

## Dynamic Adaptive Disaster Simulation: Developing A Predictive Model of Emergency Behavior Using Cell Phone and GIS Data<sup>1</sup>

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George Mason University, January 31st, 2011

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## Outline



- Introduction
  - Motivation
  - Previous Work
  - Contributions
- Our approach
  - Modeling Process
  - Calibration
- □ Validation, Results, and Discussion
- Conclusions and Future Work

## **Why Model Populations?**

### Hurricane Katrina

- No comprehensive information on population movement
  - 70,000 left in New Orleans
- Resources distributed inefficiently
  - High ground areas (Superdome)
  - "fascinating phenomena"



http://www.nerdylorrin.net/jerry/Katrina/KatrinaSuperdome.html



http://media.myfoxphilly.com/slideshows/katrina/1/lg/Fuel%20station %20damaged%20by%20Hurricane%20Katrina,%20Biloxi,%20Mississippi.htm



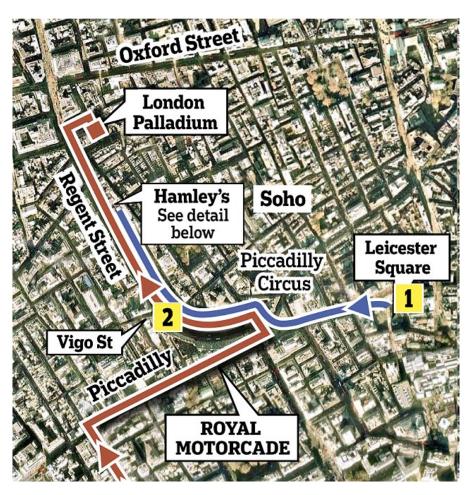
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## Why Model Populations? (cont.)



- London attacks on royal couple
  - Civil disorder was not reliably tracked
    - 100-person mob allowed to disrupt motorcade
    - Alternative routes were possible



http://www.dailymail.co.uk/news/article-1337478/ROYAL-CAR-ATTACK-Blunder-left-Camilla-cowering-hit-ribs-protestors.html

## Existing Methods of Population Modeling

### □ Agent-based modeling

### Flow/continuum-based modeling



http://1.bp.blogspot.com/\_pgrjV7xqqVY/R1mDQIMZqsI/AAAAAAAAAAGM/MPBJzQI6DY4/s400/netlogo.gif



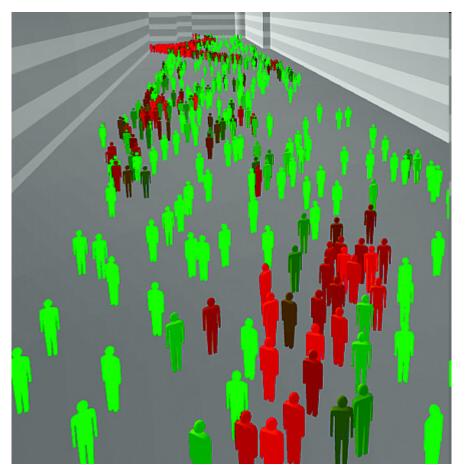
Treuille, A., Cooper, S., and Popovic, Z. (2006). Continuum crowds. ACM Transactions on Graphics, Vol. 25, Issue 3, pp. 1160-1168.

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## **Challenges in Disaster Modeling**





http://www.siemens.com/innovation/pool/de/Publikationen/Zeitschriften\_pof/pof\_herbst\_2009/virt\_real/personenstrom/pof209\_virt\_personen2.jpg

- Restricted to pre-programmed scenarios
- Based on speculations and assumptions
  - > 25-40% difference in predicted evacuation times

Online validation and data incorporation are difficult

### Dynamic Data-Driven Application Systems (DDDAS)

Better for real-time, adaptive predictions (Darema 2006)

## **The WIPER Project**

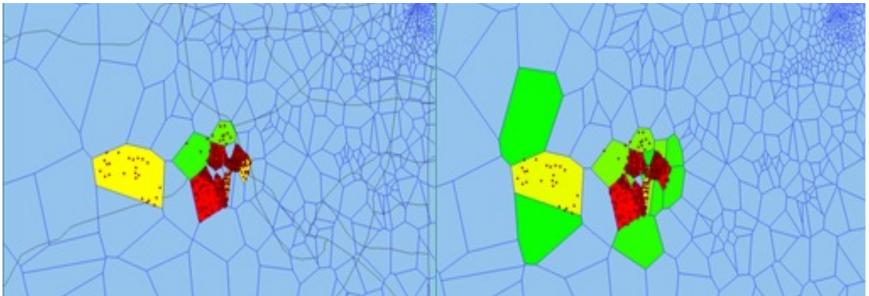


- □ Wireless Phone-based Emergency Response System
  - Cell phones used as dynamic data source

### □ Simulation and Prediction

- Pedestrian and vehicle agents
- Basic movements: flee, flock, jam

□ More work is needed (model complexity, adapt to scenarios)



## Contributions



### Developed Dynamic Adaptive Disaster Simulation (DADS)

Proof-of-concept

### DDDAS concepts

- ♦ Adapts to specific scenarios
- Continuously refines predictions
- Can incorporate data
  - Geographic Information System (GIS)
  - Streaming real-time cell phone location data
  - Tested on synthetic cell phone data
- Netlogo language and modeling environment, version 4.1.1
  - Used GIS extension

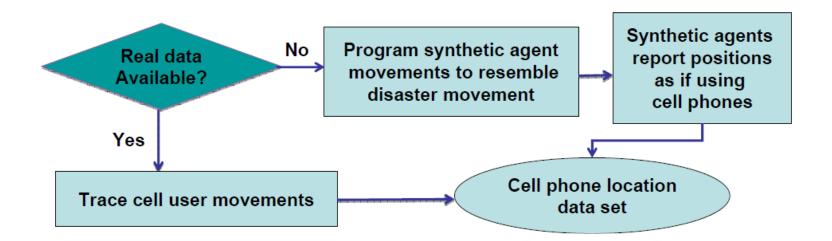
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### **System Architecture**





## **Agent Types**



### Synthetic agents

- Move as people would during a disaster
- Represent real cell phone users
  - ♦ Movements generate synthetic cell phone data
  - Necessary because of nondisclosure agreements (NDA) with cell phone company—I am a minor

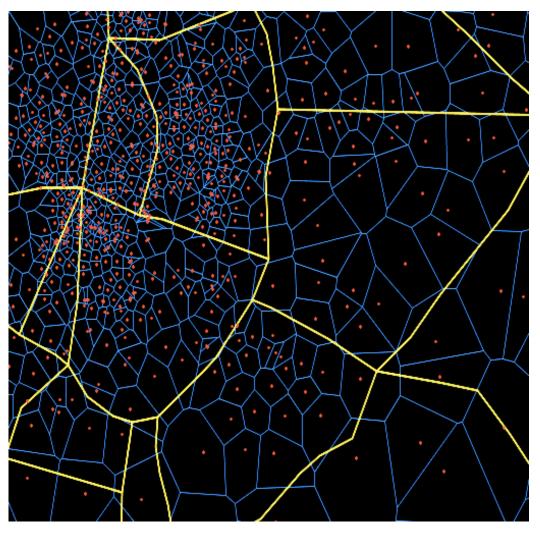
### Predictive agents

- > Move according to inferences drawn from cell phone data
- Represent predictions of future population movement
  - Attempting to adaptively model disaster evacuation
  - Emergency management can be conducted based on agents' predictions

## **Modeling Environment: GIS Space**



### □ Place names removed to maintain anonymity



## **Cell Phone Data**



20070127 | 000400 | @6f19d5 | @fafd42 | 10004 | 20070127 | 000600 | @69a50b | @fafd42 | 10004 | 20070127 | 000600 | @31f919 | @fafd42 | 10004 | 20070127 | 000700 | @570f5c | @fafd42 | 10004 | 20070127 | 000700 | @e940a6 | @fafd42 | 10893 | 20070127 | 000800 | @3e97cd | @fafd42 | 10893 | 20070127 | 000900 | @a620f5 | @fafd42 | 1005 | 20070127 | 000900 | @687ae0 | @fafd42 | 10011 | 20070127 | 001000 | @2297d7 | @fafd42 | 10011 |



http://googlephonetracking.com/

- Networks must be able to constantly track cell phones
  - Call Data Records (CDR)
  - Accuracy varies

# Phone-integrated GPS technology



http://tuberose.com/Graphics/cell%20tower.jpeg

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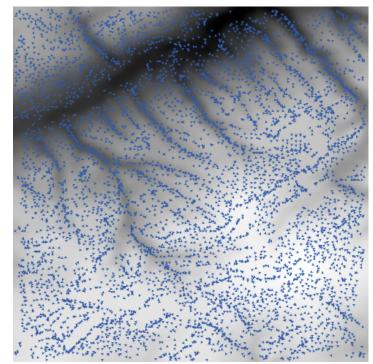
## **Modeling Approach**



□ Dynamic Potential Fields or elevation fields (Park 2009)

- Agents move from high to low potential
- Conceptually portrayed as a terrain of varying elevations
- Used for both synthetic data and DADS itself

□ Use fluid-like agents (Helbing 2002)



Wilensky, U. (2006). NetLogo GIS Gradient Example. Center for Connected Learning and Computer-Based Modeling, Northwestern University, Evanston, IL.

### **Example**

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## **Implementing Modeling Approach**



- □ Elevation field represented as matrix (Wilensky 2006)
  - Each element represents a patch of ground
  - Convolve the matrix with kernels:

 $\begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ -1 & -1 & -1 \end{bmatrix}$ (1) $\begin{bmatrix} -1 & 0 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 1 \end{bmatrix}$ (2)

For each of two gradient matrices:

- Calculate aspect:  $a(x, y) = \arctan(y/x)$
- Done in Netlogo, with GIS Extension

Agents continuously set headings to match aspect of patch

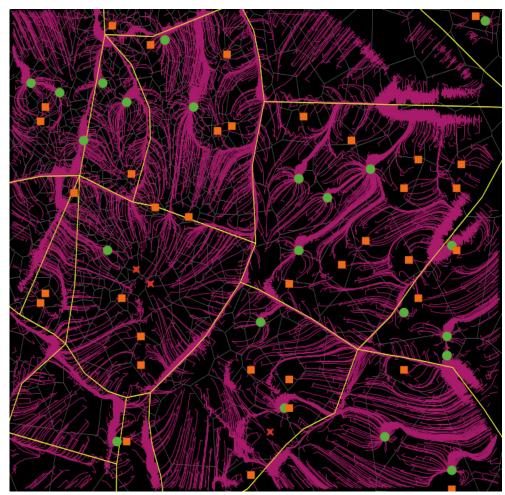
## **Generating Synthetic Data**



### Synthetic elevation field

- > Types of regions in a scenario
  - Disaster (fixed)
  - Dangerous (random)
  - Safe (random)
  - Roads (fixed)
- □ 3200+ synthetic agents
  - Realistic pedestrian speeds
- Random scenarios
  - Example

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Road

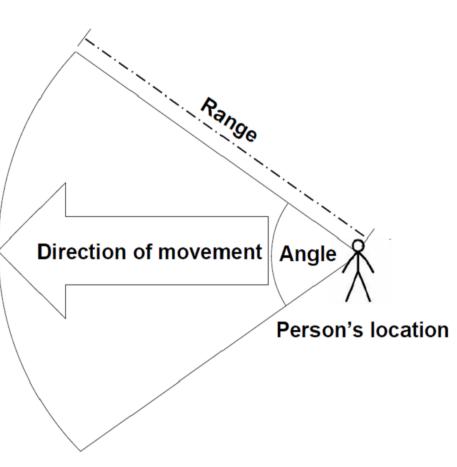
X Disaster Iocation Dangerous

location

Safe

## **Conducting Inference on Data**





### Uses "vision cone" (Torrens 2007)

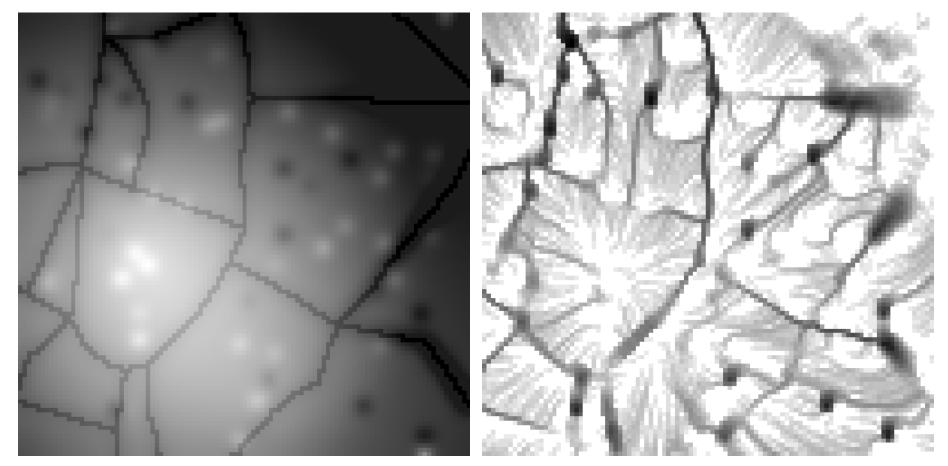
□ Used as "cone of inference"

- Patches inside the cone are inferred to be attractive
  - When a synthetic agent moves, decrease predictive elevation of patches
  - Generate a field of predictive elevations
- DADS predictive agents move on predictive elevation field
  - Represent prediction of future locations of cell phone users

Example

## **Conducting Inference on Data (cont.)**

- Problem becomes that of "reconstructing" a reasonable predictive elevation field
  - > Must accurately capture factors influencing movement



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## **Summary of Methods**

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- □ Generate synthetic elevation field
  - Synthetic agents move on it to produce synthetic data
- □ Conduct inference as location data streams in
  - Generate predictive elevation field
- Predictive agents move on predictive field
  Represent predictions of population movement
  Example

## **Measuring Simulation Quality**

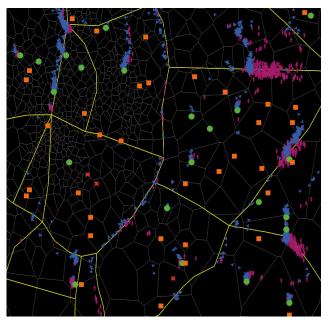


□ Manhattan distance metric (Schoenharl 2008)

Compare synthetic vs. predictive agents

•  $p_i$  and  $q_i$  represent numbers of each agent type at each cell tower

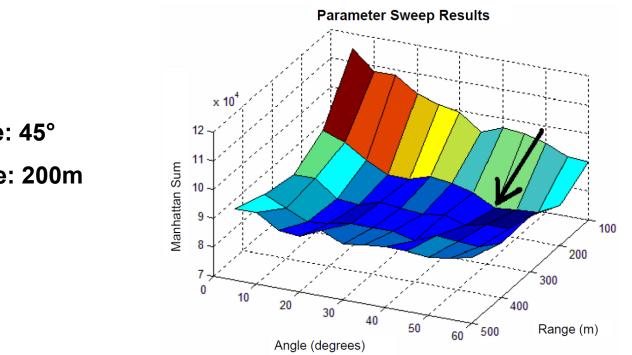
□ Smaller Manhattan distance = more accurate simulation



## **Experimental Setup and Results**



- □ Identified optimal values for vision cone angle and range
- □ Multi-resolution approach
  - Coarse, then finer parameter sweeps
  - Compared predictions of all possible parameter pairs
    - Evaluated in three random scenarios



Best angle: 45°

Best range: 200m

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## Validation



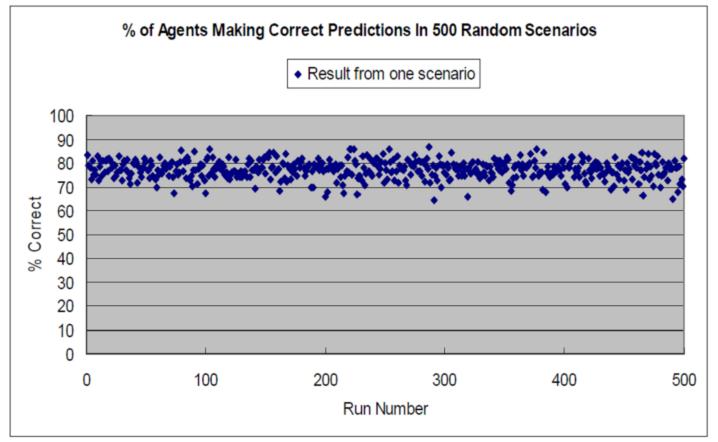
### Verifies that a model "is a reasonably accurate representation of the real world" (Xiang et al. 2005)

- Internal Validation
  - Measures stability
- Predictive Validation
  - Measures predictive accuracy
- Other Validation Methods
  - Online validation is an "open research question" (Schoenharl 2007)

### **Internal Validation**



- □ 500 runs; measured final predictions (75 minutes in advance)
  - Different randomly generated scenario each time

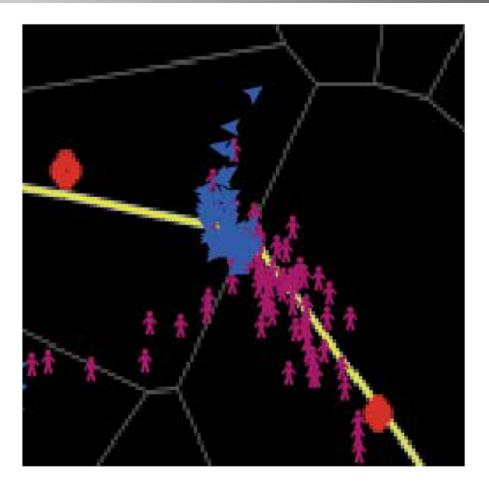


### Mean correct percentage: 77.30%; standard deviation: 3.87%

## **Predictive Validation**

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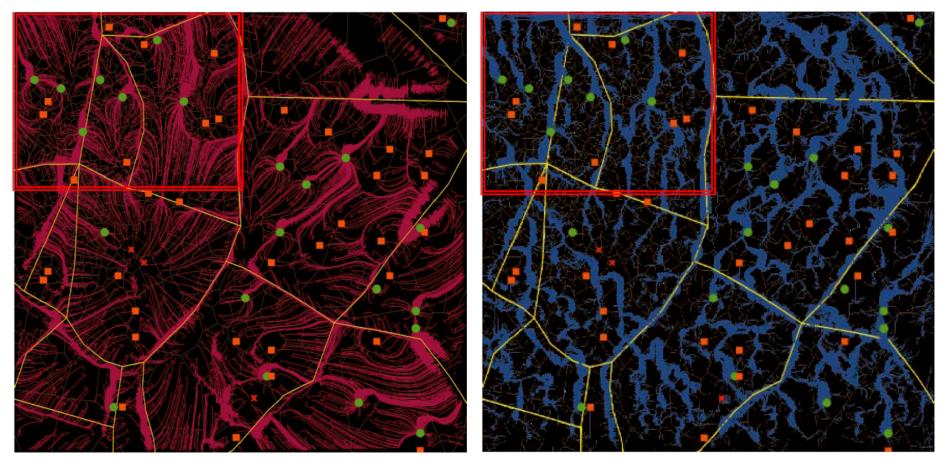
- An average of 77.30% of predictive agents made correct predictions, 75 minutes in advance
- Disadvantage of quantitative validation
  - Predictions are only correct if in the correct serving cell
- Qualitative validation is necessary



## **Qualitative Predictive Validation**



### □ Compare paths taken by synthetic/predictive agents



### **Synthetic**

### **Predictive**

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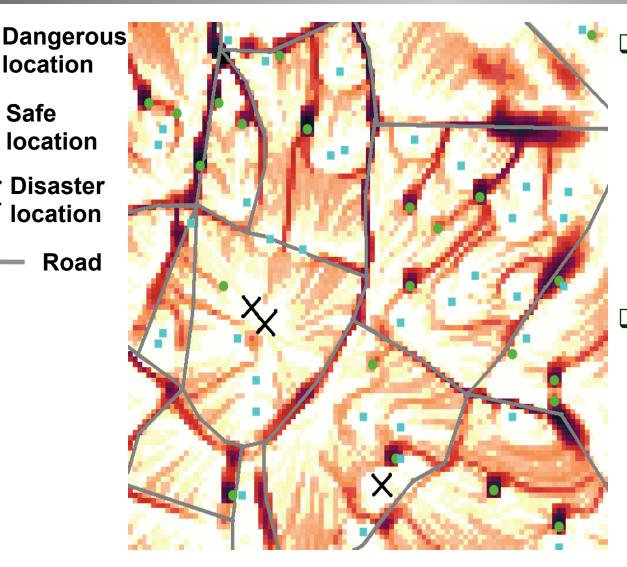
## **Qualitative Predictive Validation**



location Safe location

Disaster location

Road



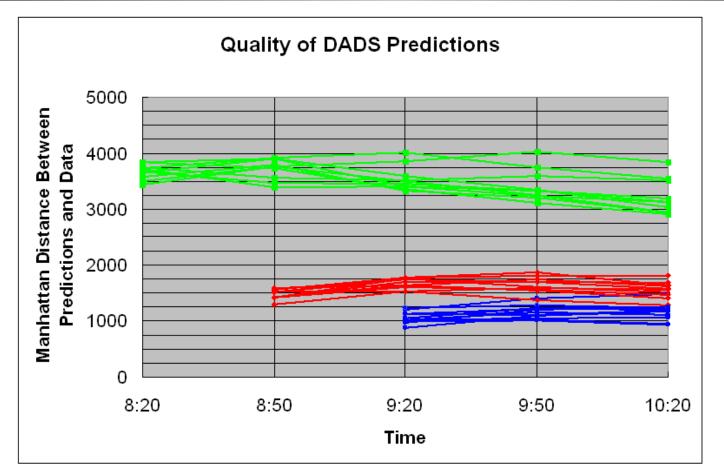
□ Shapes of attractive and repulsive regions

- > Can indicate disaster type
  - Can use to refine simulation

Future Work

□ Road networks > Usable for traffic simulations

## **Ability to Improve Predictions**



□ Predictions of population locations at 10:20 a.m.

Green lines: at 8:05, red lines: at 8:35, blue lines: at 9:05

#### □ Reflects the DDDAS concept of dynamically updating simulations

### **Discussion**



- Assumptions— "a model is only as good as the assumptions on which it is based"
  - Homogeneous agents
  - No crowd dynamics
  - Synthetic data
  - No restrictions on agent vision or movement



http://blog.creativecurator.com/wp-content/uploads/2010/05/cctv-fire.jpg

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## **Summary**



# □ DADS uses streaming cell phone location data to simulate and predict population movement in disasters

- Makes use of emergent intelligence
- Can analyze historical data
  - Study tool for past disasters
- GPS will further increase utility
- Demonstrates DDDAS
  - > Adapts to specific scenarios and constantly improves

### □ Validated on synthetic data

- Predictively and internally valid
- Provides useful inferences in situations like Katrina
  - Helpful in evacuations, even if disaster disables cell service

## **Important Considerations**



- Distinguish evacuees from responders
  - Potentially misleading data
  - Heterogeneous agents?
- □ Cell phones as a data source
  - Limited power supply
    - Could be problematic in the long term
  - Cell towers vulnerable
    - Earthquakes
    - Volcanic ash
  - GPS technology
  - How quickly can we draw accurate inferences?



http://www.wirelessestimator.com/t\_content.cfm?pagename=Hurricane%20lke%20telecom



http://chicagoist.com/2006/10/25/south\_loop\_building\_fire.php

### Future Work—DADS Itself



- □ Test on real cell phone location data
  - > Allow for adjustment of data reception
    - DDDAS concept—sensor adjustment
- □ Further assess modeling techniques
  - Increased realism
    - Agent heterogeneity
    - Crowd dynamics
- □ More sophisticated methods of parameterization
- **Explore more ways to use cell phone data** 
  - Examine call volume, distribution, location, etc.

## **Future Work—Population Simulation**



### Large-scale

- Modeling citywide or global movement patterns in other situations
- □ Small-scale
  - Modeling individual behavior
  - Depicting movement and/or evacuation in a building

# Tool for study as well as prediction



http://thecityfix.com/files/2010/06/public\_transport\_rome.jpg



http://cdn.wn.com/pd/cd/97/038669f9ba7cf8fcc73da99f5699\_grande.jpg

### **Future Work—Adaptive Simulation**





- Simulations designed to adapt to streaming data
- □ Modeling landslides in China
  - Caused by dams, mining, and deforestation
- Better sensor networks enable this sort of technology
  - DADS is an example

http://globalvoicesonline.org/2010/08/09/china-zhouqu-landslide-a-man-made-disaster/



- "The Tradeoff of Confidentiality and Access" (NRC 2007)
  - Must sacrifice precision for privacy
    - WIPER—aggregated by voronoi cell
  - How much precision is needed?
    - ♦ DADS appears to require precision
    - "Naïve realism"

### Possible solutions

- > Opt-out (Johnston 2010) or opt-in?
  - How many cell phones are needed?
- Aggregation and other tactics
- "Data enclaves" and legal or licensing systems



## Thank you!

### **Questions?**

### A paper describing this research has been accepted for SpringSim/ADS 2011, and can be found at the following URL: http://www.nd.edu/~dddas/Papers/papers.html

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