# notes for CS419

an introduction to graph algorithms

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primary references:

(Baase) computer algorithms
(McHugh) algorithmic graph theory
(Cormen/Leiserson/Rivest) intro to algorithms
(M. Atallah) alg'ms & theory of comp'n handbook

# part I: introduction

- graph theory
  - history and motivation
  - graphs and variants
  - representations
  - basic terminology
  - overview
- algorithmic techniques
  - recursion
  - divide and conquer
  - dynamic programming
  - the greedy method
  - backtracking
  - approximation
  - transformation
  - probabilistic methods
  - integer programming
  - tree-based search (\*)
  - geometric algorithms (\*)
- data structures
  - stacks and queues
  - priority queues
  - Fibonacci heaps

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# part II: basic graph algorithms

- traversal and connectivity
  - depth first search
  - breadth first search
  - biconnectivity
  - strong connnectivity
  - randomized connectivity
- minimum spanning trees
  - Boruvka, Dijkstra, Kruskal
- distance and shortest paths
  - unweighted single source
  - single source: Dijkstra
  - digraph single source digraph: Bellman-Ford
  - digraph all pairs: Floyd-Warshall
- acyclic digraphs
  - topological sorting
  - pert analysis
- tours
  - Eulerian tours
  - Hamiltonian cycles
- isomorphism and generation

# part III: advanced graph algorithms

- matchings and network flows
  - bipartite matching
  - matching (also randomized)
  - network flow
  - min cut (also randomized)
  - min cost flows
  - multi-commodity flows
- drawing and planarity
- colouring
  - vertex colouring
  - edge colouring
  - approximation algorithms
  - chordal graphs and lexicographic BFS
  - perfect graphs
- other hard problems and approximation algorithms

## part IV: selected topics

- treewidth
- dynamic graph algorithms
- parallel graph algorithms
- geometric graph algorithms

#### References

- [A] M. Atallah (ed.), Algorithms and Theory of Computation Handbook, CRC Press (1999) ISBN 0-8493-2649-4
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- [Go] M.C. Golumbic, Algorithmic Graph Theory and Perfect Graphs, Academic Press (1980) ISBN 0-12-289260-7 [specialized research primer, out of print]
- [M] J.A. McHugh, Algorithmic Graph Theory, Prentice Hall (1990) ISBN 0-13-023615-2 [out of print]
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- [We] D. West, Introduction to Graph Theory, Prentice Hall (1996) ISBN 0-13-227828-6 [grad level]
- [Wi] R. Wilson, Introduction to Graph Theory (4th ed.) Longman (1996) ISBN 0-582-24993-7 [inexpensive graph theory primer]

# 1-1: basic concepts terminology

• graph, vertex, edge, order (n), size (m), trivial graph, adjacent, endpoint, adjacency set (neighbourhood) ADJ(S), isolated,

- **degree**, min(G), max(G), degree sequence,
- Thm (deg sum)  $\sum d(v) = 2|E|$
- (spanning/induced) subgraph, union, edge sum, complement,
- path, trail, walk, cycle, circuit, length, connected, component, disconnected, cut-vertex, bridge, biconnected, block (bicomponent), vertex/edge connectivity, k-connected, k-edge-connected, distance, eccentricity, center, radius, diameter,
- **digraph** initial/terminal vertex, adjacent to/from, in/out-degree, di-path/cycle/..., semi-path/cycle/ldots, strongly connected, reachable, transitive closure
- graph variants multi-, weighted-, loop-, mixed graph

# special (di-)graphs

• acyclic, dag, tree, forest, complete, regular, hamlitonian, eulerian, (complete) bipartite, (complete) k-partite (multipartite),

# graphs as models

• assignment problem, data flow diagrams

# isomorphism

- invariants (degree sequence, cycle lengths, etc)
- complexity? n!, backtracking, automorphism,

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# 1-2: representations

• static adj. list, adj. matrix, edge list, incidence matrix

## 1-3: bipartite graphs

- matching, alternating paths, **Hall's thm**, **odd cycle thm**,
- linear recognition (bfs)

# 1-4: regular graphs

- m = rn, regular supergraph thm,
- König thm: G bipartite implies  $\chi'(G) = \Delta(G)$  (implies PRBM)
- 1-5: maximum matching algorithms later
- 1-6: planar graphs later
- 1-7: eulerian graphs later
- 1-8: hamiltonian graphs later

# Chapter 2: Algorithmic Techniques later

read on your own, make sure you recall all techniques

# Topic 1: graph traversal

- process of visiting each vertex in graph
- many graph algorithms require traversal
- most general version: traversal with list (below)
- most common variants: BFS, DFS

- FIFO list (queue): version\* of breadth first search

- LIFO list (stack): version\*\* of depth first search

- recursive version: depth first search

<sup>\*</sup> only need add never-enqueued neighbours

<sup>\*\*</sup> push neighbours in reverse order

```
graph_traverse(G)
  for each vertex v do
    visited[v] <- FALSE</pre>
  endfor
  for each vertex v do
    if not visited[v] then
      component_traverse(v) endif
  endfor
end_graph_traverse
component_traverse(s)
  list <- {s}
  while not empty(list) do
    remove_from(list,t)
    if not visited[t] then
      visit(t)
      for each neighbour w of t do (*)
        add_to(list,w) endfor
     endif
   endwhile
end_component_traverse
component_DFS(s)
  visit(s)
  for each neighbour t of s do
    if not visited[t] then
      DFS(t) endif
  endfor
end_component_DFS
```

#### • analysis:

- adjacency matrix representation

\* space 
$$\Theta(n^2)$$

\* time 
$$\Theta(n + \sum_{x} n = n + n^2)$$
  $\Theta(n^2)$ 

- adjacency list representation

\* space 
$$\Theta(n + \sum_{x} d(x) = n + 2m)$$
  $\Theta(n + m)$ 

\* time 
$$\Theta(n + \sum_{x} d(x) = n + 2m)$$
  $\Theta(n + m)$ 

# • applications:

<ul><li>connected components</li></ul>	any traversal
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#### ordered trees and traversal

- ordered trees
- recursion and procedure call tree
- ordered tree traversal (pre/in/post/level orders)

#### traversal trees

• for any graph, a component traversal defines an ordered tree, and the edges of the graph associated with the component can be partitioned based on the tree, as follows:

- breadth first (node level is graph distance)

\* tree edge to child

\* forward edge to child of previous node at same level

\* cross edge within level

- depth first

\* tree edge to child

\* back edge to ancestor (other than parent)

• dfs: pre-order tree traversal

• bfs: level-order tree traversal

• given a bfs/dfs tree, show the possible other edges

# biconnected components

•  $cut\ vertex$  removal increases number of components for some other x, y, on every x-y-path

• biconnected graph connected and no cut vertex

• bicomponent maximal biconnected subgraph

• rooted tree tree with one root vertex

• in a rooted tree,

-z child of v

-ancestor of v any vertex on v-to-root path

- proper ancestor ancestor other than vertex itself

-z descendant of v iff v ancestor of z

- proper descendant descendant other than vertex itself

- parent of v neighbour of v on v-to-root path

iff v parent of z

 $\bullet$  v cut vertex iff, w.r.t. dfs tree

root: more than one subtree

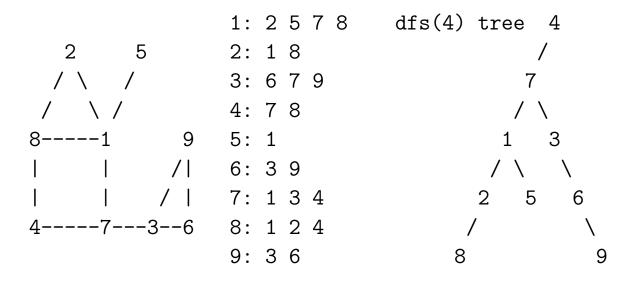
not root: child subtree has no back edge to proper v-ancestor

# finding bicomponents via depth first search

- ullet algorithm: for each v, for each child w, keep track of furthest back edge from w-subtree
- in component traversal, each v is encountered deg(v) times
- in dfs component traversal,
  - -1st w encounter: tree edge created from parent
  - last w encounter: subtree traversed, back up to parent
- how to implement above algorithm using dfs?
  - -1st encounter of child w of parent v
    - \* recurse from w
  - last encounter of w, just before backing up to v
    - \* check whether v cuts off w-subtree
  - maintain num, back, parent for each v
    - \* update back when backedge 1st encountered
    - \* update back when backing up
  - maintain edge stack
    - \* push edge when edge 1st encountered
    - \* pop edges when cutpoint discovered
  - Warning: this algorithm differs slightly from that of Baase (2nd ed'n) She pushes each back-edge once and each treeedge twice; edges are popped until the edge initially on top reappears; this avoids the need for maintaining a parent array.

```
bicomponents() (*version RBH99, differs from Baase*)
  empty stack; dfn <- 0
  for all v do
    parent[v] <- 0; num[v] <- 0; back[v] <- n+1
  for all v do
    if num[v]=0 then bidfs(v)
end_bicomponents
bidfs(v)
  inc(dfn); num[v] <- dfn; back[v] <- dfn</pre>
  for each neighbour w do
    if num[w]=0 then
                                           (*1st w encounter*)
      push [vw]; parent[w] <- v</pre>
                                                 (*tree edge*)
      bidfs(w)
      (* backup up from w to v*)
      if back[w] >= num[v] then (*v root or cuts off w? yes*)
        print 'new bicomponent'
        repeat: pop and print edge
        until popped edge is [vw]
                                   (*v root or cuts off w? no*)
      else
        back[v] <- min {back[v],back[w]}</pre>
      (*end backup from w to v*)
    elsif num[w]<num[v] and w<>parent[v] then (*back edge*)
      push [vw]; back[v] <- min {num[w],back[v]}</pre>
end bidfs
```

• example trace: execute bidfs(4) on the graph below, assuming no previous bidfs() calls (answer on the next page)



• correctness?

the truth is out there



• complexity?

(our/Baase versions)

- time: constant for each vertex/edge encounter  $\Theta(c_1 n + c_2 \sum_{v} deg(v) = c_1 n + 2c_2 m) = \Theta(n + m)$
- space: assume adjacency list representation
  - \* graph, arrays of size n, edge stack, runtime stack
  - \* edge stack: O(m) since each edge pushed once (our version) at most twice (Baase version)
  - \* runtime stack: O(n) since at most n constant size activation records

$$* \Theta(n+m) + \Theta(n) + O(m) + O(n) = \Theta(n+m)$$

```
back[1 2 3 4 5 6 7 8 9]
                          * * * 1 * * * * *
bidfs(4)
4} tree[47]
4} bidfs(7)
                          * * * 1 * * 2 * *
4} 7} tree[71]
4} 7} bidfs(1)
                          3 * * 1 * * 2 * *
4} 7} 1} tree[12]
4} 7} 1} bidfs(2)
                          3 4 * 1 * * 2 * *
4} 7} 1} 2} tree[28]
4} 7} 1} 2} bidfs(8)
                          3 4 * 1 * * 2 5 *
4} 7} 1} 2} 8} back[81]
                         3 4 * 1 * * 2 3 *
4} 7} 1} 2} 8} back[84] 3 4 * 1 * * 2 1 *
4} 7} 1} 2} backup noout 3 1 * 1 * * 2 1 *
4} 7} 1} backup noout
                          1 1 * 1 * * 2 1 *
4} 7} 1} tree[15]
                          1 1 * 1 6 * 2 1 *
4} 7} 1} bidfs(5)
4} 7} 1} backup
                          out[15]
4} 7} backup noout
                          1 1 * 1 6 * 1 1 *
4} 7} tree[73]
4} 7} bidfs(3)
                          1 1 7 1 6 * 1 1 *
4} 7} 3} tree[36]
4} 7} 3} bidfs(6)
                          1 1 7 1 6 8 1 1 *
4} 7} 3} 6} tree[69]
4} 7} 3} 6} bidfs(9)
                          1 1 7 1 6 8 1 1 9
4} 7} 3} 6} 9} back[93]
                          1 1 7 1 6 8 1 1 7
4} 7} 3} 6} backup noout 1 1 7 1 6 7 1 1 7
4} 7} 3} backup
                          out [93] [69] [36]
4} 7} backup
                          out [73]
                          out [84] [81] [28] [12] [71] [47]
4} backup
```

# lexicographic bfs

- in ordinary bfs, the order in which neighbours of a vertex are placed on the queue is arbitrary and so gives no information on the structure of the graph
- in lexbfs, the queue of vertices is replaced with a queue of vertex subsets; at all times, the subsets partition the collective neighbourhood of vertices already visited;
- the label of each queue element is its neighbourhood of already visited vertices; vertices are labelled in decreasing order as they are visited; queue elements are maintained in lexicographically decreasing order, namely the 1st item in the queue is lexicographically largest
- each iteration, an arbitrary vertex is removed from the first set of the queue, and the queue partition is refined
- $lexicographic \ order$  for finite integer subsets  $S_j$  and  $S_k$  is defined as follows:

sets are lex'ly equal if and only if they are equal set  $S_j$  is lexly less than  $S_k$  iff the largest integer which is in exactly one of  $S_j$ ,  $S_k$  is in  $S_k$ .

• e.g.:  $\{1\} < \{1,2,4\} < \{3,4\}$ 

```
lexbfs (store lexicographic labels of each vertex)
for each vertex v do
  lexLabel[v] <- { }</pre>
  lexorder[v] <- 0 endfor</pre>
for p <- n downto 1 do (* p is priority *)</pre>
  v <- any unvisited vertex with lex'ly largest label
  lexorder[v] <- n+1-p (*put v next in lexbfs order*)</pre>
  for each unvisited nbr w of v do
    add p to lexLabel[w] endfor endfor
lexbfs (refine a queue of sets in lexicographic order)
for each vertex v do
  lexorder[v] <- 0 endfor</pre>
Q <- one set containing all vertices, status old
for j <- 1 to n do (*don't need p*)
  v <- remove any vertex from first set in Q
  lexorder[v] <- j</pre>
  for each unvisited nbr x of v do
    X <- set in Q containing x
    W <- set in Q preceding X
    remove x from X
    if W is not new then
      insert in Q a new set {x} preceding X
    else
      add x to W endif endfor
  for each new set do
    change its status to old endfor endfor
```

# lexbfs and chordal graphs

- chordal graph (a.k.a. triangulated): no induced  $C_{k\geq 4}$
- simplicial vertex: neighbourhood induces a clique
- simplicial (elimination) ordering:  $(v_1, \ldots, v_n)$  with  $v_j$  simplicial in  $G[\{v_1, \ldots, v_j\}]$
- $\bullet$  chordal  $\iff$  graph has simplicial ordering
- chordal iff lexbfs order is simplicial order
- can check whether order simp'l in linear time, so linear chordal graph recognition
- simplicial orders also lead to optimal chordal graph colouring/clique size/generation

# other chordal graph results

• graph *perfect*: for each vertex induced subgraph, chromatic number equals size of largest clique

- chordal graphs are perfect
- complements of chordal graphs (co-chordal) graphs are perfect
- chordal graph  $\iff$  intersection graph of subtrees of a tree
- chordal graphs have O(n) maximal cliques

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# $\Theta(n+m)$ time/space simp'l order recognition

A[u]: list with rep'ns; vertices which must see u

```
for all v do
   A[v] <- { } endfor
for j <- n downto 2 do
   v <- j'th vertex in the lexbfs order
   U <- nbrs of v which precede v in lexbfs order
   if U non-empty then
        u <- vertex in U with highest position in lexbfs order
        add U - {u} to A[u] endif
   if A[v] - nbr[v] <> { } then return NO endif endfor
return YES
```

- correctness: u is simplicial in  $G[v_1, \ldots, u]$
- complexity:
  - line -2 can be done in O(|A[v]| + |nbr[v]|) time/space using an array of size n initialized to all zero, and reset to all zero after each test
  - $-\sum |A[u]| < \sum |nbr[v]|$
  - total time/space  $\Theta(|V| + \sum |nbr[v]| + \sum |A[v]|) = \Theta(n+m)$

# chordal graph algorithms

- linear recognition
  - lexBFS (queue of sets)
  - verify simp'l order
- max'l cliques
  - each has form  $\{v\} \cup \operatorname{Prev}(v)$
  - at most n
  - can output in linear time
- linear colour/clique
  - simp'l order
  - greedy: in forward order, each col(v) < smallest colour not in col(Prev(v))
- linear clique cover/stable set
  - simp'l order
  - greedy: in reverse order, each clique  $< -\{v\} \cup \text{Prev}(v)$

Prev(v): all previous, in simp'l order, neighbours of v

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# Topic: MST algorithms

- Boruvka, Prim/Dijkstra, Kruskal algorithms
- P/D implementations
  - naive  $O(n^3)$
  - store best edge for each fringe vertex  $O(n^2)$
  - as previous, use priority queue (heap imp'n)  $O(m \lg n)$
  - as previous, (Fibonacci heap imp'n)  $O(m+n\lg n)$  [did not cover]
- Kruskal implementations
  - sort all edges first (not so good: why?)
  - priority queue (heap imp'n)  $O(m \lg m)$
  - sideline: union/find trees
  - -n w-unions, m finds take  $O(m \lg n)$
  - -n w-unions, m c-finds take  $O(m \lg^* n)$
- REFERENCES: my 204 notes (on the web)

## randomized graph algorithms

• r.a.: has some randomized step: among a collection of items, pick one randomly

- classical cs r.a.: qsort (WC  $\Theta(n^2)$  AC  $\Theta(n \lg n)$ )
- mot'n: if perform repeatedly, guarantees AC performance
- classical graph r.a.: min cut

def'n cut (of connected graph): edge set, removal disconnects graph

• deterministic alg'ms use network flow

input: connected graph with n vertices, edges labelled output: a minimal cut (not nec. minimum) repeat n-2 times:

- e <- \*random\* edge of remaining loopless multigraph
  contract e (remove loops; leave parallel edges)
  cut <- edges between two remaining vertices</pre>
- references: Randomized Algorithms, by Motwani/Raghavan

# graph drawing

- active research area
- graph *drawing*: embedding of a graph on some surface, usually the plane, s.t.
  - vertex  $\leftrightarrow$  point
  - edge  $\leftrightarrow$  continuous line joining two vertex points
  - lines of adjacent edges intersect only at the point of incidence
  - lines of non-adjacent edges intersect at most once (and if so, at a crossing point, and not at a 'touching' or tangent point)
- ullet crossing number  $\nu(G)$ : over all drawings, minimum number of line crossings
- planar graph:  $\nu(G) = 0$
- $\nu(G) \leq k$ ? NP-complete (for fixed k?)
- G planar? polynomial (linear time)

## graph drawing: math background

- closed continuous line partitions plane (inside/outside)
- $K_{3,3}$  and  $K_5$  non-planar
- homeomorphic graphs: can be obtained from same graph by inserting vertices of degree two into edges
- Kuratowski's thm: planar iff no subgraph homeomorphic to  $K_5$  or  $K_{3,3}$
- related thm: planar iff no subgraph contractible to  $K_5$  or  $K_{3,3}$
- face: (roughly) region of planar embedding
- Euler's thm: for planar embedding of connected graph, n+f=m+2
- Corollary: for planar graph,  $m \leq 3n 6$
- $thickness\ t(G)$ : min number of planar graphs which can be superimposed to form G
- $t(G) \ge \frac{m}{3n-6}$
- planar embedding of connected planar graph G has geometric dual graph  $G^*$ :
  - vertex of  $G^* \leftrightarrow$  face of G
  - edge of  $G^* \leftrightarrow \text{join}$  two faces of G which share an edge of G
  - this concept goes back to Euclid, and is important in computational geometry

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## a planarity testing algorithm

- simple  $O(n^3)$  divide and conquer algorithm [Graph Drawing, di Battista et. al.] p74
- Hopcroft and Tarjan implementation: O(n)

algorithm is\_planar(G,C)

Input: biconnected G with m  $\leq$  3n-6 and separating cycle C Output: yes/no

- 1. find pieces of G with respect to C
- 2. for each piece P of G that is not a path
  - (a) P' <- graph obtained by adding P to C
  - (b) C' <- cycle of P' obtained from C by replacing the portion of C btwn 2 consec. attachments with a path of P btwn them
  - (c) if not is\_planar(P',C') then return NO
- 3. compute interlacement graph I of pieces
- 4. return is\_bipartite(I)