

Krzysztof Janowicz, UCSB and University of Vienna (PI)

Pascal Hitzler, K-State (Co-PI)

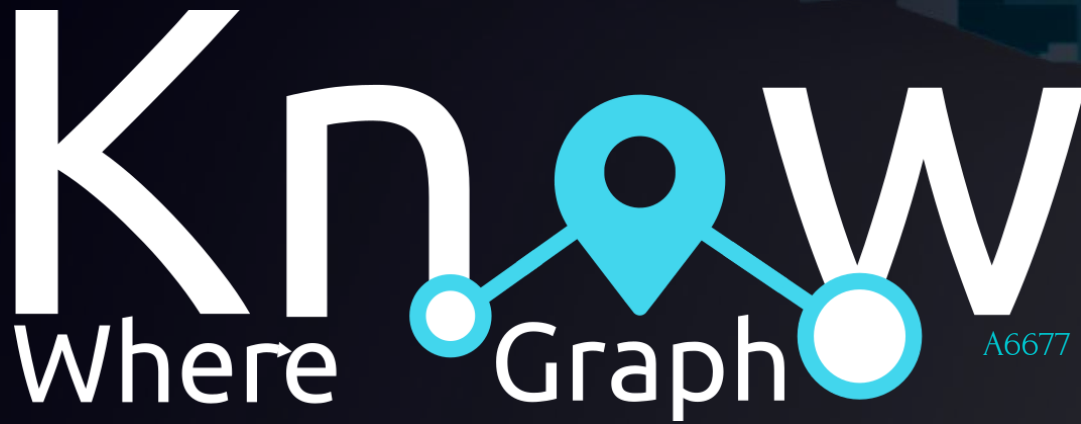
Wenwen Li, ASU (Co-PI)

Mark Schildhauer, NCEAS (Co-PI)

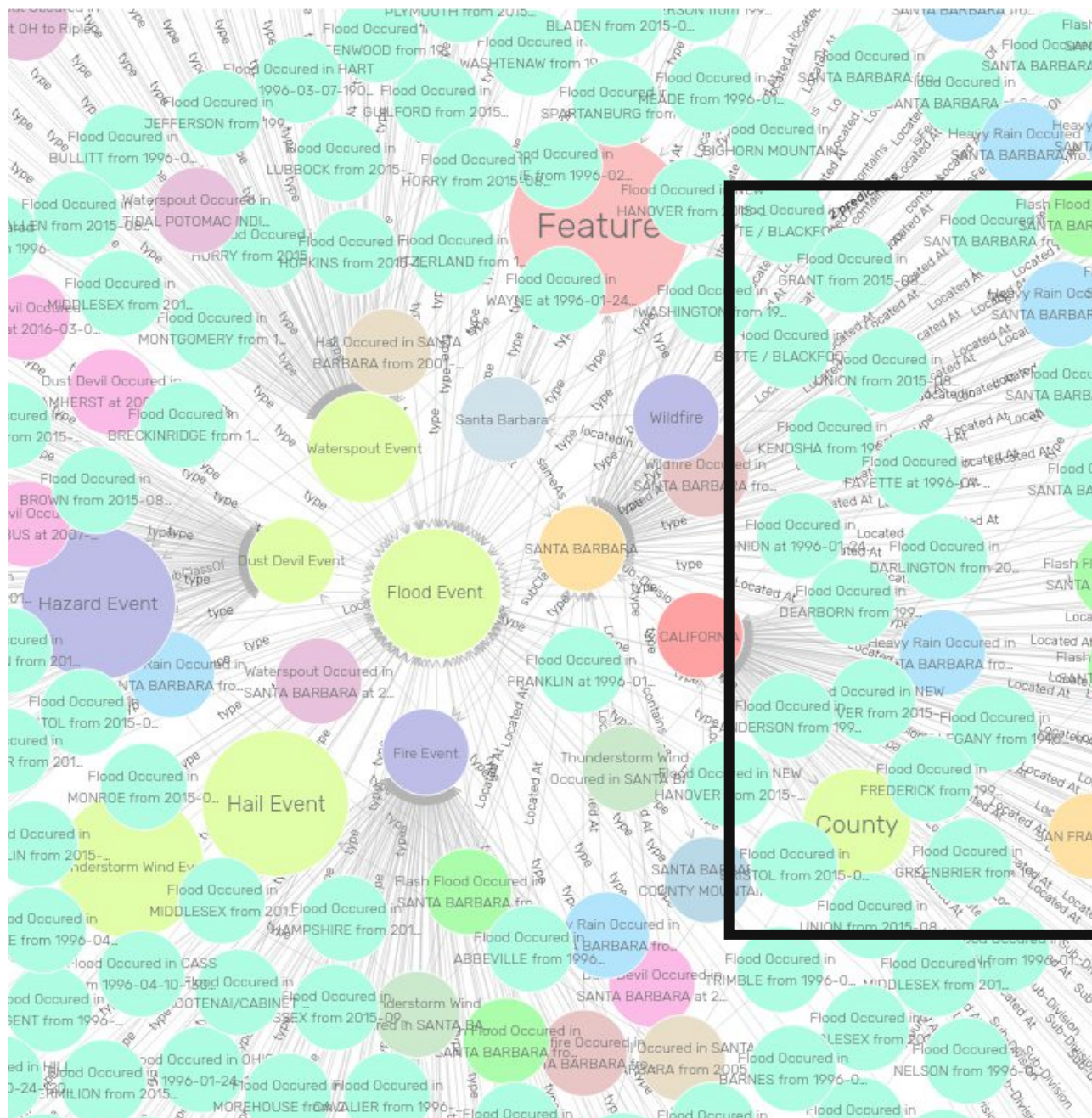
Dean Rehberger, MSU (Co-PI)

KnowWhereGraph and Beyond: Can Knowledge Graphs provide Situational Awareness for Epidemic Intelligence?

EIOS Nov. 2022



We provide area briefings within seconds for any region on Earth



WHAT WHY



Data Acquisition Bottleneck

The typical data science project spends a **majority** of all resources on data discovery, entry, cleaning, and integration— instead of actual analysis



80%

Processing
Data



20%

Deriving
Insights



Our Value Proposition

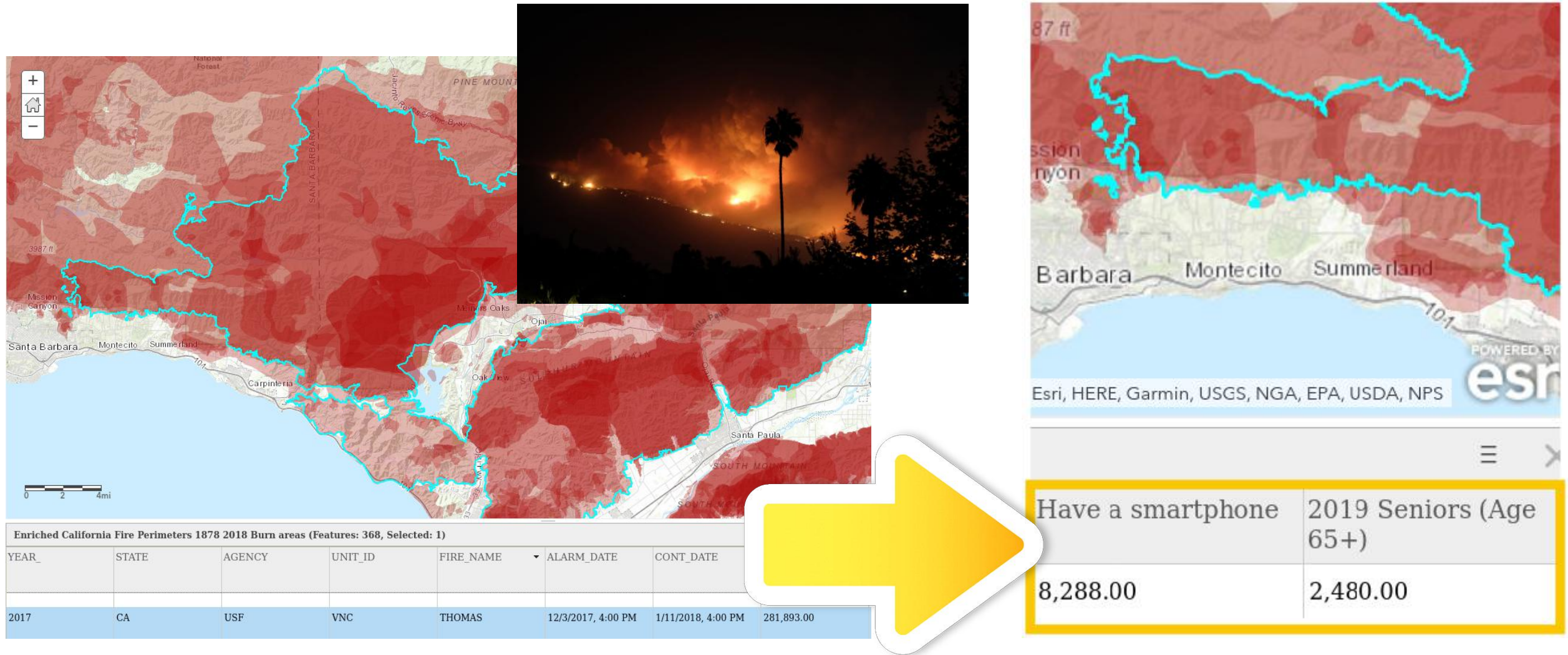
KnowWhereGraph aims at **providing area briefings within seconds** for any region on Earth to answer questions such as

- ***"What is here?"***
- ***"What happened here before?"***
- *"Who knows more?"*
- *"How does it compare to other regions or previous events?"*

By doing so, we assist decision-makers and data scientists in rapidly enriching their data with millions of connected, up-to-date facts at the human-environment interface to gain the situational awareness required for good decision-making.

Our ultimate goal is to flip the 80/20 bottleneck on its head.

Overcoming the 80/20 Bottleneck with (Esri's) GeoEnrichment



Enrich (your own) data by adding (demographic) facts about **your spatial area of interest**

GeoEnrichment Limitations



Pros

- **On-demand**
- **Well-curated**
- **Apportioned**
- **GIS-ready**



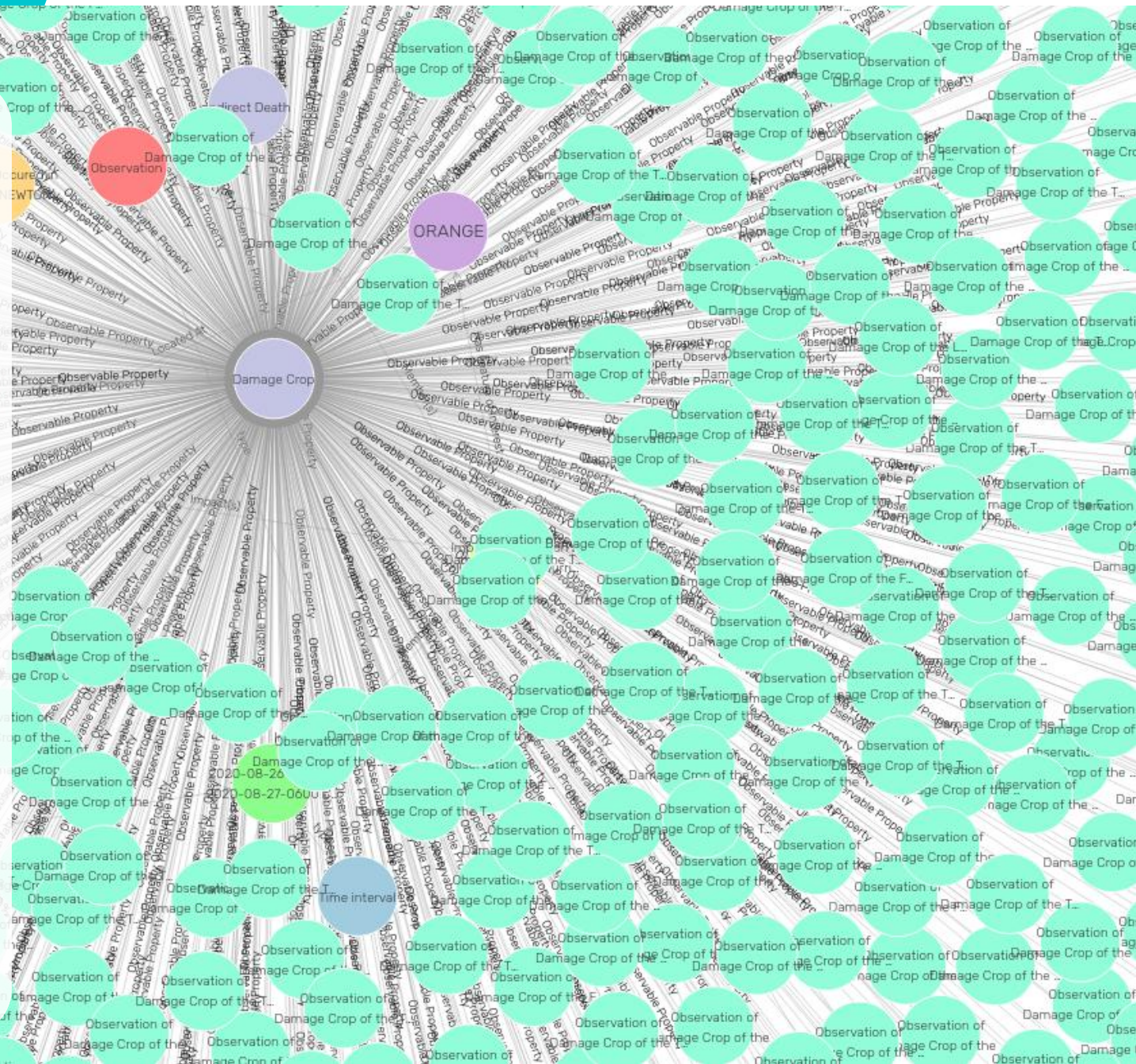
Cons

- **Pre-defined categories**
- **Essentially closed data silos**
- **Flat, tabular data**
- **Not always up-to-date**
- **Does not scale**
- **Limited support for automated integration**

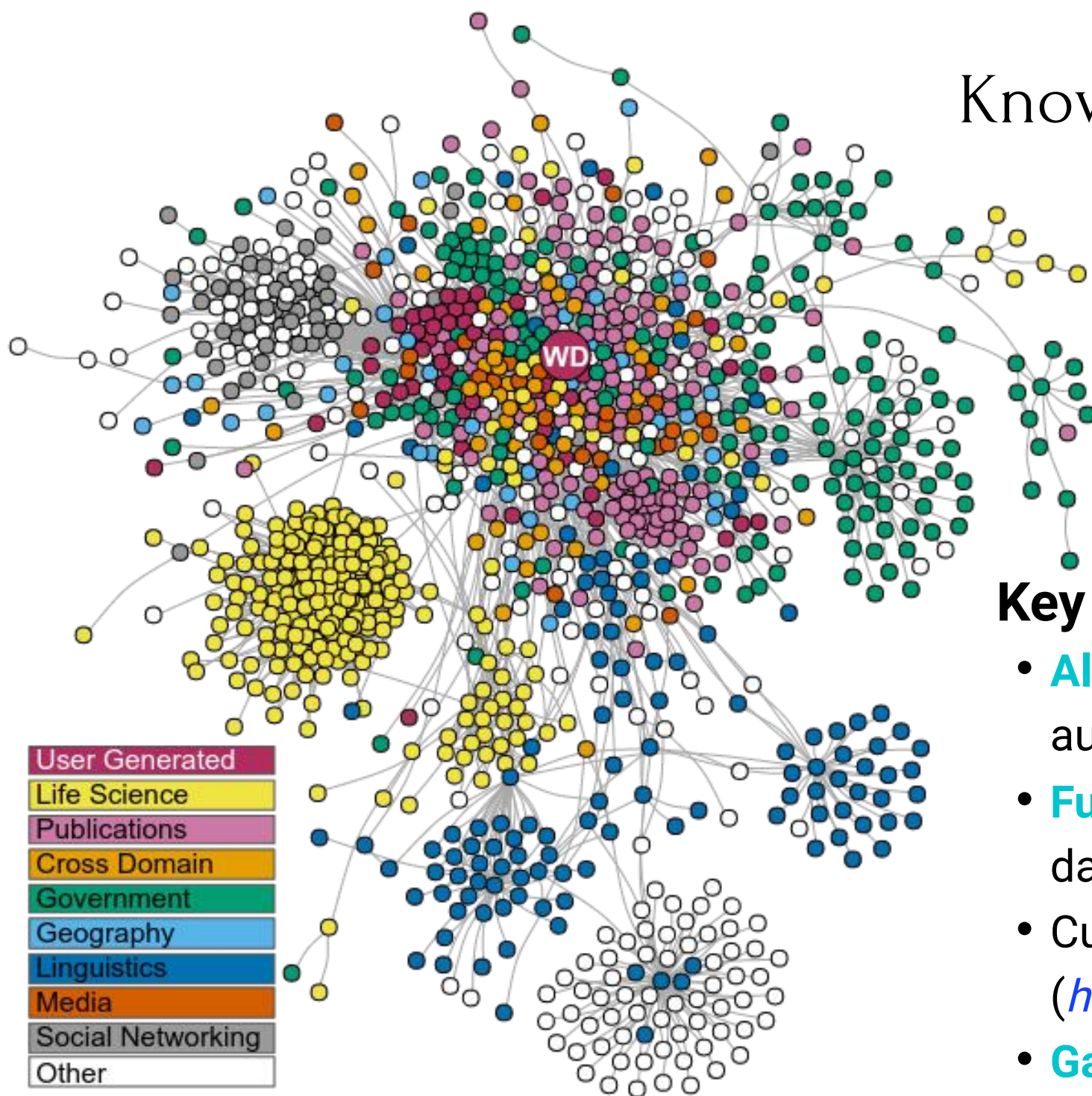
Knowledge Graphs Connect Data Across Themes

Knowledge Graphs

- **Emphasize relationships** over (in addition to) attributes
- **Connect data** (not datasets) **across themes** and silos
- Break apart the data-metadata distinction by **making data smart**, self-descriptive
- **Machine-readable**, rich semantics enables **conflation** and inference
- While **places** are key nodes in KG, we **know almost nothing** about them



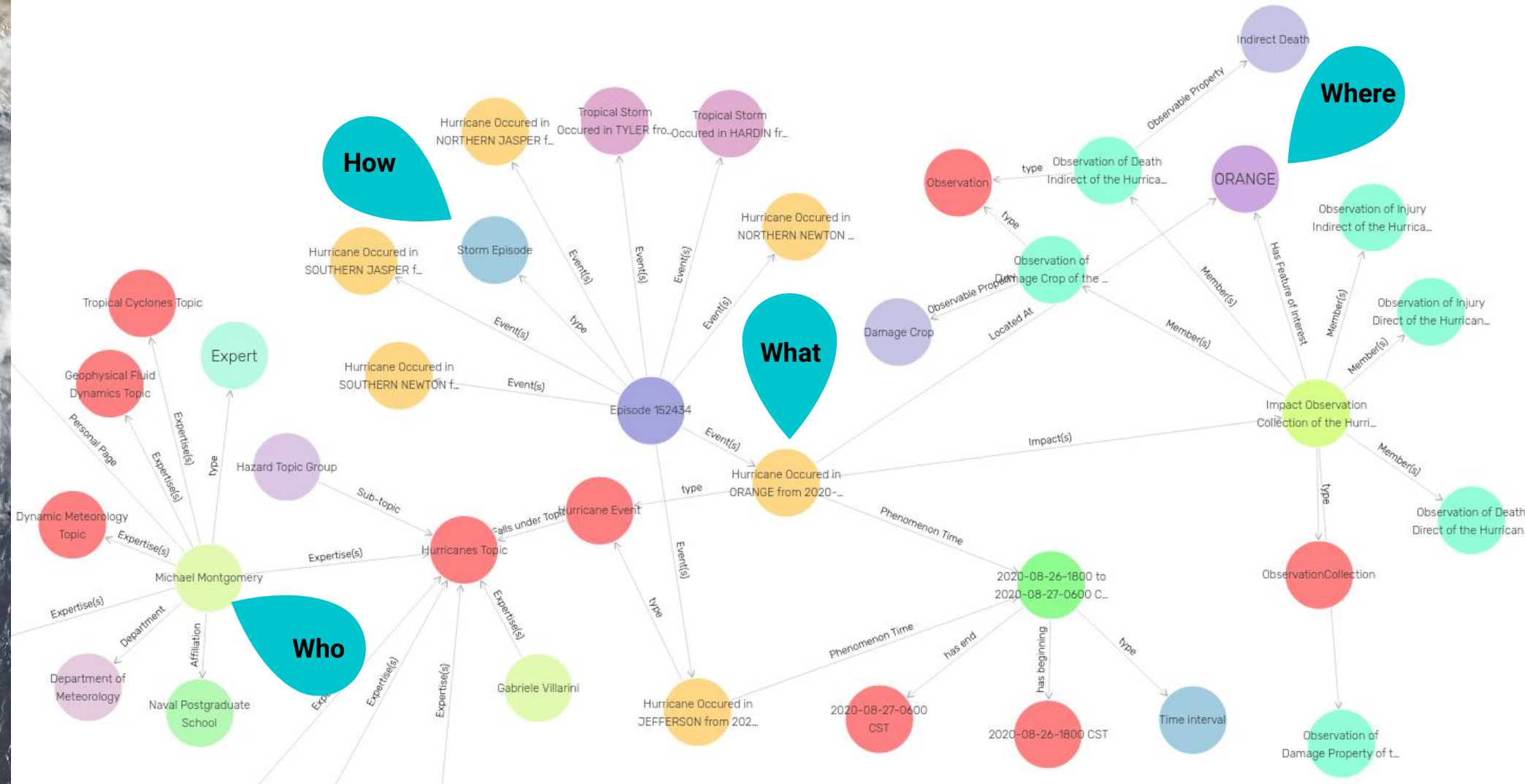
Knowledge Graph Ecosystem



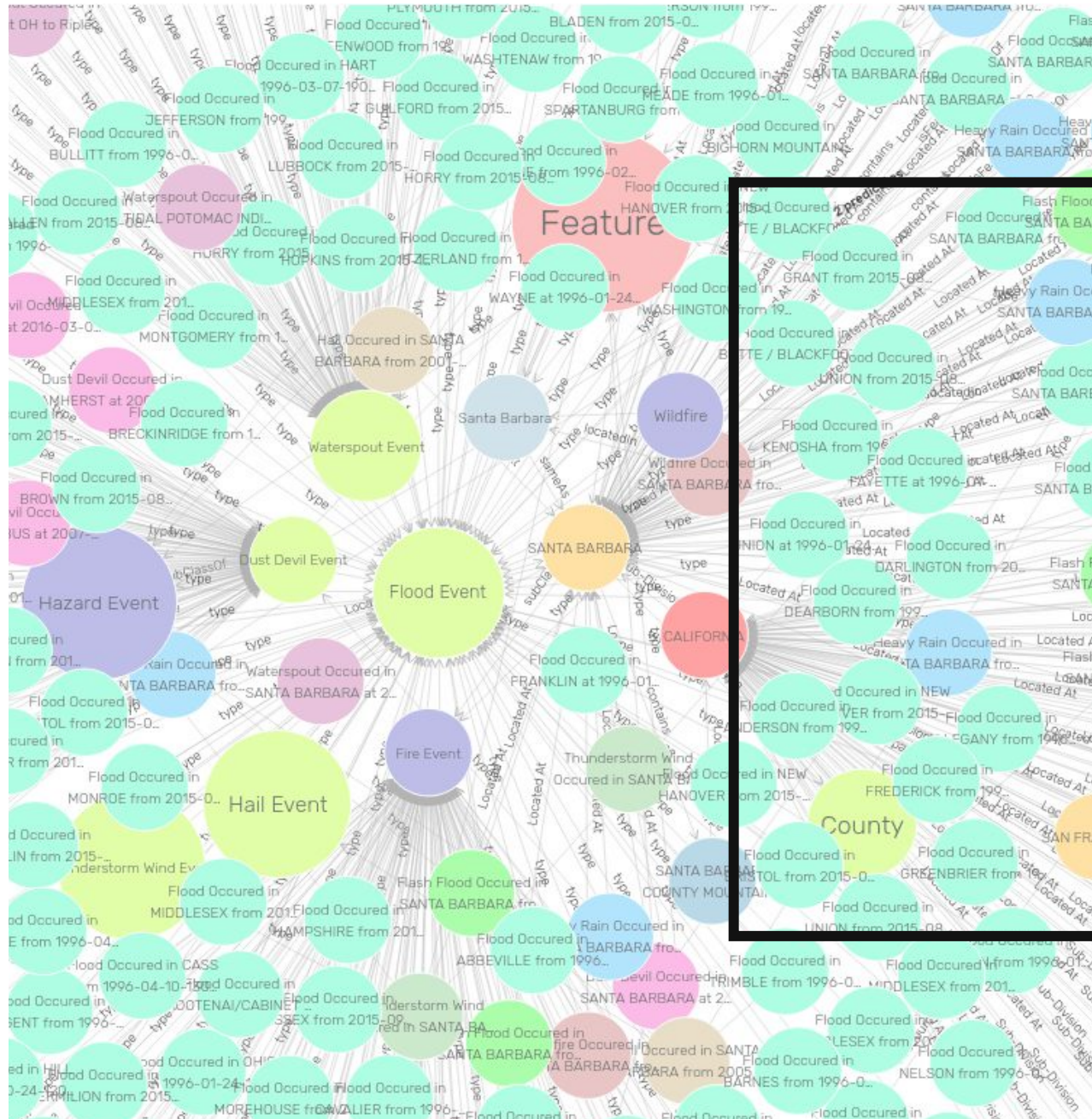
Key Ideas

- **Align vocabularies** across graphs (semi-automatically)
- **Fuse data** across graphs to access billions of data items at your **fingertips**
- Currently we establish links to **SPOKE** (<https://spoke.ucsf.edu/>) and **UFOKN**
- **Gaining a more holistic understanding!**

KnowWhereGraph = Knowledge Graph + GeoEnrichment



We are the first to **combine geoenrichment with knowledge graphs** to power **environmental intelligence** applications



WHO
BENEFITS

Pilots and Prototypes

Know
Where Graph



Farm to Table Supply Chain & Sustainability

Enhance the sustainability, efficiency, and safety of consumer food supply in collaboration with the *Food Industry Association*.



Land Valuation and Risk of Default

Driver-based land potential assessment for model based valuation and risk assessment for agricultural credit applications & loan portfolio monitoring.



Community Lifeline System Backbone

Provide a data library for time-critical situational awareness assessment by delivering data for an affected area with over 25 data layers in seconds



Humanitarian Aid

Apply our technologies to the humanitarian supply chain needs of *Direct Relief* during the COVID-19 crisis and help them to find experts.

Environmental Intelligence

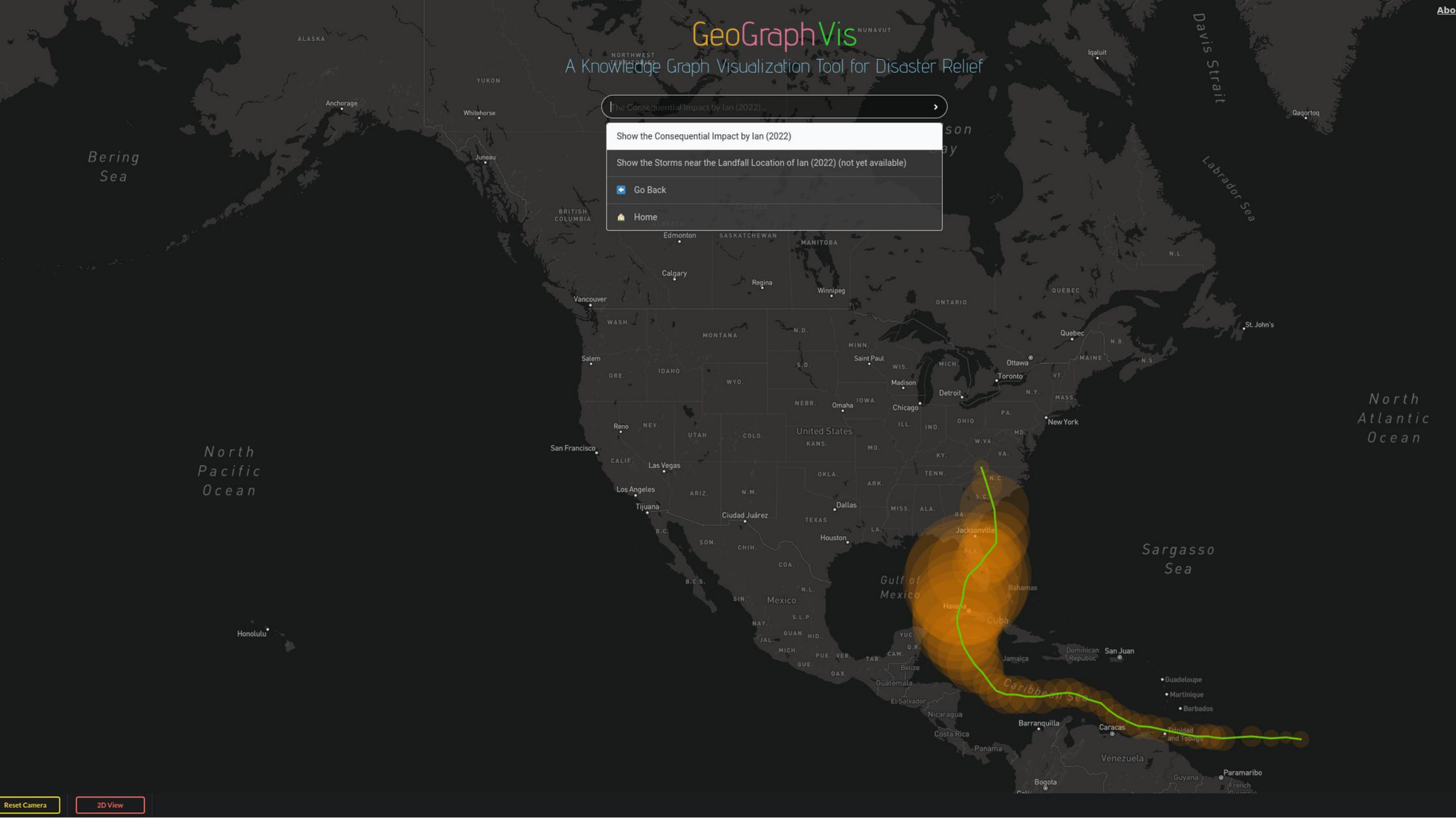
The Consequential Impact by Ian (2022)

Show the Consequential Impact by Ian (2022)

Show the Storms near the Landfall Location of Ian (2022) (not yet available)

Go Back

Home



GeoGraphVis

A Knowledge Graph Visualization Tool for Disaster Relief

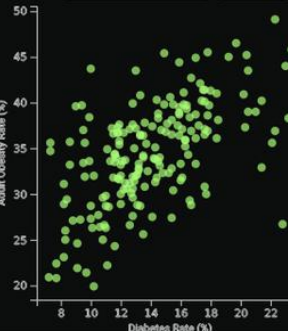
More Details about Ian (2022) ... [Chicago](#) [Cleveland](#) [Toledo](#) [Detroit](#) [Buffalo](#) [Rochester](#) [Toronto](#) [Ottawa](#) [Halifax](#) [Boston](#) [New York](#) [Philadelphia](#) [Washington](#) [Richmond](#) [Raleigh](#) [Virginia Beach](#) [Atlanta](#) [Birmingham](#) [Memphis](#) [Louisville](#) [Cincinnati](#) [St. Louis](#) [Kansas City](#) [Wichita](#) [Tulsa](#) [Fort Worth](#) [Dallas](#) [Houston](#) [San Antonio](#) [Austin](#) [Lubbock](#) [Amarillo](#) [Santa Fe](#) [Phoenix](#) [Tucson](#) [Mexicali](#) [Tijuana](#) [Ensenada](#) [Golfo de Santa Clara](#) [Ciudad Juárez](#) [La Paz](#) [Culiacán](#) [Durango](#) [Mazatlán](#) [Torreón](#) [Monterrey](#) [Reynosa](#) [Zacatecas](#) [Guadalajara](#) [Colima](#) [Xalapa](#) [Campeche](#) [Chetumal](#) [Belize](#) [Guatemala](#) [San Pedro Sula](#) [Honduras](#) [El Salvador](#) [Guatemala City](#) [Havana](#) [Cape](#) [St. Lucie](#) [Miami](#) [Marsh Harbour](#) [Bahamas](#) [Cuba](#) [Guantanamo](#) [Cap-Haitien](#) [Haiti](#) [Dominican Republic](#) [Puerto Rico](#) [Marigot](#) [Saint Kitts and Nevis](#) [Antigua and Barbuda](#) [Saint Lucia](#) [Jamaica](#) [Bermuda](#) [Sargasso Sea](#) [Gulf of Mexico](#) [Caribbean](#)

Find Experts...

2-Variable Scatter Plot

x-axis: Diabetes Rate Linear

y-axis: Adult Obesity Rate Linear



Disaster Type

Hurricane

Reset Camera

2D View

Layers & Statistics of Ian (2022)

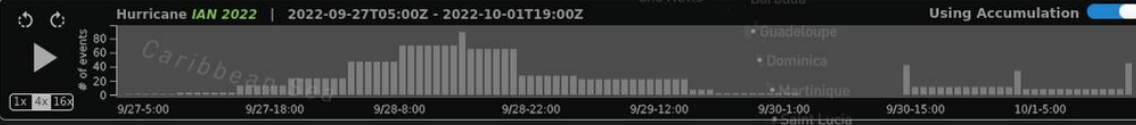
Storm Impact Layers

of Reports 191

Health Factor Layers

- Diabetes Rate
- Adult Obesity Rate
- Mentally Unhealthy Days
- Mental Health Provider Rate
- Low birthweight Rate
- Uninsured Rate
- Annual Mammogram Rate
- Flu Vaccination Rate
- Food Environment Index
- Injury Deaths

Uninsured Rate (%)



Home



The 30,000 Foot View

KnowWhereGraph serves 8+ different types of globally unique place and **region identifiers** and 10+ different thematic layers containing **millions of facts about these regions**.

- Discrete, Hierarchical Global Grid System
S2 Cells (L9 globally, L11 USA, L14 CA, **L16 for urban** (*))
- Global Administrative Areas
- US Federal Judicial District (*)
- National Weather Zones
- FIPS Codes
- DMA (*)
- ZIP (*)
- Climate Division

Our region identifiers are linked to Wikidata/Linked Data Cloud.

A team/customer that uses our identifiers gets access to billions of triples outside in addition to ~1.2B from the current KWG.

KnowWhere Graph (*)

(*) and KnowWhen

About 12-15 Billion triples (possibly spatiotemporally-scoped) extracted and interlinked across 28 data sources at the interface between humans and the environment from 16 organizations (e.g., gov. agencies, universities, and NGOs). More data is being added constantly. Final size: 30-50B

KnowWhere Graph (*)

(*) and KnowWhen

Feature

Place

County

State

z.06098

z.06092

z.06091

z.06095

z.06094

z.06090

z.06093

z.06096

z.06097

z.06099

z.06100

z.06101

z.06102

z.06103

z.06104

z.06105

z.06106

z.06107

z.06108

z.06109

z.06110

z.06111

z.06112

z.06113

z.06114

z.06115

z.06116

z.06117

z.06118

z.06119

z.06120

z.06121

z.06122

z.06123

z.06124

z.06125

z.06126

z.06127

z.06128

z.06129

z.06130

z.06131

z.06132

z.06133

z.06134

z.06135

z.06136

z.06137

z.06138

z.06139

z.06140

z.06141

z.06142

z.06143

z.06144

z.06145

z.06146

z.06147

z.06148

z.06149

z.06150

z.06151

z.06152

z.06153

z.06154

z.06155

z.06156

z.06157

z.06158

z.06159

z.06160

z.06161

z.06162

z.06163

z.06164

z.06165

z.06166

z.06167

z.06168

z.06169

z.06170

z.06171

z.06172

z.06173

z.06174

z.06175

z.06176

z.06177

z.06178

z.06179

z.06180

z.06181

z.06182

z.06183

z.06184

z.06185

z.06186

z.06187

z.06188

z.06189

z.06190

z.06191

z.06192

z.06193

z.06194

z.06195

z.06196

z.06197

z.06198

z.06199

z.06200

z.06201

z.06202

z.06203

z.06204

z.06205

z.06206

z.06207

z.06208

z.06209

z.06210

z.06211

z.06212

z.06213

z.06214

z.06215

z.06216

z.06217

z.06218

z.06219

z.06220

z.06221

z.06222

z.06223

z.06224

z.06225

z.06226

z.06227

z.06228

z.06229

z.06230

z.06231

z.06232

z.06233

z.06234

z.06235

z.06236

z.06237

z.06238

z.06239

z.06240

z.06241

z.06242

z.06243

z.06244

z.06245

z.06246

z.06247

z.06248

z.06249

z.06250

z.06251

z.06252

z.06253

z.06254

z.06255

z.06256

z.06257

z.06258

z.06259

z.06260

z.06261

z.06262

z.06263

z.06264

z.06265

z.06266

z.06267

z.06268

z.06269

z.06270

z.06271

z.06272

z.06273

z.06274

z.06275

z.06276

z.06277

z.06278

z.06279

z.06280

z.06281

z.06282

z.06283

z.06284

z.06285

z.06286

z.06287

z.06288

z.06289

z.06290

z.06291

z.06292

z.06293

z.06294

z.06295

z.06296

z.06297

z.06298

z.06299

z.06300

z.06301

z.06302

z.06303

z.06304

z.06305

z.06306

z.06307

z.06308

z.06309

z.06310

z.06311

z.06312

z.06313

z.06314

z.06315

z.06316

z.06317

z.06318

z.06319

z.06320

z.06321

z.06322

z.06323

z.06324

z.06325

z.06326

z.06327

z.06328

z.06329

z.06330

z.06331

z.06332

z.06333

z.06334

z.06335

z.06336

z.06337

z.06338

z.06339

z.06340

z.06341

z.06342

z.06343

z.06344

z.06345

z.06346

z.06347

z.06348

z.06349

z.06350

z.06351

z.06352

z.06353

z.06354

z.06355

KnowWhere Graph (*)

(*) and KnowWhen

About 12-15 Billion triples (possibly spatiotemporally-scoped) extracted and interlinked across 28 data sources at the interface between humans and the environment from 16 organizations (e.g., gov. agencies, universities, and NGOs). More data is being added constantly. Final size: 30-50B

KnowWhere Graph

<https://www.knowwheregraph.org/graph/>

Thematic Datasets					Place-Centric Datasets		
Dataset Name/ Theme	Source Agency	Key Attributes	Spatial Coverage	Temporal Coverage	Place-Centric Dataset	Defining Authority	Spatial Coverage
Soil Properties	USDA	soil type, farmland class	Targeted regions in US	Current	S2 Cells	Google	Lvl 9 (Global), Lvl 13 (US),
Wildfires	USGS, USDA, USFS, NIFC	wildfire type, burn severity, num. acres burned, contained date	US	1984–current	Global Administrative Regions	University of Berkeley, Museum of Vertebrate Zoology and the International Rice Research Institute	Global
Earthquakes	USGS	magnitude, length, width, geometry	Global (mag. over 4.5)	2011-01-01 to 2022-01-18			
Climate Hazards	NOAA	injuries, deaths, property damages	US	1950–2022			
Expert - Covid-19 Mobility	Direct Relief (DR)	name, affiliation, expertise	Global	2021	US Federal Judicial District	DoJ, ESRI	US
Expert - General	KWG, UC System, DR, Semantic Scholar	name, affiliation, expertise with spatiotemporal scopes	Global	unlimited	National Weather Zones	NOAA	US
Cropland Types	USDA	crop types (raster data)	US	2008-2021	FIPS Codes	NRCS	US
Air Qual. Obs.	U.S. EPA	AQI value, CO concentration	US	1980–2022	Designated Market Area	Nielen	US
Smoke Plumes	NOAA	daily smoke plumes extent	US	2010-2022	ZIP	ZCTA	US
Climate Observations	NOAA	temperature, precipitation, PDSI, PHSI	US	1950 - 2022	Climate Division	NOAA	US
Disaster Declaration	FEMA	designated area, program, amount approved, program designated date	US	1953 - 2022	Census Metropolitan Area	US Census	US
Smoke Plume Extents	NOAA	Smoke extent	US	2017 - 2022	Drought Zone	NDMC, USDA, NOAA	US
BlueSky Forecasts	Bluesky	PM10, PM5	US	2022-03-07	Geographic Name Information System	USGS	US
Transportation (highway network)	DOT	road type, road length, road sign	US	2014			
Public Health	CDC, US Census	below poverty level percent, diabetes age adjusted 20 plus percent, obesity age adjusted 20 plus percent	US	2017			
Social Vulnerability	CDC/ATSDR	social vulnerability index	US	2018			
Hurricane Tracks	NOAA	max wind speed, min pressure	US	1851-2020			

[This overview is always outdated]

Classical Knowledge Representation and Reasoning

F. Baader, W. Nutt

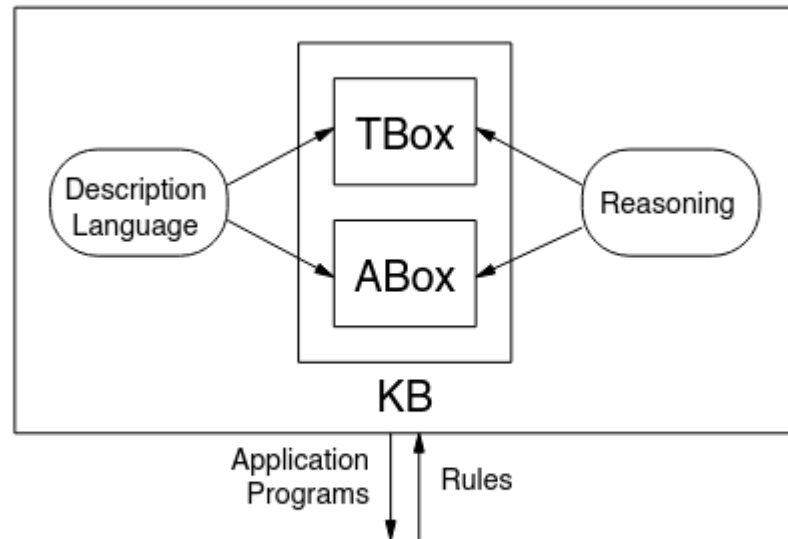
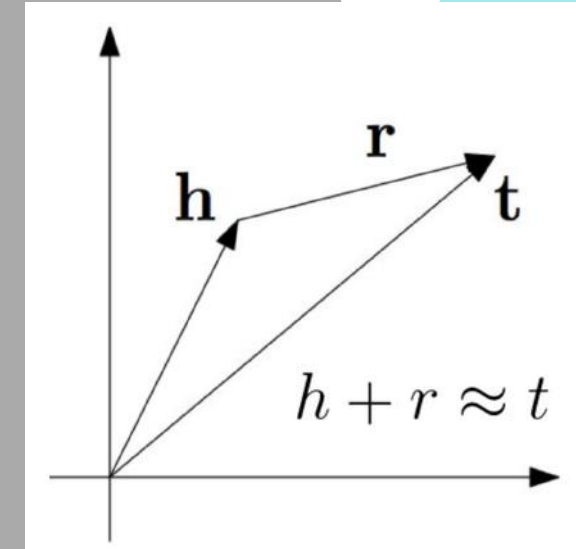


Fig. 2.1. Architecture of a knowledge representation system based on Description Logics.

- Family of **satisfiability**-based reasoning services
 - **(Concept)** subsumption: $KB \models C \sqsubseteq D$
 - **Instance** checking: $KB \models C(a)$
- Requires (almost) **no data**
- **No uncertainty** at the cost of the **inability to handle noise**

Representation Learning and Its Downstream Tasks



- E.g., learning **embeddings** as low(er)-dimensional **vector** (space) representations of entities and their relations
 - **Link prediction**: predicting unknown/missing statements between (two) entities.
 - **Node classification**: predict the class membership of unlabeled entities
- Requires a lot of (unbiased/representative) data
- **Great ability to handle uncertainty** and **noise**

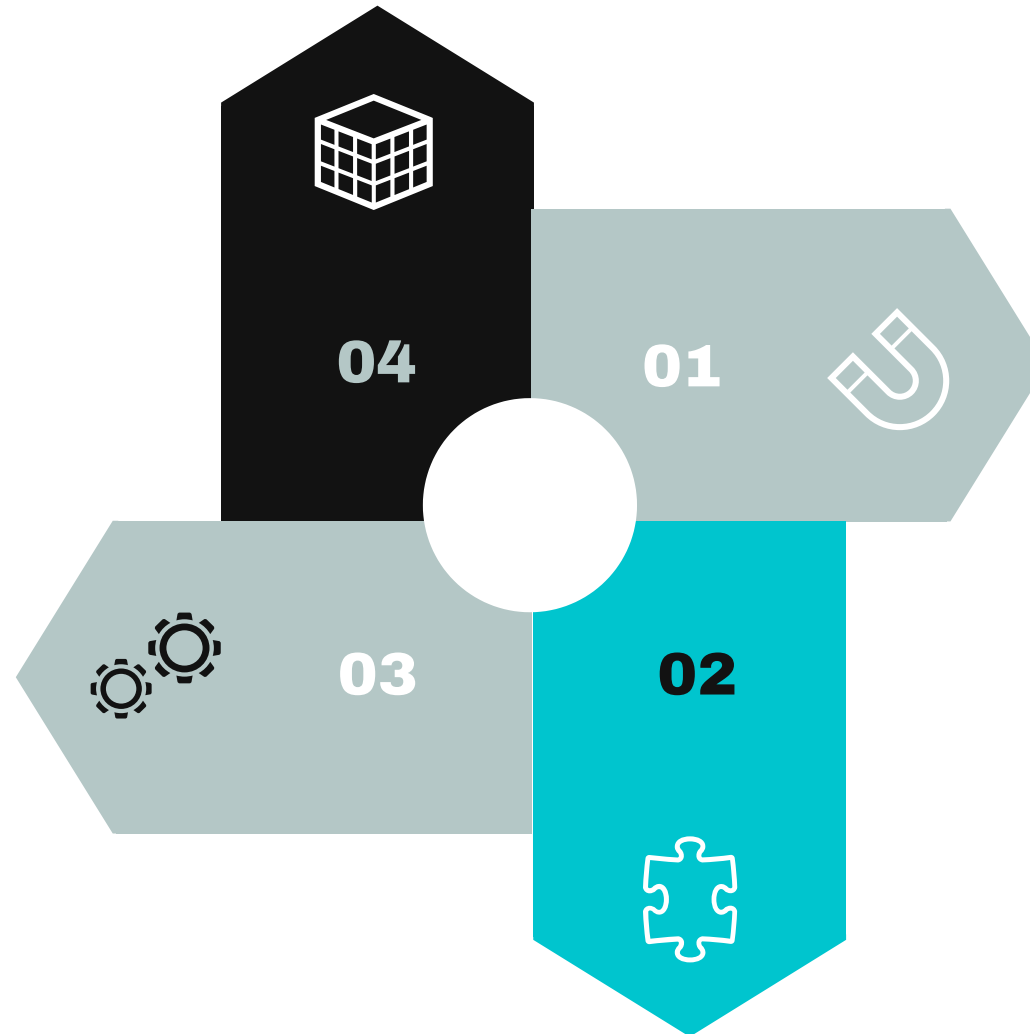
KWG and Epidemic Intelligence from Open Sources (EIOS)?

AI/ML Ready

KWG data (down to the level of individual observations) is directly ready for usage as features in AI/ML applications.

Not Another Portal

Instead of being the next software-based portal (remember Flash?), we serve smart data ready to use with off-the-shelf software.



Finding the right people

We use ML technology to infer expertise and recommend experts across topics and geographies.

Cross Domain

Querying from KWG into SPOKE allows crosswalks from the place, to its population, local exposure, down to the pathogen and treatment.

Team and Partnership



Team



Randall Barker
Senior Personnel
IN10T



Bruno Basso
Senior Personnel
MSU



Ling Cai
Student at STKO Lab
Department of Geography
UC Santa Barbara



Ty Fitzpatrick
Senior Personnel
Esri



Catherine Foley
Senior Personnel
MSU



Zhining Gu
PhD Student
Arizona State University



Tim Murphy
Senior Personnel
Esri



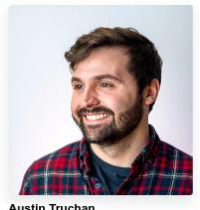
Paulina Oliva
Senior Personnel
USC



Andrew Schroeder
VP of Research and Analysis
Direct Relief



Yuanyuan Tian
PhD Student
Geography
School of Geographical Sciences and
Urban Planning
Arizona State University



Austin Truchan
Designer
MSU



Dalia Varanka
Research Scientist
U.S. Geological Survey



Kitty Currier
Postdoc
UCSB



Frank Davenport
Research Scientist
UCSB Climate Hazards Center



Anthony D'Onofrio
Developer
MSU



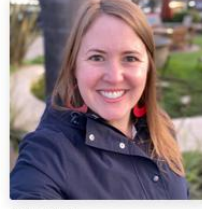
Tony Howser
Senior Personnel
UCSB



Matt Jones
Senior Personnel
UCSB



Zifeng Liu
Ph.D. Student
STKO Lab
Department of Geography
UC Santa Barbara



Alica Sheill
Postdoc Manager
Matrix
Michigan State University



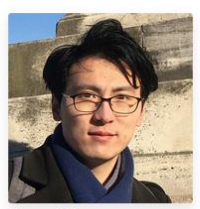
Meilin Shi
PhD Student
UCSB



Cogan Shimizu
Postdoc
K-State



Sizhe Wang
PhD Student
The School of Computing, Informatics,
and Decision Systems Engineering
Arizona State University



Zhangyu Wang
Graduate Student
UCSB



Dawn Wright
Chief Scientist
Esri and Professor of Geography and
Oceanography
Oregon State University



Karen Doehner
UCSB



Andrew Eells
Research Assistant
DaSe Lab
Kansas State University



Colby Fisher
Senior Personnel
Hydronos Labs, Oliver Wyman



Anna Lopez-Carr
Monitoring and Evaluation Specialist
Research and Analysis Group
Direct Relief



Myles McHugh
Senior Personnel
Hydronos Labs



Bryce Mecum
Science Software Engineer
NCEAS, UCSB



David Smith
Soil Scientist
Hydronos Labs, Oliver Wyman
New York



Shirley Stephen
Postdoc
UCSB



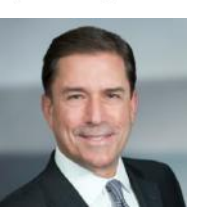
Thomas Thelen
Software Engineer
NCEAS, UCSB



Joseph Zalewski
Ph.D. Student
DaSe Lab
Kansas State University



Rui Zhu
Postdoc
UCSB



Mike Matheis
Oliver Wyman



PIs



Krzysztof Janowicz
Principal Investigator
GeoInformatics
UC Santa Barbara, USA
University of Vienna, Austria



Pascal Hitzler
Computer Science
Kansas State University



Mark Schildhauer
Environmental Sciences
NCEAS, UCSB



Wenwen Li
Urban Planning and Geoinformatics
Arizona State University



Dean Rehberger
Humanities
Michigan State University

Former Team



Sella Gonzalez
Head of Software Development
Matrix
Michigan State University



Gengchen Mai
Ph.D. Candidate
Space and Time for Knowledge
Organization Lab
Department of Geography
UC Santa Barbara



Scott Robinson
Partner
Oliver Wyman's Commodity and Risk
Practices



Lynn Usery
Senior Scientist
U.S. Geological Survey



Lu Zhou
Postdoc
K-State