# Reverse geocoding and implications for geospatial privacy

Paul Zandbergen
Department of Geography
University of New Mexico

#### Outline

- Geospatial privacy
- Geocoding / reverse geocoding
- Experimental design
- Results and Conclusion

## **Geospatial Privacy**

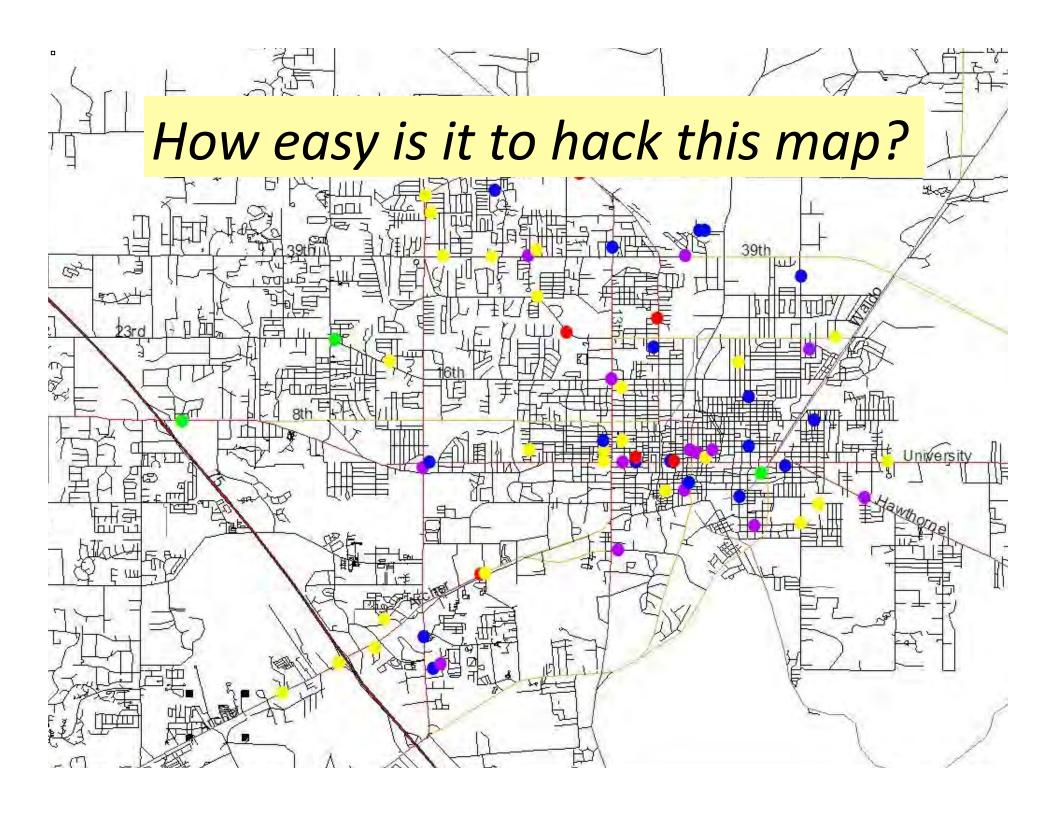
 Under increasing threat from new technologies, high resolution data and easy-to-use tools





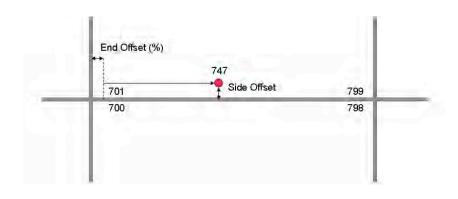


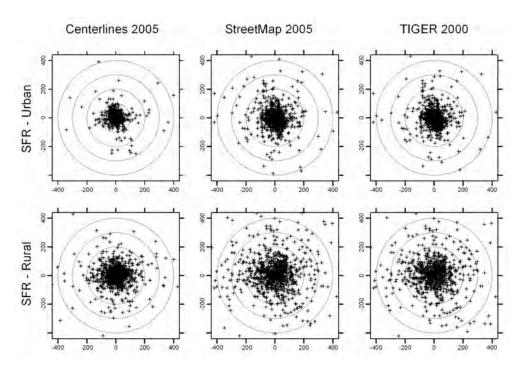




## Geocoding

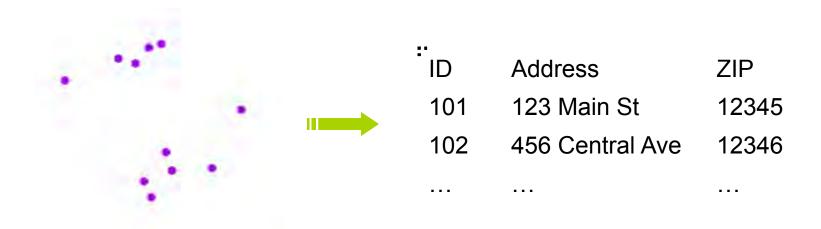
- Widely employed
- Well understood
- Substantial errors





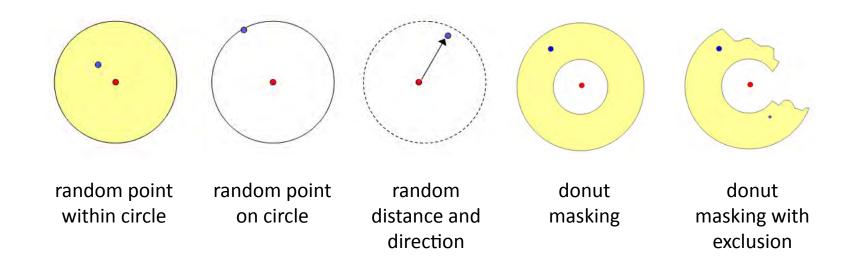
#### Tool of a hacker: Reverse geocoding

- Geocoding in reverse
- Relative easy, relatively new
- Key tool for "hacking" published maps
- Not well understood

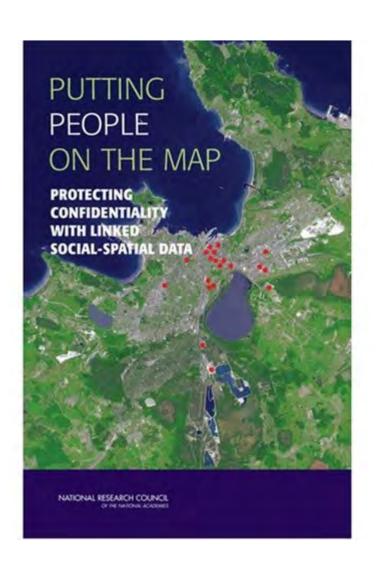


#### How to protect spatial confidentiality?

- Geographic masking
- But how to do this most effectively?
- Need better understanding of reverse geocoding



#### Review of the State of the Art



Panel of confidentiality issues arising from the integration of remotely sensed and self-identifying data

"No known technical strategy [....] for managing linked spatial-social data adequately resolves conflicts among the objectives of data linkage, open access, data quality, and confidentiality protection across datasets and across uses." (Conclusion 3)

National Research Council, 2007

#### Research Questions

What are the capabilities of reverse geocoding to identify individuals from published locations?

How does this vary with the methods employed for geocoding and reverse geocoding?

How does this vary with population density?

## **Experimental Design**

- Actual building locations with known addresses
  - Travis County, TX (Austin)
  - Sample of 2,500 residential locations
  - Stratified across 5 population density classes
- Geocode using 5 different geocoders
- Reverse geocode using 3 different techniques
- Determine accuracy of reverse geocoding

## **Address Points**



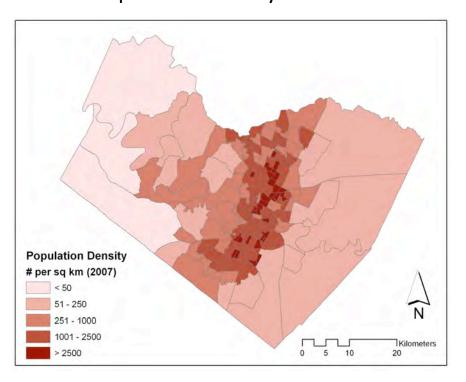


#### Geocoders

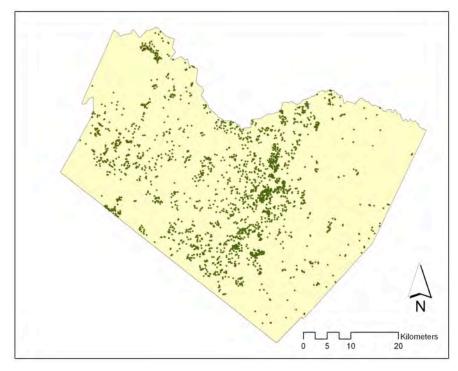
- TeleAtlas Address Points (commercial)
- TeleAtlas Streets (commercial)
- Google Maps (free API)
- StreetMap USA Pro 2007 in ArcGIS
- Geolytics 2007 (using TIGER 2007data)

## Study Area – Travis County, TX

#### **Population Density Zones**



#### Sample of residential address points



## Reverse Geocoding & Accuracy

- Original residential address points
  - Snap to nearest residential building
- TeleAtlas reverse street geocoding
  - Submit for commercial processing
- Google Maps reverse geocoding
  - Free API
- Accuracy of reverse matches
  - 1. Perfect match (street name and number)
  - 2. Close match (number within 10)
  - 3. Same street only

# Results – Match Rates (%)

Geocoding Technique	Population density (people/km²)					
	< 50	50 to 250	250 to 1000	1000 to 2500	> 2500	Total
TeleAtlas AP	43.6	70.2	91.8	92.8	93.6	78.4
TeleAtlas Street	93.2	92.8	98.0	99.8	96.8	96.1
Google Maps	92.2	95.6	98.2	99.0	96.2	96.2
StreetMap Pro	81.2	83.6	95.8	99.0	95.8	91.1
Geolytics	77.0	80.2	92.4	96.2	90.2	87.2
Combined	31.4	59.0	86.0	89.6	86.8	70.6

#### Results – Same Street Matches

		Reverse Geocoding Technique			
		Austin AP	Google Maps	TeleAtlas Street	
	Austin AP	100.0	96.7	90.1	
000	TeleAtlas AP	99.5	92.0	89.6	
	Google Maps	99.2	32.0	90.5	
Tech	TeleAtlas Street	88.2	92.5	99.5	
nique	StreetMap Pro	89.1	76.8	82.5	
	Geolytics	54.5	54.7	56.9	

#### Results – Close Reverse Matches

		Reverse Geocoding Technique			
		Austin AP	Google Maps	TeleAtlas Street	
Geocoding Technique	Austin AP	100.0	95.5	23.1	
	TeleAtlas AP	99.1	91.8	24.3	
	Google Maps	98.8	28.6	24.3	
	TeleAtlas Street	69.8	63.2	92.5	
	StreetMap Pro	60.8	42.9	44.0	
	Geolytics	33.8	27.4	29.3	

#### Results – Perfect Reverse Matches

		Reverse Geocoding Technique			
		Austin AP	Google Maps	TeleAtlas Street	
	Austin AP	100.0	94.2	9.1	
000	TeleAtlas AP	97.9	91.8	9.0	
	Google Maps	97.2	27.8	9.1	
Tech	TeleAtlas Street	18.7	9.4	55.0	
nique	StreetMap Pro	17.0	7.3	8.2	
	Geolytics	7.1	2.3	2.6	

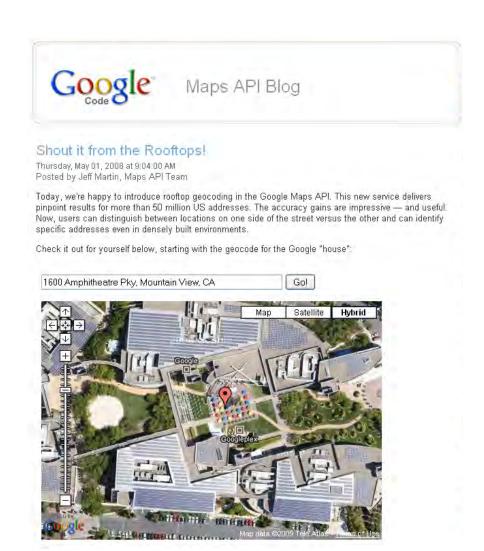
#### Effect of Population Density – Percent Perfect Matches

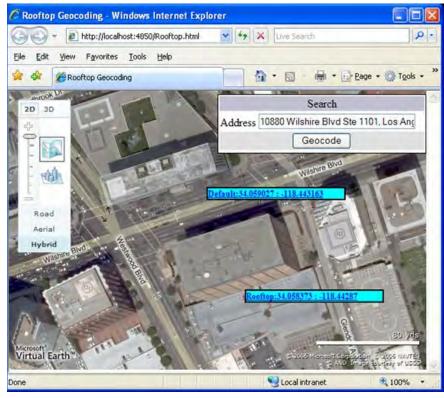
Geocoding	Reverse	Population density (people/km²)				
		< 50	50 to 250	250 to 1000	1000 to 2500	> 2500
StreetMap Pro	Austin AP	22.9	16.9	15.3	16.1	17.3
Geolytics	Austin AP	14.0	7.1	7.0	5.4	6.5
Austin AP	TeleAtlas Street	2.5	6.8	11.6	11.4	8.3
TeleAtlas Street	TeleAtlas Street	51.6	63.7	56.7	53.8	49.8
Google Maps	TeleAtlas Street	4.5	6.4	12.3	11.2	7.4
TeleAtlas Street	Google Maps	9.6	9.8	8.8	9.6	9.4
Geolytics	Google Maps	1.9	2.0	2.6	2.5	2.3

## Results Summary

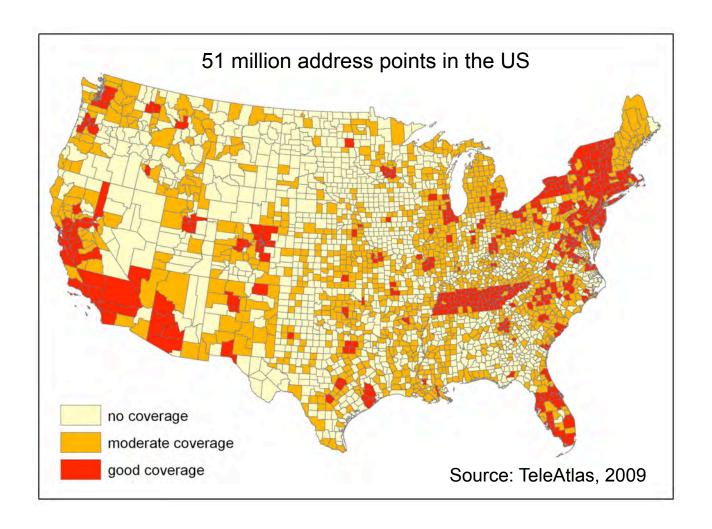
- Accuracy of reverse geocoding varies greatly
- Building level (reverse) geocoding is typically most accurate
- Street geocoding is quite noisy
  - Easy to get the right street
  - Very few perfect matches
- Accuracy is substantially improved if knowledge of the original geocoding technique is available!
- No clear pattern with population density

#### Rooftop Geocoding in Google Maps and Virtual Earth





#### Commercial Address Points - TeleAtlas



Licensed to Google, Virtual Earth, ArcGIS Business Analyst, Pitney Bowes / Group 1 / MapInfo

## Reverse Geocoding

Reverse Geocoding Example using the GReverseGeocder v1.0.7

The example below uses the getPlacemarkProperty function to get the "PostalCode" Copyright 2008, Nico Goeminne nicogoeminne at gmail.com



Reverse geocoding now supported in Google Maps and Microsoft Virtual Earth

Also supported in latest version of ArcGIS – requires some customization or ArcWeb services

Numerous free easy-to-use online utilities

#### Conclusions

- Accuracy of reverse geocoding
  - Varies greatly with geocoder / reverse geocoder combination
  - Between 2% and 98% perfect reverse matches
- Knowledge of original geocoding method is critical
  - noisy results from street geocoding can be reverse coded
- Trends:
  - Address points are the new standard in geocoding
  - Reverse geocoding is relatively easy
- Techniques to protect privacy may need to assume a worstcase scenario: very high resolution address data

#### **Future Research**

- Replicate in other study areas
- Examine urban/rural gradients more closely
- Experiment with different masking techniques
- Develop a framework for spatial  $\kappa$ -anonymity

## Acknowledgements

- National Science Foundation
- UNM Research Allocation Committee
- American Civil Liberties Union