A Need-Finding Study with Users of Geospatial Data **CHI23**

Parker Ziegler



Sarah E. Chasins

EPIC Data Lab Retreat • April 17, 2023

Topography of Mt. Tamalpais, Marin County, USA





Check out the CHI'23 paper





Ok, but hold up, Parker. What is geospatial data? (And why study how domain experts work with it?)



Background Geospatial Data

Earth's surface.

Geospatial data describes the location and attributes of phenomena on the



Background Geospatial Data

👈 🛛 💽 A Need-Finding Study with User × 🛛 +

attributes

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23.924	5,073	21,205	Blount County, Alab	0500000US01009	4
26.188	898	3,429	Bullock County, Alab	0500000US01011	5
26.726	1,777	6,649	Butler County, Alaba	0500000US01013	6
29.619	13,202	44,572	Calhoun County, Ala	0500000US01015	7
32.757	4,449	13,582	Chambers County, A	0500000US01017	8
21.77	2,359	10,836	Cherokee County, Al	0500000US01019	9
24.866	4,262	17,140	Chilton County, Alab	0500000US01021	0
18.386	980	5,330	Choctaw County, Ala	0500000US01023	11
27.48	2,562	9,323	Clarke County, Alaba	0500000US01025	2
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22.622	1,320	5,835	Cleburne County, Al	0500000US01029	4
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24.817	7,394	29,794	Elmore County, Alab	0500000US01051	25
33.547	4,338	12,931	Escambia County, Al	0500000US01053	26

location





Background Geospatial Data

Geospatial data is everywhere today.







Satellite Imagery

Environmental Sensor Networks

OpenStreetMap



Domain Experts and Geospatial Data

Earth and Climate Science







Data Journalism



Domain Experts and Geospatial Data



Earth and Climate Science



Social Sciences



Data Journalism

← → C O A https:// Department of Geography Franklin College of Arts and Science UNIVERSITY OF GEORGIA

单 🏾 🍨 CyanoKhoj-India

CyanoKhoj | India

This dashboard is designed for quick analysis of CyanoHABs and Water Quality Assessment using Sentinel-3 imagery for select Indian waterbodies.

1) Select Waterbod

•••

Ukai_Dam_Gujarat 🌲

2) Select Cloud Mask And Non-Water Area Flag

✓ Mask Cloud Cover and Non-Water Are
 2018-09-01
 2018-09-30

✓ Filter Map to center
Apply Filter

3) Select an image (dated)

S3A_20180918T045719_20180918T050019 ≑

4) Select Visualisation

Cyanobacteria_Cell_Density 🌲

CCD Map from Chl-a absorption and phycocyanin absorption





Domain Experts and Geospatial Data



Earth and **Climate Science**



Social Sciences



Data Journalism

🛑 🛑 🛑 🐞 🥌 CyanoKhoj-India

CyanoKhoj | India

This dashboard is designed for quick analysis of CyanoHABs and Water Quality Assessment using Sentinel-3 imagery for select Indian waterbodies.

1) Select Waterbody

Ukai_Dam_Gujarat

2) Select Cloud Mask And Non-Water Area Flag Mask Cloud Cover and Non-Water Are

2018-09-01 2018-09-30

Filter Map to center

Apply Filter

3) Select an image (dated

S3A_20180918T045719_20180918T050019

4) Select Visualisatio

Cyanobacteria_Cell_Density 4

CCD Map from Chl-a absorption and phycocyanin



Barriers to working with geospatial data are high.



Barriers to working with geospatial data are high.



Example

Geographic Information Systems

• Require significant background in geospatial data theory



Cartography



Databases

Statistics

HCI research^{1, 2, 3} has shown that GISs are especially difficult for non-geographers to learn and use.

> 1. Traynor, C. and Williams, M.G. Why are geographic information systems hard to use? *Conference* Companion on Human Factors in Computing Systems (1995).

2. Traynor, C. & Williams, M. G. End users and GIS: a demonstration is worth a thousand words. in Your wish is my command: programming by example 115–134 (Morgan Kaufmann Publishers Inc., 2001). 3. Haklay, M. (Muki) & Skarlatidou, A. Human-Computer Interaction and Geospatial Technologies – Context. in Interacting with Geospatial Technologies 1–18 (John Wiley & Sons, Ltd, 2010). doi:10.1002/9780470689813.ch1.



Barriers to working with geospatial data are high.

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Example

Jupyter Notebooks

Programming Systems Geospatial programming abstractions are increasingly common in Python, R, and JavaScript



mapbox

 Must develop proficiency with programming languages and environments

geopandas



Research has yet to explore the specific obstacles **domain experts** face in their work with geospatial data.



GIS Usability

Computational Notebooks



Design Software

mapbox

Geospatial Analysis and Visualization Libraries



Analysis Visualization

Data Discovery Data Transformation Analysis Representation



Contribution

The goal of this research is to **identify the computing needs of domain expert geospatial data users**.



Roadmap



Roadmap



15

We conducted a contextual inquiry study with 25 participants.



Earth and Climate Science







Data Journalism



Interdisciplinary





We conducted a contextual inquiry study with 25 participants.



Number of Participants



Study Design

Session Structure and Analysis

• 50–70 minute open-task observations

- Followed by **semi-structured** post-interviews
- Inductive **thematic** analysis



Roadmap



Roadmap



1. Study Design

• 4. Introducing cartokit

Findings

We identified 12 challenges across five phases of participants' work with geospatial data.

Data Discovery

Solving Geospatial Data Constraints

Data Transformation

Aligning Geospatial Datasets

Topological Errors

Reducing Resolution to Improve Performance

Data Subsetting and Caching

Analysis

Identifying Geospatial Operators Understanding Geospatial Operator Semantics Visibility of Geometry in Programming Environments

Analysis Representation

Reproducing Geospatial Analyses

Creating Informal Program Representations

Visualization

Sketching Cartographic Variants

Geospatial Information in Design Software



preprocessing.



Reprojection

.

Resampling

Participants needed to transform datasets to a shared spatial and temporal reference for analysis, but alignment required complex



Clipping





Temporal Aggregation



PE2's Task. Develop a model to predict groundwater withdrawal.

Spatial Resolution



MOD16





USDA-NASS

500m

4638.3m



Temporal Interval

Geographic Extent



Global

Monthly

Conterminous U.S.



Conterminous U.S.



PE2's Task. Develop a model to predict groundwater withdrawal.

Spatial Resolution



Temporal Interval

Geographic Extent



Global

Monthly

Conterminous U.S.



Conterminous U.S.



PE2's Task. Develop a model to predict groundwater withdrawal.

Spatial Resolution



Temporal Interval

Yearly

Accumulate 🕂

Geographic Extent

Global

y Yearly

Accumulate

Conterminous U.S.

Conterminous U.S.



PE2's Task. Develop a model to predict groundwater withdrawal.

Spatial Resolution



Temporal Interval

Geographic Extent

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· Accumu	late	Clip	
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y		Conterminous U.S.	Kansas
26			



Aligning geospatial datasets required participants to have significant about the datasets themselves.

Server Toolbox Ready to Use Toolbox **Spatial Analyst Toolbox** Spatial Statistics Toolbox ... +35 More

Bitwise Left Shift Kriging Raster Calculator Iso Cluster Unsupervised Fuzzy Overlay Zonal Histogram Darcy Flow ... +200 More

Identify the correct sequence of transformations among hundreds of operators

fluency in geospatial data theory in addition to contextual information

Expected

Actual





Determine when selected **transformations** produced **undesirable results**



Findings

We identified 12 challenges across five phases of participants' work with geospatial data.

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Sketching Cartographic Variants

Geospatial Information in Design Software



Participants wanted to visualize their data using many different **cartographic representations** to:

- Identify the map type that represented their data most effectively
- Produce tangible artifacts for collaborators to evaluate











PJ5 created over **20 draft maps** for a story on biased predictive policing algorithms.





Choropleth and Dot Density





Gridded Heat Map



Gridded Heat Map with Bar Charts



Producing most map variants required going through **the entire analysis and visualization pipeline**.



Census Tracts

Counties

Additional Data Transformation

Across Multiple Tools



Participants tried to speed up the drafting process in creative ways. One common technique involved screenshotting in-progress maps.

Participant E5



Participant J6





Participant S2



Participants tried to **speed up** the drafting process in creative ways. One common technique involved **screenshotting in-progress maps**.

Participant J6

Screenshots

-ayouts



Screenshotting came with limitations.

 Only allowed users to capture cartographic changes within a map type rather than across map types
 Once a final map design was chosen, participants had to reproduce the selected draft in code



Roadmap



2. Findings

1. Study Design

3. Design Opportunities

• 4. Introducing cartokit

Roadmap



2. Findings

1. Study Design

• 4. Introducing cartokit

Design Opportunities

We synthesized **six design opportunities** for designers and developers of geospatial analysis and visualization systems.

Solving Geospatial Data Constraints

Opportunity 1. Participants struggled to find geospatial data satisfying complex spatial and temporal constraints (Section 5.1). While many could describe their constraints succinctly, satisfying them involved constructing bespoke workflows to combine, align, and simplify their raw datasets (Section 5.2). These challenges suggest an opportunity for tools that (1) offer alternative programming abstractions to express data constraints and (2) infer geospatial data queries and transformations from constraints.

Assistive Tools for Constructing Geospatial Analysis Pipelines

Opportunity 2. Participants could describe the target outputs of their geospatial analyses but struggled to construct pipelines to produce them (Section 5.3). This suggests an opportunity for tools that (1) accept noncode specifications of analysis intent, (2) synthesize analysis programs that satisfy specifications, and (3) support users in editing programs.

Opportunity 3. Participants relied on running operators and manually inspecting outputs to understand operator semantics (Section 5.3.2). This was computationally expensive and time-consuming, suggesting an opportunity for tools that surface information on operator semantics without requiring execution across entire inputs.

Reproducible, Shareable **Geospatial Workflows**

Opportunity 4. Participants using GISs struggled to create reproducible, shareable geospatial workflows (Section 5.4.2). Limitations in existing history interfaces made it difficult to recover information on the current analysis state or revisit past analysis decisions (Section 5.4.1). These struggles suggest opportunities for tools that (1) support efficient search through system history and (2) distill history into a portable and executable representation.

Exploring the **Cartographic Design** Space

Opportunity 5. Participants wanted to visualize their geospatial data using multiple cartographic representations, but transitioning between representations required engineering each one from scratch (Section 5.5.1). This suggests an opportunity for cartographic design tools that reduce the viscosity [8] of switching between map types.

Opportunity 6. Many participants used direct manipulation design software to visualize geospatial data. These tools discard all geographic information, making it difficult to refactor an analysis once visualization work has begun (Section 5.5.2). This suggests an opportunity for tools that (1) bridge geospatial analysis and cartographic design and (2) maintain the underlying geospatial data representation of graphical elements while supporting direct manipulation.



Design Opportunities

"viscosity" of map type transitions.

vega-lite



plot



ggplot2

Possible Solution. Grammar of Graphics

Opportunity. Cartographic design tools could focus on reducing the

Restrict geospatial file formats, data models, and map types \Rightarrow Could not express many of the maps participants made



Design Opportunities

Opportunity. Cartographic design tools could **pair programmatic** and direct manipulation paradigms for map construction.



Sketch-n-Sketch

Edit **source** or **output** and propagate edits **bidirectionally** ⇒ Design maps using **direct** manipulation while giving access to program representations



Roadmap



2. Findings

1. Study Design

• 4. Introducing cartokit

Roadmap



3. Design Opportunities

2. Findings

1. Study Design

A direct manipulation programming system for interactive cartography on the web.











Introducing cartokit A direct manipulation programming system for interactive cartography on the web.



Edit programs through direct manipulation of the output





A direct manipulation programming system for interactive cartography on the web.





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Synthesize the program to produce the output map





A direct manipulation programming system for interactive cartography on the web.

> Support sophisticated geospatial data transformations









A direct manipulation programming system for interactive cartography on the web.

Core Challenge. Transforming data to achieve different cartographic representations.







Demo

Can we reproduce this graphic¹ from the Washington Post using cartokit?

1. Will global warming make temperature less deadly? Washington Post <u>https://www.washingtonpost.com/climate-</u> environment/interactive/2023/hot-cold-extreme-temperaturedeaths/.

Change in deaths linked to temperature

Projected average for 2080-2099, compared to a world without additional emissions



Globe shows RCP4.5 scenario. Antarctica left blank because it has no permanent human population.

Source: Climate Impact Lab via Human Climate Horizons. Method detailed in Carleton et. al., 2022.



