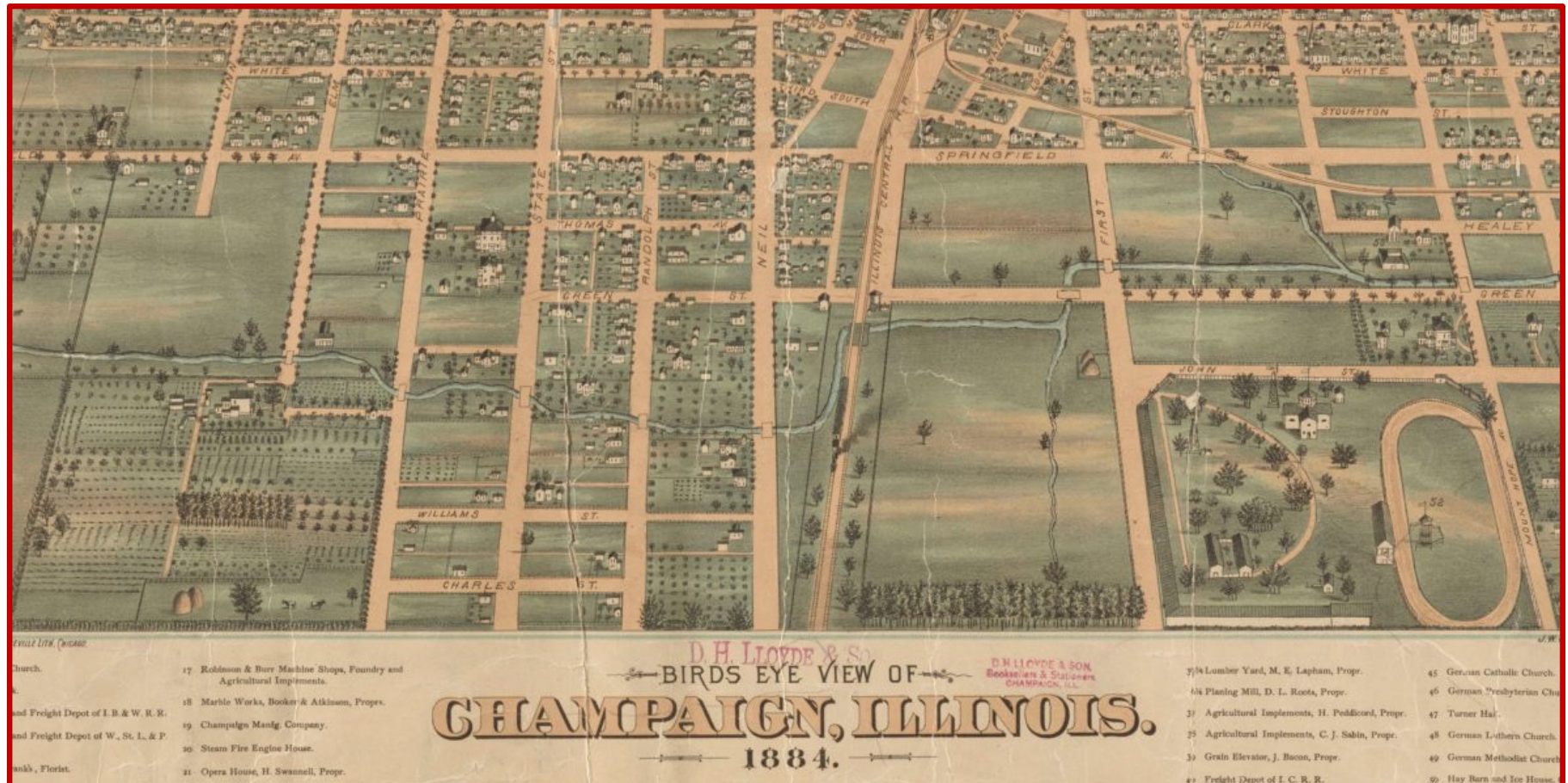

Automating Human Geospatial Intelligence

Kevin C.C. Chang

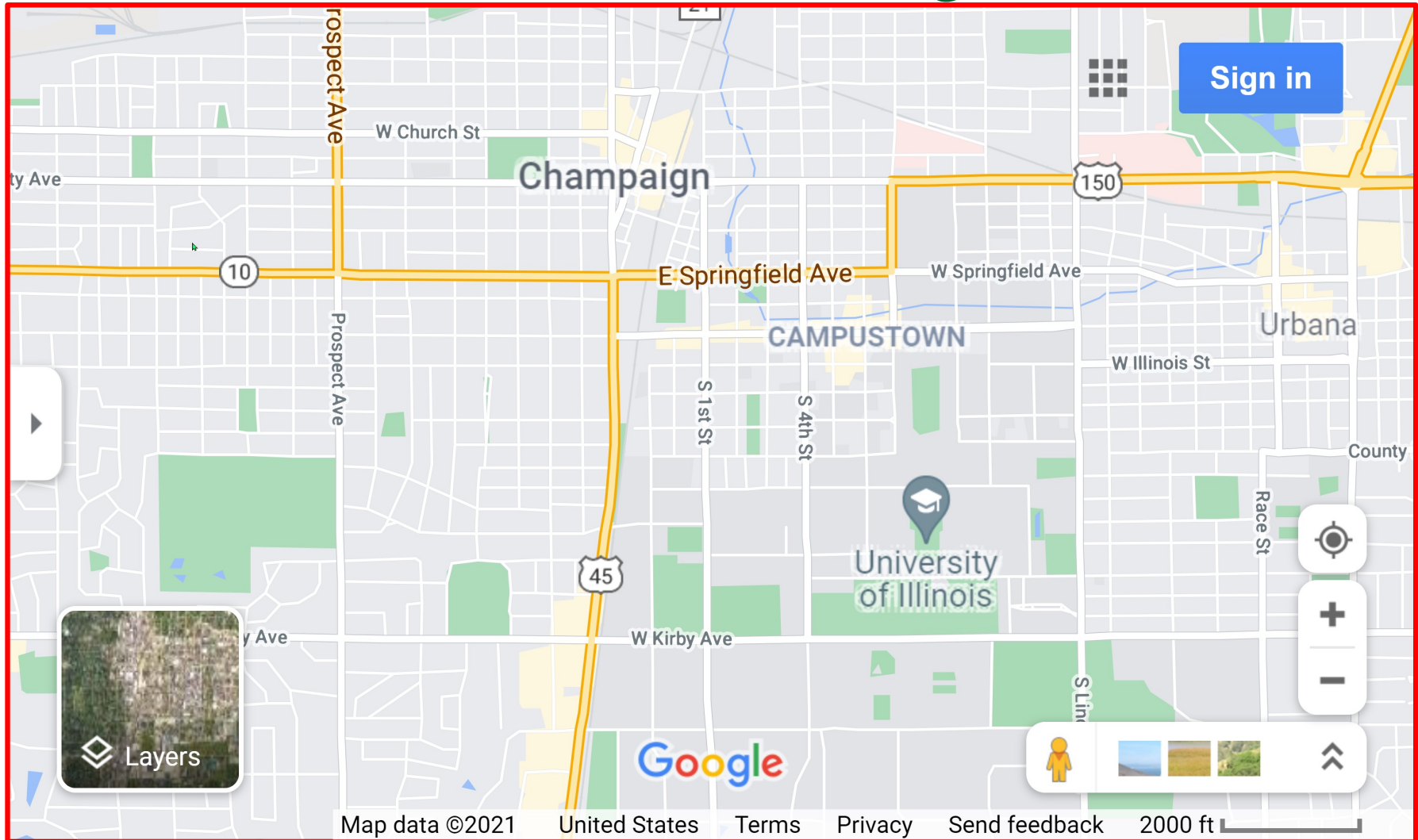


What about people? Where were people? Who live there?



What about people?

Human GEOINT is still lacking!



While Physical GEOINT advances, Human GEOINT is largely lagging.

- Automatic large-scale sensing makes rich and up-to-date Physical GEOINT possible.
- But human GEOINT is mostly relying on **manual, periodical survey** like census.

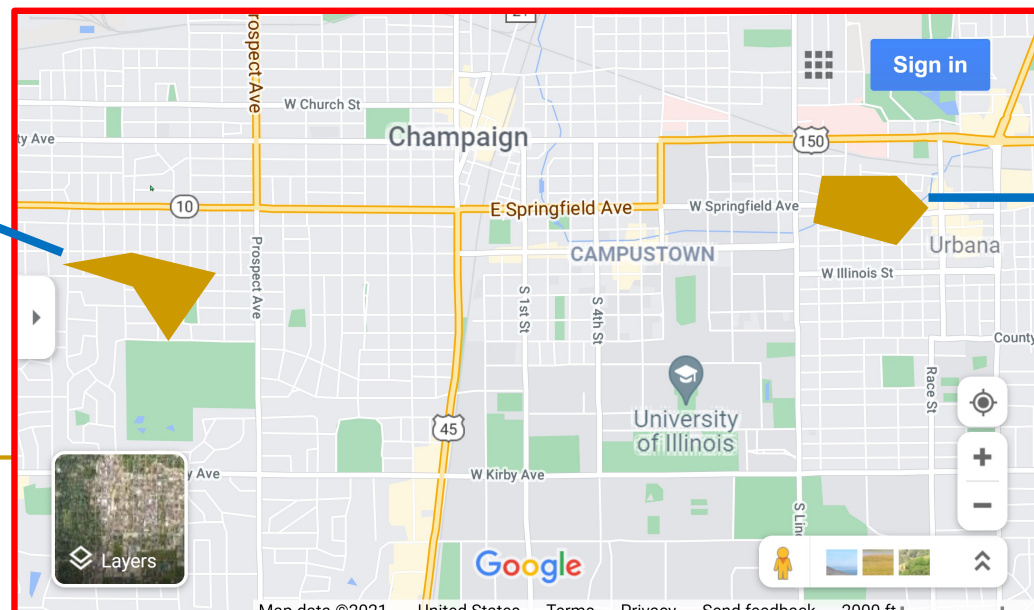


Limitations of current Human GEOINT

- **Outdated** (as soon as published)
- **Coarse-Grained** (census tract or block)
- **Static** (average/aggregate, not “this time”)
- **Passive** (relying on inhabitant corporation)

Goals: Can we automate Human GEOINT, to **actively** and **dynamically** acquire **fine-grained, up-to-date** information about humans in a place?

- **(Q1) Neighborhood Detection:** Where are various neighborhoods located in the city?
- **(Q2) Neighborhood Profiling:** What characterizes each neighborhood (e.g., race, education)?



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The time has come: Big Data that is GEO-tagged

- Google Street View
- Microsoft Building Footprints
- Open government data- Data.Gov everywhere.
- Social media- 500m tweets per day in 2021.

Preliminary Study: The City of Chicago

- Six zip code area in Chicago.
 - Downtown, midtown, suburbs.

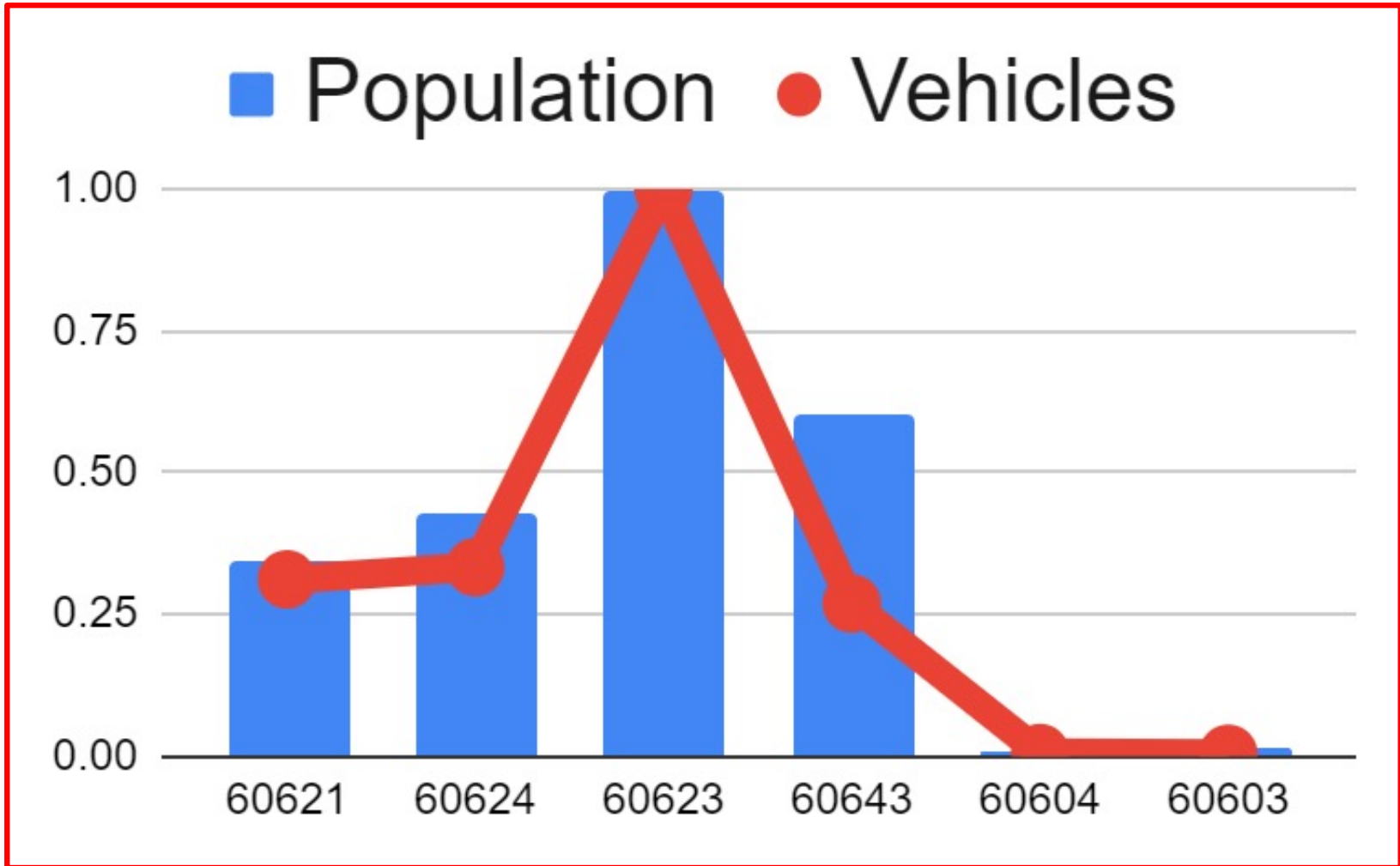
US Census ACS, from 2019

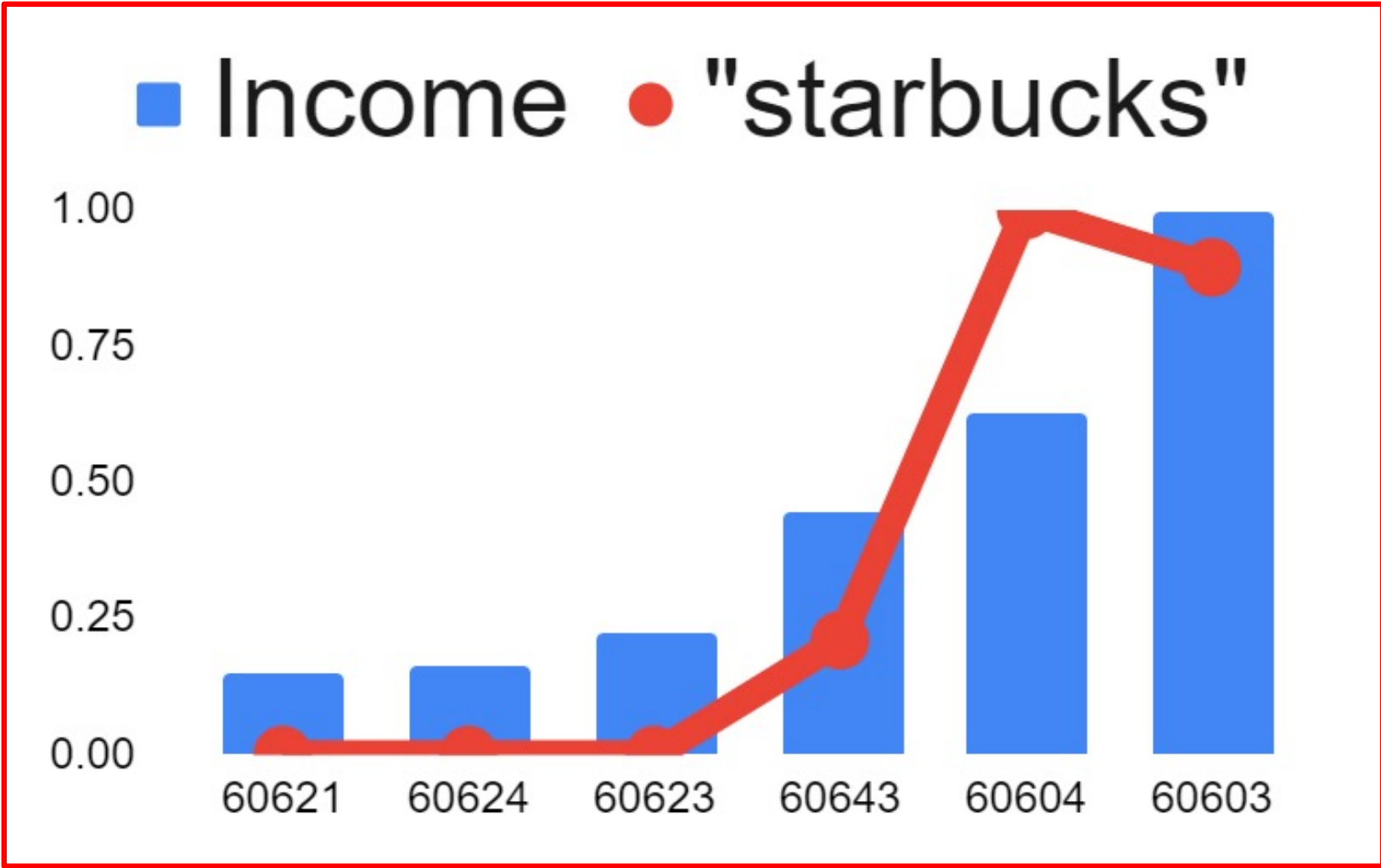
Zip Code	60621	60624	60623	60643	60604	60603
Population	28018	34829	81283	48887	823	1052
Income \$	22158	23429	32460	65631	91750	146250
College %	10.2	9.5	10.3	37.3	91.9	95.2
White %	2.2	4.5	38.0	23.3	64.6	64.1
Asian %	0.2	0.2	0.1	0.3	29.2	32.7

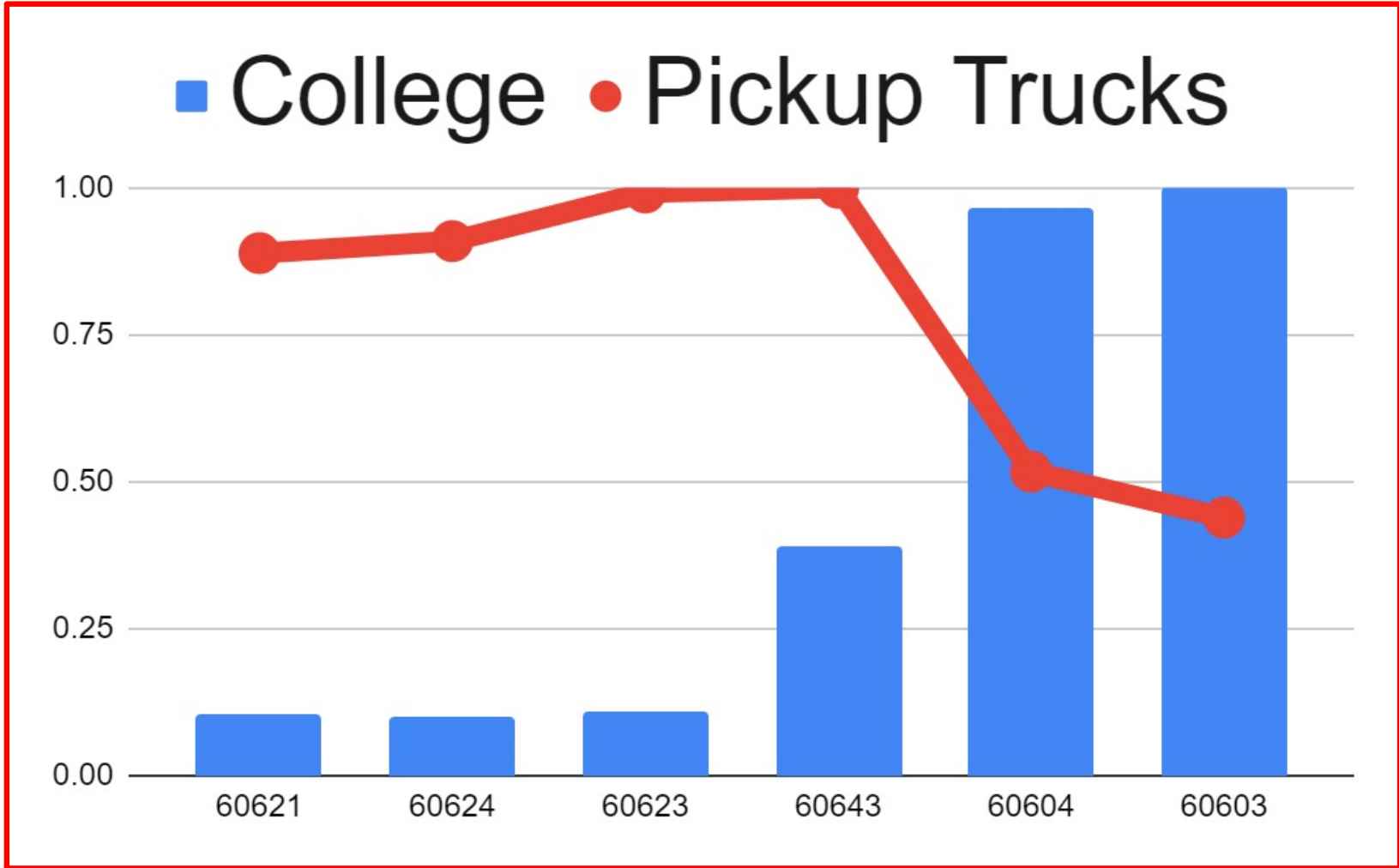
Data Collected and Processed

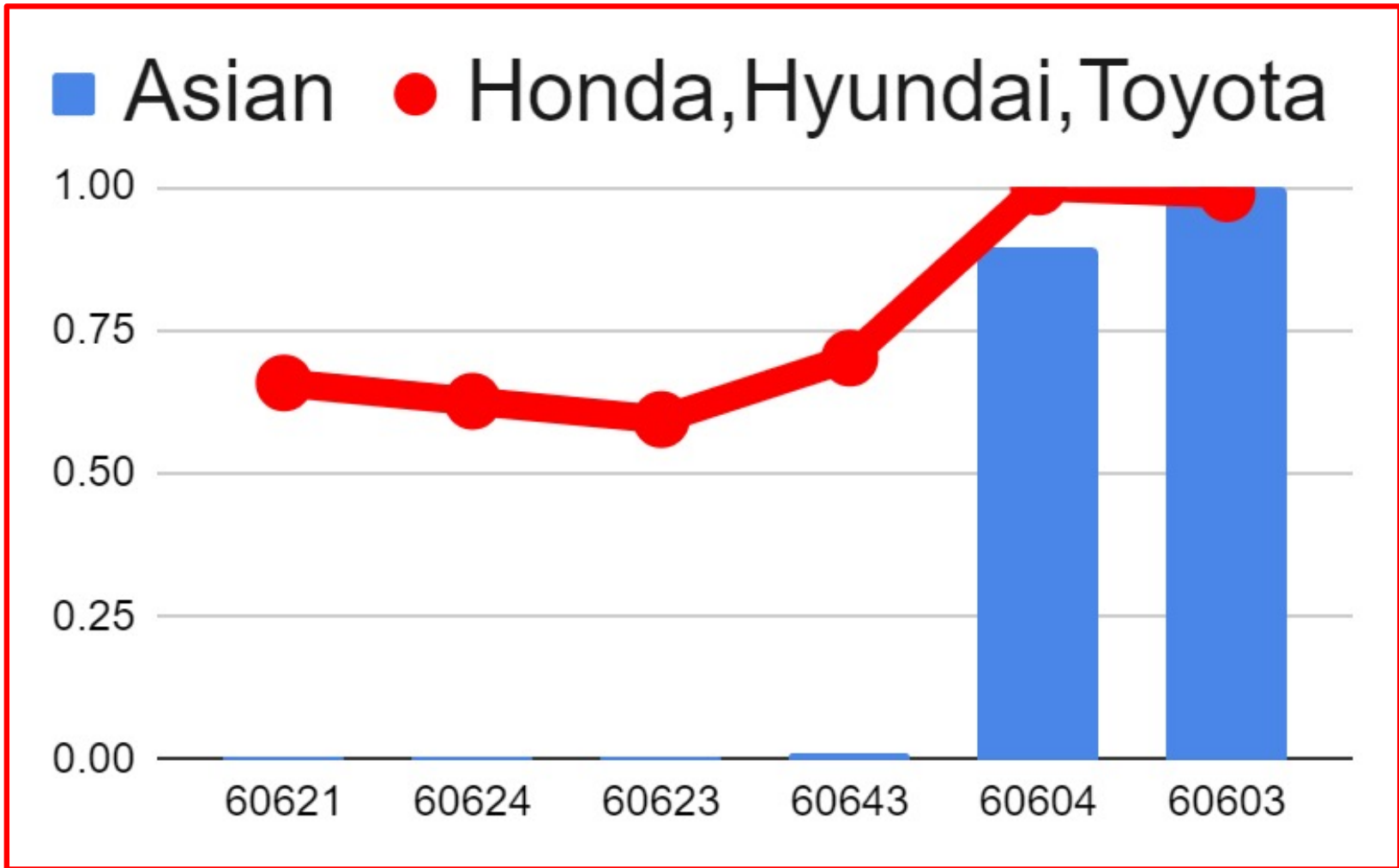
- **Street View** images: 241K images 2021.
 - Extracted 62814 vehicles and other features.
- **Tweets**: 25K from Archive.Org 2011 – 2020.
- **Crime** incidents: 941K records 2001 – 2020.
- **Building addresses**: 40K records (OpenStreetMap).
- **Building footprints** (Microsoft)
- Can data “predict” the human world?

Interestingly, we found
strong features that are
highly indicative for *every*
target latent property.









So we submitted a proposal abstract.

- Task 1: Data Source Discovery
- Task 2: Geo Data Alignment
- Task 3: Feature Extraction and Integration
- Task 4: Prediction Model Construction

Mission Imperative: MI#1 Assured Positioning, Navigation, Timing, & Targeting (to extend GEOINT foundation products with automatic Human GEOINT).

Application Areas: Cross Cutting—*Foundation* (to build foundation for Human GEOINT), *Geospatial Cyber* (to use cyber data).

Title: Automating Human Geospatial Intelligence

1. BAA number: HM0476-20-BAA-0001
Topic Number: 3
Title: NGA Research Innovative Seedlings
2. Lead Organization: University of Illinois at Urbana-Champaign
3. Type of Organization: OTHER EDUCATIONAL
4. Other Team Members/sub-contractors: None
5. Proposal Title: Automating Human Geospatial Intelligence
6. Principle Investigator: Kevin Chen-Chuan Chang
7. Technical Point of Contact: Kevin Chang, Professor, Computer Science Department, 201 N. Goodwin Avenue, Urbana, Illinois 61801, Phone: (217) 244-2919, Email: kcchang@illinois.edu
8. Administrative Point of Contact: Robin Beach, Director, Pre-Award, Sponsored Programs Administration, 1901 S. First Street, Suite A, Champaign, Illinois 61820-7406, Phone: (217) 333-2187, E-mail: spa@illinois.edu
9. Award Instrument Requested: grant
10. Award Type Requested: cost-contract—no fee (Grants, Cooperative Agreements)
11. Total Rough Order of Magnitude: \$250,000
12. Date Abstract Submitted: August 19, 2021
13. Taxpayer Identification Number (TIN): 37-60000511
14. Data Universal Numbering System (DUNS) Number: 04-154-4081
15. Commercial and Government Entity (CAGE) Code: 4B808



Thank You!

