DATA-INTENSIVE GEOSPATIAL SEMANTICS

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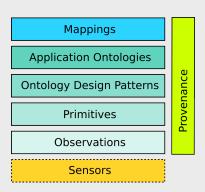
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THE THREE V'S OF BIG GEO-DATA

- Volume: The size of the involved data, their multi-dimensional nature, as well as their inter-linkage which creates a global graph. E.g., Volunteered Geographic Information, Location-based Social Networks, sensor networks, high resolution remote sensing data, complex transportation simulations.
- Variety: The number of heterogeneous sources and type of data is increasing as well. Combining social media with authoritative sources and integrating different formats such as video, audio, photo, and text allows a more holistic analysis but raises new issues in data integration.
- Velocity: Big Data is not only about large amounts of data but also the **speed** at which data is **created** and **updated**. A rapidly increasing number of data sources deliver near real-time data which poses new challenges for stream reasoning and rule systems, which data do we **keep**?

ONTOLOGICAL FRAMEWORK TO SUPPORT VARIETY



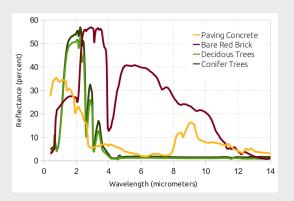
- Observation-driven ontology engineering to foster semantic heterogeneity
- If ontologies are too lightweight to restrict meaning, we can still anchor them in observations (and thus provide provenance)
- May require a symbol grounding level for observation procedures
- Ontologies should be about communication and not about replacing numerical models; please do not try to develop an universal ontology for rivers, mountains, forests, and so forth

SAMPLING BIG GEO-DATA



- Spatial data: 20,765/3,247,409 POI and 64/71 types (OpenStreetMap)
- Temporal data: 440,939 check-ins by 35,745 users to 150,300 POI of 408 different types (WhrrI); from more than 3 million check-ins per day
- **Thematic data**: 218,760 geo-referenced Wikipedia articles and 287,210 geo-referenced travel blog entries.

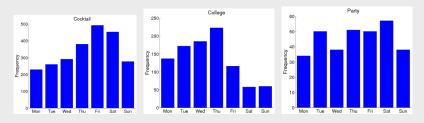
INTRODUCING SEMANTIC SIGNATURES



- Combine numerical (statistical) models and data with ontologies to derive local (personal) primitives (reifications)
- Analogy to Spectral Signatures used in Remote Sensing
- Multiple spectral bands → multiple semantic bands

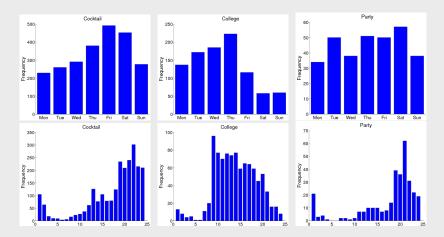
SEMANTIC SIGNATURES – TEMPORAL BANDS

■ When you are is what you are



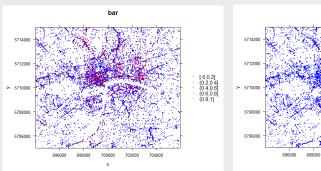
- Locations types and log-in patterns from the Location-based Social Network Whrrl
- Day-band from Semantic Signatures
- Local Reifications (Primitives): e.g., Weekend vs. Workday
- We used them to automatically compute missing types of Whrrl POIs

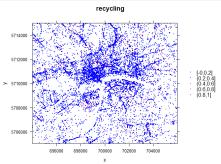
SEMANTIC SIGNATURES – DAY AND HOUR-BASED BANDS



Multiple semantic bands may be required to distinguish between feature types. We can add a hour-band in addition to the day-band

SEMANTIC SIGNATURES – SPATIAL-SEMANTIC BAND

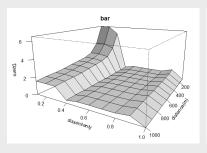


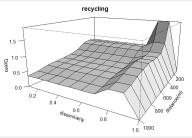


- POIs plotted by **similarity** to bar and recycling in OSM data, London, UK
- Local Reifications (Primitives): e.g., Uniform and Clumped
- Bars (and similar features) tend to clump together
- Recycling (and similar features) are rather uniformly distributed

SEMANTIC SIGNATURES – SPATIAL-SEMANTIC BAND

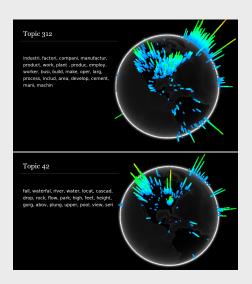
■ Where you are is what you are





- *D_{zero}* **measures** the **likeliness** of features of a certain **type** to co-occur within a specific **semantic and spatial range**.
- User support: generate **recommendations**, and **clean up** data based on **type likelihood**. 'How likely is a recycling center directly next to an existing one?'

SEMANTIC SIGNATURES – THEMATIC BAND

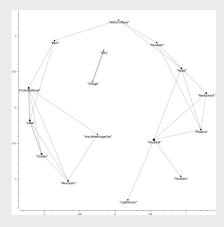


- A thematic band can be computed out of unstructured text using latent Dirichlet allocation (LDA); data source Wikipedia and Travelblogs
- Non-georeferenced plain text is often still geo-indicative
- Different types (taken from DBPedia) of geographic features have different,
 diagnostic topics associated to them (out of 500 topics)

SEMANTIC SIGNATURES – THEMATIC BAND







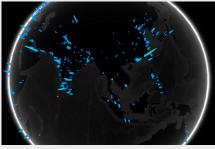
■ City topics: 204>450>104>282>267>497>443>484>277>97>...

Town topics: 425>450>419>367>104>429>266>69>204>308>...

Mountain topics: 27>110>5>172>208>459>232>398>453>183>...

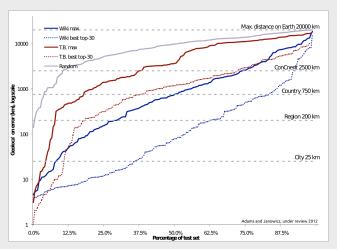
DIALOC - NEW DATA SOURCES FOR OLD QUESTIONS





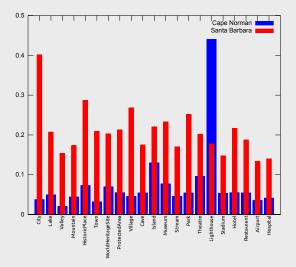
- Use **plain text**, not low-level image analysis as data source to estimate geographic locations from images
- 'market food street narrow dense populated asia economy air conditioning smog fog humid warm building construction skyscrapers skyline shipping export channel harbor transportation tram city advertisement'

DIALOC - LOCATION ESTIMATION



In about 70% of all queries, DiaLoc excludes 99.9% of the land-surface of the Earth.

DIALOC - GEOGRAPHIC FEATURE TYPES



■ Semantic signatures used to infer which **type of feature** is described.

SUMMARY AND LESSONS LEARNED

- Big (Geo)-Data requires small, local theories (microtheories)
- Develop geo-ontology design patterns, not domain ontologies
- Developed patterns in a community process (GeoVoCamp)
- Mine and learn ontological primitives from observation data
- Semantic signatures as one methodology to learn primitives
- Make the domain expert the **knowledge engineer**
- Late assignment of classes (e.g., River)
- Ontology alignment and mapping to connect local theories
- What is the **80-20** rule of geospatial semantics?
- Cross-media validation using different semantic bands