

# From GPS and Google Maps to Spatial Computing

IEEE Intl. Conf. on Contemporary Computing (IC3), IIIT, Delhi, India, Aug., 2015

GMU Intl. Symposium on Spatiotemporal Computing, VA, USA, July 2015.

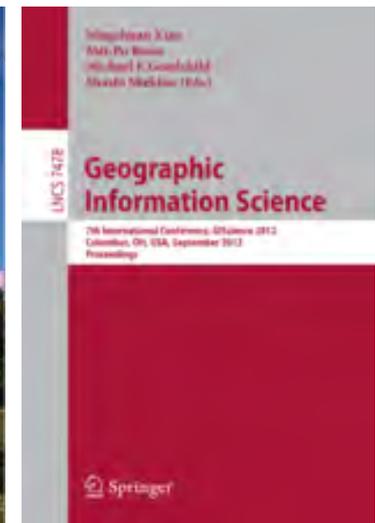
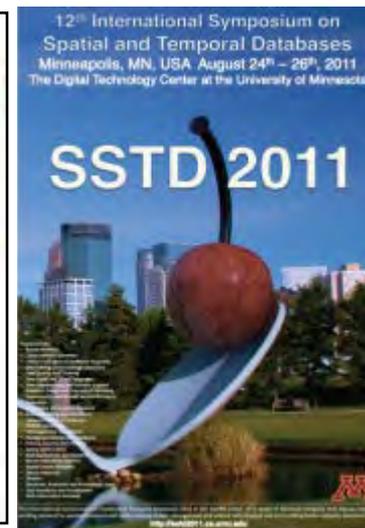
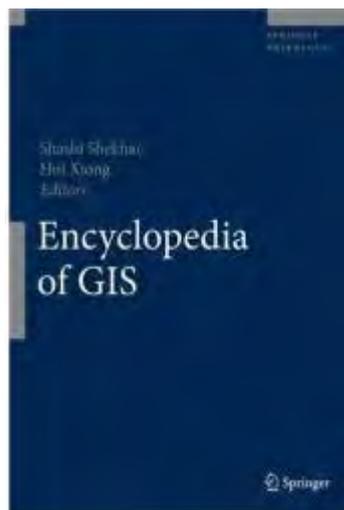
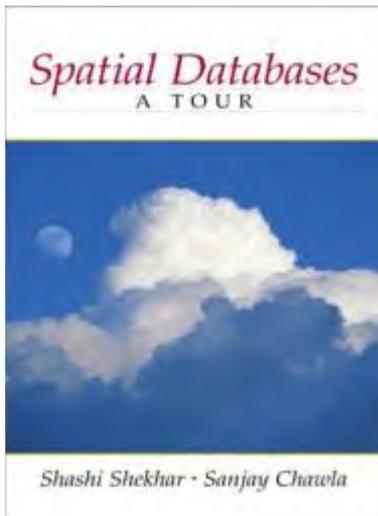
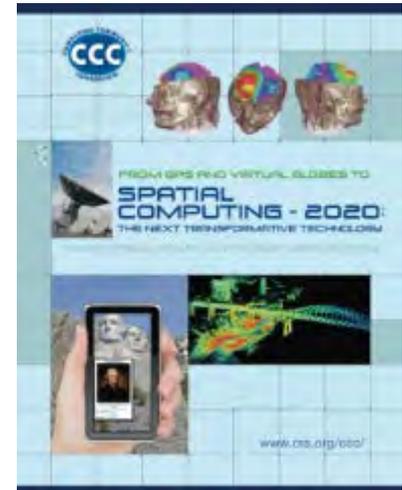
## Shashi Shekhar

McKnight Distinguished University Professor

Department of Computer Science and Eng.

University of Minnesota

[www.cs.umn.edu/~shekhar](http://www.cs.umn.edu/~shekhar)



Acknowledgements: Work supported by NSF (grant IIS-1320580) and USDOD (grant HM0210-13-1-0005).

# Courses

## CSCI 5715: From GPS and Virtual Globes to Spatial Computing

Map of students online at Coursera.org



[www.coursera.org/course/spatialcomputing](http://www.coursera.org/course/spatialcomputing)

UNIVERSITY OF MINNESOTA  
Driven to Discover™

### From GPS and Google Maps to Spatial Computing

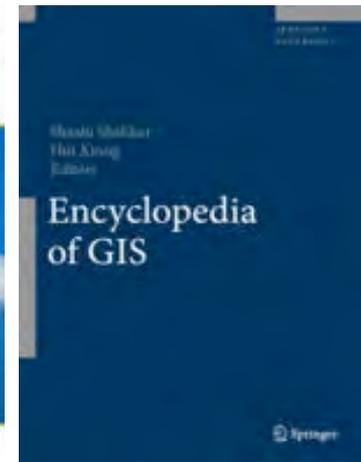
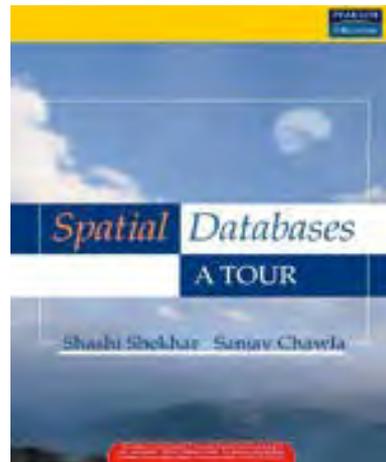
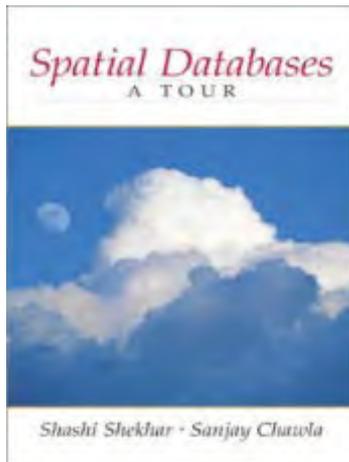
This course introduces concepts, algorithms, programming, theory and design of spatial computing technologies such as global positioning systems (GPS), Google Maps, location-based services and geographic information systems. Learn how to collect, analyze, and visualize your own spatial datasets while avoiding common pitfalls and building better location-aware technologies.

Preview Lectures



## CSCI 8715: Spatial Databases

[www.spatial.cs.umn.edu/Courses/Fall13/8715](http://www.spatial.cs.umn.edu/Courses/Fall13/8715)



## Alumni in Academia



## Current Students



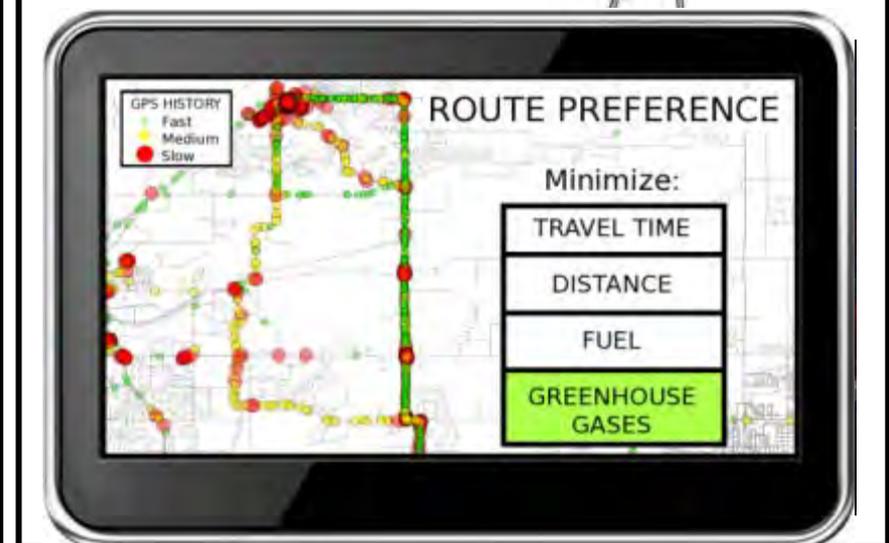
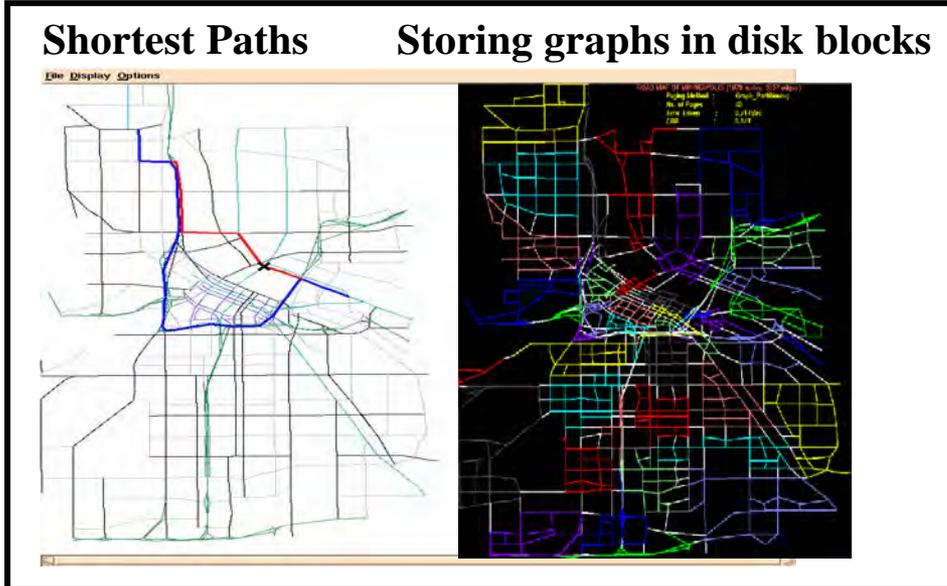
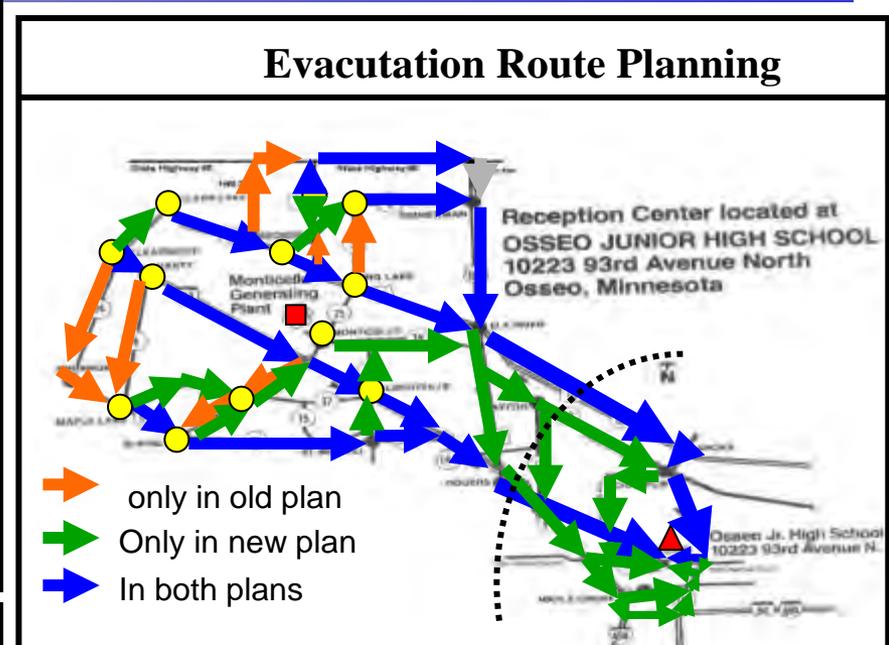
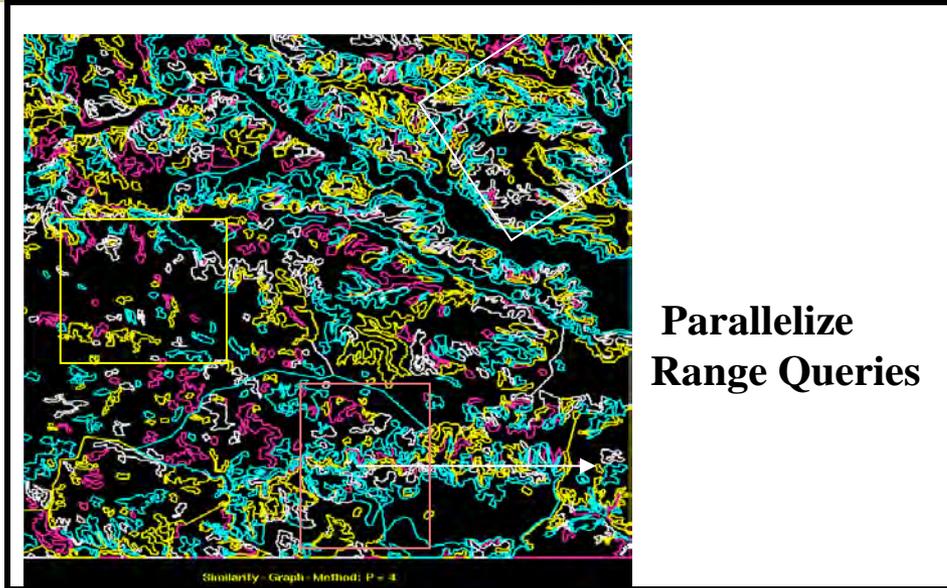
## Alumni in Industry



## Alumni in Government Agency



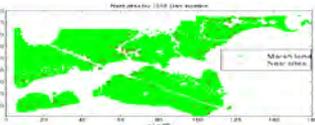
# Research Theme 1: Spatial Databases



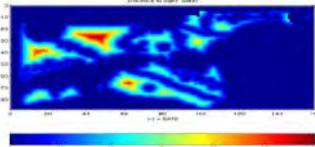
# Theme 2 : Spatial Data Mining

## Location Prediction: nesting sites

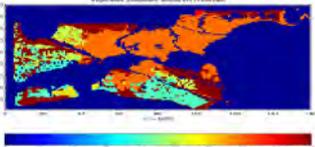
Nest locations



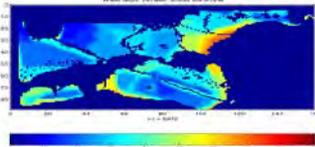
Distance to open water



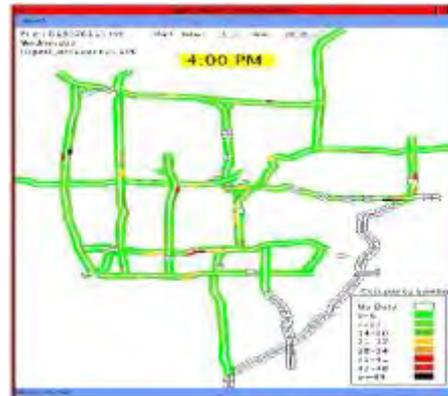
Vegetation durability



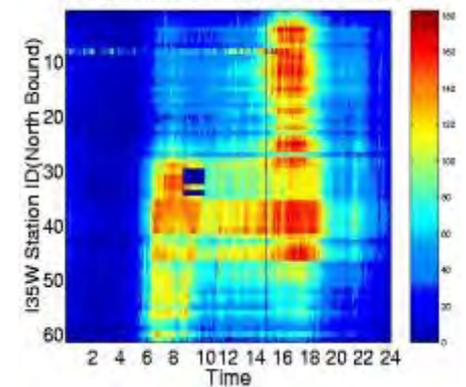
Water depth



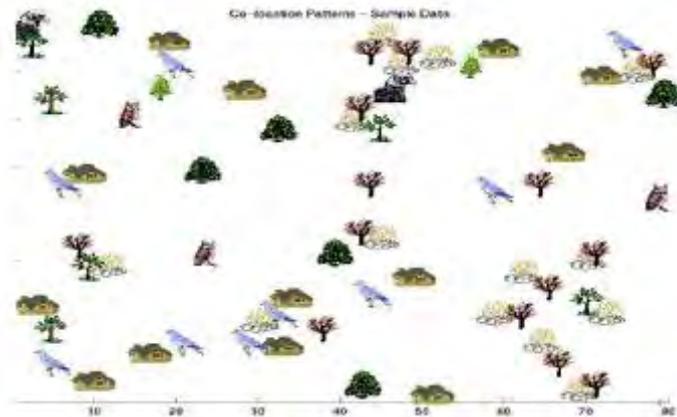
## Spatial outliers: sensor (#9) on I-35



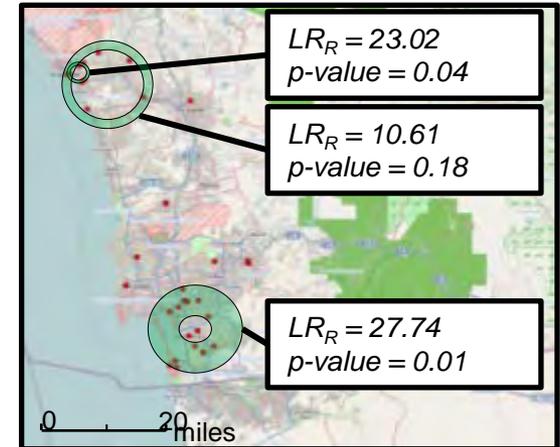
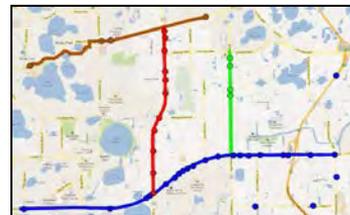
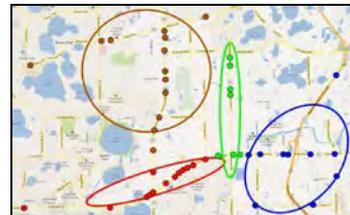
Average Traffic Volume (Time v.s. Station)



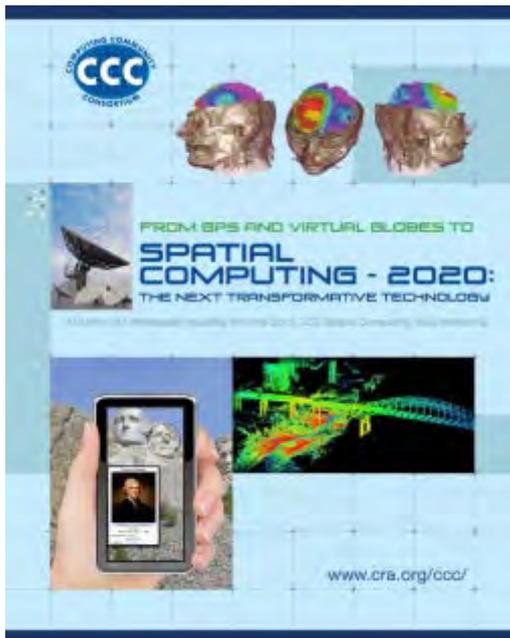
## Co-location Patterns



## Spatial Concept Aware Summarization



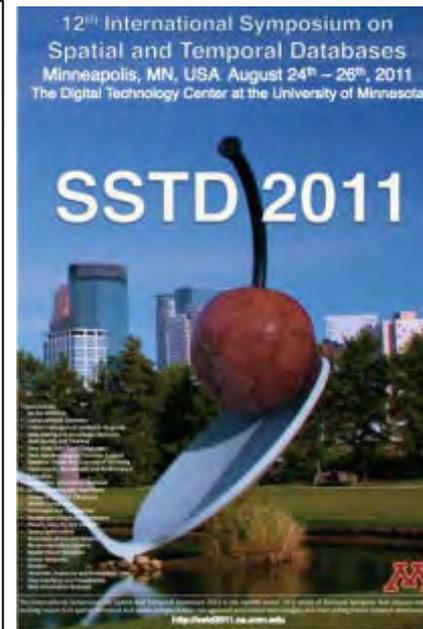
# Recent Professional Activities



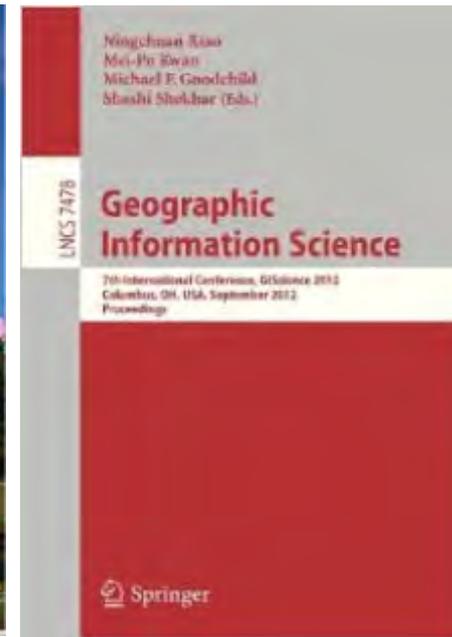
Spatial Computing  
Visioning Workshop  
Computing Community  
Consortium (CCC)



Geoinformatica  
Journal



Symposium on Spatial  
and Temporal  
Database 2011

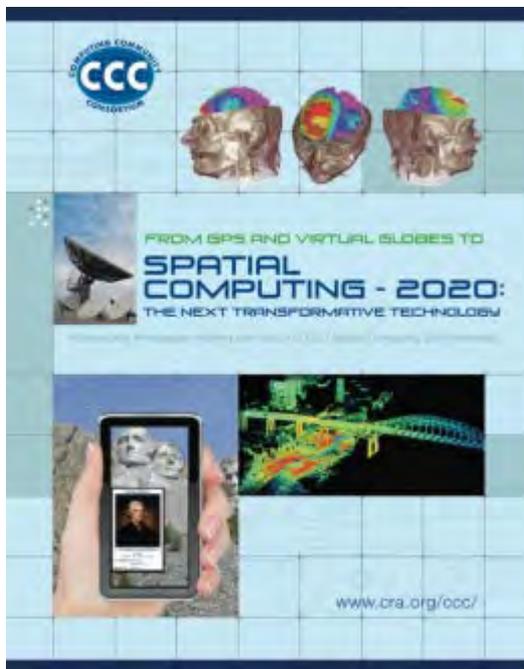


GIScience  
Conference 2012

# Sources

- From GPS and Virtual Globes to Spatial Computing 2020, CCC Report, 2013. [www.cra.org/ccc/visionsing/visionsing-activities/spatial-computing](http://www.cra.org/ccc/visionsing/visionsing-activities/spatial-computing)

- With few slides on work from presenter's group  
Identifying patterns in spatial information: a survey of methods, Wiley Interdisc. Reviews: Data Mining and Know. Discovery , 1(3):193-214, May/June 2011. (DOI: 10.1002/widm.25).



# Outline



- Introduction
  - Spatial Computing Audience: Niche => Everyone
  - Spatial Computing 2020 - Workshop
- GPS
- Location Based Services
- Spatial Statistics
- Spatial Database Management Systems
- Virtual Globes
- Geographic Information Systems
- Conclusions

# What is Spatial Computing?

- Transformed our lives through understanding spaces and places
  - Examples: localization, navigation, site selection, mapping,
  - Examples: spatial context, situation assessment (distribution, patterns), ...



Smarter Planet



# The Changing World of Spatial Computing

	Last Century	Last Decade
<b>Map User</b>	Well-trained few	Billions
<b>Mappers</b>	Well-trained few	Billions
<b>Software, Hardware</b>	Few layers, e.g., Applications: Arc/GIS, Databases: SQL3/OGIS	Almost all layers
<b>User Expectations &amp; Risks</b>	Modest	Many use-case & Geo-privacy concerns

# It is widely used by Government!

## Geospatial Information and Geographic Information Systems (GIS): An Overview for Congress



**Table I. Members of the Federal Geographic Data Committee (FGDC)**

Dept. of Agriculture	Environmental Protection Agency
Dept. of Commerce	Federal Emergency Management Agency
Dept. of Defense	General Services Administration
Dept. of Energy	Library of Congress
Dept. of Health and Human Services	National Aeronautics and Space Administration
Dept. of Housing and Urban Development	National Archives and Records Administration
Dept. of the Interior (Chair)	National Science Foundation
Dept. of Justice	Tennessee Valley Authority
Dept. of State	
Dept. of Transportation	Office of Management and Budget (Co-Chair)

# It is only a start! Bigger Opportunities Ahead!

McKinsey Global Institute

Big data: The next frontier for innovation, competition, and productivity

The study estimates that the use of personal location data could save consumers worldwide more than **\$600 billion annually by 2020**. Computers determine users' whereabouts by tracking their mobile devices, like cellphones. The study cites smartphone location services including Foursquare and Loopt, for locating friends, and ones for finding nearby stores and restaurants.

But the biggest single consumer benefit, the study says, is going to come from time and fuel savings from location-based services — tapping into real-time traffic and weather data — that **help drivers avoid congestion and suggest alternative routes**. The location tracking, McKinsey says, will work either from drivers' mobile phones or GPS systems in cars.

**The New York Times**

Published: May 13, 2011

New Ways to Exploit Raw Data May Bring Surge of Innovation, a Study Says

# CCC Visioning Workshop: Making a Case for Spatial Computing 2020

[http://cra.org/ccc/spatial\\_computing.php](http://cra.org/ccc/spatial_computing.php)



## Computing Community Consortium

We support the computing research community in creating compelling research visions and the mechanisms to realize these visions.

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ACTIVITIES

RESOURCES

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GO

### Funded Visioning Activities

Disaster Management   SEES IT   HealthIT   Interactive Tech   Architecture   XLayer   Robotics   Learning Tech  
Open Source   Cyber Physical Systems   Global Development   Theoretical CS   Big Data Computing   NetSE  
**Spatial Computing**

## From GPS and Virtual Globes to Spatial Computing-2020

### About the workshop

This workshop outlines an effort to develop and promote a unified agenda for Spatial Computing research and development across US agencies, industries, and universities. See the original workshop proposal [here](#).

### *Spatial Computing*

Spatial Computing is a set of ideas and technologies that will transform our lives by understanding the physical world, knowing and communicating our relation to places in that world, and navigating through those places.

The transformational potential of Spatial Computing is already evident. From Virtual Globes such as Google Maps and Microsoft Bing Maps to consumer GPS devices, our society has benefitted immensely from spatial technology. We've reached the point where a hiker in Yellowstone, a schoolgirl in DC, a biker in Minneapolis, and a taxi driver in Manhattan know precisely where they are, nearby points of interest, and how to reach their destinations. Large

## Logistics

**Date:** Sept. 10th-11th, 2012

**Location:** Keck Center

**Hotel:** Liaison Hotel

## Steering Committee

Erwin Gianchandani

Hank Korth

## Organizing Committee

Peggy Agouris, George Mason University

Walid Aref, Purdue University

Michael F. Goodchild, University of California - Santa Barbara

# Workshop Participants

## Academia

**Peggy Agouris**, George Mason University  
**Divyakant Agrawal**, University of California Santa Barbara  
**Cecilia Aragon**, University of Washington  
**Walid G. Aref**, Purdue University  
**Elisa Bertino**, Purdue University  
**Henrik Christensen**, Georgia Institute of Technology  
**Isabel Cruz**, University of Illinois at Chicago  
**Michael R. Evans**, University of Minnesota  
**Steven Feiner**, Columbia University  
**Jie Gao**, Stony Brook University  
**Michael Goodchild**, University of California Santa Barbara  
**Sara Graves**, University of Alabama Huntsville  
**Rajesh Gupta**, University of California San Diego  
**Chuck Hansen**, University of Utah  
**Stephen Hirtle**, University of Pittsburgh  
**Krzysztof Janowicz**, University of California Santa Barbara  
**John Jensen**, University of South Carolina  
**Daniel Keefe**, University of Minnesota  
**John Keyser**, Texas A&M University  
**Craig A. Knoblock**, Information Sciences Institute

**Hank Korth**, Lehigh University  
**Benjamin Kuipers**, University of Michigan  
**Vipin Kumar**, University of Minnesota  
**Richard Langley**, University of New Brunswick  
**Chang-Tien Lu**, Virginia Tech  
**Dinesh Manocha**, University of North Carolina  
**Edward M. Mikhail**, Purdue  
**Harvey Miller**, University of Utah  
**Joe Mundy**, Brown University  
**Dev Oliver**, University of Minnesota  
**Rahul Ramechandran**, UA Huntsville  
**Norman Sadeh**, CMU  
**Shashi Shekhar**, University of Minnesota  
**Daniel Z. Sui**, Ohio State  
**Roberto Tamassia**, Brown University  
**Paul Torrens**, University of Maryland  
**Shaowen Wang**, University of Illinois at Urbana-Champaign  
**Greg Welch**, University of North Carolina  
**Ouri E. Wolfson**, University of Illinois at Chicago  
**Mike Worboys**, University of Maine  
**May Yuan**, University of Oklahoma  
**Avidesh Zakhor**, University of California Berkeley

>30 Universities

## Industry

**Mark Abrams**, ESG  
**Mohamed Ali**, Microsoft  
**Lee Allison**, Arizona Geological Survey  
**Virginia Bacon Talati**, Computer Science and Telecommunications Board (CSTB)  
**Ramon Caceres**, AT&T Research  
**Vint Cerf**, Google  
**Jade DePalacios**, Naval Postgraduate School  
**Jon Eisenberg**, Computer Science and Telecommunications Board (CSTB)  
**Tom Erickson**, IBM  
**Erwin Gianchandani**, CCC  
**Eric Hoel**, ESPI  
**Xuan Liu**, IBM  
**Siva Ravada**, Oracle  
**Jagan Sankaranarayanan**, NEC Labs  
**Lea Shanley**, Wilson Center  
**Kevin Pomfret**, Centre for Spatial Law and Policy

14 Organizations

## Government

**Nabil Adam**, DHS  
**Vijay Atluri**, NSF  
**David Balshaw**, NIH/NIEHS  
**Budhendra Bhaduri**, ORNL  
**Kelly Crews**, NSF  
**Beth Driver**, NGA  
**Walton Fehr**, USDOT  
**Myron Gutmann**, NSF  
**Susanne Hambrusch**, NSF  
**Michelle Heacock**, NIH/NIEHS  
**Clifford Jacobs**, NSF  
**Farnam Jahanian**, NSF  
**Todd Johannesen**, NGA  
**Thomas Johnson**, NGA  
**Henry Kelly**, OSTP  
**Alicia Lindauer**, USDOE  
**Keith Marzullo**, NSF  
**John L. Schnase**, NASA  
**Jim Shine**, Army Research  
**Raju Vatsavai**, ORNL  
**Eric Vessey**, NSA  
**Howard D. Wactlar**, NSF  
**Tandy Warnow**, NSF  
**Nicole Wayant**, Army Research  
**Mark Weiss**, NSF  
**Maria Zemankova**, NSF  
**Li Zhu**, NIH/NCI

12 Agencies

# Workshop Highlights

## Agenda

- Identify fundamental research questions for individual computing disciplines
- Identify cross-cutting research questions requiring novel, multi-disciplinary solutions



## Organizing Committee

- Peggy Agouris, George Mason University
- Walid Aref, Purdue University
- Michael F. Goodchild, University of California - Santa Barbara
- Erik Hoel, Environmental Systems Research Institute (ESRI)
- John Jensen, University of South Carolina
- Craig A. Knoblock, University of Southern California
- Richard Langley, University of New Brunswick
- Ed Mikhail, Purdue University
- Shashi Shekhar, University of Minnesota
- Ourl Wolfson, University of Illinois
- May Yuan, University of Oklahoma



# Workshop Highlights

## Pull Panel: National Priorities, Societal Applications of Spatial Computing

**Chair:** Henry Kelly, OSTP

### Members

US-DoD: Eric Vessey

US-DoD: Todd Johanesen

NIH/NIEHS: Michelle Heacock

NASA: John L Schnase

DHS: Nabil Adam

NSF EarthCube: Clifford Jacobs

DOT: Walton Fehr

DOE: Alicia Lindauer

## Push Panel: Spatial Computing (SC) Platform Trends, Disruptive Technologies

**Chair:** Dinesh Manocha, UNC

### Members:

Graphics & Vision: John Keyser, TAMU

Interaction Devices: Steven Feiner, Columbia University

LIDAR : Avideh Zakhor, UCB

GPS Modernization: Mark Abrams, Advisor to USG

Cell Phones: Ramon Caceres, AT&T

Indoor Localization: Greg Welch, UNC

Internet Localization: Rajesh Gupta, UCSD

Cloud Computing: Divyakant Agarwal, UCSB

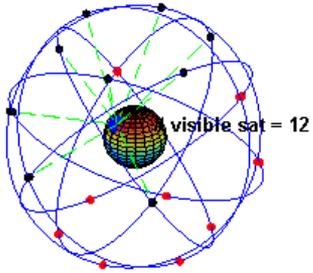
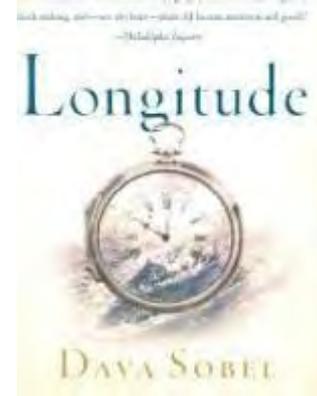
# Outline



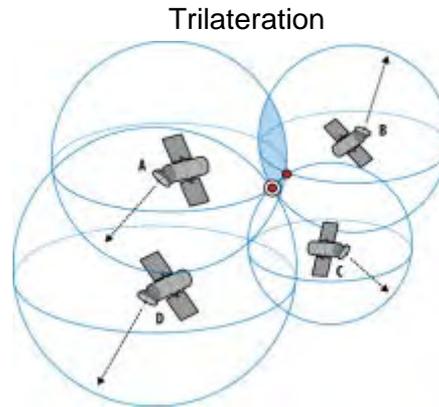
- Introduction
- **GPS**
  - **Outdoors => Indoors**
- Location Based Services
- Spatial Statistics
- Spatial Database Management Systems
- Virtual Globes
- Geographic Information Systems
- Conclusions

# Global Positioning Systems (GPS)

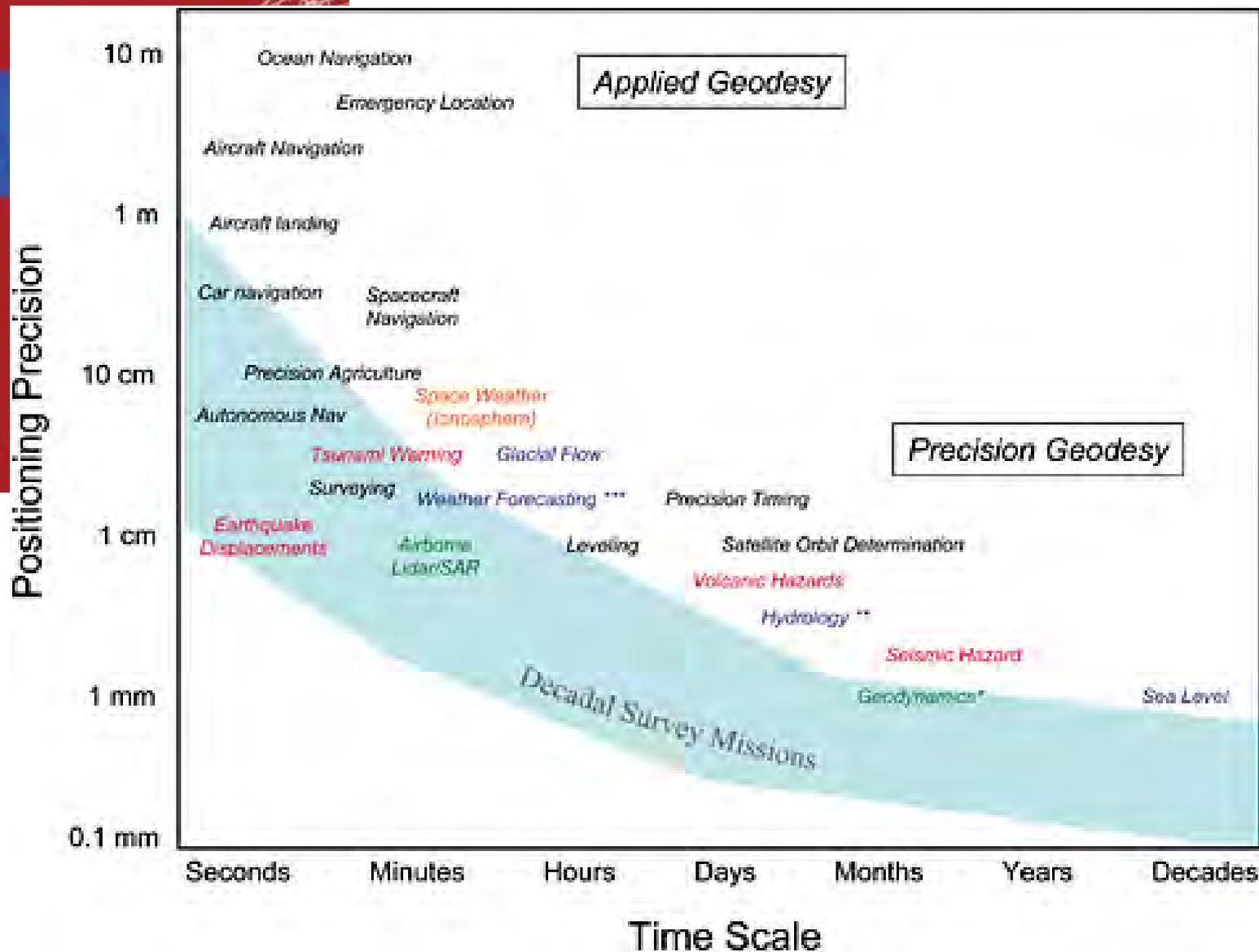
- Positioning ships
  - Latitude f(compass, star positions)
  - Longitude: dead-reckoning => marine chronometer
  - Longitude prize (1714), accuracy in nautical miles
- Global Navigation Satellite Systems
  - Infrastructure: satellites, ground stations, receivers, ...
  - Use: Positioning (sub-centimeter), Clock synchronization



[http://en.wikipedia.org/wiki/Global\\_Positioning\\_System](http://en.wikipedia.org/wiki/Global_Positioning_System)



<http://answers.oreilly.com/topic/2815-how-devices-gather-location-information/>



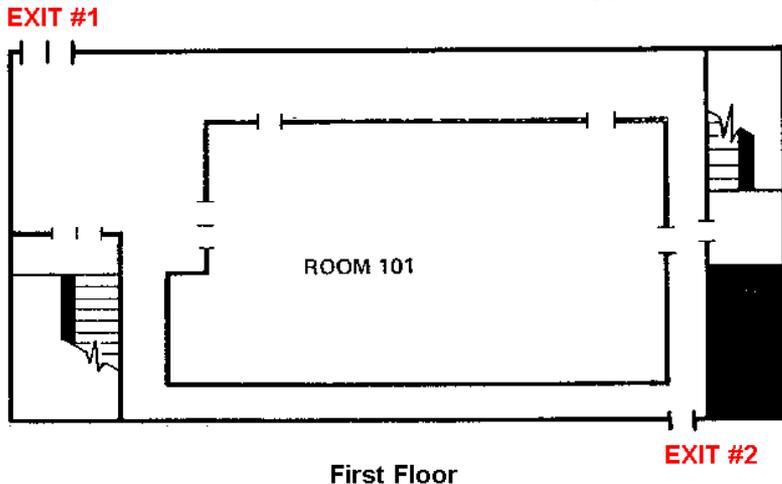
# Trends: Localization Indoors and Underground

- GPS works outdoors, but,
  - We are indoors 90% of time!
  - Ex. malls, hospitals, airports, etc.
  - Indoor asset tracking, exposure hotposts, ...

<http://www.mobilefringe.com/products/square-one-shopping-center-app-for-iphone-and-android/>



- Leveraging existing indoor infrastructure
  - Blue Tooth, WiFi, Cell-towers, cameras, Other people?
- How to model indoors for navigation, tracking, hotspots, ...?
  - What are nodes and edges ?



Get In-Store  
Notifications

WiFi Localization



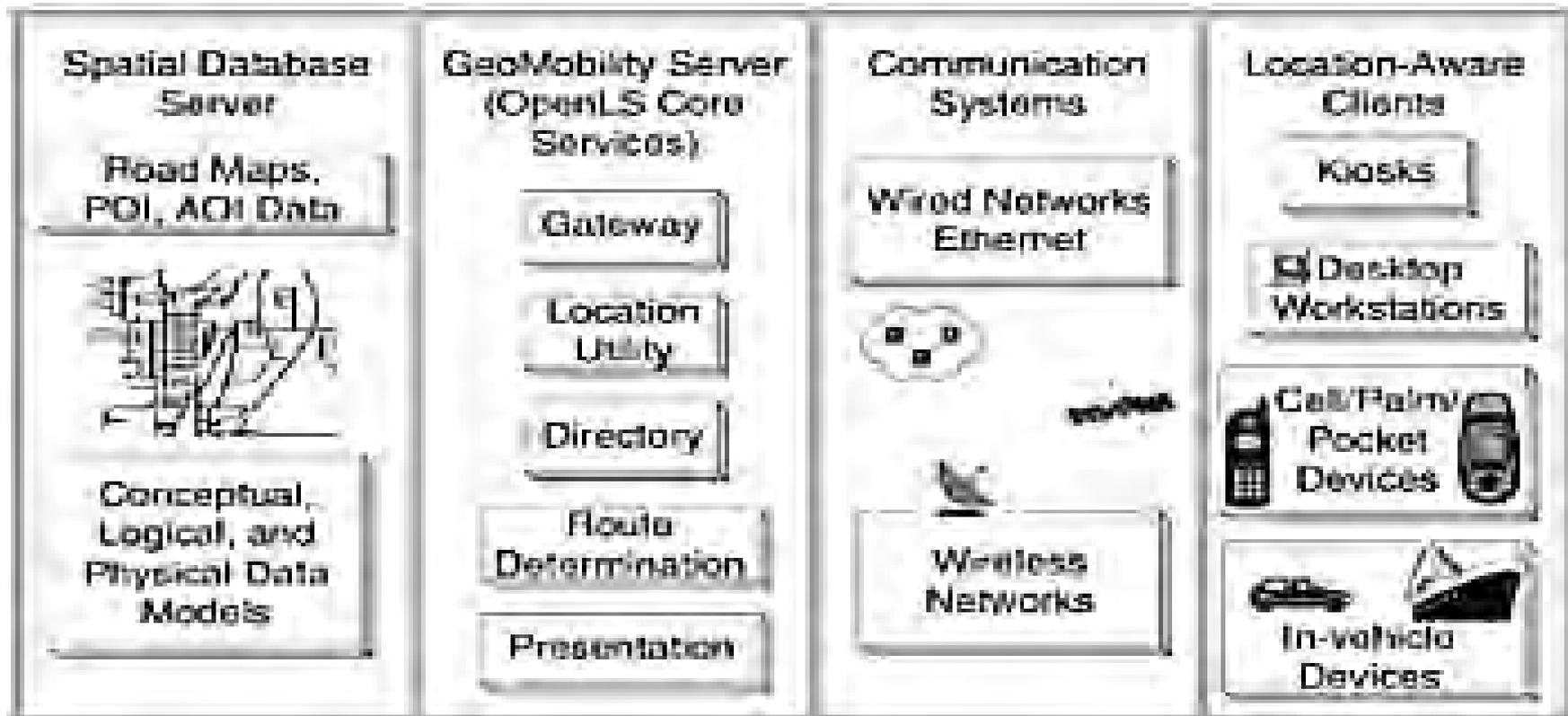
# Outline



- Introduction
- GPS
- Location Based Services
  - Queries => Persistent Monitoring
- Spatial Statistics
- Spatial Database Management Systems
- Virtual Globes
- Geographic Information Systems
- Conclusions

# Location Based Services

- Open Location Services: Queries
  - Location: Where am I? (street address, <latitude, longitude>)
  - Directory: Where is the nearest clinic (or doctor)?
  - Routes: What is the shortest path to reach there?



# Next Generation Navigation Services

- ❑ Eco-Routing
- ❑ Best start time
- ❑ Road-capacity aware

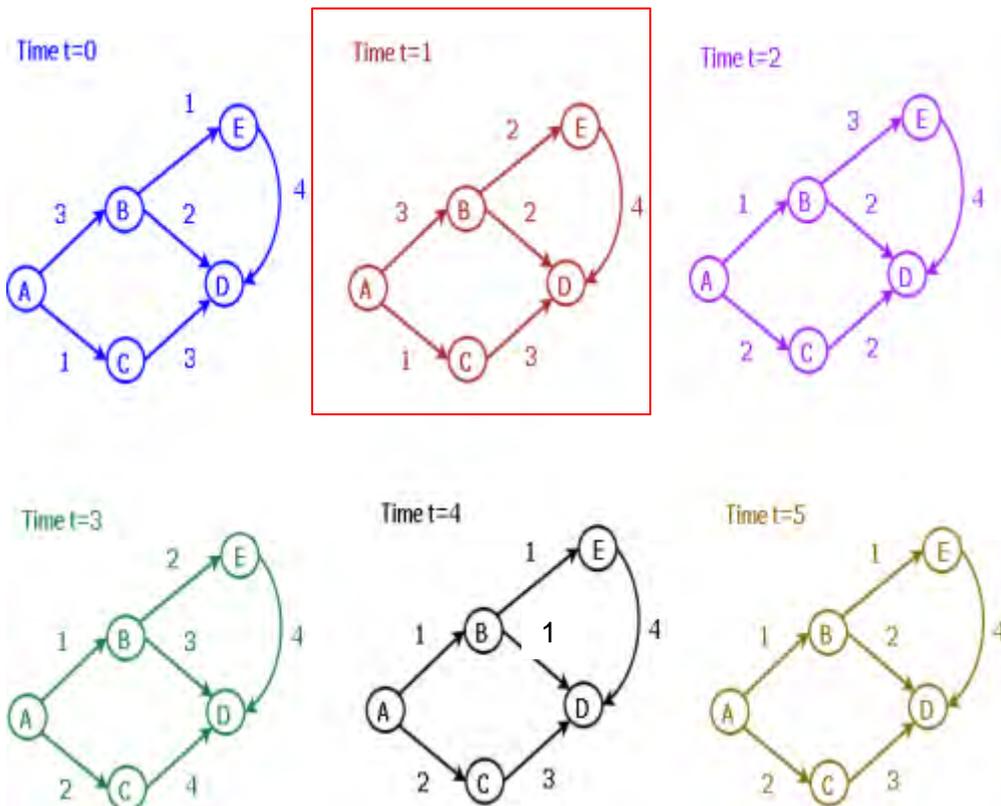


Static	Time-Variant
Which is the shortest travel time path from downtown Minneapolis to airport?	

# Routing Challenges: Lagrangian Frame of Reference

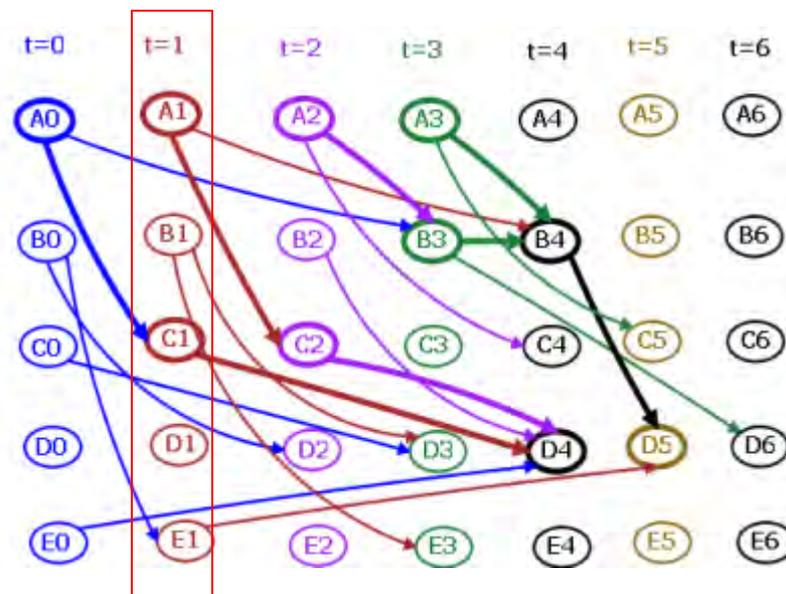
Q? What is the cost of Path  $\langle A,C,D \rangle$  with start-time  $t=1$  ? Is it 3 or 4 ?

## Snapshots of a Graph



Path	T = 0	T = 1	T = 2	T = 3
$\langle A,C,D \rangle$	4	3	5	4
$\langle A,B,D \rangle$	6	4	4	3

## Lagrangian Graph



**Details:** A Critical-Time-Point Approach to All-Start-Time Lagrangian Shortest Paths: A Summary of Results, (w/ V. Gunturi et al.), Proc. Intl. Symp. on Spatial and Temporal Databases, Springer LNCS 6849, 2011. Complete results accepted for the IEEE Transactions on Knowledge and Data Engineering.

# Spatio-temporal Graphs: Computational Challenges

## Ranking changes over time

Violates stationary assumption in  
Dynamic Programming

Time	Preferred Routes
<b>7:30am</b>	<b>Via Hiawatha</b>
<b>8:30am</b>	<b>Via Hiawatha</b>
<b>9:30am</b>	<b>via 35W</b>
<b>10:30am</b>	<b>via 35W</b>

## Waits, Non FIFO Behavior

Violate assumption of Dijkstra/A\*

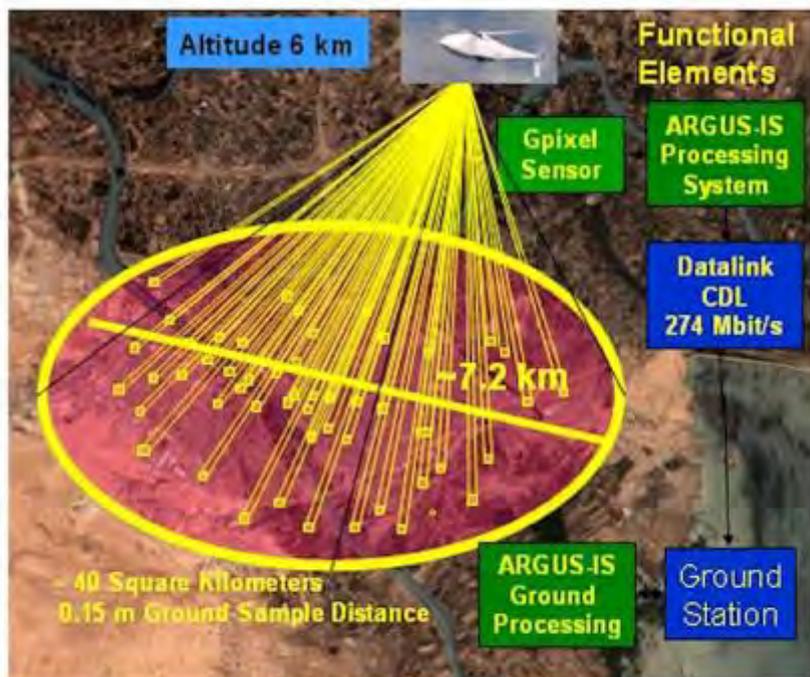
Time	Route	Flight Time
<b>8:30am</b>	<b>via Detroit</b>	<b>6 hrs 31 mins</b>
<b>9:10am</b>	<b>direct flight</b>	<b>2 hrs 51 mins</b>
<b>11:00am</b>	<b>via Memphis</b>	<b>4 hrs 38mins</b>
<b>11:30am</b>	<b>via Atlanta</b>	<b>6 hrs 28 mins</b>
<b>2:30pm</b>	<b>direct flight</b>	<b>2 hrs 51 mins</b>

\*Flights between Minneapolis and Austin (TX)

**Details:** A Critical-Time-Point Approach to All-Start-Time Lagrangian Shortest Paths: A Summary of Results, (w/ V. Gunturi et al.), Proc. Intl. Symp. on Spatial and Temporal Databases, Springer LNCS 6849, 2011. Complete results accepted for the IEEE Transactions on Knowledge and Data Engineering.

# Trends: Persistent Geo-Hazard Monitoring

- Environmental influences on our health & safety
  - air we breathe, water we drink, food we eat
- Surveillance
  - **Passive > Active > Persistent**
  - **How to economically cover all locations all the time ?**
  - Crowd-sourcing, e.g., smartphones, tweets,
  - Wide Area Motion Imagery



# Outline



- Introduction
- GPS
- Location Based Services
- **Spatial Statistics**
  - From Mathematical (e.g., hotspot)
  - To Spatial (e.g., hot features)
- Spatial Database Management Systems
- Virtual Globes
- Geographic Information Systems
- Conclusions

# Hotel That Enlivened the Bronx Is Now a 'Hot Spot' for Legionnaires'

By WINNIE HU and NOAH REMNICK AUG. 10, 2015

## Contaminated Cooling Towers

Five buildings have been identified as the potential source of the Legionnaires' disease outbreak in the South Bronx.

- Possible sources of Legionnaires' outbreak
- Additional sites found with legionella bacteria
- Locations of people with Legionnaires'



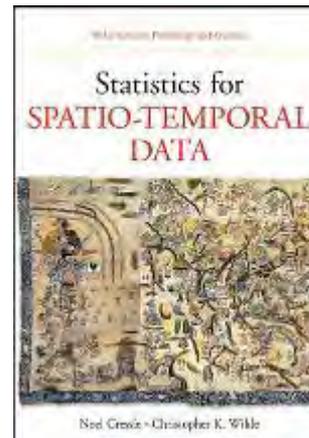
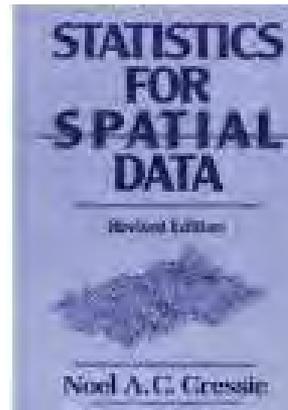
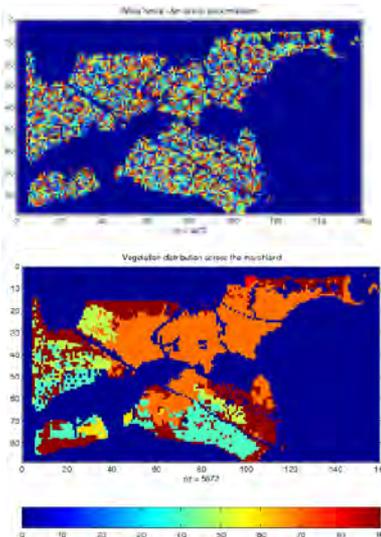
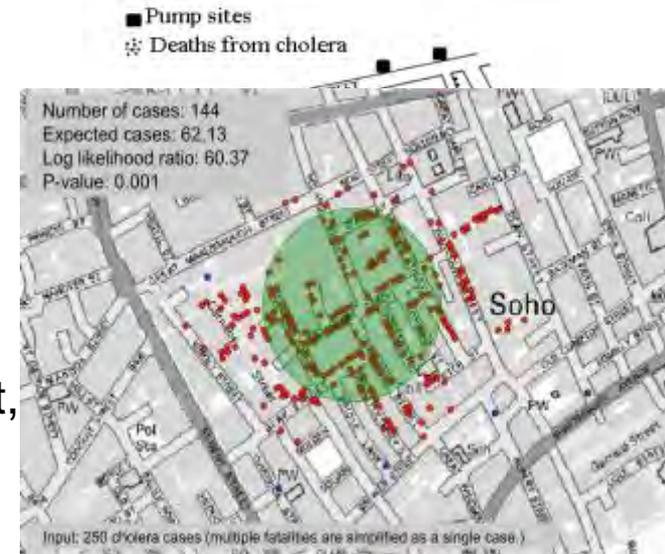
The Opera House Hotel is at the center of the outbreak. Edwin J. Torres for The New York Times

Source: New York Mayor's Office

By The New York Times

# Spatial Statistics: Mathematical Concepts

- Spatial Statistics
  - Quantify uncertainty, confidence, ...
  - Is it significant?
  - Is it different from a chance event or rest of dataset?
    - e.g., SaTScan finds circular hot-spots
- Model Auto-correlation, Heterogeneity, Edge-effect,
  - Point Process, e.g., Ripley's K-functions, SatScan
  - Geo-statistics, e.g., Kriging, GWR
  - Lattice-based models



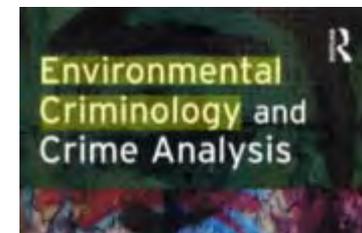
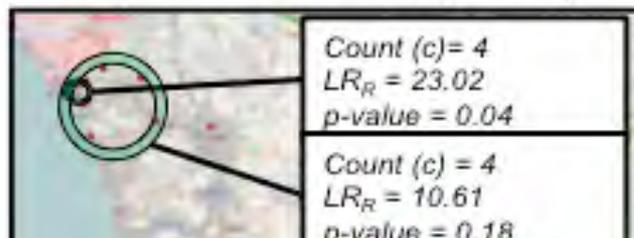
# Semantic Gap between Spatial and Machine Learning

- Representation choices beyond Linear Algebra
- Environmental Criminology
  - Routine Activities Theory, Crime Pattern Theory, Doughnut Hole pattern
- Formulation: **rings**, where **inside** density is significantly higher than **outside** ...

Input



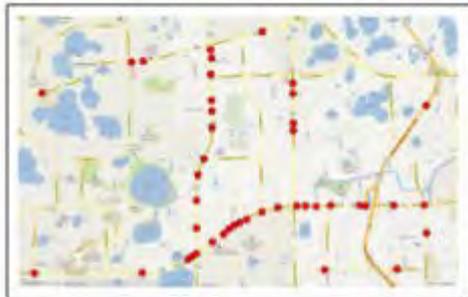
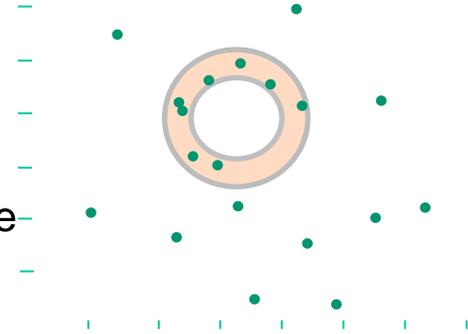
– Output: Ring Shaped Hotspot Detection (RHD)



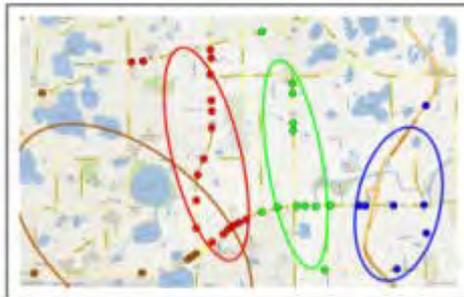
Mathematics	Concepts	Relationships
Sets	Set Theory	Member, set-union, set-difference, ...
Vector Space	Linear Algebra	Matrix & vector operations
Euclidean Spaces	Geometry	Circle, <b>Ring</b> , Polygon, Line_String, Convex hull, ...
Boundaries, Graphs, Spatial Graphs	Topology, Graph Theory, Spatial graphs, ...	Interior, boundary, Neighbor, <b>inside</b> , <b>surrounds</b> , ..., Nodes, edges, paths, trees, ... Path with turns, dynamic segmentation, ...

# Trends: Spatial-Concept Aware Patterns

- Spatial Concepts
  - Natural geographic features, e.g., rivers, streams, ...
  - Man-made geographic features, e.g., transportation network
  - Spatial theories, e.g., environmental criminology – doughnut hole
- Spatial-concept-aware patterns
  - Hotspots: Circle => Doughnut holes
  - Hot-spots => Hot Geographic-features



(a) Input



(b) Crimestat K-means with Euclidean Distance



(c) Crimestat K-means with Network Distance

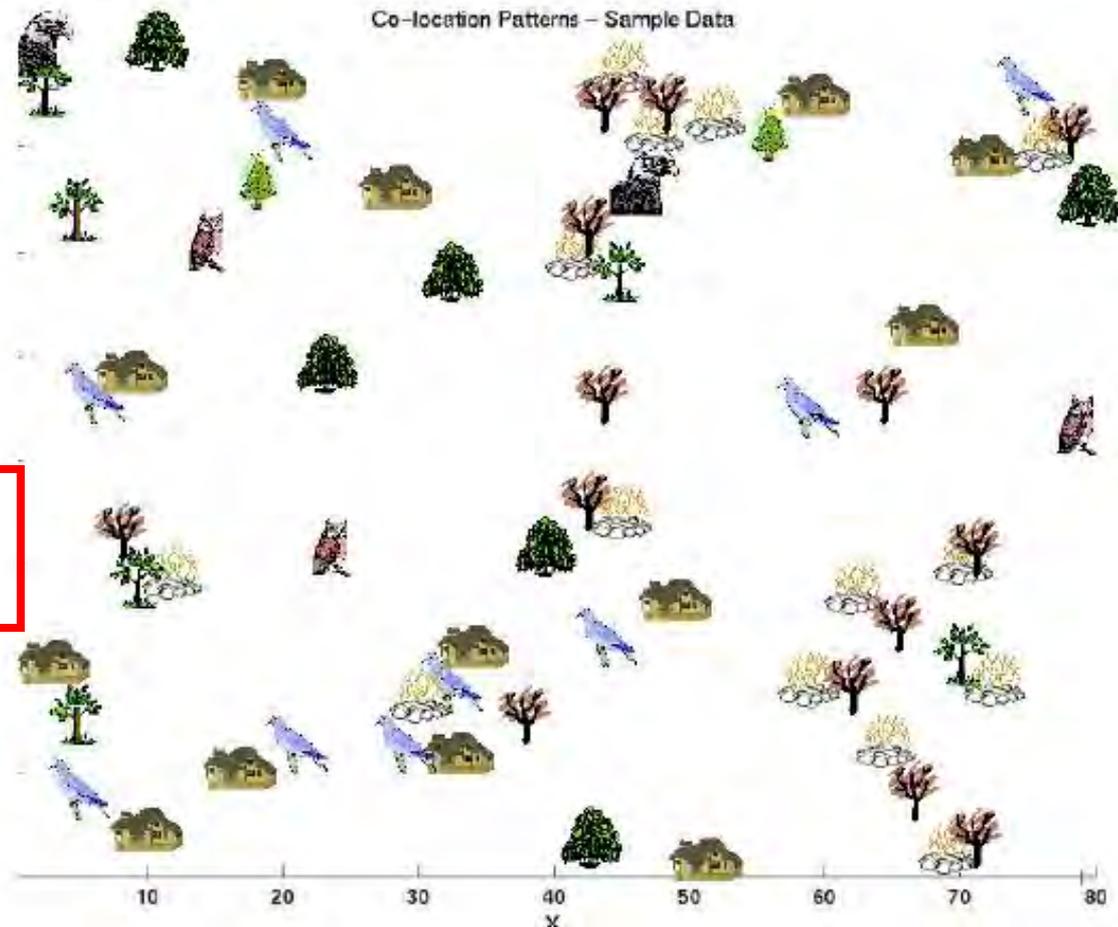


(d) KMR

**Details:** A K-Main Routes Approach to Spatial Network Activity Summarization, (w/ D. Oliver et al.)  
IEEE Transactions on Knowledge and Data Engineering, 26(6):1464-1478, 2014.

# Co-locations/Co-occurrence

- Given: A collection of different types of spatial events
- Find: Co-located subsets of event types



**Details:** Discovering colocation patterns from spatial data sets: a general approach, (w/ H. Yan et al.), IEEE Transactions on Knowledge and Data Engineering, 16(12), Dec. 2004.

# Fast Algorithms to Mine Colocations from Big Data

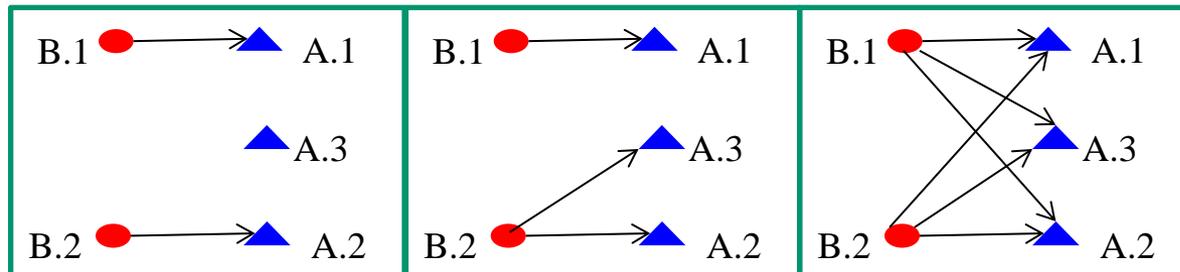
**Participation ratio**  $pr(f_i, c)$  of feature  $f_i$  in colocation  $c = \{f_1, f_2, \dots, f_k\}$ :  
 fraction of instances of  $f_i$  with feature  $\{f_1, \dots, f_{i-1}, f_{i+1}, \dots, f_k\}$  nearby  
 (i.e. within a given distance)

**Participation index**  $PI(c) = \min\{pr(f_i, c)\}$

## Properties:

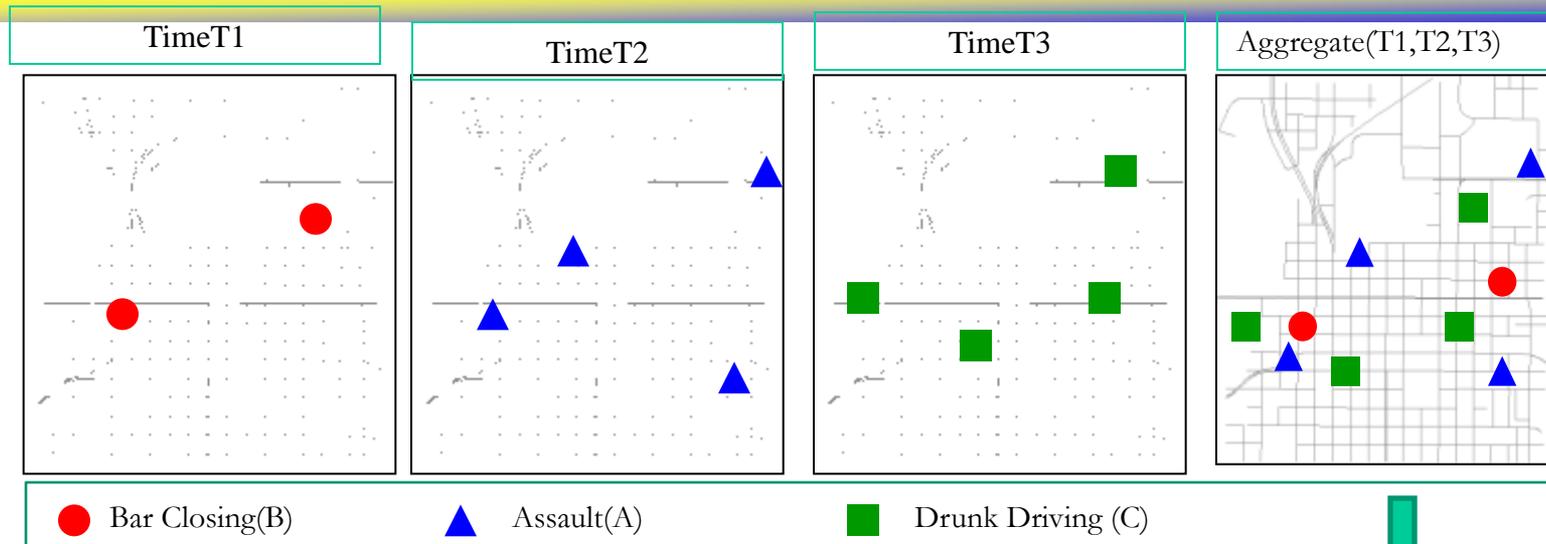
- (1) **Computational:** Non-monotonically decreasing like support measure  
 Allows scaling up to big data via pruning
- (2) **Statistical:** Upper bound on Cross-K function

## ■ Comparison with Ripley's K-function (Spatial Statistics)

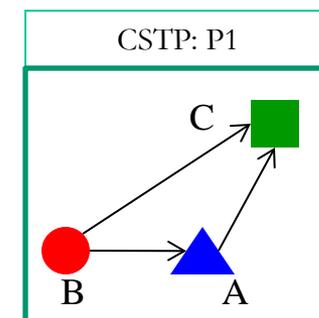


<b>K-function (B → A)</b>	$2/6 = 0.33$	$3/6 = 0.5$	$6/6 = 1$
<b>PI (B → A)</b>	$2/3 = 0.66$	1	1

# Cascading spatio-temporal pattern (CSTP)

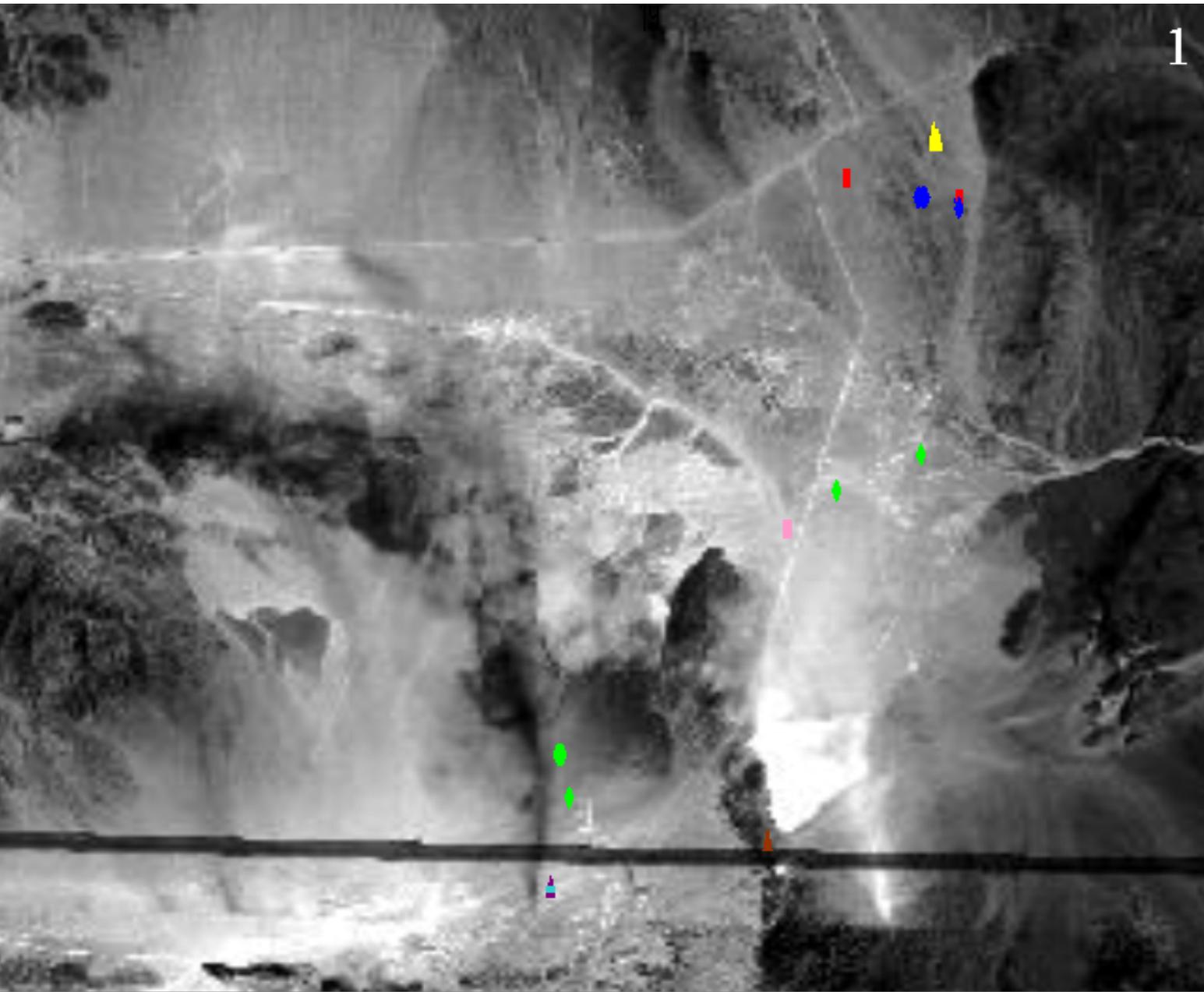


- ❑ *Input:* Urban Activity Reports
- ❑ *Output:* CSTP
  - ❑ *Partially ordered* subsets of ST event types.
  - ❑ Located together in space.
  - ❑ Occur in *stages* over time.
- ❑ Applications: Public Health, Public Safety, ...



**Details:** Cascading Spatio-Temporal Pattern Discovery, (w/ P. Mohan et al.), IEEE Transactions on Knowledge and Data Engineering, 24(11), Nov. 2012.

# MDCOP Motivating Example : Input



- Manpack stinger  
(2 Objects)



- M1A1\_tank  
(3 Objects)



- M2\_IFV  
(3 Objects)



- Field\_Marker  
(6 Objects)

- T80\_tank  
(2 Objects)



- BRDM\_AT5  
(enemy) (1 Object)



- BMP1  
(1 Object)





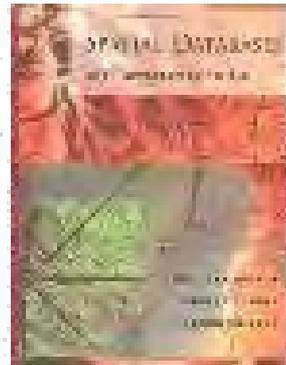
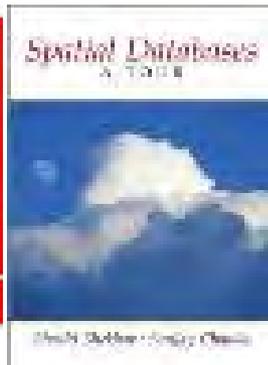
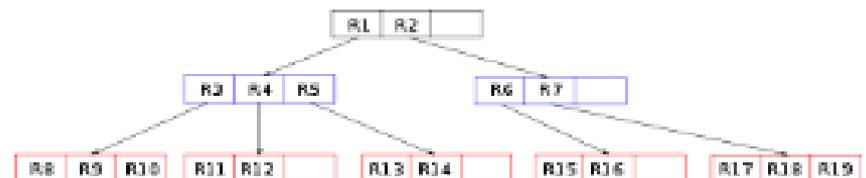
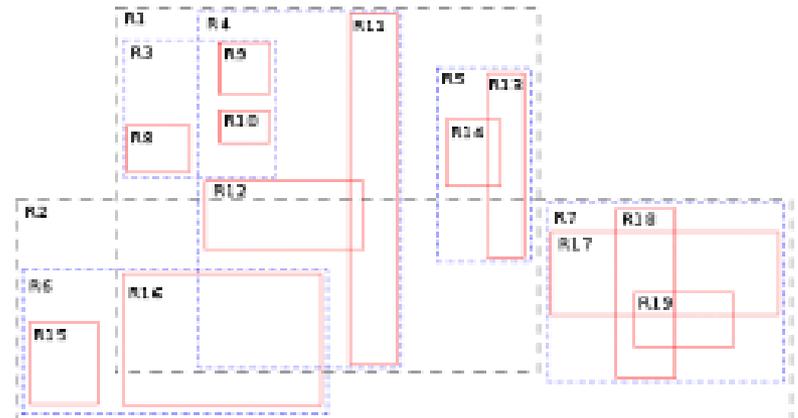
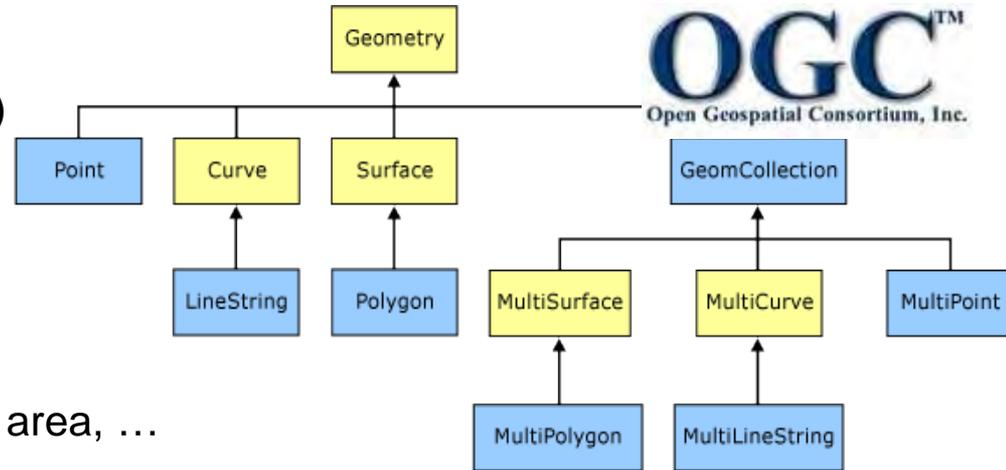
# Outline



- Introduction
- GPS
- Location Based Services
- Spatial Statistics
- **Spatial Database Management Systems**
  - Scalability => Privacy
- Virtual Globes
- Geographic Information Systems
- Conclusions

# Spatial Databases for Geometry

- Dice, Slide, Drill-down, Explore, ...
  - Closest pair( school, pollution-source)
  - Set based querying
- Reduce Semantic Gap
  - Clumsy code for inside, distance, ...
  - 6 data-types
  - Operations: inside, overlap, distance, area, ...
- Scale up Performance
  - Data-structures: B-tree => R-tree
  - Algorithms: Sorting => Geometric



# Challenge: Privacy vs. Utility Trade-off



- Check-in Risks: Stalking, GeoSlavery, ...
- Ex: Girls Around me App (3/2012), Lacy Peterson [2008]
- Others know that you are not home!



The Girls of Girls Around Me. It's doubtful any of these girls even know they are being tracked. Their names and locations have been obscured for privacy reasons. (Source: [Cult of Mac, March 30, 2012](#))

# Challenge: Geo-privacy, geo-confidentiality, ...

- Emerging personal geo-data
  - Trajectories of smart phones, gps-devices, life-trajectories and migrations, ...
- Privacy: Who gets my data? Who do they give it to? What promises do I get?
- Socio-technical problem
  - Need policy support
  - Challenges in fitting location privacy into existing privacy constructs (i.e HIPPA, Gramm-Leach-Bliley, Children's Online Privacy Protection Act)
- Groups interested in Geo-Privacy
  - Civil Society, Economic Entities, Public Safety ,Policy Makers

**Table 4.2: Geo-privacy Policy Conversation Starters**

**GEOTARGETED**  
ALERTS AND WARNINGS

1. Emergencies are different (E-911)
2. Differential geo-privacy can improve safety (E-911 → PLAN, CMAS)
3. Send apps to data, not vice-versa (e.g., eco-routing)
4. Transparent transactions for location traces for increased consumer confidence
5. Responsible entities for location traces (Credit-bureau/census, HIPPA++ for responsible parties)



# Outline

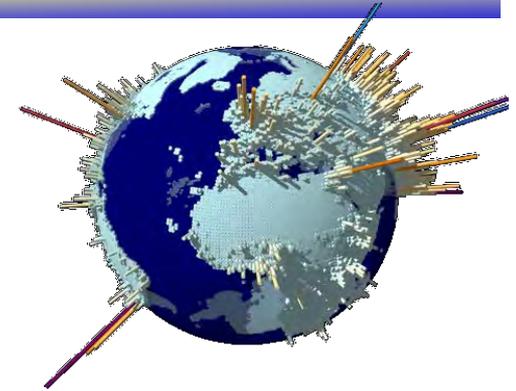


- Introduction
- GPS
- Location Based Services
- Spatial Statistics
- Spatial Database Management Systems
- Virtual Globes & VGI
  - Quilt => Time-travel & Depth
- Geographic Information Systems
- Conclusions

# Virtual Globes & Volunteered Geo-Information

- Virtual Globes

- Visualize Spatial Distributions, Patterns
- Visual drill-down, e.g., fly-through
  - Change viewing angle and position
  - Even with detailed Streetview!



- Volunteered Geo-Information

- Allow citizens to make maps & report
- Coming to public health!
- People's reporting registry (E. Brokovich)
- [www.brockovich.com/the-peoples-reporting-registry-map/](http://www.brockovich.com/the-peoples-reporting-registry-map/)



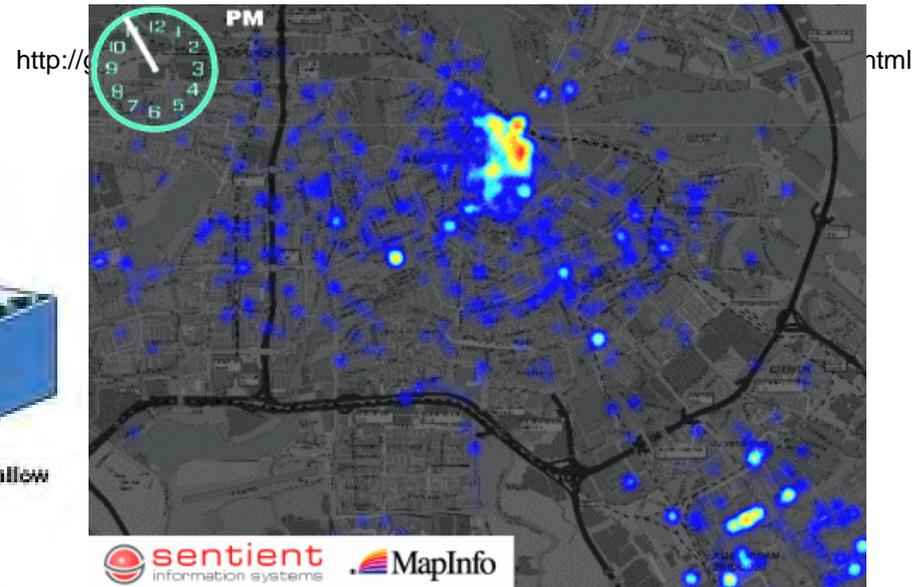
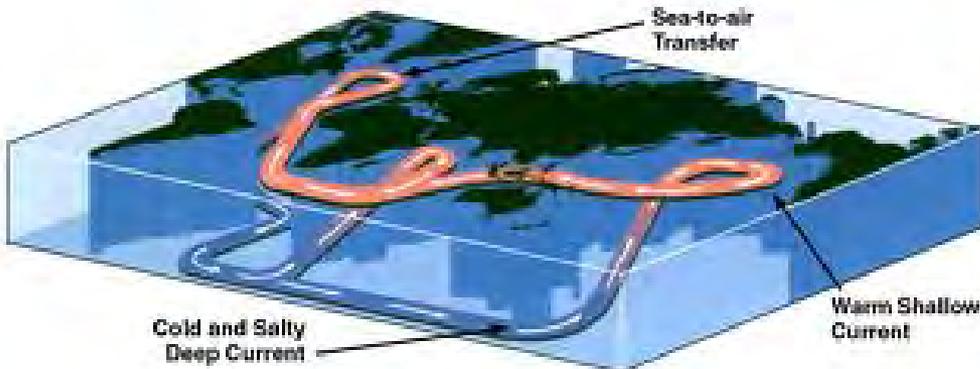
# Virtual Globes in GIS Education

- Coursera MOOC: From GPS and Google Earth to Spatial Computing
  - 21,844 students from 182 countries (Fall 2014)
  - 8 modules, 60 short videos, in-video quizzes, interactive examinations, ...
  - 3 Tracks: curious, concepts, technical
  - Flipped classroom in UMN on-campus course



# Opportunities: Time-Travel and Depth in Virtual Globes

- Virtual globes are snapshots
- How to add time? depth?
  - Ex. Google Earth Engine, NASA NEX
  - Ex. Google Timelapse: 260,000 CPU core-hours for global 29-frame video
- How may one convey provenance, accuracy, age, and data semantics?
- What techniques are needed to integrate and reason about diverse available



# Outline



- Introduction
- GPS
- Location Based Services
- Spatial Statistics
- Spatial Database Management Systems
- Virtual Globes
- **Geographic Information Systems**
  - Geo => Beyond Geo
- Conclusions

# Geographic Information Systems & Geodesy

- **GIS:** An umbrella system to
  - capture, store, manipulate, analyze, manage, and present diverse geo-data.
  - SDBMS, LBS, Spatial Statistics, ...
  - Cartography, Map Projections, Terrain, etc.
  - Q? How to model time? Spatio-temporal?
- **Reference Systems**
  - Which countries in North Korea missile range?
  - 3D Earth surface displayed on 2D plane
  - Spherical coordinates vs. its planar projections
  - Q? What are reference systems for time?



## North Korea's missiles

At least 1,000 of various types, according to South Korea's defense ministry



### Key arsenal

**Taepodong-2** First successful launch December 12, 2012 (Unha-3 rocket based on same system)



**Taepodong-1** Tested 1998 (failed)



**Rodong** Operational



**Scud-C** Operational



**Scud-B** Operational



500 km

1,500 km

300 km

500 km

4,000 km

6,000 km

Source: AFP/KOM/Global Security

AFP



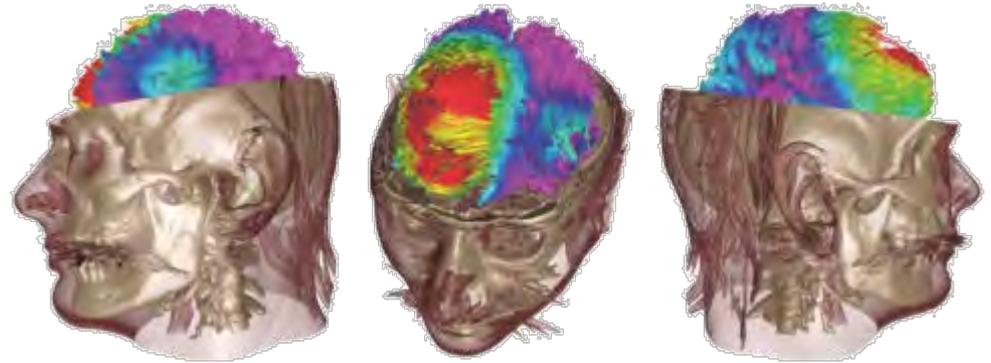
# Opportunities: Beyond Geographic Space

- Spaces other than Earth
  - Challenge: reference frame?
- Ex. Human body
  - What is Reference frame ?
    - Adjust to changes in body
    - For MRIs, X-rays, etc.
  - What map projections?
  - Define path costs and routes to reach a brain tumor ?

Outer Space	Moon, Mars, Venus, Sun, Exoplanets, Stars, Galaxies
Geographic	Terrain, Transportation, Ocean, Mining
Indoors	Inside Buildings, Malls, Airports, Stadiums, Hospitals
Human Body	Arteries/Veins, Brain, Neuromapping, Genome Mapping
Micro / Nano	Silicon Wafers, Materials Science



<http://convergence.ucsb.edu/issue/14>



# Outline



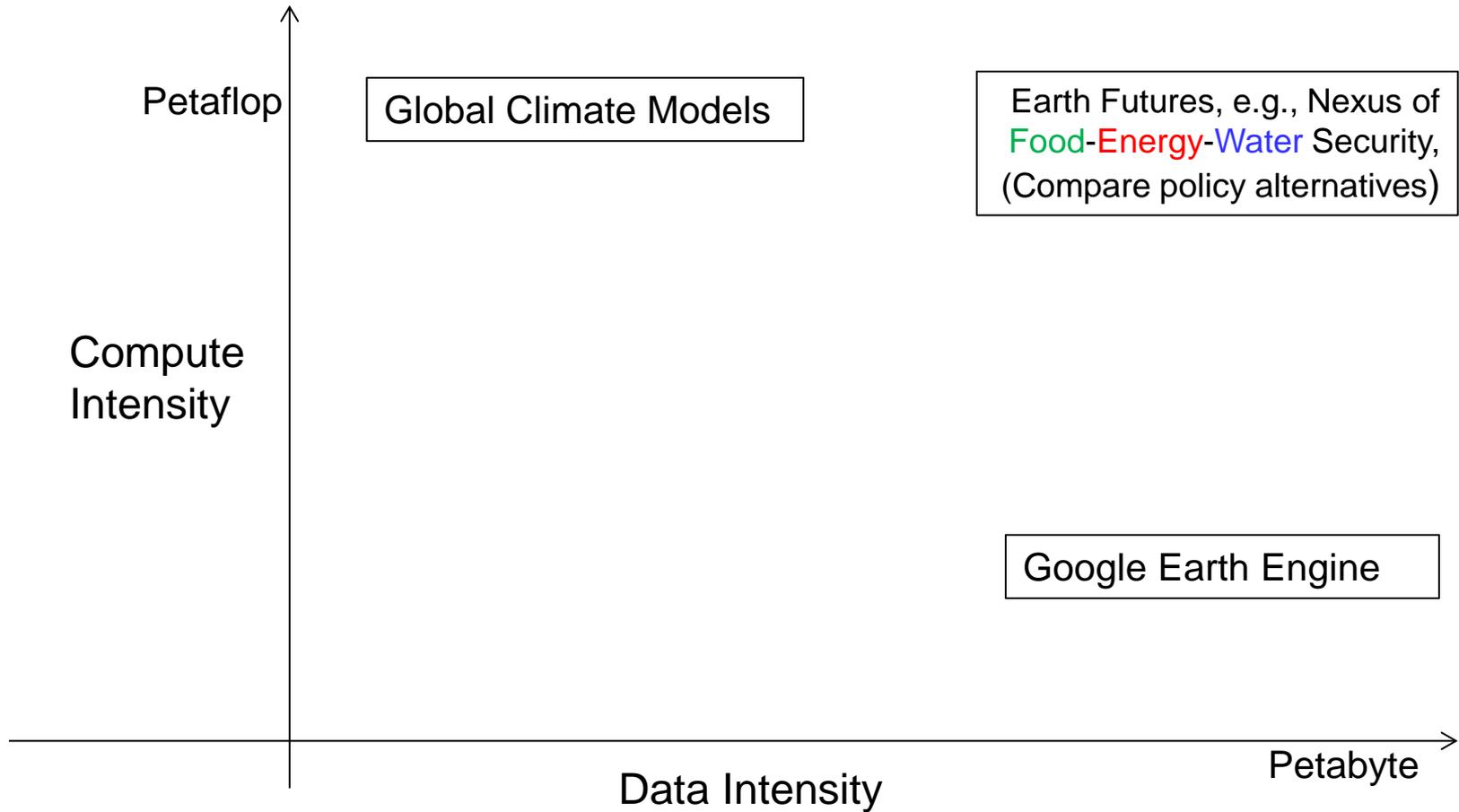
- Introduction
- GPS
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- Geographic Information Systems
- **Conclusions**

# Nexus of Food, Energy, Water Security

- USA:
  - NSF: INFEWS, \$70M in FY16
  - Reports from OSTP, NIC, USDOE, ...
- **Spatial computing is essential**
  - Water census (USGS)
  - Local sourcing, virtual water trade
  - Precision Agriculture, ...
- Excerpt “Satellites provide the big picture”, **Science**, 349(6249):684-685, Aug. 14<sup>th</sup>, 2015.

Satellite observations have revolutionized our understanding of [hydrology](#), [water availability](#), and [global change](#), while catalyzing modern advances in [weather](#), [flood](#), [drought](#), and [fire prediction](#) in ways that would not have occurred with relatively sparse ground-based measurements alone. Earth-observing satellites provide the necessary “big-picture” spatial coverage, as well as the regional-to-global understanding essential for improving predictive models and informing policy-makers, resource managers, and the general public.

# High-Capability Computing & Future of Virtual Globes



# Recommendations

- Spatial Computing has transformed our society
  - It is only a beginning!
  - It promises an astonishing array of opportunities in coming decade
- However, these will not materialize without support
- Universities
  - Institutionalize spatial computing
    - GIS Centers, a la Computing Centers of the 1960's
  - Incorporate spatial thinking in STEM curriculum
    - During K-12, For all college STEM students?
- Government
  - Increase support spatial computing research
  - Larger projects across multiple universities
  - Include spatial computing topics in RFPs
  - Include spatial computing researchers on review panels
  - Consider special review panels for spatial computing proposals



# Panel: Spatio-Temporal (ST) Computing Questions

- 13. **What is missing from ... research agenda?** What can be achieved in ... 5 years?
- 7. What are the major obstacles ... ?
- 6. What are promising data models for managing ST data?
- 8. Is it appropriate to model the temporal domain as 4<sup>th</sup> dimension?
- 4. What are the latest advances in ST computing?
- 14. What is the way to educate the next generation workforce with ST knowledge?

Source: A Critical-Time-Point Approach to All-Start-Time Lagrangian Shortest Paths: A Summary of Results, Proceedings of the Symposium on Spatial and Temporal Databases, Springer LNCS 6849, 2011:74-91. (Complete results accepted for IEEE Transactions on Knowledge and Data Eng.)

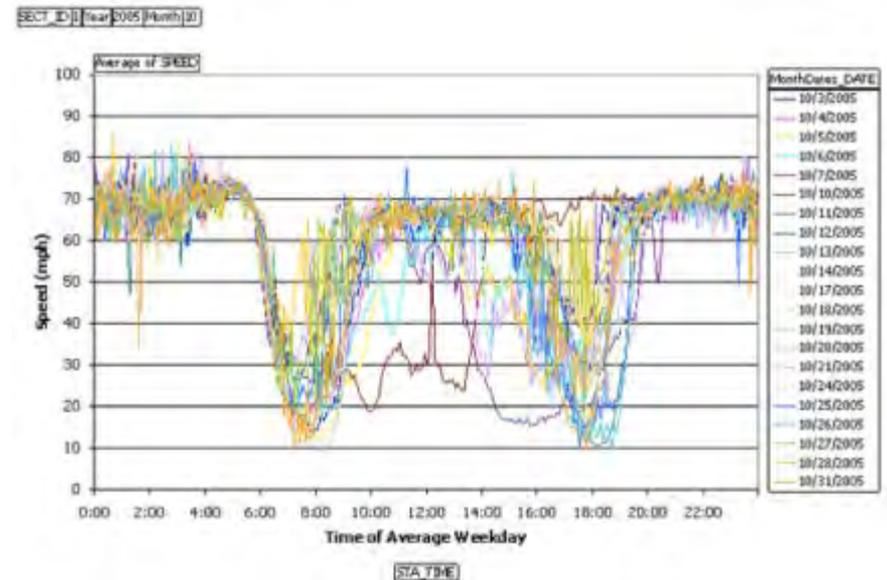
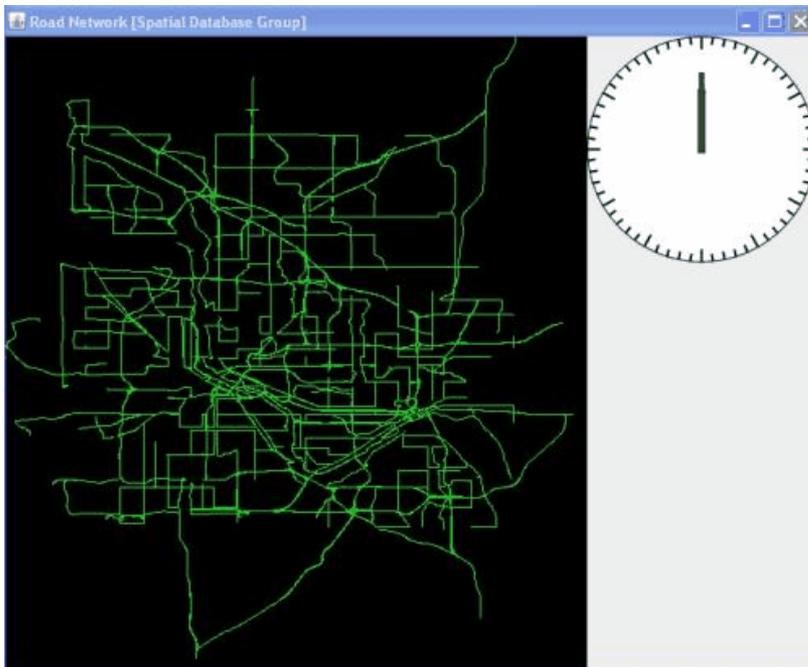
# Dynamic Nature of Transportation Network



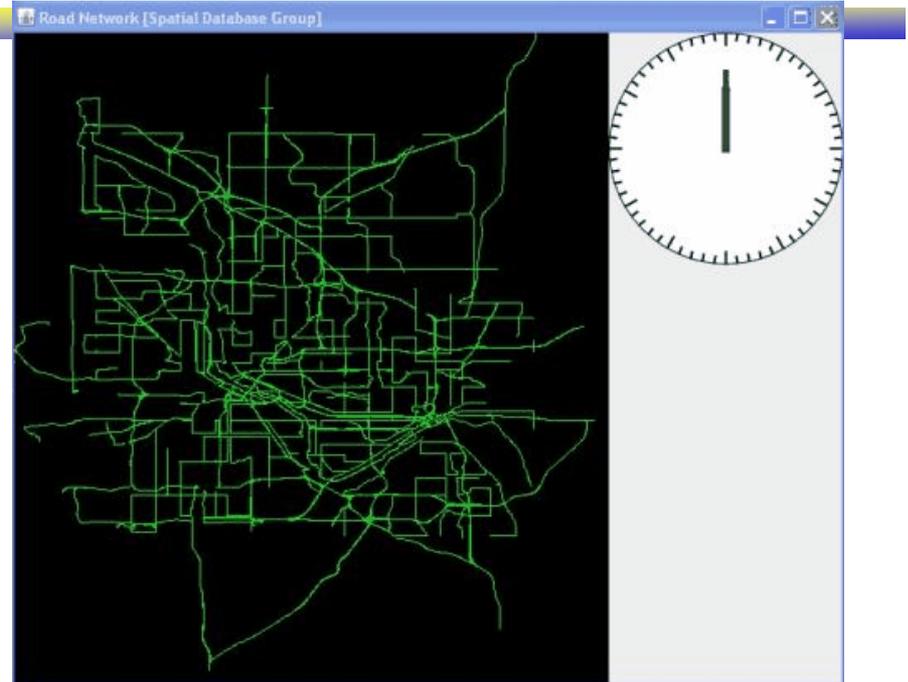
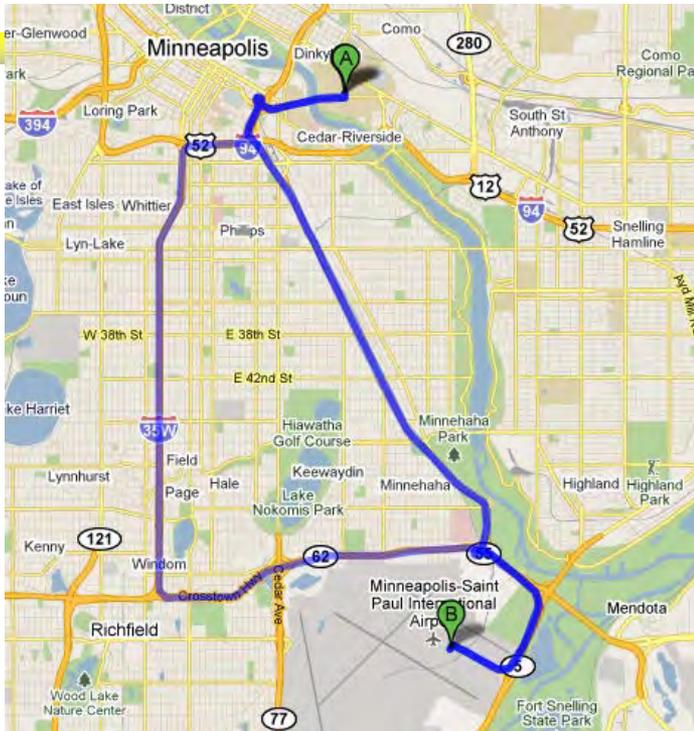
Traffic during non-rush hours



Traffic during Rush hours



# Problem Instance



## INPUT:

- **Source:** University of Minnesota
- **Destination:** MSP Airport
- **Time Interval** 7:00am ---  
12:00noon

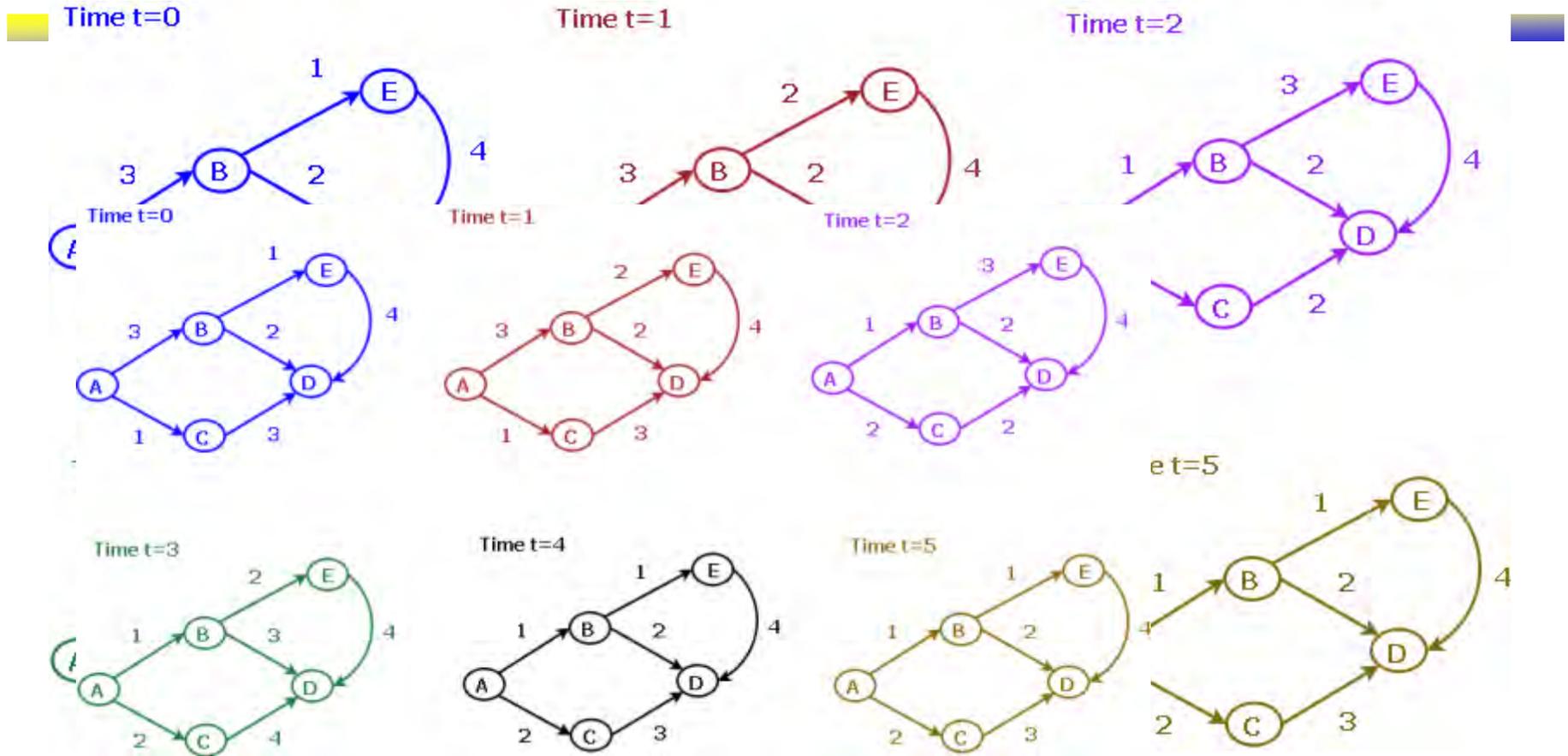
## OUTPUT:

<b>Time</b>	<b>Preferred Routes</b>
<b>7:30am</b>	<b>Via Hiawatha</b>
<b>8:30am</b>	<b>Via Hiawatha</b>
<b>9:30am</b>	<b>via 35W</b>
<b>10:30am</b>	<b>via 35W</b>

# Panel: Spatio-Temporal (ST) Computing Questions

- 13. What is missing from ... research agenda? What can be achieved in ... 5 years?
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# Challenge: Lagrangian Frame of Reference



Q? What is cost of Path  $\langle A, C, D \rangle$  start time=2 ?

➤ Is it 4 or 5 ??

Path	T = 0	T = 1	T = 2	T = 3
$\langle A, C, D \rangle$	4	3	5	4
$\langle A, B, D \rangle$	6	5	4	3

# Challenges

## Non Stationarity ranking of paths

Time	Preferred Routes
7:30am	Via Hiawatha
8:30am	Via Hiawatha
9:30am	via 35W
10:30am	via 35W

➤ Violation of stationary assumption dynamic programming

## Non FIFO Behavior

Time	Route	Flight Time
8:30am	via Detroit	6 hrs 31 mins
9:10am	direct flight	2 hrs 51 mins
11:00am	via Memphis	4 hrs 38mins
11:30am	via Atlanta	6 hrs 28 mins
2:30pm	direct flight	2 hrs 51 mins

\*Flight schedule between Minneapolis and Austin (TX)

➤ Violates the no wait assumption of Dijkstra/A\*

# Panel: Spatio-Temporal (ST) Computing Questions

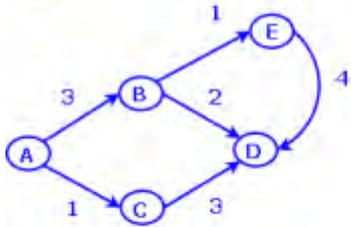


- 13. What is missing from ... research agenda? What can be achieved in ... 5 years?
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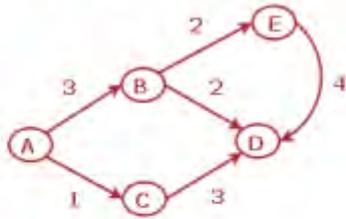
# Naïve Solution (1/2)

## Snapshot Model

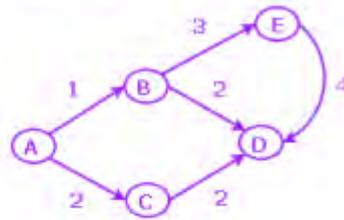
Time t=0



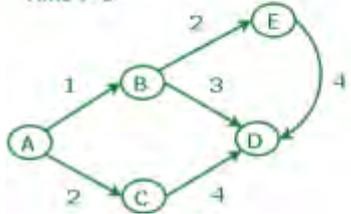
Time t=1



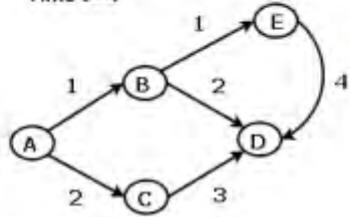
Time t=2



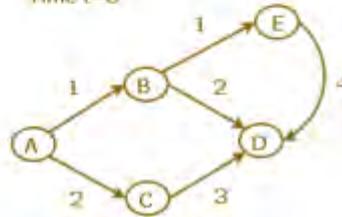
Time t=3



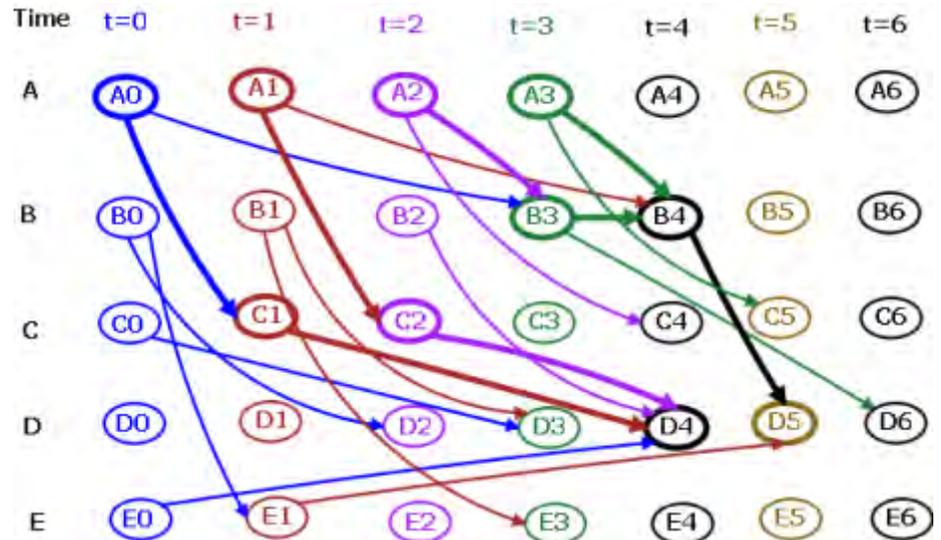
Time t=4



Time t=5



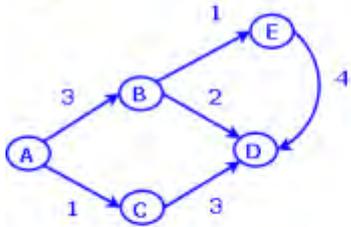
Time expanded Graph (TEG)



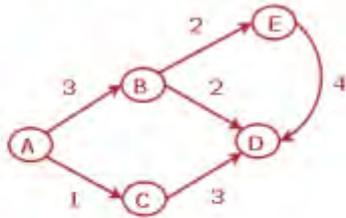
# Naïve Solution (1/2)

## Snapshot Model

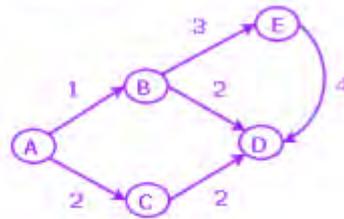
Time t=0



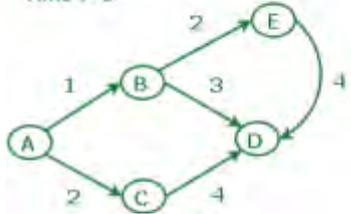
Time t=1



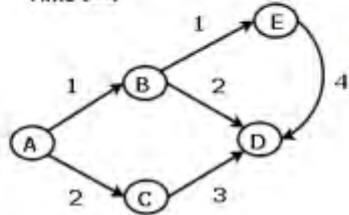
Time t=2



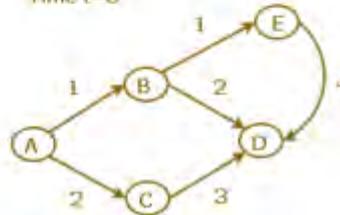
Time t=3



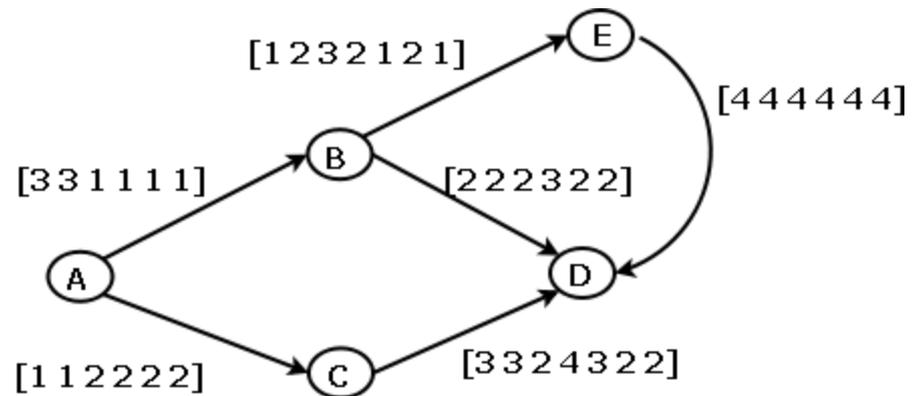
Time t=4



Time t=5



## Time aggregate graph (TAG)



# Panel: Spatio-Temporal (ST) Computing Questions



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# Handling Non-FIFO Behavior (Earliest Arrival Time Series Transformation)

Consider Travel time series

[3 3 1 1]  
t= 0 1 2 3

STEP 1

Add time index to  
travel time to  
get arrival time series

Arrival time series

[3+0 3+1 1+2 1+3]  
t= 0 1 2 3



[3 4 3 4]

STEP 2

Compare each value to the ones  
on its right and choose the min

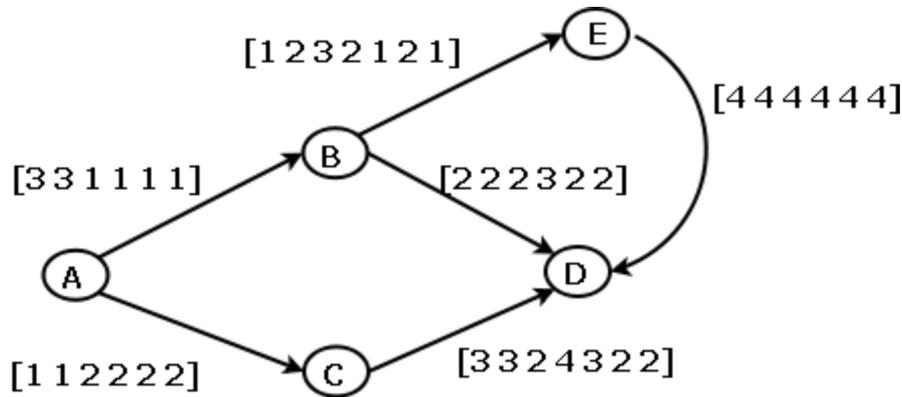
[3 3 3 4]  
Earliest Arrival  
time series

## Observation:

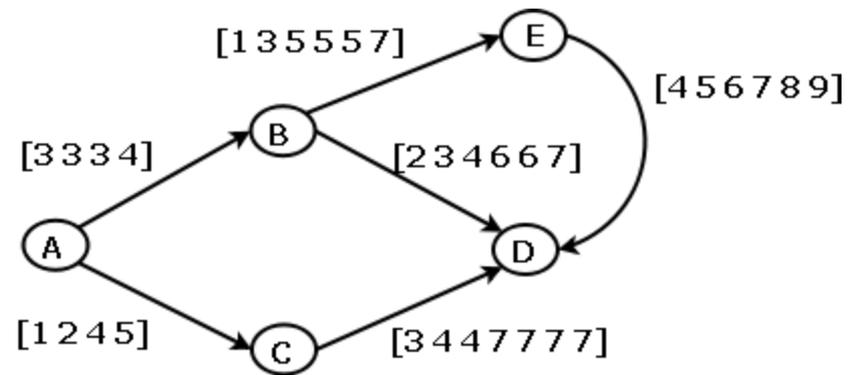
➤ Earliest arrival time series is FIFO in nature.

# Travel-Time vs. Earliest Arrival Time

Travel-Time-Series



Earliest Arrival Time -Series

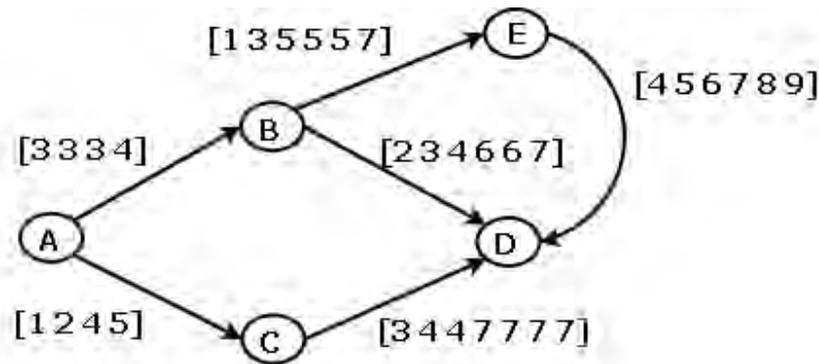


## Observation:

- Earliest arrival time series is FIFO in nature.
- Computing fastest path is easier with earliest arrival time-series
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# Basic concepts for *Critical-time-point*

- **Non-critical times:** Path ranking can't change.
- **Critical-time-points:** Time point where path ranking may change.



Candidate Paths  
between A and D

<A,C,D>

<A,B,D>

<A,B,E,D>

Path Functions for time [0 3]

[4 4 7 7]

[6 6 6 6]

t=2

[9 9 9 9]

The ranking changes at t=2.  
Thus t=2 becomes a  
*Critical time point*

- Path ranking cannot change at *non critical-time-points*.

# Panel: Spatio-Temporal (ST) Computing Questions



- 13. What is missing from ... research agenda? What can be achieved in ... 5 years?
- 7. What are the major obstacles ... ?
- 4. What are the latest advances in ST computing?
- 6. What are promising data models for managing ST data?
- 8. Is it appropriate to model the temporal domain as 4<sup>th</sup> dimension?
- 14. What is the way to educate the next generation workforce with ST knowledge?

# Q13. ST Education

- Tutorials
- Articles in Encyclopedia of GIS
  - Springer's (M. Yuan, K. Stewart, ...) , AAG's
- Survey Papers, book chapters
- Books
  - Choro-chronous (EU project), Moving Object Databases (Guting), Trajectory Processing (Y. Zheng), ...
- Courses
- Degree programs
  - Interdisciplinary graduate programs
  - NSF NRT (previously IGERT)