

# Directed Graphs

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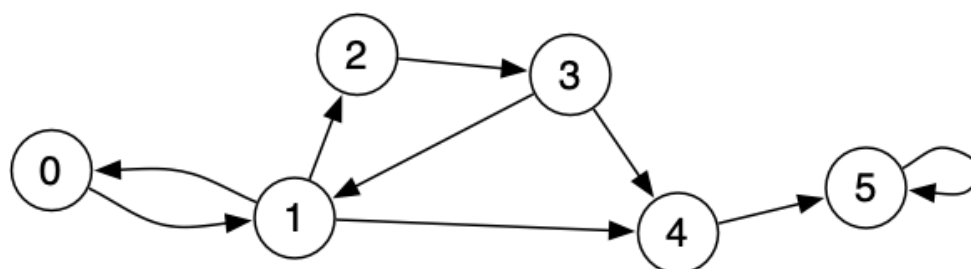
- Directed Graphs (Digraphs)
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## ❖ Directed Graphs (Digraphs)

Reminder: **directed graphs** are ...

- graphs with  $V$  vertices,  $E$  edges  $(v,w)$
- edge  $(v,w)$  has **source**  $v$  and **destination**  $w$
- unlike undirected graphs,  $v \rightarrow w \neq w \rightarrow v$

Example digraph:



## ❖ Digraph Applications

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Potential application areas:

Domain	Vertex	Edge
Web	web page	hyperlink
scheduling	task	precedence
chess	board position	legal move
science	journal article	citation
dynamic data	malloc'd object	pointer
programs	function	function call
<b>make</b>	file	dependency

## ❖ ... Digraph Applications

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Problems to solve on digraphs:

- is there a directed path from  $s$  to  $t$ ? (transitive closure)
- what is the shortest path from  $s$  to  $t$ ? (shortest path)
- are all vertices mutually reachable? (strong connectivity)
- how to organise a set of tasks? (topological sort)
- which web pages are "important"? (PageRank)
- how to build a web crawler? (graph traversal)

## ❖ Transitive Closure

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Problem: computing **reachability** (**reachable**( $G, s, t$ ))

Given a digraph  $G$  it is potentially useful to know

- is vertex  $t$  reachable from vertex  $s$ ?

Example applications:

- can I complete a schedule from the current state?
- is a malloc'd object being referenced by any pointer?

## ❖ ... Transitive Closure

One possibility to implement a reachability check:

- use **hasPath(G,s,t)** (itself implemented by DFS or BFS algorithm)
- feasible only if  $reachable(G,s,t)$  is an infrequent operation

What about applications that frequently check reachability?

Would be very convenient/efficient to have:

$$reachable(G,s,t) \equiv G.tc[s][t]$$

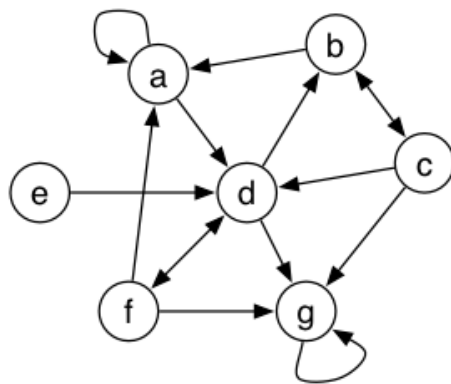
$tc[][]$  is called the **transitive closure** matrix

- $tc[s][t]$  is 1 if there is a path from  $s$  to  $t$ , 0 otherwise

Of course, if  $V$  is large, then this may not be feasible either.

## ❖ ... Transitive Closure

The **tc**[ ][ ] matrix shows all directed paths in the graph



	a	b	c	d	e	f	g
a	1	0	0	1	0	0	0
b	1	0	1	0	0	0	0
c	0	1	0	1	0	0	1
d	0	1	0	0	0	1	1
e	0	0	0	1	0	0	0
f	1	0	0	1	0	0	1
g	0	0	0	0	0	0	1

*adjacency matrix*

	a	b	c	d	e	f	g
a	1	1	1	1	0	1	1
b	1	1	1	1	0	1	1
c	1	1	1	1	0	1	1
d	1	1	1	1	0	1	1
e	1	1	1	1	0	1	1
f	1	1	1	1	0	1	1
g	0	0	0	0	0	0	1

*reachability matrix*

Question: how to build **tc**[ ][ ] from **edges**[ ][ ]?

## ❖ ... Transitive Closure

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**Goal:** produce a matrix of reachability values

Observations:

- $\forall s, t \in \text{vertices}(G): (s, t) \in \text{edges}(G) \Rightarrow tc[s][t] = 1$
- $\forall i, s, t \in \text{vertices}(G): (s, i) \in \text{edges}(G) \wedge (i, t) \in \text{edges}(G) \Rightarrow tc[s][t] = 1$

In other words

- **$tc[s][t] = 1$**  if there is an edge from  $s$  to  $t$  (path of length 1)
- **$tc[s][t] = 1$**  if there is a path from  $s$  to  $t$  of length 2 ( $s \rightarrow i \rightarrow t$ )



## ❖ ... Transitive Closure

Extending the above observations gives ...

An algorithm to convert **edges** into a *tc*

```
makeTC(G):  
|  tc[][] = edges[][]  
|  for all i ∈ vertices(G) do  
|  |  for all s ∈ vertices(G) do  
|  |  |  for all t ∈ vertices(G) do  
|  |  |  |  if tc[s][i]=1 ∧ tc[i][t]=1 then  
|  |  |  |  |  tc[s][t]=1  
|  |  |  |  end if  
|  |  |  end for  
|  |  end for  
|  end for
```

This is known as [Warshall's algorithm](#)

## ❖ ... Transitive Closure

How it works ...

After copying `edges[][]`, `tc[s][t]` is 1 if  $s \rightarrow t$  exists

After first iteration ( $i=0$ ), `tc[s][t]` is 1 if

- either  $s \rightarrow t$  exists or  $s \rightarrow 0 \rightarrow t$  exists

After second iteration ( $i=1$ ), `tc[s][t]` is 1 if any of

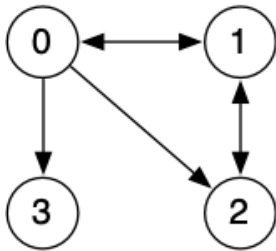
- $s \rightarrow t$  or  $s \rightarrow 0 \rightarrow t$  or  $s \rightarrow 1 \rightarrow t$  or  $s \rightarrow 0 \rightarrow 1 \rightarrow t$  or  $s \rightarrow 1 \rightarrow 0 \rightarrow t$

After the  $V^{th}$  iteration, `tc[s][t]` is 1 if

- there is a directed path in the graph from  $s$  to  $t$

## ❖ ... Transitive Closure

Tracing Warshall's algorithm on a simple graph:



Graph

	[0]	[1]	[2]	[3]
[0]	0	1	1	1
[1]	1	0	1	0
[2]	0	1	0	0
[3]	0	0	0	0

Initially

	[0]	[1]	[2]	[3]
[0]	0	1	1	1
[1]	1	1	1	1
[2]	0	1	0	0
[3]	0	0	0	0

After first iteration

	[0]	[1]	[2]	[3]
[0]	1	1	1	1
[1]	1	0	1	0
[2]	1	1	1	1
[3]	0	0	0	0

After second iteration

No change  
on any  
following  
iterations

## ❖ ... Transitive Closure

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Cost analysis:

- storage: additional  $V^2$  items (but each item may be 1 bit)
- computation of transitive closure:  $V^3$
- computation of **reachable()**:  $O(1)$  after generating **tc[ ][ ]**

Amortisation: need many calls to **reachable()** to justify setup cost

Alternative: use DFS in each call to **reachable()**

Cost analysis:

- storage: cost of Stack and Set during DFS calculation
- computation of **reachable()**:  $O(V^2)$  (for adjacency matrix)

## ❖ Digraph Traversal

Same algorithms as for undirected graphs:

```
depthFirst(G,v):
```

```
|  mark  $v$  as visited  
|  for each  $(v,w) \in \text{edges}(G)$  do  
|  |  if  $w$  has not been visited then  
|  |  |  depthFirst( $w$ )  
|  |  end if  
|  end for
```

```
breadthFirst(G,v):
```

```
|  enqueue  $v$   
|  while queue not empty do  
|  |   $curr = \text{dequeue}$   
|  |  if  $curr$  not already visited then  
|  |  |  mark  $curr$  as visited  
|  |  |  enqueue each  $w$  where  $(curr,w) \in \text{edges}(G)$   
|  |  end if  
|  end while
```

## ❖ Example: Web Crawling

**Goal:** visit every page on the web

**Solution:** breadth-first search with "implicit" graph

```
webCrawl(startingURL):
|   mark startingURL as alreadySeen
|   enqueue(Q,startingURL)
|   while not isEmpty(Q) do
|       currPage=dequeue(Q)
|       visit currPage
|       for each hyperLink on currPage do
|           if hyperLink not alreadySeen then
|               mark hyperLink as alreadySeen
|               enqueue(Q,hyperLink)
|           end if
|       end for
|   end while
```

**visit** scans page and collects e.g. keywords and links

## ❖ PageRank

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**Goal:** determine which Web pages are "important"

**Approach:** ignore page contents; focus on hyperlinks

- treat Web as graph: page = vertex, hyperlink = di-edge
- pages with many incoming hyperlinks are important
- need to computing "incoming degree" for vertices

Problem: the Web is a *very* large graph

- approx.  $10^{10}$  pages,  $10^{11}$  hyperlinks

## ❖ ... PageRank

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Assume for the moment that we could build a graph ...

Naive PageRank algorithm:

```
PageRank(myPage):  
|   rank=0  
|   for each page in the Web do  
|   |   if linkExists(page,myPage) then  
|   |       rank=rank+1  
|   |   end if  
|   end for
```

Note: requires **inbound** link check (normally, we check outbound)



## ❖ ... PageRank

$V$  = # pages in Web,  $E$  = # hyperlinks in Web

Costs for computing PageRank for each representation:

Representation	linkExists(v,w)	Cost
Adjacency <b>matrix</b>	<code>edge[v][w]</code>	$1$
Adjacency <b>lists</b>	<code>inLL(list[v],w)</code>	$\cong E/V$

Not feasible ...

- adjacency matrix ...  $V \cong 10^{10} \Rightarrow$  matrix has  $10^{20}$  cells
- adjacency list ...  $V$  lists, each with  $\cong 10$  hyperlinks  $\Rightarrow 10^{11}$  list nodes

So how to really do it?

## ❖ ... PageRank

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The random web surfer strategy.

Each page typically has many outbound hyperlinks ...

- choose one at random, without a **visited[]** check
- follow link and repeat above process on destination page

If no visited check, need a way to (mostly) avoid loops

Important property of this strategy

- if we randomly follow links in the web ...
- ... more likely to re-discover pages with many inbound links

## ❖ ... PageRank

Random web surfer algorithm ...

```
curr=random page, prev=null
for a long time do
|   if curr not in array rank[] then
|       rank[curr]=0
|   end if
|   rank[curr]=rank[curr]+1
|   if random(0,100) < 85 then // with 85% chance ...
|       prev=curr                // ... keep crawling
|       curr=choose hyperlink from curr
|   else
|       curr=random page // avoid getting stuck
|       prev=null
|   end if
end for
```

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## ❖ ... PageRank

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Above is a very rough approximation to reality.

If you want the details ...

- The Anatomy of a Large-Scale Hypertextual Web Search Engine  
<https://research.google/pubs/pub334/>
- The PageRank Citation Ranking: Bringing Order to the Web  
<http://ilpubs.stanford.edu:8090/422/1/1999-66.pdf>

And the background ...

- Authoritative Sources in a Hyperlinked Environment  
<https://dl.acm.org/doi/pdf/10.1145/324133.324140>

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