
2	24	38	0	14
3	10	24	38	0

src
"from"

intermediate nodes: 0 1 2 3





We now have all of the shortest-distances between all (src, dst) pairs!

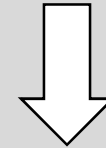
dst
"to"

	0	1	2	3
0	0	14	28	42
1	38	0	14	28
2	24	38	0	14
3	10	24	38	0

intermediate nodes: 0 1 2 3

What's the big-O runtime?

And what about the *paths*?



0	0	0	0
1	1	1	1
2	2	2	2
3	3	3	3

How could we keep track of the paths as we go...?!

FW is popular *in practice*...



Kevin Burke to me

[show details](#) 10/27/10

[← Rep](#)

I think the key takeaway from the toll problem is, if **Floyd Warshall** doesn't solve your problem, clearly you didn't run **Floyd Warshall** enough times.

I'd like to submit the revised ACM problem solving approach:

Old

Try **Floyd Warshall**

If that doesn't work consider other algorithms

New

Try **Floyd Warshall**

Try running **Floyd Warshall** n times

If that still doesn't solve your problem then consider other algorithms

Thanks, Kevin!

May your weekend
have $O(2^N)$ runtime!

Join us for LAC hours Friday!

Step 4

check each *src* to each *dst* THROUGH 3

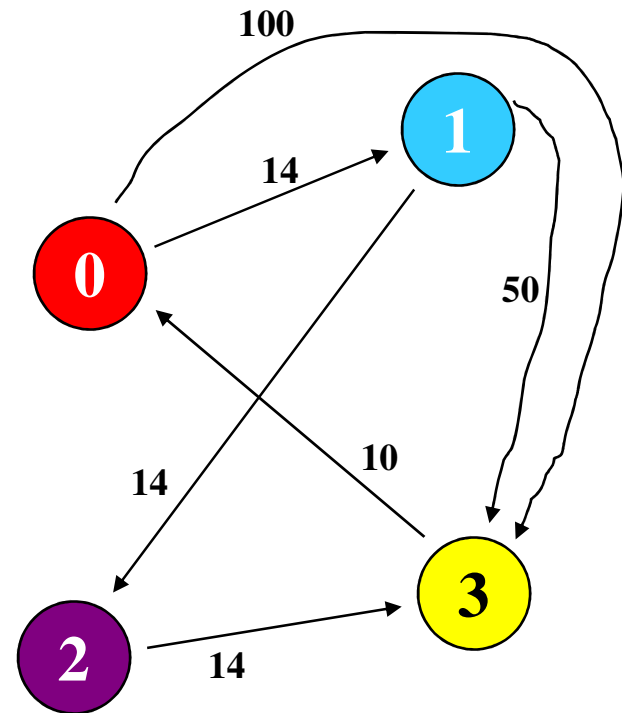
3 entries will change – which?

dst
"to"

	0	1	2	3
0	0	14	28	42
1	inf	0	14	28
2	inf	inf	0	14
3	10	24	38	0

src
"from"

intermediate nodes: 0 1 2 3



Done!

$O(\quad)$?

the path?

Demo!

dst
"to"

	1	2	3	4
1	0	14	28	42
2	38	0	14	28
3	24	38	0	14
4	10	24	38	0

4 intermediate node(s)

$T[k][src][dst]$

$$= \min \begin{cases} \text{lose } k \\ T[k-1][src][dst] \\ T[k-1][src][k] + T[k-1][k][dst] \\ \text{use } k \end{cases}$$

Whole algorithm!

Step 2 check each *src* to each *dst* THROUGH ②

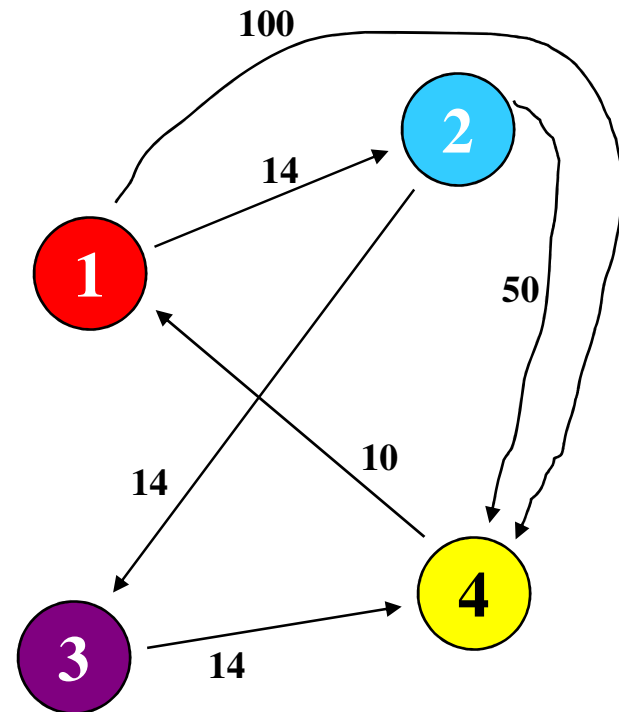
3 entries will change – which?

dst
"to"

	①	②	③	④
①	0	14	inf	100
②	inf	0	14	50
③	inf	inf	0	14
④	10	24	inf	0

src
"from"

1 intermediate node(s)
①



Step 3 check each *src* to each *dst* THROUGH 3

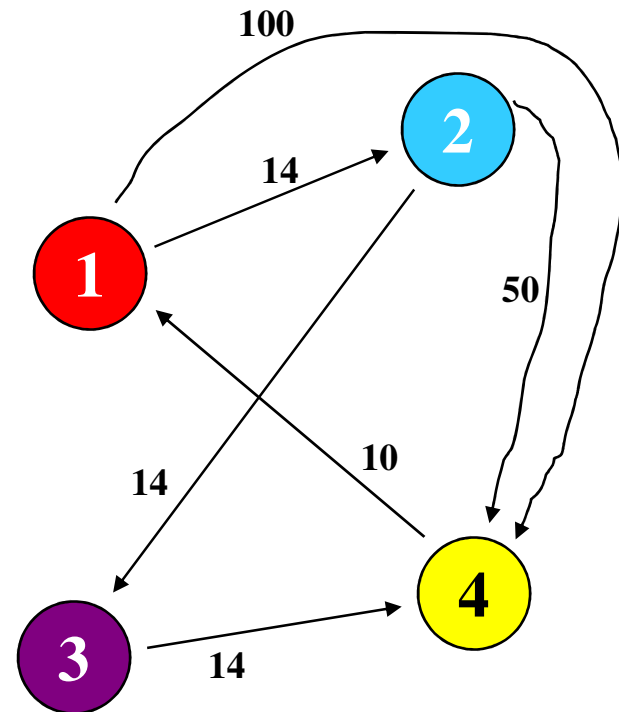
2 entries will change – which?

dst
"to"

	1	2	3	4
1	0	14	28	64
2	inf	0	14	50
3	inf	inf	0	14
4	10	24	38	0

src
"from"

2 intermediate node(s)
1 2



Step 3 check each *src* to each *dst* THROUGH 4

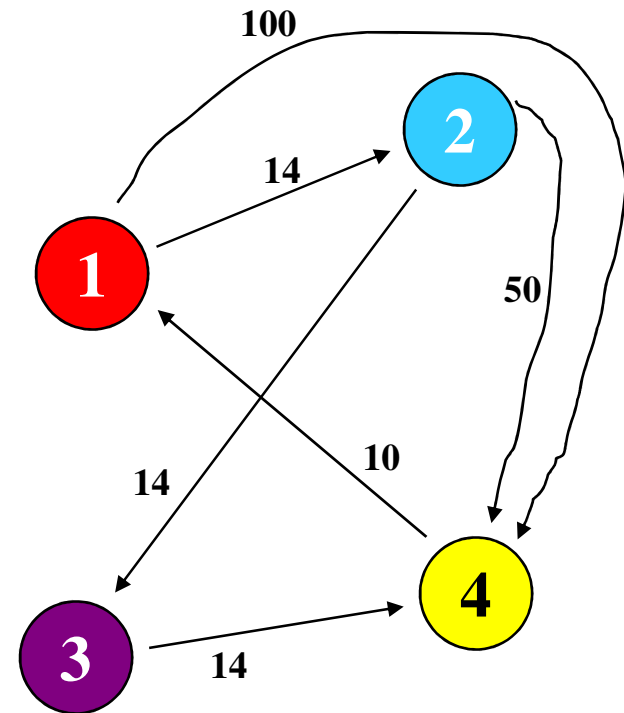
3 entries will change – which?

dst
"to"

	1	2	3	4
1	0	14	28	42
2	inf	0	14	28
3	inf	inf	0	14
4	10	24	38	0

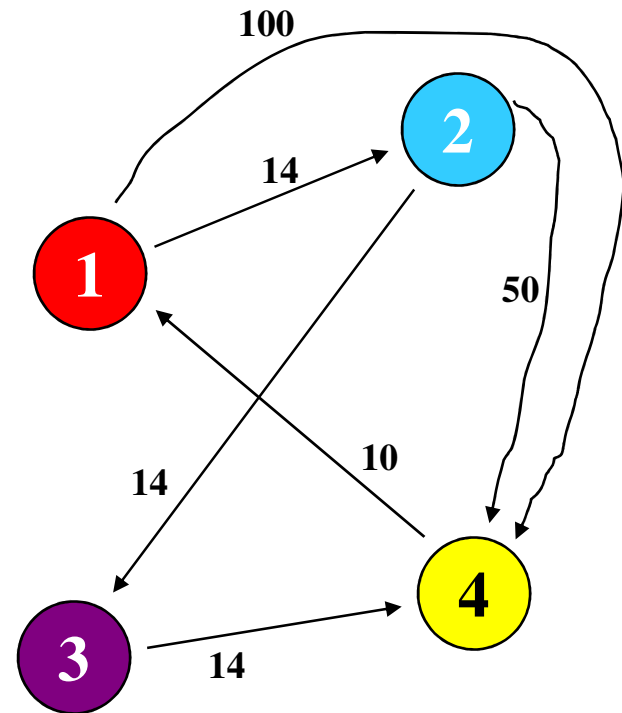
src
"from"

3 intermediate node(s)
1 2 3



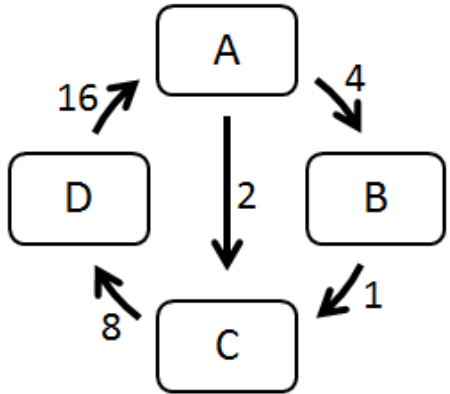
map2.txt

0	14	28	42
38	0	14	28
24	38	0	14
10	24	38	0



Edge	Cost	Through A	Cost	Through B	Cost	Through C	Cost	Through D	Cost
A → A	0	A → A → A	0	A → B → A	0	A → C → A	0	A → D → A	0
A → B	4	A → A → B	4	A → B → B	4	A → C → B	4	A → D → B	4
A → C	2	A → A → C	2	A → B → C	2	A → C → C	2	A → D → C	2
A → D	∞	A → A → D	∞	A → B → D	∞	A → C → D	10	A → D → D	10
B → A	∞	B → A → A	∞	B → B → A	∞	B → C → A	∞	B → D → A	25
B → B	0	B → A → B	∞	B → B → B	0	B → C → B	∞	B → D → B	0
B → C	1	B → A → C	∞	B → B → C	1	B → C → C	1	B → D → C	1
B → D	∞	B → A → D	∞	B → B → D	∞	B → C → D	9	B → D → D	9
C → A	∞	C → A → A	∞	C → B → A	∞	C → C → A	∞	C → D → A	24
C → B	∞	C → A → B	∞	C → B → B	∞	C → C → B	∞	C → D → B	28
C → C	0	C → A → C	0	C → B → C	0	C → C → C	0	C → D → C	0
C → D	8	C → A → D	8	C → B → D	8	C → C → D	8	C → D → D	8
D → A	16	D → A → A	16	D → B → A	16	D → C → A	16	D → D → A	16
D → B	∞	D → A → B	20	D → B → B	20	D → C → B	20	D → D → B	20
D → C	∞	D → A → C	18	D → B → C	18	D → C → C	18	D → D → C	18
D → D	0	D → A → D	0	D → B → D	0	D → C → D	0	D → D → D	0

map1.txt



SOLUTION

	$Y \rightarrow X \rightarrow X$ or $X \rightarrow X \rightarrow Y$	No change
	$X \rightarrow X$ or $X \rightarrow Y \rightarrow X$	No change (already 0)

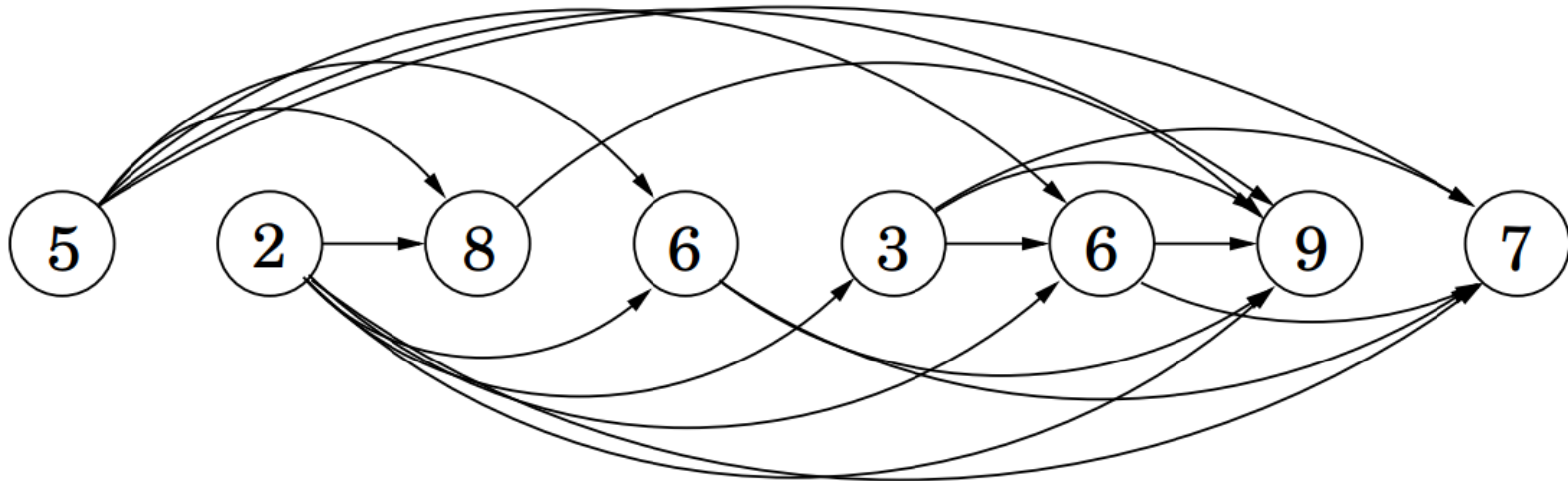
MISS example

Length of max path
to get to this node.

1	1	2	2	2	3	4	4
---	---	---	---	---	---	---	---

Parents?

5 2 8 6 3 6 9 7



MISS example

Length of max path to get to this node.

1	1	2	2	2	3	4	4
---	---	---	---	---	---	---	---

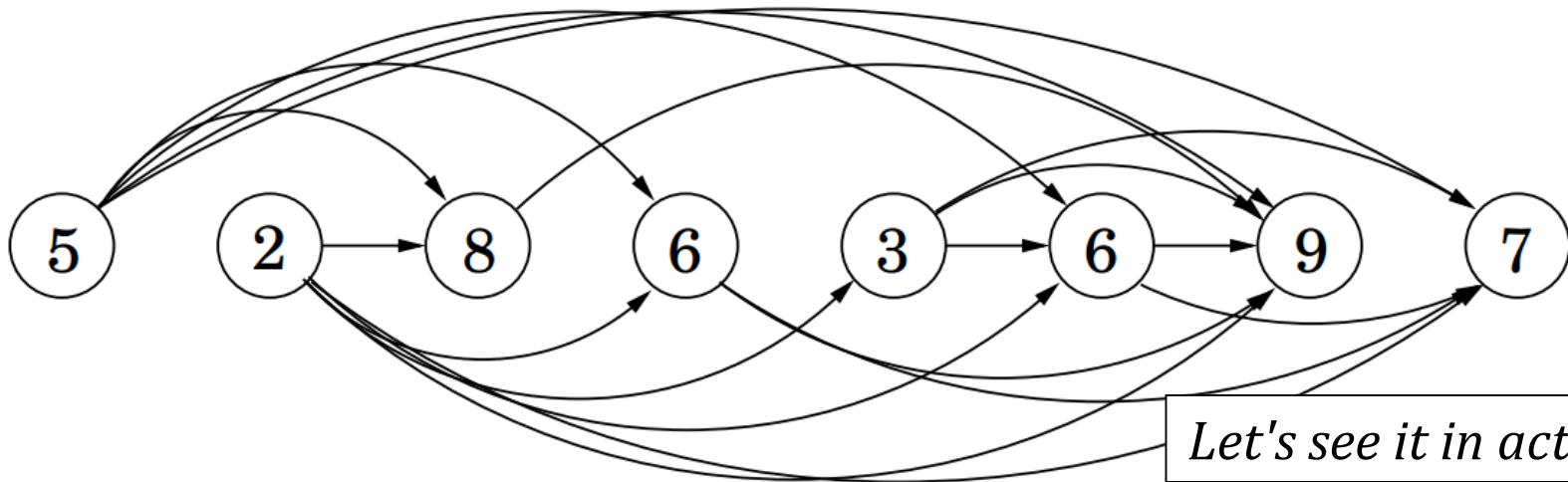
-1 -1 0 0 1 4 5 5

Parents

5 2 8 6 3 6 9 7

0 1 2 3 4 5 6 7

indices



hw11 problem 3: big-O

(Part 3) **big-O** analysis

All functions are equal; some functions
are more equal than others...

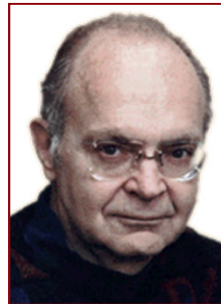
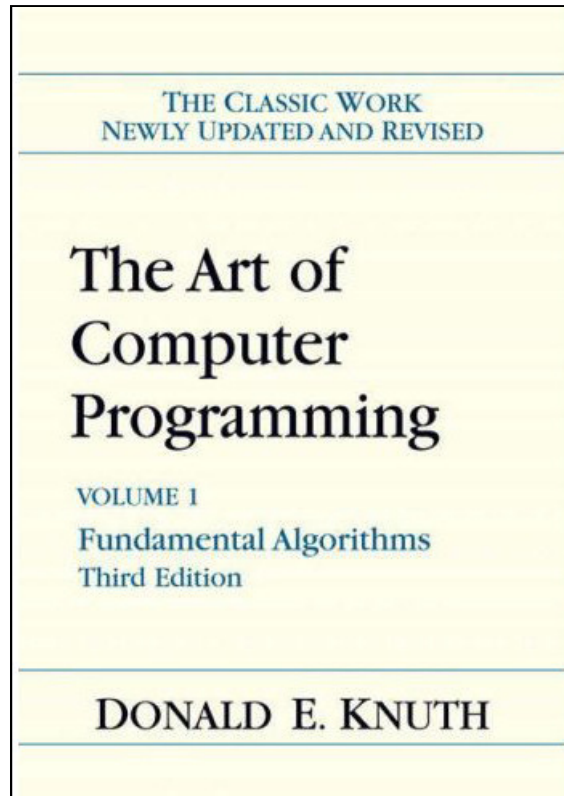
George **big-O**rwell

hw11 problem 3: big-O

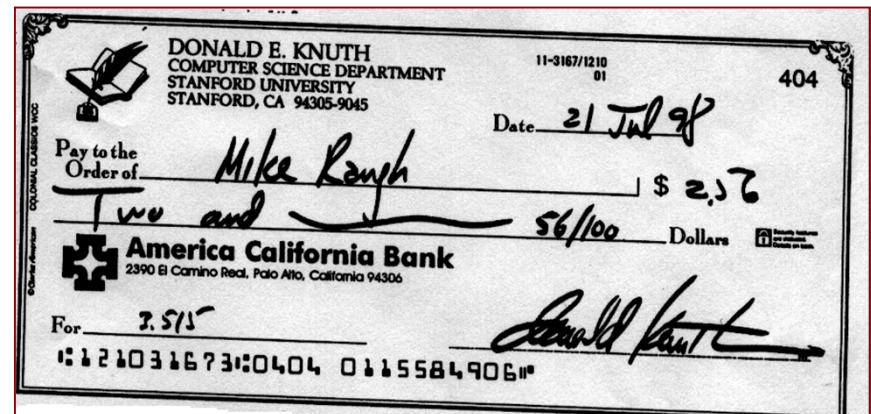
(Part 3) big-O analysis

All functions are equal; some functions are more equal than others...

George **big-O**rwell



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Considered an expert at writing **compilers**, Knuth started to write a book about compiler design in 1962, and soon realized that the scope of the book needed to be much larger. In June 1965, Knuth finished the first draft of what was originally planned to be a single volume of twelve chapters. His hand-written manuscript was 3,000 pages long: he had assumed that about five hand-written pages would translate into one printed page, but his publisher said instead that about 1½ hand-written pages translated to one printed page. This meant the book would be approximately 2,000 pages in length. At this point, the plan was changed: the book would be published in seven volumes, each with just one or two chapters. Due to the growth in the material, the plan for Volume 4 has since expanded to include Volumes 4A, 4B, 4C, and possibly 4D.