

CS 170: Algorithms

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Static Course Webpage. (cs170.org)

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... ([Staff](#))

CS 170: Algorithms

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Will mostly use piazza. ([Link.](#))

CS 170: Algorithms

Static Course Webpage. (cs170.org)

... ([Staff](#))

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Homeworks will be turned on gradescope. Latex ok, scanning handwritten stuff ok.

CS 170: Algorithms

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A puzzle.

Does a list have a cycle?

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Does a list have a cycle?

Access to list is a pointer to the “first element.”

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Mark first node.

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While next cell not marked, go to next cell.

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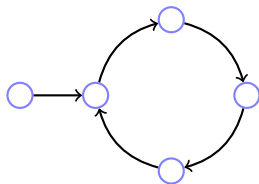
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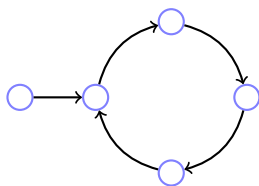
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First node not in cycle!

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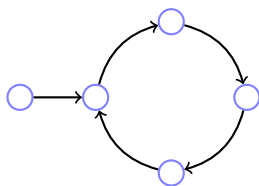
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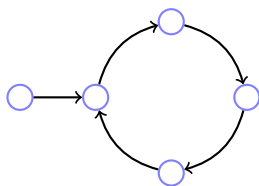
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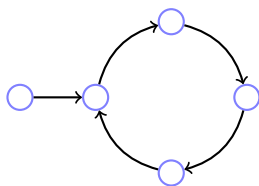
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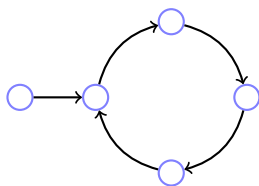
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Construct example.

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Two ptrs: Step: advance ptr 1 **twice**,

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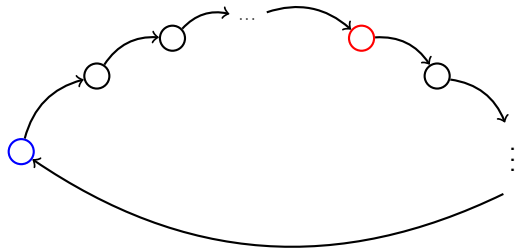
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If ever at the same place, report cycle.

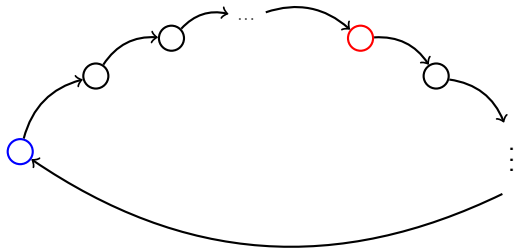
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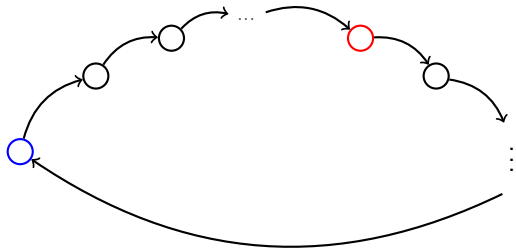
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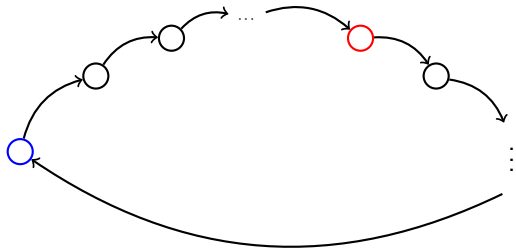


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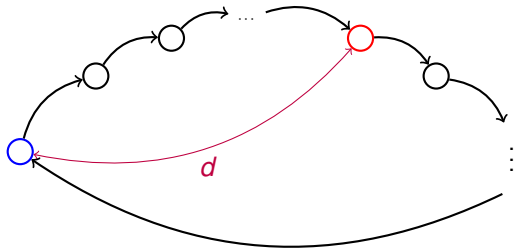
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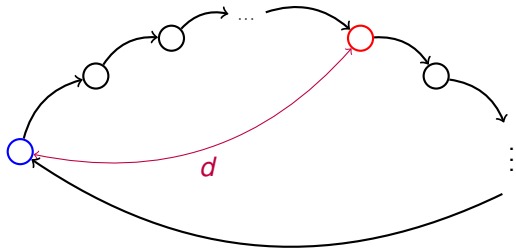
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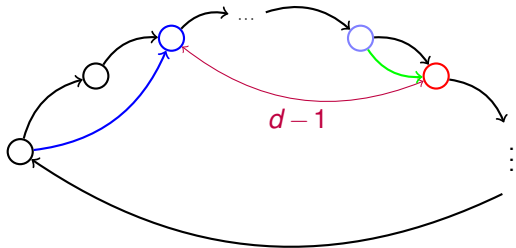
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d decreases every step.

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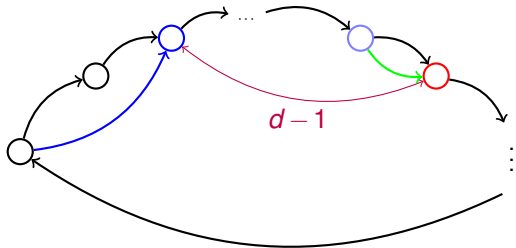
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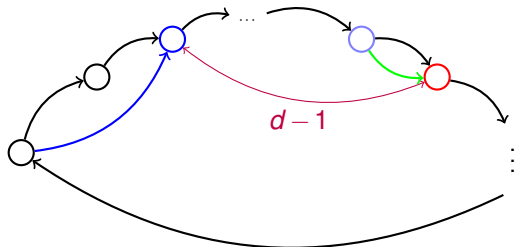
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Runtime: n steps to cycle

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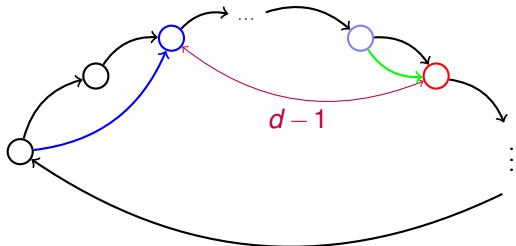
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Runtime: n steps to cycle n steps to catch up.

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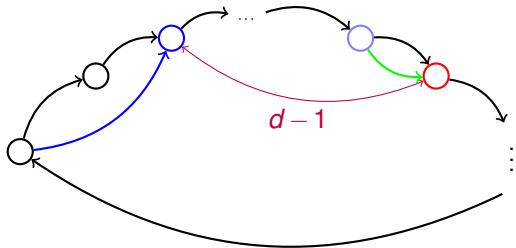
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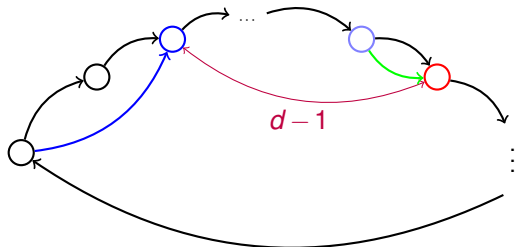
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Additional storage: two pointers.

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Puzzles ..

Solutions

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which...

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Is this a useful process?

Algorithms for the Human Genome Project

Reconstruct DNA...

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ACTGAAACTGAGTAGATA....

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Read first, then next, then next, ...3.1 billion times...

.. slow... error prone...

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Page Rank.

Problem: What is good on the web?

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Problem: What is good on the web?

Website that pays search engine most?

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Random Surfer Model (Brin-Page):

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New Model for user: google.

Algorithms...

Driving Directions

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Driving Directions
Airline Scheduling

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Compiling

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Driving Directions
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...(at Akamai, maintaining fastest, cheapest least loaded caches)

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Calculating: 300 BC through Middle Ages in Europe.

I – one

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I – one

V – five

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X – ten

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C – one hundred

Calculating: 300 BC through Middle Ages in Europe.

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C – one hundred

D – five hundred

Calculating: 300 BC through Middle Ages in Europe.

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D – five hundred

M – a thousand

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Calculating: 300 BC through Middle Ages in Europe.

I – one

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X – ten

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D – five hundred

M – a thousand

VIII – eight

Calculating: 300 BC through Middle Ages in Europe.

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DCLXXI – five hundred plus a hundred plus fifty plus ten plus ten..

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Multiply roman numbers?

Modern system.

From India, via Al Khwarizmi.

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Mayans (base 20):

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Mayans (base 20): dots (ones) and underlines (fives).

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Babylonions (base 60):

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From India, via Al Khwarizmi.

He also described recipes for adding, multiplying, solving quadratics, computing digits of π ..

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The input representation for modern computers and communication.

Writing to propagating..

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..but Fibonacci popularized its use.

Italian mathematician (1170-1250) who traveled to learn the
Hindu-Arab math.



Fibonacci numbers.

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Can we do better?

Better Algorithm.

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Let's try it.

From demo: Size matters.

How many bits in the representation of F_n ?

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$\log_2 F_n$

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At most $O(n^2)$.

Demo: polynomial.

Doubling size, made fast fib grow by factor of roughly four.

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$c(2n)^2$

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Not true for exponential algorithms. Squares runtime!

Asymptotic Analysis.

Used $O(n)$ for number of additions, rather than $n - 2$.

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2^{694n} versus $O(n^2)$.

For 2^{694n} , doubling n , squares run time.

For $O(n^2)$, doubling n , multiplies run time by four.

Refreshing Asymptotic Notation.

Ignore constant factors.

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Upper bound.

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Formally, for positive functions g, f from integers to reals,
 $g(n) = O(f(n))$, if there is a constant c where $g(n) \leq cf(n)$.

More asymptotic notation.

Ω notation.

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$g(n) = \Theta(f(n))$ if $g(n) = O(f(n))$ and $g(n) = \Omega(f(n))$.

Al Khwarizmi: Arithmetic.

Addition: $O(n)$

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	1	2	3	4	5	6	7	8	9
+	9	2	1	2	3	7	6	9	1
<hr/>									

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Addition: $O(n)$

							1	1		
	1	2	3	4	5	6	7	8	9	
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<hr/>								8	0	

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				0	1	1	1	1	
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$$\begin{array}{r} 1\ 0\ 0\ 0\ 0\ 1\ 1\ 1\ 1 \\ \ 1\ 2\ 3\ 4\ 5\ 6\ 7\ 8\ 9 \\ +\ 9\ 2\ 1\ 2\ 3\ 7\ 6\ 9\ 1 \\ \hline 1\ 0\ 4\ 4\ 6\ 9\ 4\ 4\ 8\ 0 \end{array}$$

Time: $O(n)$

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Can we do better?

Need to look at the numbers to add them...

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	1	0	0	0	0	1	1	1	1	
		1	2	3	4	5	6	7	8	9
+	9	2	1	2	3	7	6	9	1	
<hr/>										
	1	0	4	4	6	9	4	4	8	0

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More Al Khwarizmi's: algorithms.

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Addition: $O(n)$

Multiplication:

	1	2	3	4	5	6	7	8	9
×	9	2	1	2	3	7	6	9	1
<hr/>									

More Al Khwarizmi's: algorithms.

Addition: $O(n)$

Multiplication:

$$\begin{array}{r} \\ 1 \\ \times 9 \\ \hline 1 \end{array}$$


More Al Khwarizmi's: algorithms.

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			1	2	3	4	5	6	7	8	9
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<hr/>											
			1	2	3	4	5	6	7	8	9
	9	2	2	2	2	2	2	2	2	1	

.
.



n

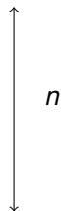
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Quick Thoughts.

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Complexity or runtimes in terms of size of representation.

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Asymptotic analysis.