Spatial Networks

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Outline

- 1. Motivation, and use cases
- 2. Example spatial networks
- 3. Conceptual model
- 4. Need for SQL extensions
- 5. CONNECT statement
- 6. RECURSIVE statement
- 7. Storage and data structures
- 8. Algorithms for connectivity query
- 9. Algorithms for shortest path

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Navigation Systems

Historical

- Navigation is a core human activity for ages!
- Trade-routes, Routes for Armed-Forces

Recent Consumer Platforms

- Devices: Phone Apps, In-vehicle, "GPS", ...
- WWW: Google Maps, MapQuest, ...

Services

- Display map around current location
- Compute the shortest route to a destination
- Help drivers follow selected route





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Location Based Services

Location: Where am I ?

- Geo-code: Place Name (or Street Address) \rightarrow <latitude, longitude>
- Reverse Geo-code: <latitude, longitude> → Place Name

• **Directory:** What is around me?

- Where is the nearest Clinic? Restaurant? Taxi?
- List all Banks within 1 mile.
- Routes: How do I get there?
 - What is the shortest path to get there?

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Limitations of Spatial Querying

OGIS Simple Feature Types

- Supports Geometry (e.g., Points, LineStrings, Polygons, ...)
- However, lack Graphs data type, shortest_path operator

Traditional SQL

- Supports select, project, join, statistics
- Lacked transitive closure, e.g., network analysis (next slide)
- SQL3 added recursion & transitive closure

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Spatial Network Analysis

- <u>Route (</u> A start-point, Destination(s))
 - What is the shortest path to get there?
 - What is the shortest path to cover a set of destinations?
- <u>Allocation (A set of service centers, A set of customers)</u>
 - Assign customers to nearest service centers
 - Map service area for each service center
- <u>Site Selection (A set of customers, Number of new service centers)</u>
 - What are best locations for new service centers ?

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Spatial Network Query Example



- 1. Find shortest path from a start-point to a destination
- 2. Find nearest hospital by driving distance
- 3. Find shortest route to deliver packages to a set of homes
- 4. Allocate customers to nearest service center

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Railway Network & Queries



- Find the number of stops on the Yellow West (YW) route.
- List all stops which can be reached from Downtown Berkeley (2)
- List the routes numbers that connect Downtown Berkeley (2) & Daly City (5)
 - Find the last stop on the Blue West (BW) route

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River Network & Queries



- 1. List the names of all direct and indirect tributaries of Mississippi river
- 2. List the direct tributaries of Colorado
- 3. Which rivers could be affected if there is a spill in North Platte river

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Spatial Networks: Three Examples



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Data Models of Spatial Networks

1. Conceptual Model

- Information Model: Entity Relationship Diagrams
- Mathematical Model: Graphs
- 2. Logical Data Model
 - Abstract Data types
 - Custom Statements in SQL
- 3. Physical Data Model
 - Storage-Structures
 - Algorithms for common operations

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Modeling Roadmaps



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An Entity Relationship Diagram



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Graph Models

- A Simple Mathematical Model
 - A graph G = (V,E)
 - V = a finite set of vertices
 - E = a set of edges model a binary relationship between vertices



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A Graph Model of River Network

- Nodes = rivers
- Edges = A river falls into another river



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Pros and Cons of Graph Models

- Strength
 - Well developed mathematics for reasoning
 - Rich set of computational algorithms and data-structures
- Weakness
 - Models only one binary relationship
- Implications



- A. Difficult to model multiple relationships, e.g., connect, turn
- B. Multiple graph models possible for a spatial network

Modeling Turns in Roadmaps

• Approach 1: Model turns as a set of connects





- Approach 2: Use hyper-edges (and hyper-graphs)
- Approach 3: Annotate graph node with turn information

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Alternative Graph Models for Roadmaps

- Choice 1:
 - Nodes = road-intersections
 - Edge (A, B) = road-segment connects adjacent road-intersections A, B
- Choice 2:
 - Nodes = (directed) road-segments
 - Edge (A,B) = turn from road-segment A to road-segment B
- Choice 3:
 - Nodes = roads
 - Edge(A,B) = road A intersects_with road B

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Quiz

Which of the following are usually not captured in common graph models of roadmaps?

- a) Turn restrictions (e.g., no U turn)
- b) Road intersections
- c) Road segments
- d) All of the above

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Data Models of Spatial Networks

- 1. Conceptual Model: Entity Relationship Diagrams, Graphs
- 2. Logical Data Model & Query Languages
 - Abstract Data types
 - Custom Statements in SQL
- 3. Physical Data Model: Storage-Structures, Algorithms

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Transitive Closure

- Consider a graph G = (V, E)
- Transitive closure(G) = G* = (V*, E*), where
 - V* = V
 - (A, B) in E* if and only if there is a path from A to B in G.



- Example
 - G has 5 nodes and 5 edges
 - G* has 5 nodes and 9 edges
 - Note edge (1,4) in G* for
 - path (1, 2, 3, 4) in G.



(a) Graph G

(c) Transitive closure (G) = Graph G

(d) Transitive closure in relation form

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- Example
 - G has 5 nodes and 5 edges
 - G* has 5 nodes and 9 edges
 - Note edge (1,4) in G* for
 - path (1, 2, 3, 4) in G.



R		
SOURCE	DEST	
1	2	
1	5	
2	3	
3	4	
5	3	

(a) Graph G

(c) Transitive closure (G) = Graph G

(b) Relation form

(d) Transitive closure in relation form

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- Example
 - G has 5 nodes and 5 edges
 - G* has 5 nodes and 9 edges
 - Note edge (1,4) in G* for
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R		
SOURCE	DEST	
1	2	
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2	3	
3	4	
5	3	

(a) Graph G



(c) Transitive closure (G) = Graph G

(b) Relation form

UNIVERSITY OF MINNESOTA Driven to Discoversm (d) Transitive closure in relation form

- Example
 - G has 5 nodes and 5 edges
 - G* has 5 nodes and 9 edges
 - Note edge (1,4) in G* for
 - path (1, 2, 3, 4) in G.



(a)	Graph G	



R		
SOURCE	DEST	
1	2	
1	5	
2	3	
3	4	
5	3	

(b) Relation form

х	
SOUR	CE DEST
1	2
1	5
2	3
3	4
5	3
1	3
2	4
5	4
1	4

(c) Transitive closure (G) = Graph G

(d) Transitive closure in relation form

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Limitations of Original SQL

- Recall Relation algebra based languages
 - Ex. Original SQL
 - Can not compute transitive closure, e.g., shortest path!

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Supporting Graphs in SQL

- Abstract Data Type (user defined)
 - SQL3
 - May include shortest path operation!
- Custom Statements
 - SQL2 CONNECT clause in SELECT statement
 - For directed acyclic graphs, e.g. hierarchies
 - SQL3 WITH RECURSIVE statement
 - Transitive closure on general graphs

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Querying Graphs: Overview

- Relational Algebra
 - Can not express transitive closure queries
- Two ways to extend SQL to support graphs
 - 1. Abstract Data Types
 - 2. Custom Statements
 - SQL2 CONNECT BY clause(s) in SELECT statement
 - SQL3 WITH RECURSIVE statement

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CONNECT BY : Input, Output



- (b) Start Node S, e.g., Missouri
- (c) Travel Direction
- Output: Transitive closure of G
 - Ex. Predecessors of S = Missouri
 - Ex. Successors of S = Missouri



 (a) Mississippi network (Y1 = Bighorn river, Y2 = Power river, P1 = Sweet water River, P2 = Big Thompson river)

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Directed Edges: Tabular Representation



 (a) Mississippi network (Y1 = Bighorn river, Y2 = Power river, P1 = Sweet water River, P2 = Big Thompson river)

Table: Falls_Into

Source	Dest
P1	Platte
P2	Platte
Y1	Yellowstone
Y2	Yellowstone
Platte	Missouri
Yellowstone	Missouri
Missouri	Mississippi
Ohio	Mississippi
Red	Mississippi
Arkansas	Mississippi

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CONNECT BY- PRIOR - START WITH

SELECT source FROM Falls_Into CONNECT BY PRIOR source = dest START WITH dest ="Missouri"

Table: Falls_Into

Source	Dest
P1	Platte
P2	Platte
Y1	Yellowstone
Y2	Yellowstone
Platte	Missouri
Yellowstone	Missouri
Missouri	Mississippi
Ohio	Mississippi
Red	Mississippi
Arkansas	Mississippi

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CONNECT BY- PRIOR - START WITH

SELECT source FROM Falls_Into CONNECT BY PRIOR source = dest START WITH dest ="Missouri"

Q? What does CONNECT BY ... PRIOR specify?

- Direction of travel
- Example: From Dest to Source
- Alternative: From Source to Dest

Table: Falls_Into

Source	Dest
P1	Platte
P2	Platte
Y1	Yellowstone
Y2	Yellowstone
Platte	Missouri
Yellowstone	Missouri
Missouri	Mississippi
Ohio	Mississippi
Red	Mississippi
Arkansas	Mississippi

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CONNECT BY- PRIOR - START WITH

<u>Choice 1:</u> Travel from Dest to Source Ex. List direct & indirect tributaries of Missouri.

SELECT source FROM Falls_Into CONNECT BY PRIOR source = dest START WITH dest ="Missouri"



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CONNECT BY- PRIOR - START WITH

<u>Choice 1:</u> Travel from Dest to Source Ex. List direct & indirect tributaries of Missouri.

SELECT source FROM Falls_Into CONNECT BY PRIOR source = dest START WITH dest ="Missouri"

<u>Choice 2:</u> Travel from Source to Dest Ex. Which rivers are affected by spill in Missouri?

> SELECT dest FROM Falls_Into CONNECT BY source = PRIOR dest START WITH source ="Missouri"

Table: Falls_Into

Source	Dest
P1	Platte
P2	Platte
Y1	Yellowstone
Y2	Yellowstone
Platte	Missouri
Yellowstone	Missouri
Missouri	Mississippi
Ohio	Mississippi
Red	Mississippi
Arkansas	Mississippi

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SELECT source FROM Falls Into			Table	e: Falls_Int
CONNECT BY PRIOR source = dest			Source	Dest
			P1	Platte
2. Add rows from Falls_Into where (R Falls Into.dest) to create table <root< td=""><td>oot_Result.so</td><td>urce = ren></td><td>P2</td><td>Platte</td></root<>	oot_Result.so	urce = ren>	P2	Platte
<u> </u>			Y1	Yellowstone
			Y2	Yellowstone
			Platte	Missouri
	Poot Pos	116	Yellowstone	Missouri
			Missouri	Mississippi
	Source	Dest	Ohio	Mississippi
		IVIISSOUII	Red	Mississippi
	rellowstone	IVIISSOUII	Arkansas	Mississippi

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SELECT source				Table	: Falls_Into	0
CONNECT BY PRIOR source = dest				ce	Dest	
			P1		Platte	
2. Add rows from Falls_Into where (R Falls Into.dest) to create table <root< p=""></root<>	P2		Platte			
			Y1		Yellowstone	
			Y2		Yellowstone	
			Platte		Missouri	
	Root Res	114	Yellow	stone	Missouri	
	Root Resi		Yellow Missou	stone uri	Missouri Mississippi	
	Root Rest	ult Dest	Yellow Missou Ohio	stone uri	Missouri Mississippi Mississippi	
	Root Resu Source	Dest Missouri	Yellow Missou Ohio Red	stone uri	Missouri Mississippi Mississippi Mississippi	
	Root Resu Source Platte Yellowstone	Dest Missouri Missouri	Yellow Missou Ohio Red Arkans	stone uri sas	Missouri Mississippi Mississippi Mississippi Mississippi	

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SELECT source FROM Falls	Table: Falls_Into							
CONNECT BY PRIOR source = dest						ce	Dest	
START WITH					P1		Platte	
2. Add rows from Falls_Into where (Root_Result.source = Falls_Into.dest) to create table <root +="" 1="" children="" level=""></root>							Platte	
Root + 1 level children							Yellowstone	
	Deet	1			Y2		Yellowstone	
Source	Dest				Platte		Missouri	
P1	Platte				Yellov	vstone	Missouri	
P2	Platte		Seed Res		Misso	ouri	Mississippi	
Y1	Yellowstone		Source	Dest	Ohio		Mississippi	
Y2	Yellowstone		Platte	Missouri	Red		Mississippi	
Platte	Missouri		Yellowstone	Missouri	Arkan	sas	Mississippi	
Yellowstone	Missouri							

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SELECT source FROM Falls Into				Table: Falls_Inte		
CONNECT BY PRIOR SOL	Source	Dest				
	oun			P1	Platte	
Add rows from Falls_Inte Falls Into.dest) to create ta	o whe able <	ere (Root+1levo <root +="" 2="" level<="" td=""><td>:hild.source = children></td><td>P2</td><td>Platte</td></root>	:hild.source = children>	P2	Platte	
			vel children	Y1	Yellowstone	
				Y2	Yellowstone	
No new rows		Source	Dest	Platte	Missouri	
can be added		P1	Platte	Yellowstone	Missouri	
into the result!		P2	Platte	Missouri Ohio	Mississippi	
		Y1	Yellowstone		Mississippi	
		Y2	Yellowstone	Red	Mississippi	
		Platte	Missouri	Arkansas	Mississippi	
		Yellowstone	Missouri			
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SELECT source FROM Falls Into			Table	e: Falls_Into
CONNECT BY PRIOR source	Source	Dest		
	P1	Platte		
2. Add rows from Falls_Into wh Falls Into.dest) to create table	iere (Root+1lchi <root +="" 2="" level<="" td=""><td>ld.source = children></td><td>P2</td><td>Platte</td></root>	ld.source = children>	P2	Platte
· · · · ·	Einal answ	VOL	Y1	Yellowstone
			Y2	Yellowstone
The query	Source		Platte	Missouri
returned all	P1		Yellowstone	Missouri
predecessors of	P2		Missouri	Mississippi
Missouri!	Y1		Ohio	Mississippi
	Y2		Red	Mississippi
	Platte		Arkansas	Mississippi
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Quiz

- Which of the following is false about CONNECT BY clause?
- a) It is only able to output predecessors, but not successors, of the start node
- b) It is able to output transitive closure of a directed graph
- c) It usually works with PRIOR and START WITH keywords
- d) None of the above

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 - SQL3 WITH RECURSIVE statement

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WITH RECURSIVE: Input, Output

• Input:

- (a) Edges of a directed graph G
- (b) Sub-queries to
 - Initialize results
 - Recursively grow results
 - Additional constraints





- Output: Transitive closure of G
 - Ex. Predecessors of a node
 - Ex. Successors of a node

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Syntax of WITH RECURSIVE Statement



AS (SELECT source, dest FROM R)

UNION

(SELECT R.source, X.dest

FROM R, X

WHERE R.dest=X.source)

Description of Result Table Initialization Query



Recursive Query to grow result

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Example Input and Output

WITH RECURSIVE X(source,dest) AS (SELECT source,dest FROM R) UNION (SELECT R.source, X.dest FROM R, X WHERE R.dest=X.source)



R						
SOURCE	DEST					
1	2					
1	5					
2	3					
3	4					
5	3					

(a) Graph G



(c) Transitive closure (G) = Graph G



x						
SOURCE	DEST					
1	2					
1	5					
2	3					
3	4					
5	3					
1	3					
2	4					
5	4					
1	4					

(d) Transitive closure in relation form

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5

3

4

3

- Initialize X by ۲ (SELECT source, dest FROM R)
- Recursively grow X by • (SELECT R.source, X.dest FROM R, X WHERE R.dest=X.source)
- Infer X(a,c) from R(a,b),X(b,c)

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- Initialize X by (SELECT source,dest FROM R)
- Recursively grow X by (SELECT R.source, X.dest FROM R, X WHERE R.dest=X.source)
- Infer X(a,c) from R(a,b),X(b,c)

Infer X(1,3) from R(1,2),X(2,3) Infer X(2,4) from R(2,3),X(3,4)

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- Initialize X by (SELECT source,dest FROM R)
- Recursively grow X by (SELECT R.source, X.dest FROM R, X WHERE R.dest=X.source)
- Infer X(a,c) from R(a,b),X(b,c)

Infer X(1,3) from R(1,2),X(2,3) Infer X(2,4) from R(2,3),X(3,4) Infer X(5,4) from R(5,3),X(3,4)

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- Initialize X by (SELECT source,dest FROM R)
- Recursively grow X by (SELECT R.source, X.dest FROM R, X WHERE R.dest=X.source)
- Infer X(a,c) from R(a,b),X(b,c)

Infer X(1,3) from R(1,2),X(2,3) Infer X(2,4) from R(2,3),X(3,4) Infer X(5,4) from R(5,3),X(3,4) Infer X(1,4) from R(1,5),X(5,4)

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- Initialize X by ۲ (SELECT source, dest FROM R)
- Recursively grow X by • (SELECT R.source, X.dest FROM R, X WHERE R.dest=X.source)
 - Infer X(a,c) from R(a,b),X(b,c)

Infer X(1,3) from R(1,2),X(2,3)Infer X(2,4) from R(2,3),X(3,4)Infer X(5,4) from R(5,3),X(3,4)Infer X(1,4) from R(1,5), X(5,4)

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Quiz

- Which of the following are true about WITH RECURSIVE clause?
- a) It is able to output transitive closure of a directed graph
- b) It usually works with an edge table
- c) It includes two SELECT statements
- d) All of the above

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Outline

- 1. Where do we use spatial networks?
- 2. Example spatial networks
- 3. Conceptual model of spatial networks
- 4. Why do we need SQL extensions
- 5. CONNECT clause
- 6. RECURSIVE statement
- 7. Data structures
- 8. Algorithms for connectivity query
- 9. Algorithms for shortest path

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Data Models of Spatial Networks

- 1. Conceptual Model : Entity Relationship Diagrams, Graphs
- 2. Logical Data Model : Abstract Data types , Custom Statements in SQL
- 3. Physical Data Model
 - Storage: Data-Structures, File-Structures
 - Algorithms for common operations

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Main Memory Data-Structures

- Adjacency matrix
 - M[A, B] = 1 if and only if edge(vertex A, vertex B) exists
- Adjacency list :
 - maps a vertex to a list of its successors



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Disk-based Tables

- Normalized tables
 - one for vertices, other for edges
- Denormalized
 - one table for nodes with adjacency lists



Node (R)				
id	x	У		
1	4.0	5.0		
2	6.0	3.0		
3	5.0	1.0		
4	3.0	2.0		
5	1.0	3.0		

Edge (S)						
source	dest	distance				
1	2	√ 8				
1	4	$\sqrt{10}$				
2	3	√ 5				
2	4	$\sqrt{10}$				
4	5	√ 5 [°]				
5	1	$\sqrt{18}$				

id	x	у	Successors	Predecessors
1	4.0	5.0	(2,4)	(5)
2	6.0	3.0	(3,4)	(1)
3	5.0	1.0	0	(2)
4	3.0	2.0	(5)	(1,2)
5	1.0	3.0	(1)	(4)

(d) Node and Edge Relations

(e) Denormalized Node Table

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File-Structures: Partition Graph into Disk Blocks

- Which partitioning reduces disk I/O for graph operations?
 - Choice 1: Geometric partition



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File-Structures: Partition Graph into Disk Blocks

- Which partitioning reduces disk I/O for graph operations?
 - Choice 1: Geometric partition
 - Choice 2: min-cut Graph Partition
 - Choice 2 cuts fewer edges and is preferred
 - Assuming uniform querying popularity across edges







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Exercise: Graph Based Storage Methods

- Consider spatial network on right
- If a disk page holds 3 nodes, which partitioning will has fewest cut-edges?
 - (a) (1, 2, 3), (4,5,6)
 (b) (2, 3, 4), (1, 5, 6)
 (c) (1, 2, 6), (3, 4, 5)
 (d) (1, 3, 5), (2, 4, 6)



0

(1,5)

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6

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 - Storage-Structures
 - Algorithms for common operations

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Algorithms

- Main memory
 - Connectivity: Breadth first search, depth first search
 - Shortest path: Dijkstra's algorithm, A*
- Disk-based
 - Shortest path Hierarchical routing algorithm

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Algorithms for Connectivity Query



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Quiz

Which of the following is false?

- a) Breadth first search visits nodes layer (i.e. generation) by layer
- b) Depth first search try a path till dead-end, then backtrack to try different paths
- c) Depth first search always performs better than breadth first searchd) None of the above

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Shortest Path Algorithms

Iterate

- Expand most promising descent node
 - Dijkstra's: try closest descendent to self
 - A* : try closest descendent to both destination and self
- Update current best path to each node, if a better path is found
- Till destination node is expanded

Dijkstra's vs. A*














Dijkstra's Algorithm

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Dijkstra's Algorithm

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Dijkstra's Algorithm

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Shortest Path Algorithms

• Iterate

- Expand most promising node
 - Dijkstra's: try closest descendent to self
 - A* : try closest descendent to both destination and self
- Update current best path to each node, if a better path is found
- Till destination node is expanded

Correct assuming

- Sub-path optimality
- Fixed, positive and additive edge costs
- A* heuristic function h(x) (estimated distance to destination from x) has two properties
 - It is an underestimate to the actual distance
 - It is consistent i.e., for every edge (x,y), h(x) <= d(x,y) + h(y) (d is the length of edge (x,y)</p>

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Shortest Path Strategies for Sec Memory

- Dijkstra's and Best first algorithms
 - Work well when entire graph is loaded in main memory
 - Otherwise their performance degrades substantially
- Hierarchical Routing Algorithms
 - Works with graphs on secondary storage
 - Loads small pieces of the graph in main memories
 - Can compute least cost routes

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Shortest Path Strategies for Sec Memory

- Key ideas behind Hierarchical Routing Algorithm
 - Fragment graphs pieces of original graph obtained via node partitioning
 - Boundary nodes nodes of with edges to two fragments
 - Boundary graph a summary of original graph
 - Contains Boundary nodes
 - Boundary edges: edges across fragments or paths within a fragment

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Shortest Path Strategies for Sec Memory

- A Summary of Optimal path in original graph can be computed
 - Using Boundary graph and 2 fragments
- The summary can be expanded into optimal path in original graph
 - Examining a fragments overlapping with the path
 - Loading one fragment in memory at a time

Shortest Path Strategies - (Illustration of the Algorithm)

- Figure 6.7(a) fragments of source and destination nodes
- Figure 6.7(b) computing summary of optimal path using
 - Boundary graph and 2 fragments
 - Note use of boundary edges only in the path computation
- Figure 6.8(a) The summary of optimal path using boundary edges
- Figure 6.8(b) Expansion back to optimal path in original graph

Hierarchical Routing Algorithm-Step 1

- Step 1: Choose Boundary Node Pair
 - Minimize COST(S,Ba)+COST(Ba,Bd)+COST(Bd,D)
 - Determining Cost May Be Non-Trivial



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Hierarchical Routing- Step 2

• Step 2: Shortest Boundary Path



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Hierarchical Routing- Step 3

- Step 3: Expand Boundary Path: (B_{a1}, B_d) -> B_{a1} Bd_{a2} B_{a3} B_{da4}...B_d
 - Boundary Edge (B_{ij}, B_j) ->fragment path (B_{i1}, N₁N₂N₃.....N_k, B_j)



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Quiz

Which of the following is false?

- a) Hierarchical routing algorithms are Disk-based shortest path algorithms
- b) Breadth first search and depth first search are both connectivity query algorithms
- c) Best first algorithm is always faster than Dijkstra's algorithm
- d) None of the above

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