

# Quantifying Semantic Uncertainties of Volunteered Geographic Information to Understand Human Travel Behaviors

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## Outline

- Motivation
- Methodology
- Experiment
- Discussion

Conventional Data Source:

Internal resource →

Household Travel Survey

- Income
- Household size
- # of daily trips



#### **Long-Distance TRAVEL LOG**

Name of person completing this log: Last Name: Your person number: (Person #s are on the Travel Diary label) Travel Day: Travel Period\*:

OP1 OP2 OP3 OP4 OP5 OP6 OP7 OP8 O No one in my household made a long-distance trip in the eight weeks prior to our travel day. If this is the case, please fill in the bubble above and return this Log with your completed Diaries.

#### INSTRUCTIONS

- Record details about all long-distance trips made by any household member during the travel period shown on the label.
- A long-distance trip is a trip made to a location 50 miles away or more from your home.
- Record each way (away from home and returning home) as a separate trip.
- If you made more than 8 long-distance trips, please record the details on a separate piece of paper.

How do I provide my Long-Distance Travel Log Information?

- Online: Enter your information at www.catravelsurvey.com. Use PIN# on the label
- Mail: Return with your completed travel diaries.
- Phone: We will call you to collect your Log and Travel Diary information, Or, you can call us at the toll free hotine number below.

Questions? Call the tell-free hotline at 1-877-261-4621

#### Lists A and B are on the back! -

Trip Departure DATE (Locations 50 miles away or more)	WHERE were you when you STARTED this trip?	WHERE did you travel TO? (Your final destination)	MAIN PURPOSE of trip the LIST A CODES	HOW MANY OTHER PEOPLE were traveling with you? (Excluding yourself)	What METHOD OF TRAVEL was used for the longest distance? UseLIST B CODES
Trip 1: Most Recent	Place Name:	Place Name:		# of people traveling with you (excluding yourself):	
Date:	Address or Nearest Cross-streets:	Address or Neurest Cross-streets:	List ONE code only	# of <u>household members</u> (excluding yourself): Which <u>household members</u> traveled?	List ONE code only
no day yr	City:State/ZP/Country:	City: State/ZIP/Country:		(use person & from diary label)  O P1 O P2 O P3 O P4 O P5 O P6 O P7 O P8	Ramember to recon
Trip 2	Place Name:	Place Name:		If of people traveling with you (excluding yourself):      If of <u>household members</u> , (excluding yourself):      Which household members traveled?      (up person A B from day (abed)      P1    P2    P3    P4    P5    P6    P7    P8	List ONE code only
Date:	Address or Neurest Cross-streets:	Address or Nearest Cross-streets:	List ONE code only		
	City:State/ZP/Country:	City: State/ZIP/Country:			
Trip 3	Place Name:	Place Name:		If of people traveling with you (excluding yourself):	List ONE code only
Date:	Address or Nearest Cross-streets:	Address or Neurest Cross-streets:	List ONE code only	# of <u>household members</u> (michaling yourself):  Which <u>household members</u> traveled? (use person #s from diary label)  P1	
	City:State/ZP/Country:	City: State/ZIP/Country:			
Trip 4	Place Name:	Place Name:		f people traveling with you (excluding yourself):	
Date:	Address or Neurest Cross-streets: Address or Neurest Cross-streets:		List ONE code only	# of <u>household members</u> (recluding yourself): Which <u>household members</u> traveled? (use person #s from dary label)	List ONE code only
	City: State/ZIP/Country:	City:State/ZP/Country:		OPIOP2 OP3 OP4 OP5 OP6 OP7 OP8	

#### LIST A CODES - TRIP PURPOSE

- 2 Business (work-related meeting / convention / seminar) 3 Combined business and pleasure
- 4 School-related activity
- 5 Visit friends / relatives / wedding / funeral
- 6 Medical
- 7 Vacation / Sightseeing
- 8 Outdoor recreation (sports, fishing, hunting, camping, boating, etc.)
- 9 Entertainment (theater, concert, sports event, gambling, etc.)
- 10 Personal business (e.g., shopping)
- 11 Drive someone else
- 12 Return home
- 97 Other (write code 97 and specify)

- NON-MOTORIZED: 1 Walk
- Bilder
- Wheelchair / Mobility Scooter
- 4 Other Non-Motorized (skateboard, etc.)

#### PRIVATE VEHICLE:

- 5 Auto / Van / Truck Driver
- 6 Auto / Van / Truck Passenger
- 7 Carpool / Vanpool
- 8 Motorcycle / Scooter / Moped

#### PRIVATE TRANSIT

9 Taxi / Hired Car / Limo 10 Rental Car / Vehicle

- LIST B CODES METHOD OF TRAVEL
- 12 Greyhound Bus
- 13 Airplane 14 Other Private Transit

#### PUBLIC TRANSIT:

- 15 Local Bus, Rapid Bus 16 Express Bus / Commuter Bus (AC Transbox
- Golden Gate Transit. etc.)
- 17 Promium Bus (Metro Orange / Silver Line)
- 18 School Bus
- 19 Public Transit Shuttle (DASH, Emery Go-Round, etc.)
- 20 AirBART / LAX FlyAway

- 11 Private Shuttle (SuperShuttle, employer, hotel, etc.) 21 Dial-A-Ride / ParaTransit (Access Services, etc.)
  - 22 Amtrak Bus 23 Other Bus
  - Rail / Subway
  - 24 BART, Metro Red / Purple Line 25 ACE Amtrak Caltrain Coaster Metrolink
  - 26 Metro Blue / Green / Gold Line. Muni Metro. Sacramento Light Rail, San Diego Sprinter / Trolley / Orange / Blue / Green, VTA Light Rail
  - 27 Street Car / Cable Car 28 Other Rail

  - 29 Ferry / Boat

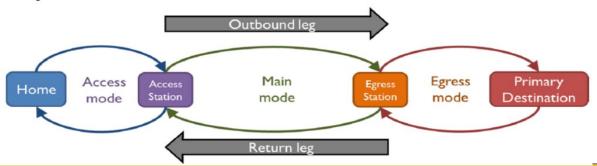






### However:

- Internal resource (e.g., household travel survey) provides data about the main outbound leg and return leg but not the access and egress portion of a tour;
- It also does not provide information about opportunities for activity participation at the access and egress stations, home location and primary destination.









How about external contexts of travels?

- Understanding of the place/neighborhood (originations, destinations, or intermediate stops)
- Data source: Volunteered Geographic Information (VGI)



















- Check-in numbers; number of reviews; ratings → Popularity
- Place Types → Functionality/Human's perception





How to quantify semantics?

BIKE RACE, REDDING, CA, -122.359935, 40.569064

HAPUNA BEACH PRINCE HOTEL, KAMUELA, HI, -155.821209, 20.002854





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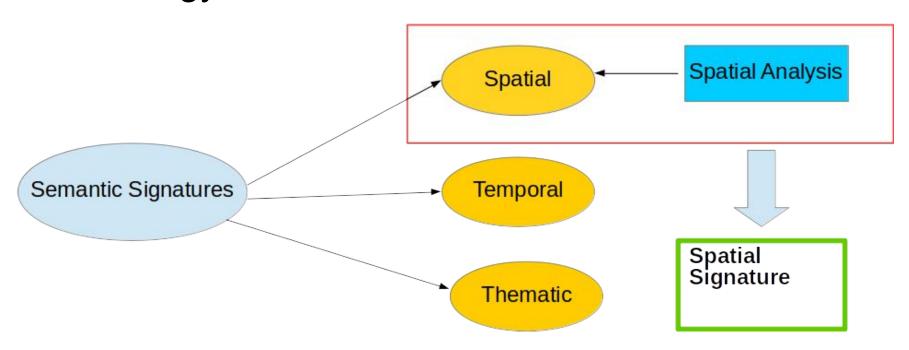
Rui Zhu

Rui Zhu









Janowicz, K., McKenzie, G., Hu, Y., Zhu, R., and Gao, So. (2018): <u>Using Semantic Signatures for Social Sensing in Urban Environments</u>. Mobility Patterns, Big Data and Transport Analytics.







## **Spatial Signatures**

- **Spatial structure** of the data belonging to a place type is used to quantify its semantics.
- Spatial statistics are applied to describe such spatial structure.
- Spatial point patterns. Spatial autocorrelation analysis, spatial interaction analysis with other geographic features, place-based analysis. → 41 statistics





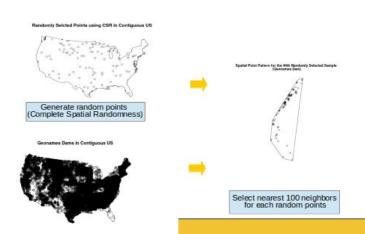


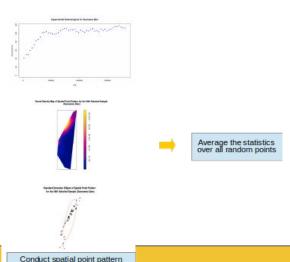
Spatial Signature -- Spatial point patterns

Intensity-based: local intensity, kernel density estimation

Distance-based: nearest-neighbor distance, Ripley's K, and standard

deviational analysis





analysis on these 100 neighbors







Spatial Signature -- Spatial point patterns -- Examples

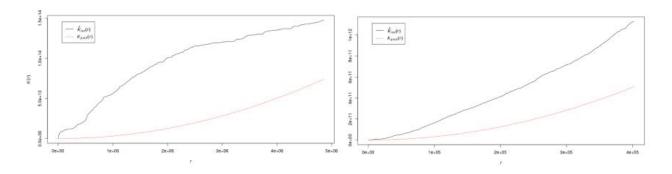


Figure 1: Ripley's K of Park (left) and Dam (right) from DBpedia Places.

Statistics: mean and std. of the deviation between theoretical can observed K curves



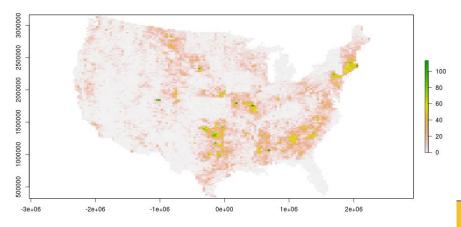




Spatial Signature -- Spatial AutoCorrelation Analysis

- Moran's I: how intensities of cells differ from their neighbors
- Semivarigram: measure the variation of cell intensities in a specific distance lag class.

Dams in GeoNames



Cell size: 36 km \* 22.2 km

Cell value: number of instances falling in the cell







Spatial Signature -- Spatial point patterns -- Examples

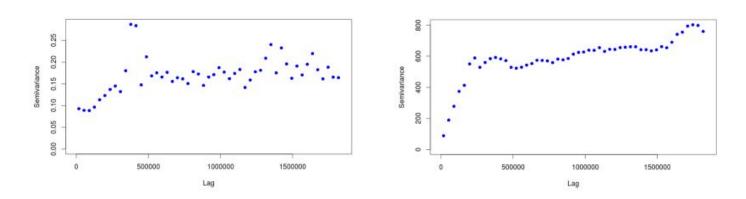


Figure 2: Experimental semivariogram of Park (left) and Dam (right) from TGN.

Statistics: mean and std. of the semivariance at first, median and last lag distance



Spatial Signature -- Spatial Interaction with Other Geographic features

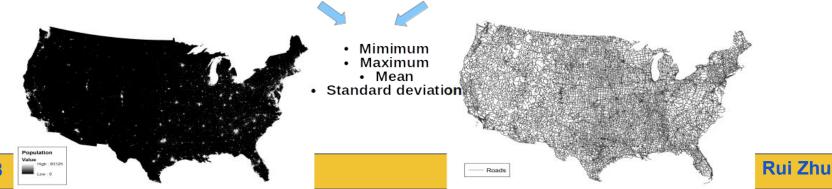
- Population
- Climate
- Road network

Population (LandScan2014)

Population for each feature point

Road Segment (Digital Chart of the World)

Distance to nearest segment for each feature point









Spatial Signature -- Spatial Interaction with Other Geographic features -- Examples

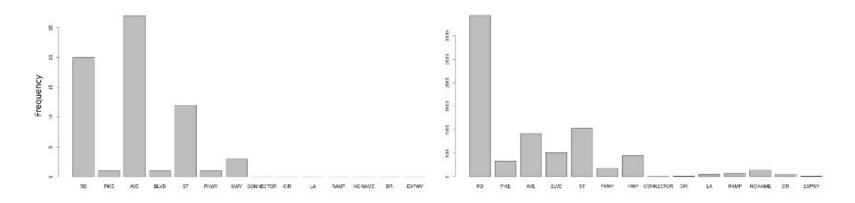


Figure 3: Histogram of road types for Amusement Park (left) and Restaurant from Google.







Spatial Signature -- Place-based statistics

In contrast to spatial statistics, they focus more on describing the *topological* and *hierarchical relations* between places.

- The number (and entropy) of distinct states (or counties) a place type occurs in;
- The number (and entropy) of adjacent states (or counties) that also contain features of the target type;







Spatial Signature -- Place-based statistics -- Examples

• Distinguishing feature types such as Glacier (which occur in eight US-states according to DBpedia) and River (which occur in all states).





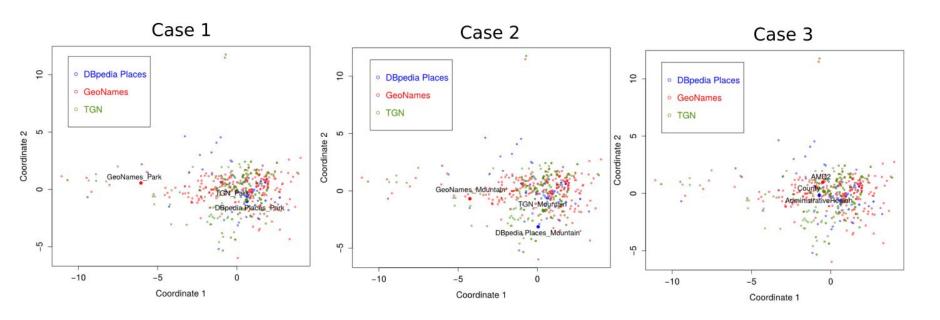
Spatial Point Pattern		Spatial Autocorrelations	Spatial Interaction with Other Geographic Features		Place-based statistics	
	Intensity			min	Number of distinct states (or counties)	
Local	Mean distance to nearest neighbor		Population	max		
	std. of distance to nearest neighbor	Global Moran's I		mean	Entropy of states (or counties)	
	Kernel density (range)		std.		Number of adjacent	
	Kernel density (bandwidth)			min of shortest distance	states (or counties) that have the same feature type	
	$Ripley's \ K \ (range)$		Road Network	max of shortest distance		
	Ripley's K (mean deviation)			mean of shortest distance	Number of distinct feature types for nearest neighbor	
	std. ellipse (rotation)	Semivariogram		std. of shortest distance		
	std. ellipse (std. along x-axis)	(first distance lag)		entropy of nearest road types  Entropy of feat types for		COMPANY OF THE PROPERTY OF THE PARKS OF THE
	std. ellipse (std. along y-axis)			$mean\\precipitation$	nearest neighbor	
Global	Intensity	Semivariogram (median distance lag)		$std. \\ precipitation$		Mean KL Divergence of the topic distribtion
			Climate	mean temperature max	LDA-based	
				std. temperature max	approach	
	Kernel density			mean temperature min		
	(range)	Semivariogram		std. temperature min		Entropy of
	Kernel density	(last distance lag)		mean water vapor pressure		the topic distribution
	(bandwidth)			std. water vapor pressure		

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# Experiment

## 1. Similarity of place types



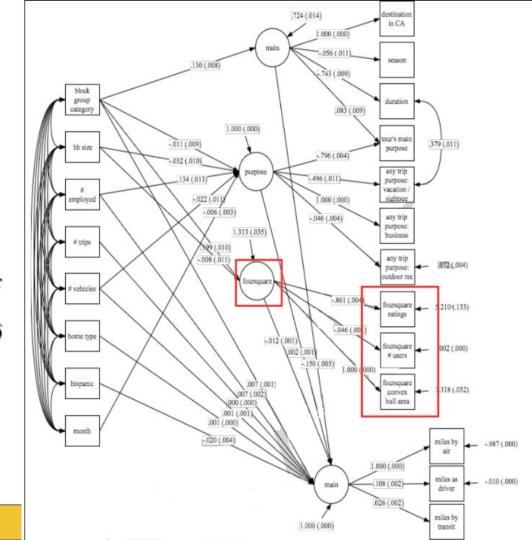
# Experiment

- 2. Modeling of the LD pattern
  - Structural Equation Model

Measurement model for y:  $y = \Lambda_y \eta + \varepsilon$ 

Measurement model for x:  $x = \Lambda_x \xi + \delta$ 

Structural model:  $\eta = B\eta + \Gamma\xi + \zeta$ 









# Summary

- External information could be incorporated to understand human travel behaviors;
- Volunteered geographic information has potentials to provide such external information;
- Quantifying the semantic uncertainty of VGI help us to understand places/neighborhoods.







## Thanks a lot!

Any questions / comments?