

CSEP 521

Applied Algorithms

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Winter 2013

Lecture 2

Announcements

- **Reading**
 - Chapter 2.1, 2.2
 - Chapter 3
 - Chapter 4
- **Homework Guidelines**
 - Prove that your algorithm works
 - A proof is a “convincing argument”
 - Give the run time for you algorithm
 - Justify that the algorithm satisfies the runtime bound
 - You may lose points for style

Announcements

- Monday, January 21 is a holiday
 - No class
- Makeup lecture, Thursday, January 17, 5:00 pm – 6:30 pm
 - UW and Microsoft
 - View off line if you cannot attend
- Homework 2 is due January 21
 - Electronic turn in only

What does it mean for an algorithm
to be efficient?

Definitions of efficiency

- Fast in practice
- Qualitatively better worst case performance than a brute force algorithm

Polynomial time efficiency

- An algorithm is efficient if it has a polynomial run time
- Run time as a function of problem size
 - Run time: count number of instructions executed on an underlying model of computation
 - $T(n)$: maximum run time for all problems of size at most n

Polynomial Time

- Algorithms with polynomial run time have the property that increasing the problem size by a constant factor increases the run time by at most a constant factor (depending on the algorithm)

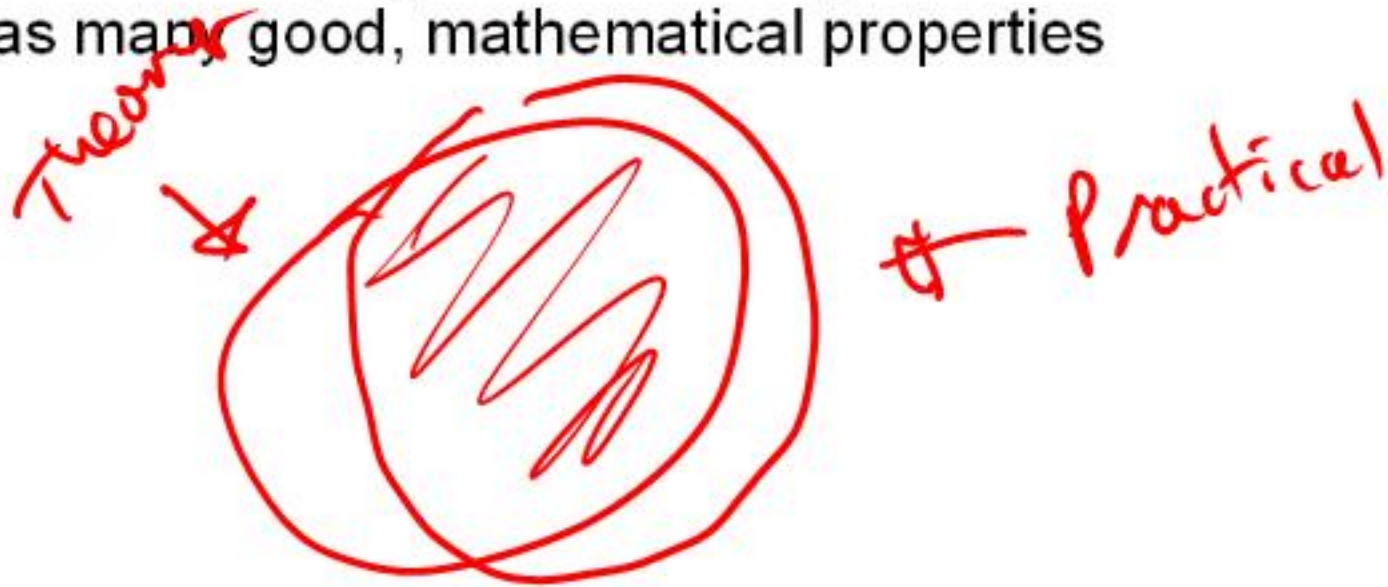
$$n^3$$

$$[2n]^3$$

$$8$$

Why Polynomial Time?

- Generally, polynomial time seems to capture the algorithms which are efficient in practice
- The class of polynomial time algorithms has many good, mathematical properties



Polynomial vs. Exponential Complexity

- Suppose you have an algorithm which takes $n!$ steps on a problem of size n
- If the algorithm takes one second for a problem of size 10, estimate the run time for the following problems sizes:

12

2 minutes

14

6 hrs

16

2 months


18

50 yrs

20

20,000 yrs

Ignoring constant factors

- Express run time as $O(f(n))$
 - Emphasize algorithms with slower growth rates
 - Fundamental idea in the study of algorithms
 - Basis of Tarjan/Hopcroft Turing Award
- 

Why ignore constant factors?

- Constant factors are arbitrary
 - Depend on the implementation
 - Depend on the details of the model
- Determining the constant factors is tedious and provides little insight

Why emphasize growth rates?

- The algorithm with the lower growth rate will be faster for all but a finite number of cases
- Performance is most important for larger problem size
- As memory prices continue to fall, bigger problem sizes become feasible
- Improving growth rate often requires new techniques

Formalizing growth rates

- $T(n)$ is $O(f(n))$ $[T : \mathbb{Z}^+ \rightarrow \mathbb{R}^+]$
 - If n is sufficiently large, $T(n)$ is bounded by a constant multiple of $f(n)$
 - Exist c , n_0 , such that for $n > n_0$, $T(n) < c f(n)$
- $T(n)$ is $O(f(n))$ will be written as:
 $T(n) = O(f(n))$
 - Be careful with this notation

Prove $3n^2 + 5n + 20$ is $O(n^2)$

show $3n^2 + 5n + 20 < 6n^2$
for $n > 5$

Let $c = 6$

Let $n_0 = 5$

Assume $n > 5$

$$3n^2 + 5n + 20 \leq 3n^2 + n^2 + n^2$$
$$\Rightarrow 5n^2 \leq 6n^2$$

$T(n)$ is $O(f(n))$ if there exist c, n_0 , such that for $n > n_0$,

$$T(n) < c f(n)$$

Order the following functions in increasing order by their growth rate

a) ~~$n \log^4 n$~~

b) ~~$2n^2 + 10n$~~

c) $2^{n/100}$

d) ~~$1000n + \log^8 n$~~

e) ~~n^{100}~~

f) ~~3^n~~

- g) ~~$1000 \log^{10} n$~~

h) ~~$n^{1/2}$~~

$1000 \log^{10} n$

$n^{1/2}$

- $1000n + \log^8 n$

- $n \log^4 n$

$2n^2 + 10n$

n^{100}

$2^{n/100}$

3^n

$$n^2 + 3n = \Omega(n)$$

Lower bounds

$$n^2 + 2n \log n = \Theta(n^2)$$

- $T(n)$ is $\Omega(f(n))$
 - $T(n)$ is at least a constant multiple of $f(n)$
 - There exists an n_0 , and $\varepsilon > 0$ such that $T(n) > \varepsilon f(n)$ for all $n > n_0$
- Warning: definitions of Ω vary
- $T(n)$ is $\Theta(f(n))$ if $T(n)$ is $O(f(n))$ and $T(n)$ is $\Omega(f(n))$

Useful Theorems

- If $\lim (f(n) / g(n)) = c$ for $c > 0$ then $f(n) = \Theta(g(n))$
- If $f(n)$ is $O(g(n))$ and $g(n)$ is $O(h(n))$ then $f(n)$ is $O(h(n))$
- If $f(n)$ is $O(h(n))$ and $g(n)$ is $O(h(n))$ then $f(n) + g(n)$ is $O(h(n))$

Ordering growth rates

- For $b > 1$ and $x > 0$
 - $\log^b n$ is $O(n^x)$
- For $r > 1$ and $d > 0$
 - n^d is $O(r^n)$

Stable Matching

Reported Results

Student	n	M / n	W / n	M / n * W / n
Stanislav	10,000	9.96	1020	10159
Andy	4,096	8.77	472	4139
Boris	5,000	10.06	499	5020
Huy	10,000	10.68	969	10349
Hans	10,000	9.59	1046	10031
Vijayanand	1,000	8.60	114	980
Robert	20,000	12.40	1698	21055
Zain	2,825	8.61	331	2850
Uzair	8,192	9.10	883	8035
Anand	10,000	9.58	1045	10011

Why is $M/n \sim \log n$?

while unmatched or
choose m -
 m - picks w .

$$\sum_{i=1}^n \frac{1}{i}$$

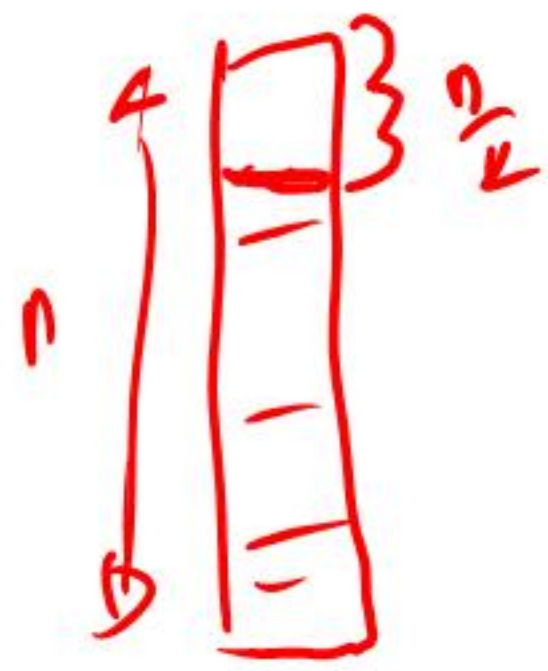


p Coupon Collector Problems

Average number of coupons
needed is $n \log n$

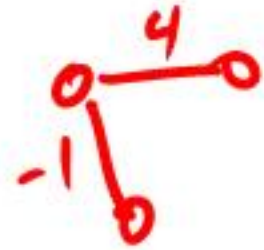
Why is $W/n \sim n / \log n$?

M's propose $n \log n$ times
each w receives $\log n$
proposals.



Graph Theory

Graph Theory

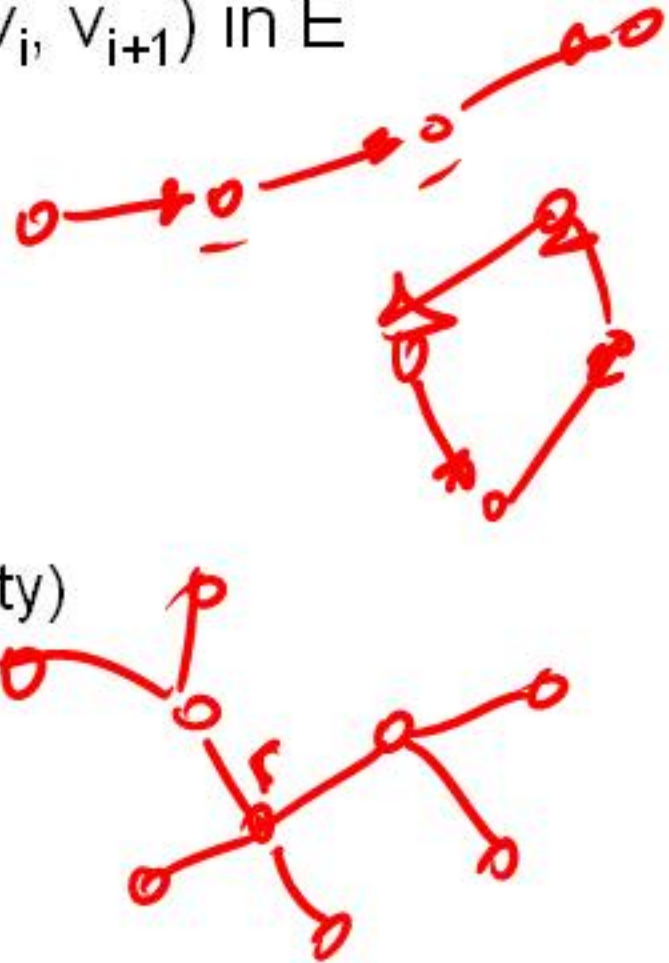


- $G = (V, E)$
 - V – vertices $|V| = n$
 - E – edges $|E| = m$
- Undirected graphs
 - Edges sets of two vertices $\{u, v\}$
- Directed graphs
 - Edges ordered pairs (u, v)
- Many other flavors
 - Edge / vertices weights
 - Parallel edges
 - Self loops



Definitions

- Path: v_1, v_2, \dots, v_k , with (v_i, v_{i+1}) in E
 - Simple Path
 - Cycle
 - Simple Cycle
- Distance
- Connectivity
 - Undirected
 - Directed (strong connectivity)
- Trees
 - Rooted
 - Unrooted



Graph search



- Find a path from s to t

$S = \{s\}$

While there exists (u, v) in E with u in S and v not in S

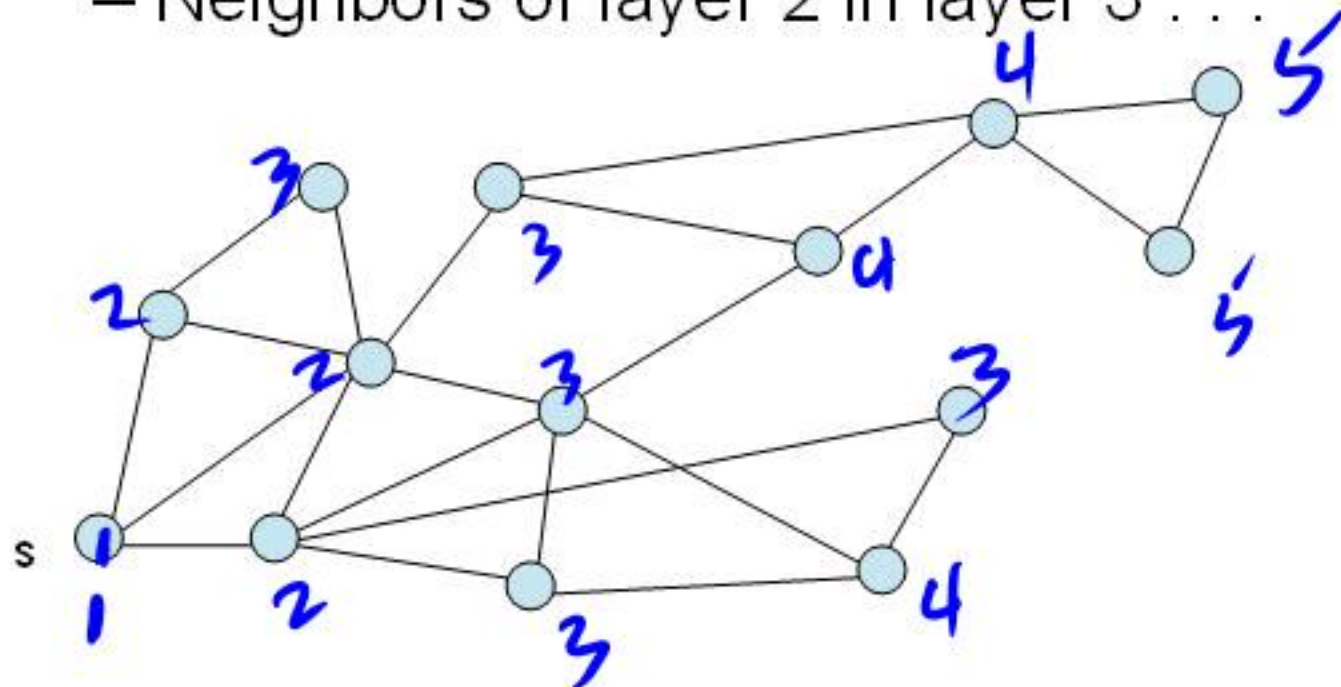
$\text{Pred}[v] = u$

Add v to S

if $(v = t)$ then path found

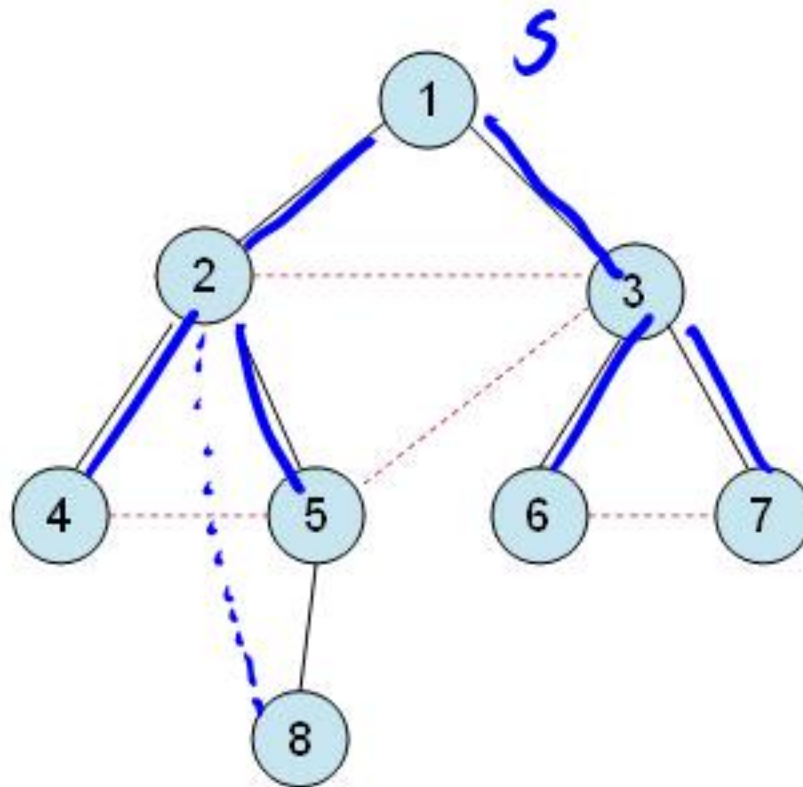
Breadth first search

- Explore vertices in layers
 - s in layer 1
 - Neighbors of s in layer 2
 - Neighbors of layer 2 in layer 3



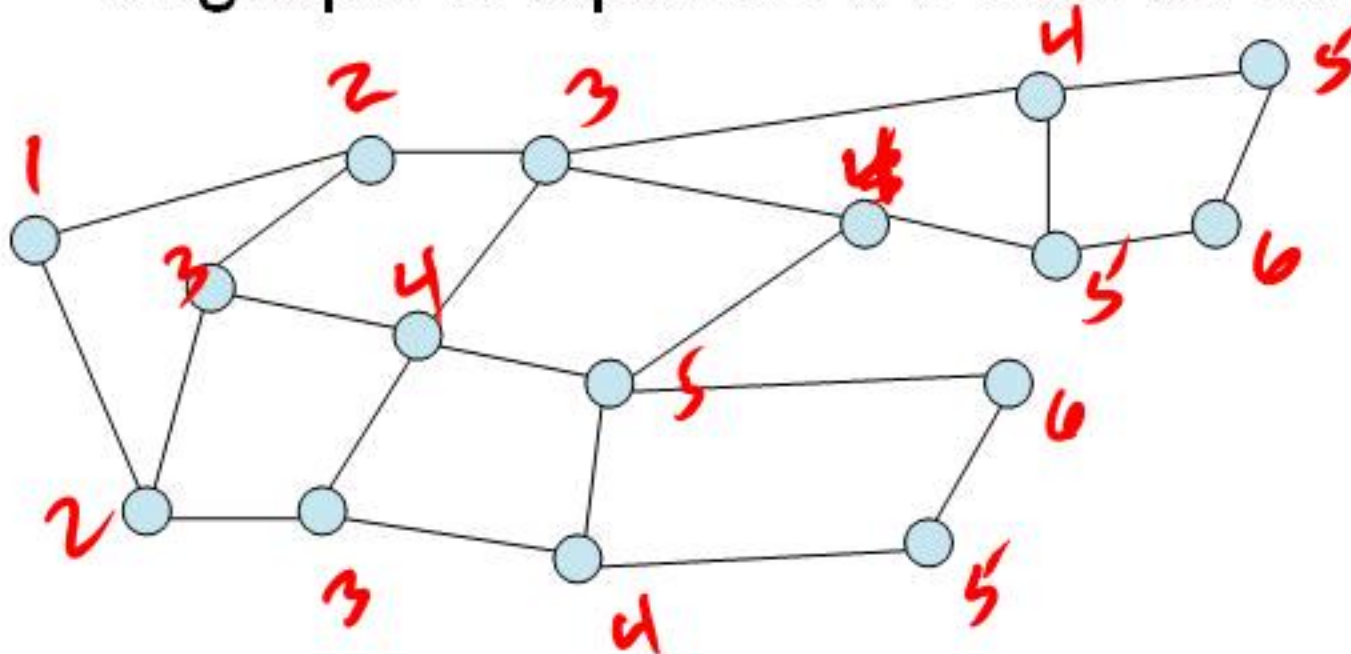
Key observation

- All edges go between vertices on the same layer or adjacent layers

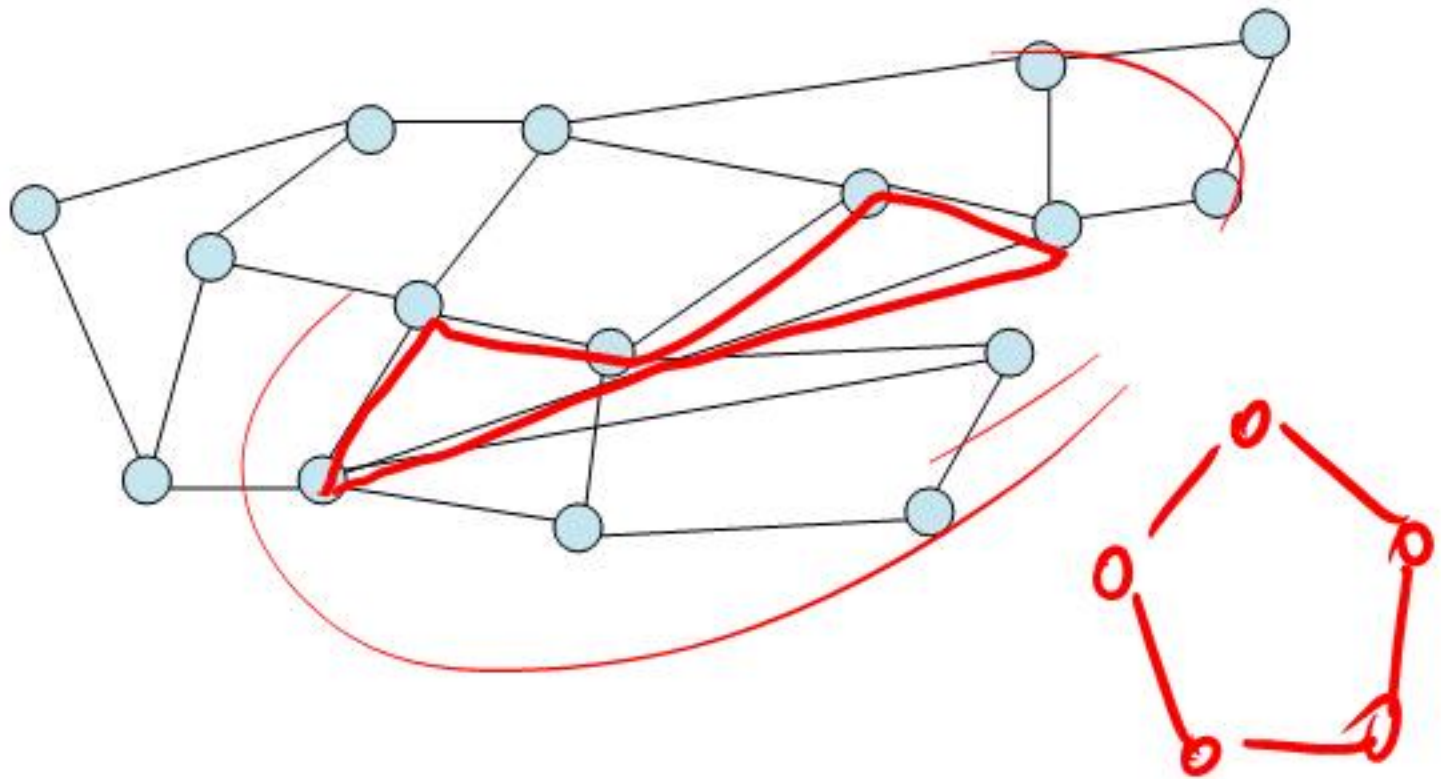


Bipartite Graphs

- A graph V is bipartite if V can be partitioned into V_1, V_2 such that all edges go between V_1 and V_2
- A graph is bipartite if it can be two colored



Can this graph be two colored?



Algorithm

- Run BFS
- Color odd layers red, even layers blue
- If no edges between the same layer, the graph is bipartite
- If edge between two vertices of the same layer, then there is an odd cycle, and the graph is not bipartite

Theorem: A graph is bipartite if and only if it has no odd cycles

Graph has an odd cycle

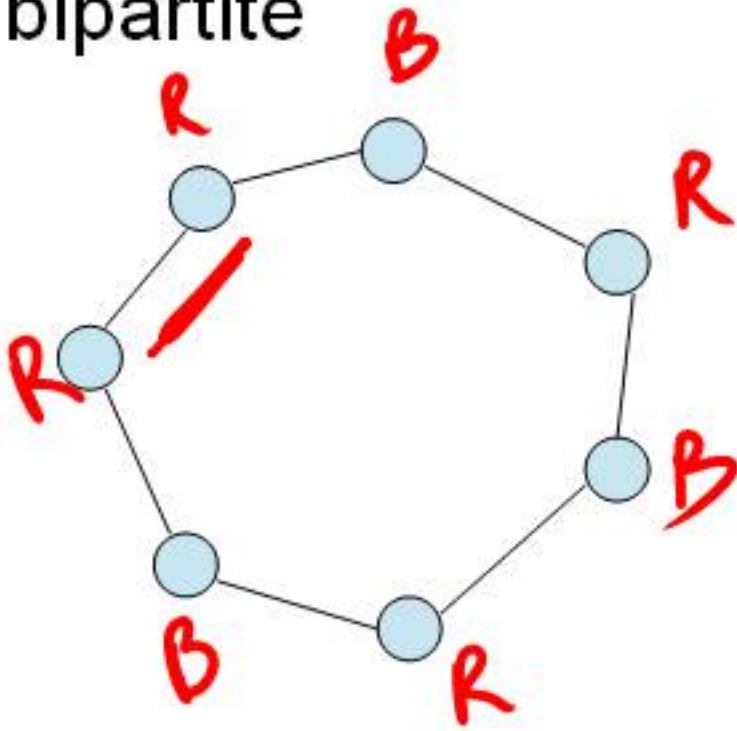
\Rightarrow Graph is not bipartite

Graph has no odd cycle

\Rightarrow Graph is bipartite

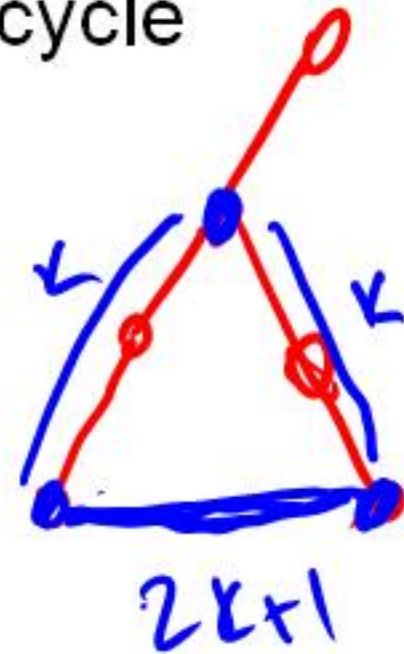
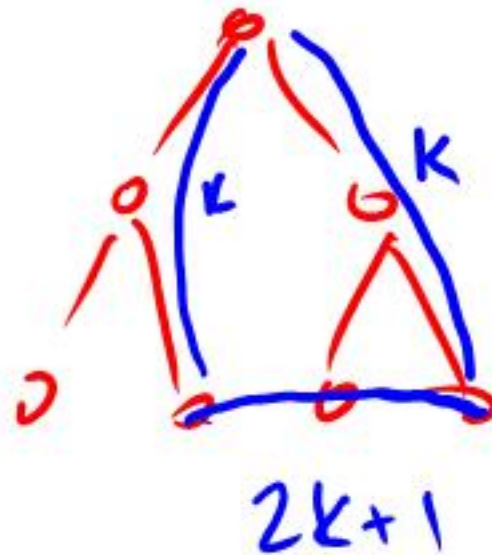
Lemma 1

- If a graph contains an odd cycle, it is not bipartite



Lemma 2

- If a BFS tree has an *intra-level edge*, then the graph has an odd length cycle



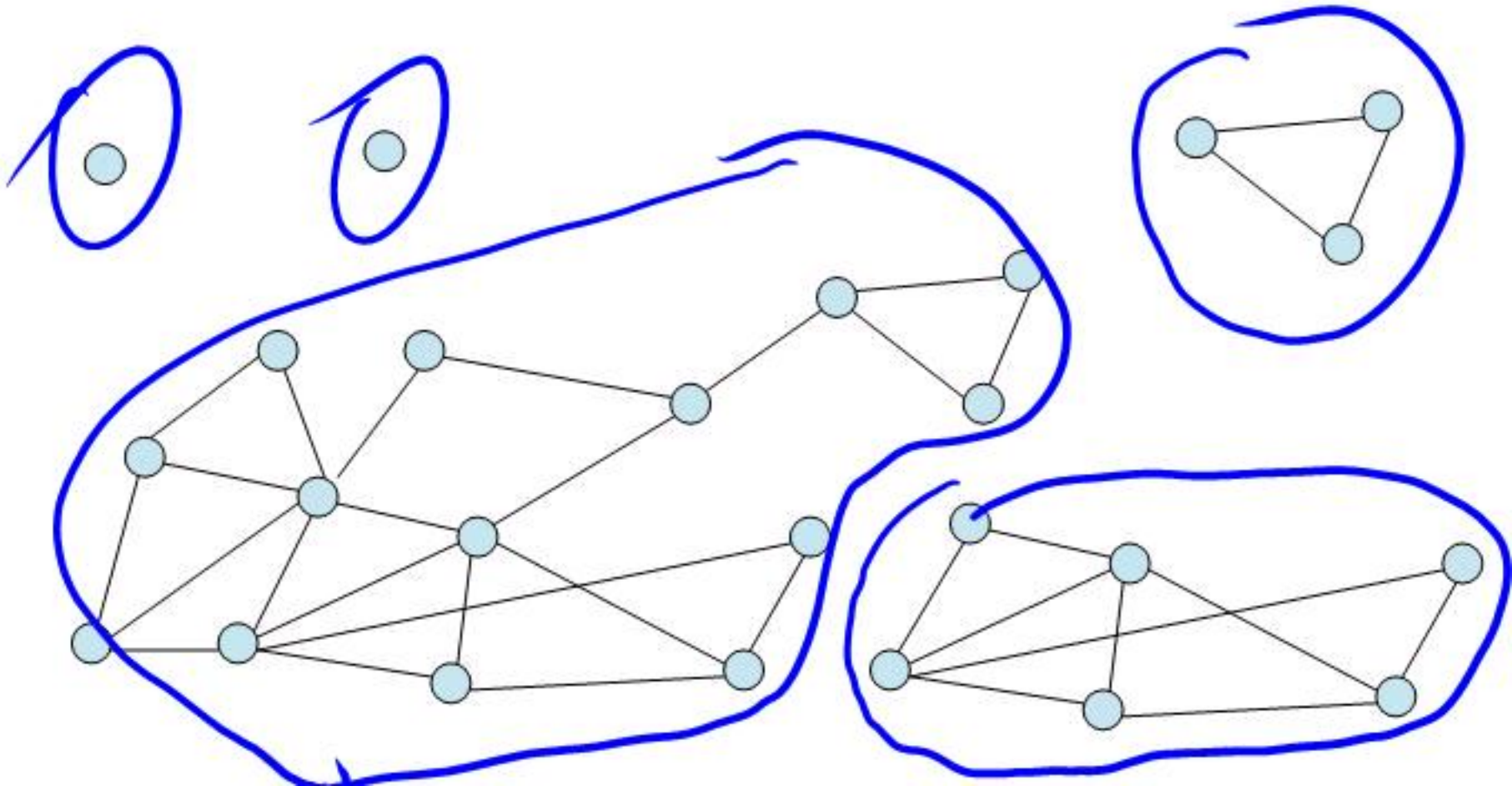
Intra-level edge: both end points are in the same level

Lemma 3

- If a graph has no odd length cycles, then it is bipartite

Connected Components

- Undirected Graphs

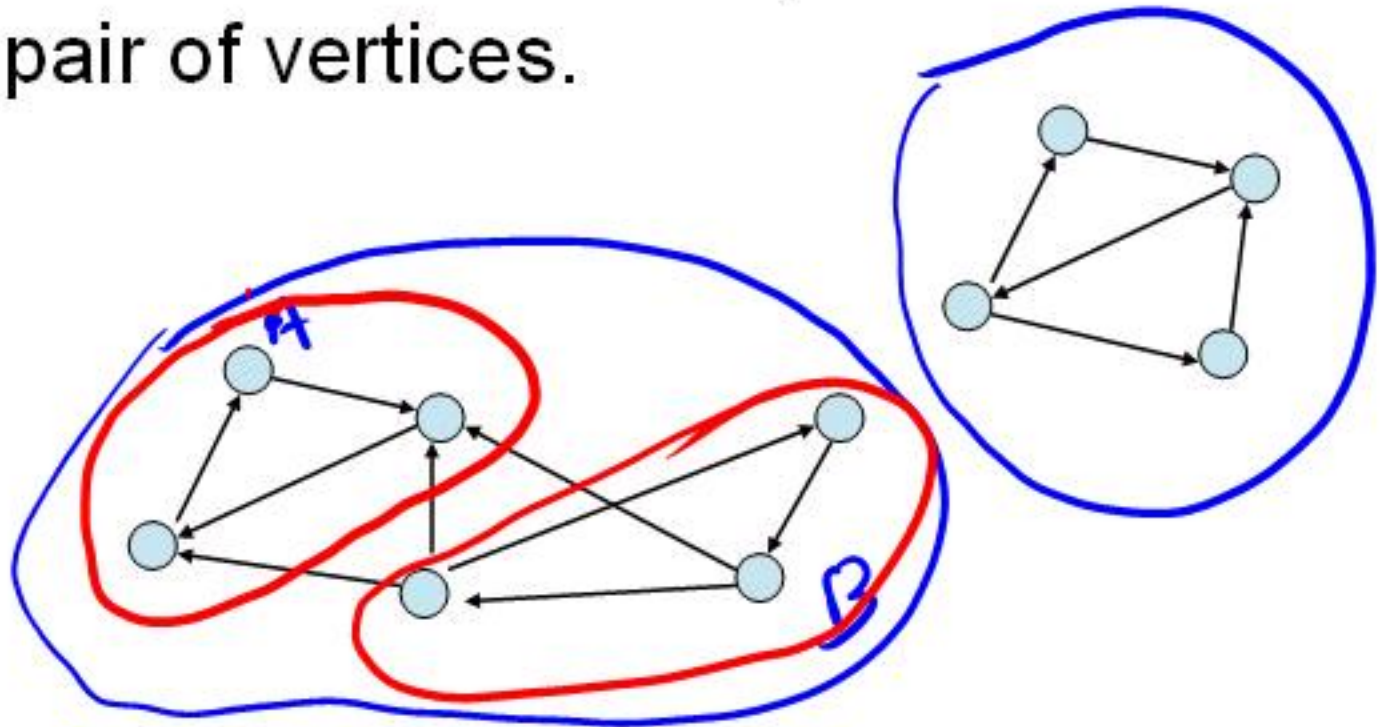


Computing Connected Components in $O(n+m)$ time

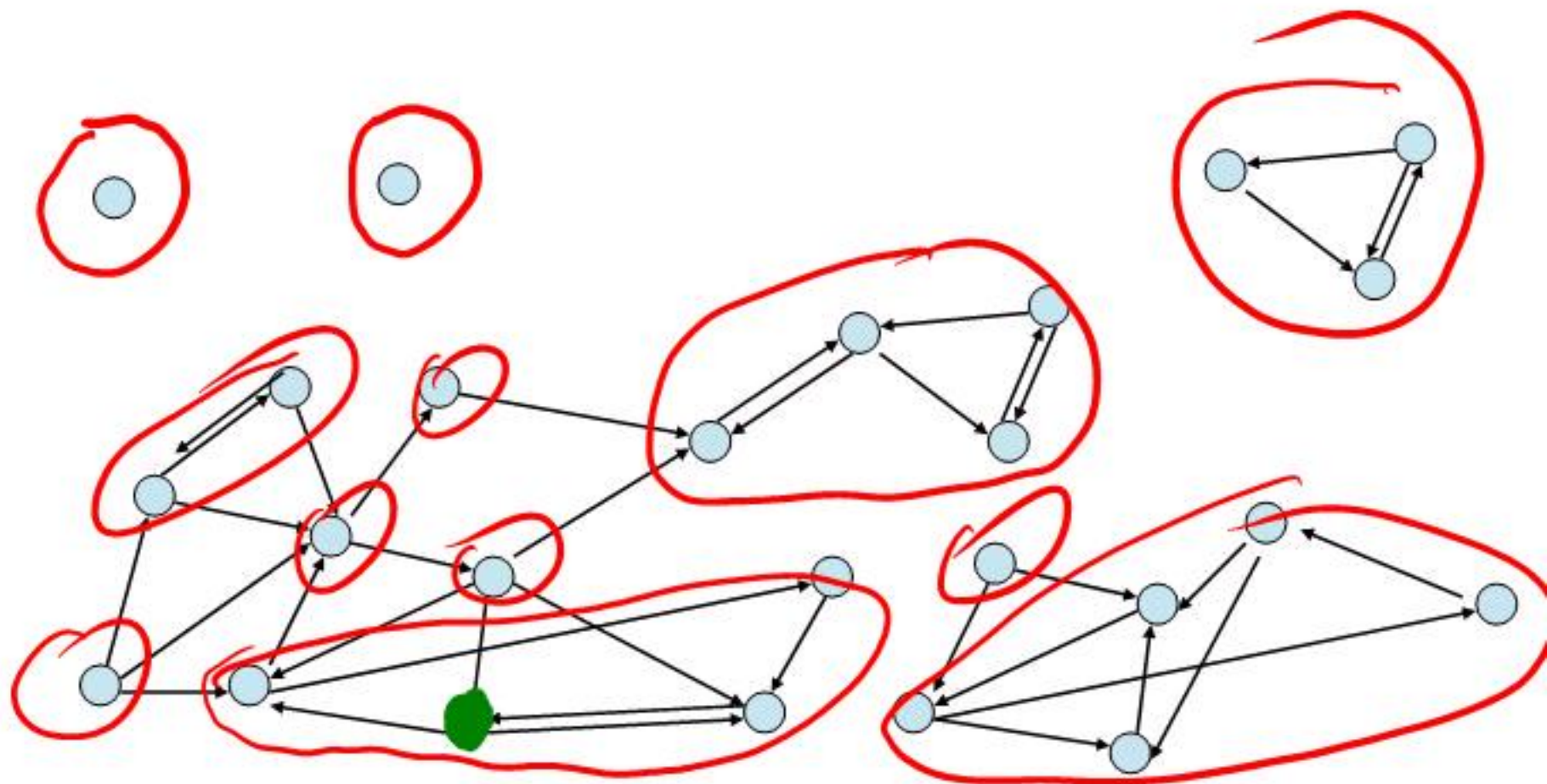
- A search algorithm from a vertex v can find all vertices in v 's component
- While there is an unvisited vertex v , search from v to find a new component

Directed Graphs

- A Strongly Connected Component is a subset of the vertices with paths between every pair of vertices.



Identify the Strongly Connected Components



Strongly connected components can be found in $O(n+m)$ time

- But it's tricky!
- Simpler problem: given a vertex v , compute the vertices in v 's scc in $O(n+m)$ time

S_1 All vertices v can reach.

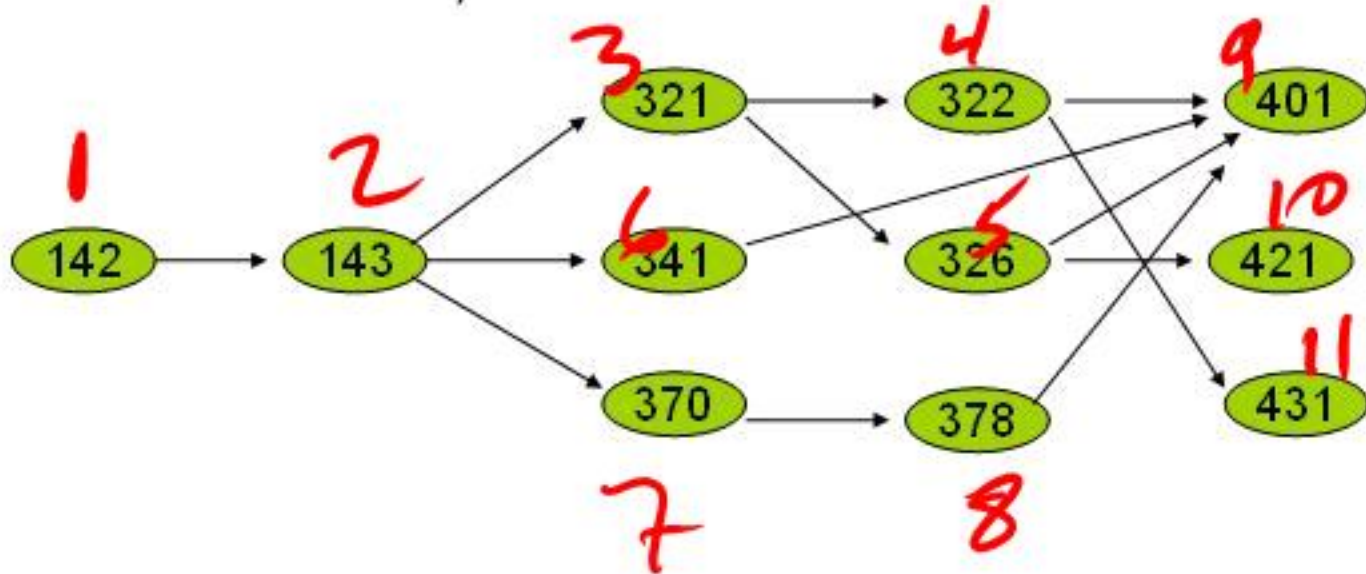
S_2 All vertices that can reach v

$$S_1 \cap S_2$$

BFS
Reverse edges
BFS

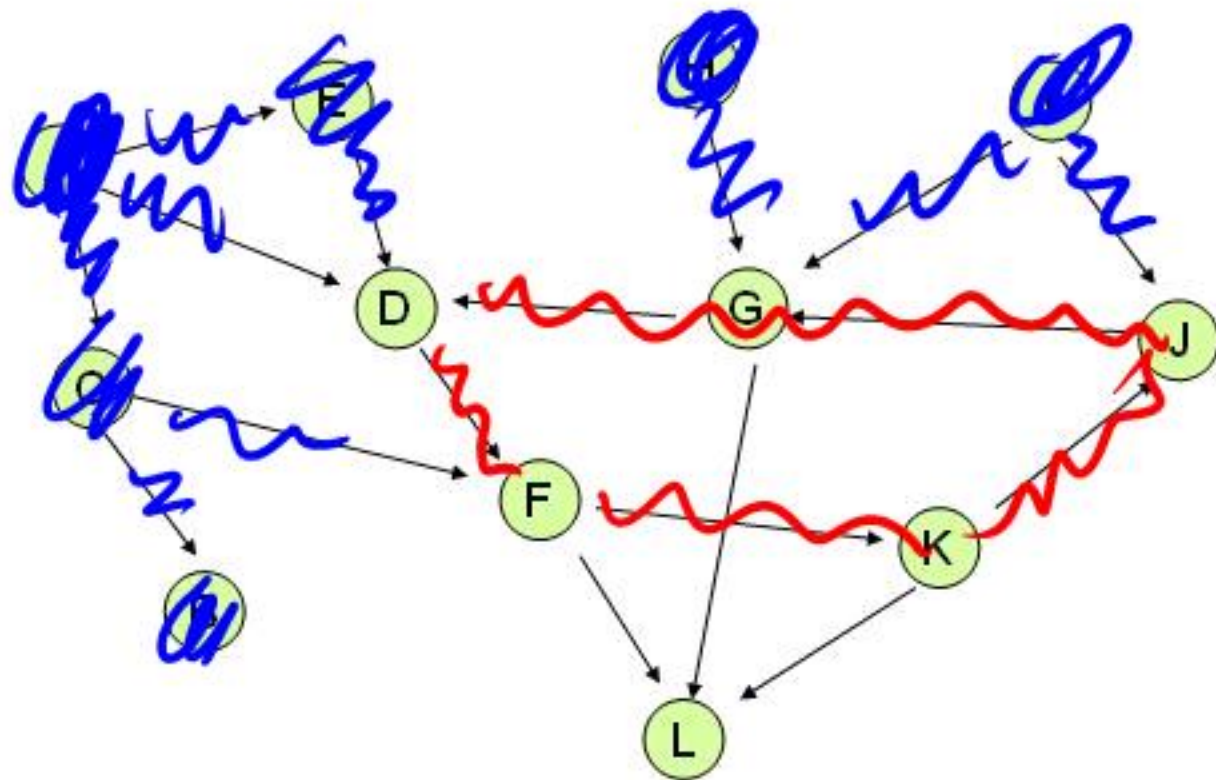
Topological Sort

- Given a set of tasks with precedence constraints, find a linear order of the tasks



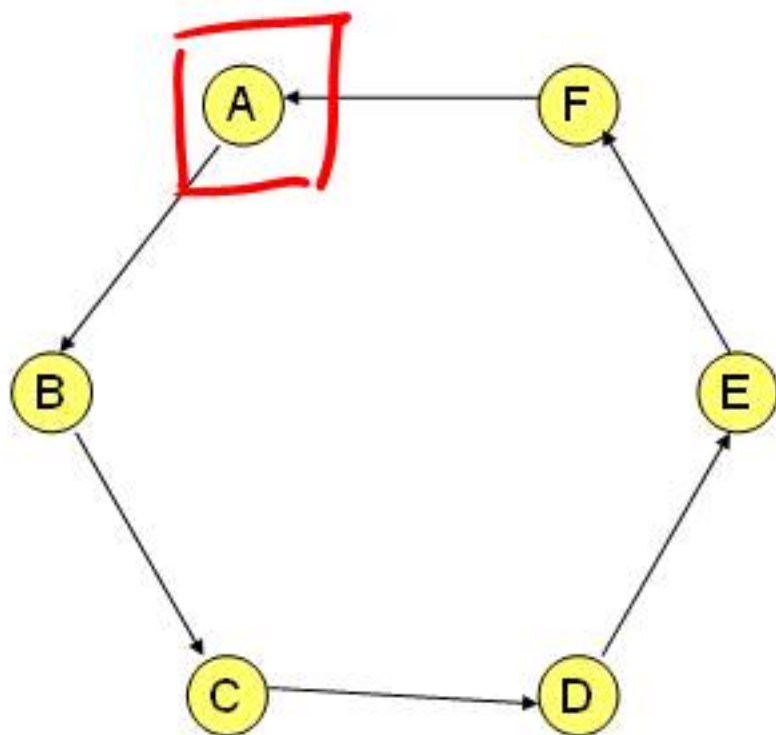
Find a topological order for the
following graph

A, H, I, E, C, B



If a graph has a cycle, there is no topological sort

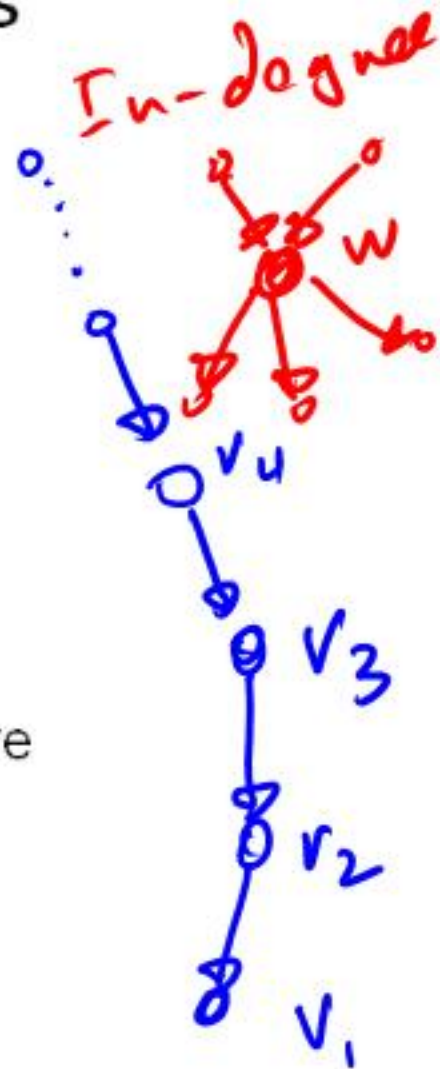
- Consider the first vertex on the cycle in the topological sort
- It must have an incoming edge



Lemma: If a graph is acyclic, it has a vertex with in degree 0

• **Proof:**

- Pick a vertex v_1 , if it has in-degree 0 then done
- If not, let (v_2, v_1) be an edge, if v_2 has in-degree 0 then done
- If not, let (v_3, v_2) be an edge . . .
- If this process continues for more than n steps, we have a repeated vertex, so we have a cycle



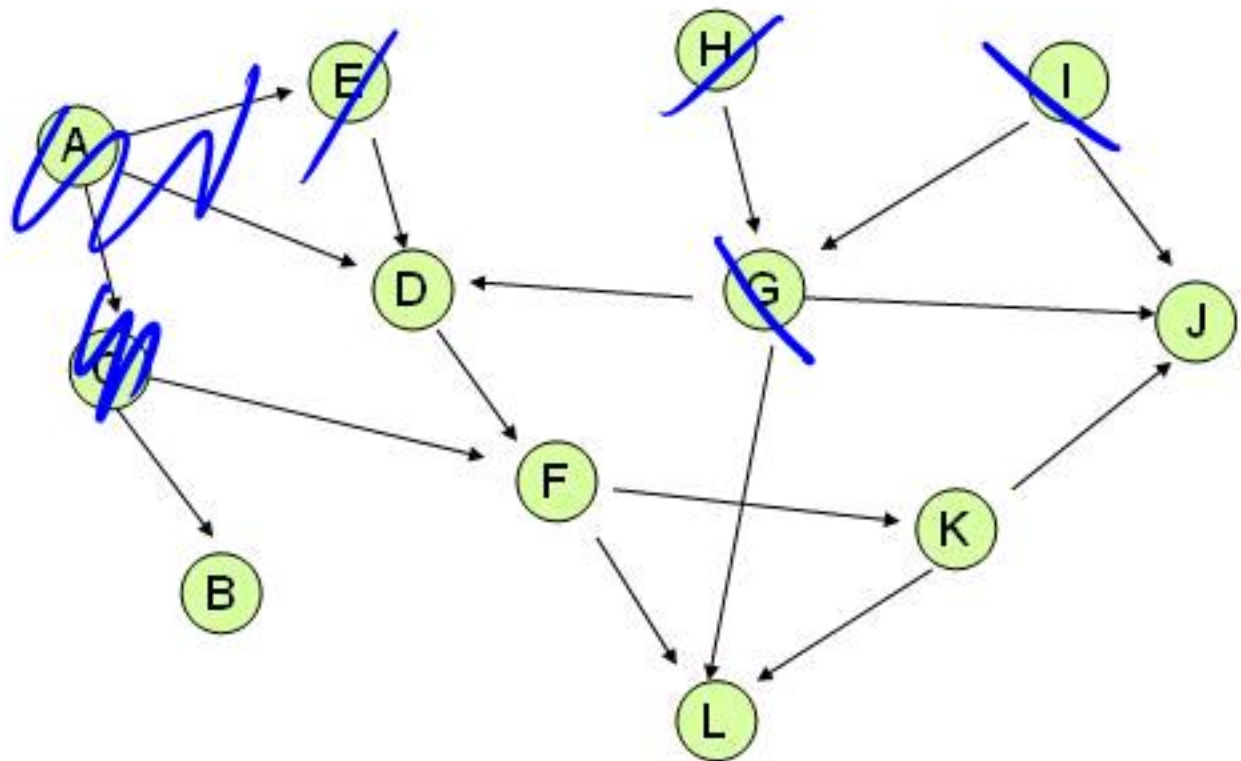
A, E, H, I, G

Topological Sort Algorithm

While there exists a vertex v with in-degree 0

Output vertex v

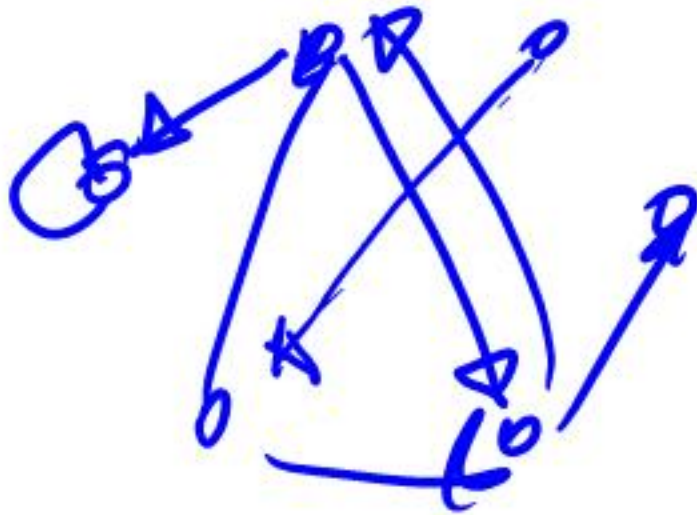
Delete the vertex v and all out going edges



Details for $O(n+m)$ implementation

- Maintain a list of vertices of in-degree 0
- Each vertex keeps track of its in-degree
- Update in-degrees and list when edges are removed
- m edge removals at $O(1)$ cost each

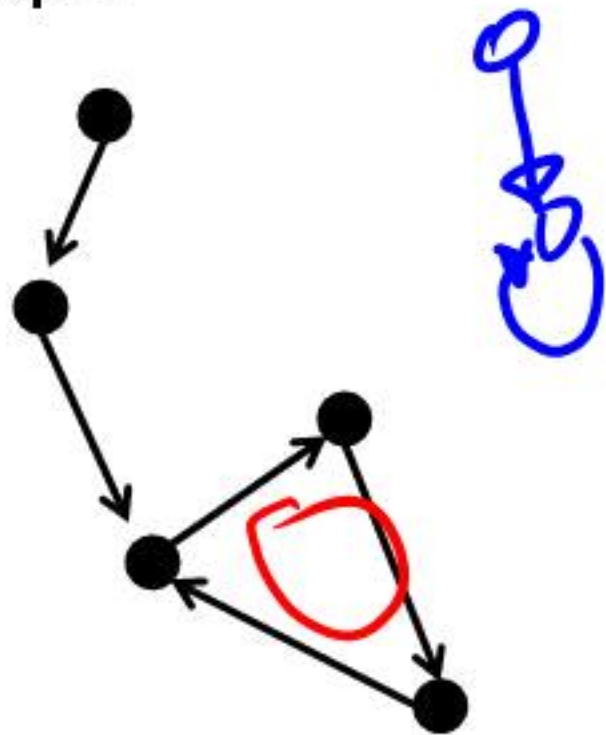
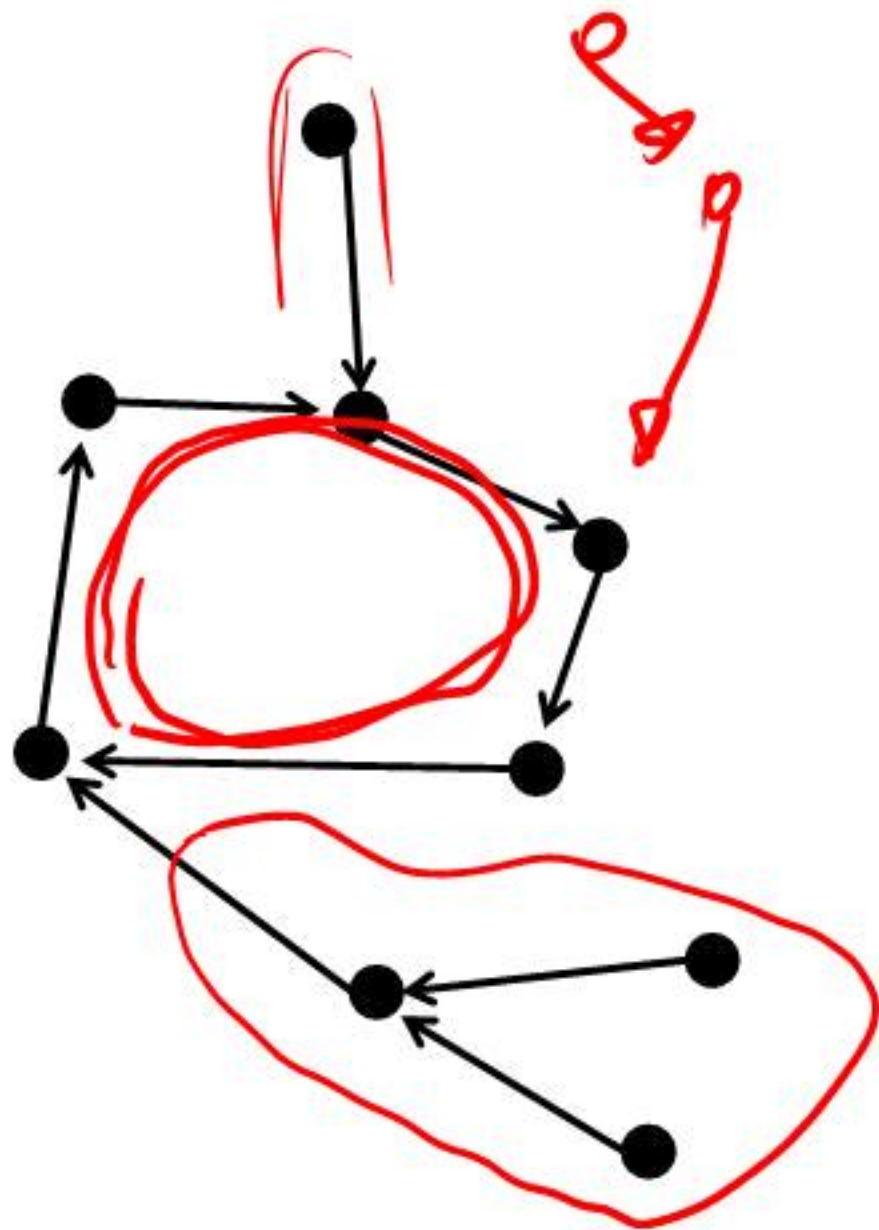
Random Graph models



Edge with
probab.
P.

$n = 100,000,000$

Random out degree one graph



Question:
What is the cycle structure as N gets large?
How many cycles?
What is the cycle length?



Greedy Algorithms