



SWADE: Smart Water Data Exchange

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Creating an Extensible Data Exchange and Analytics Sandbox for Smart Water Infrastructures



Motivation

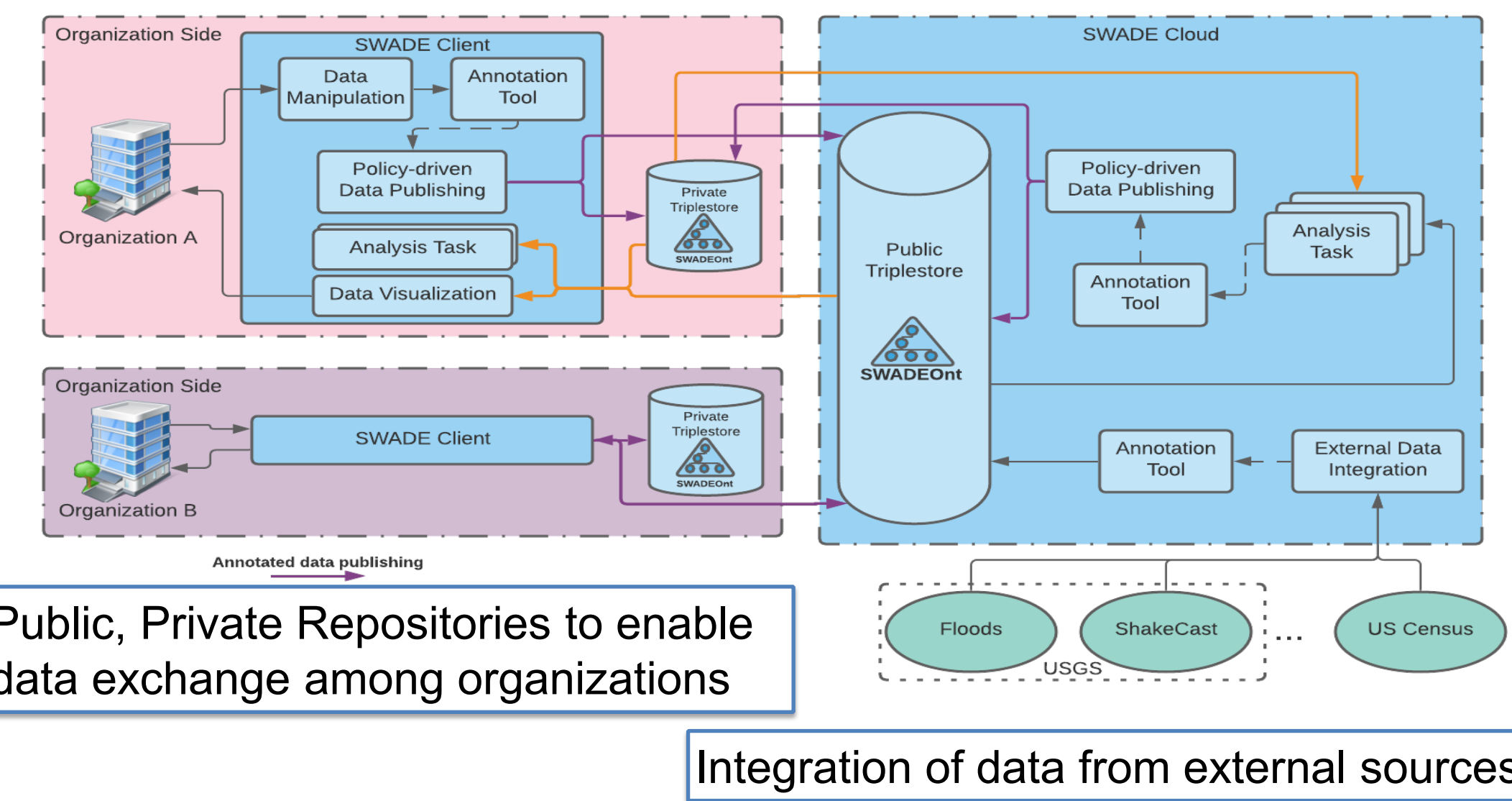
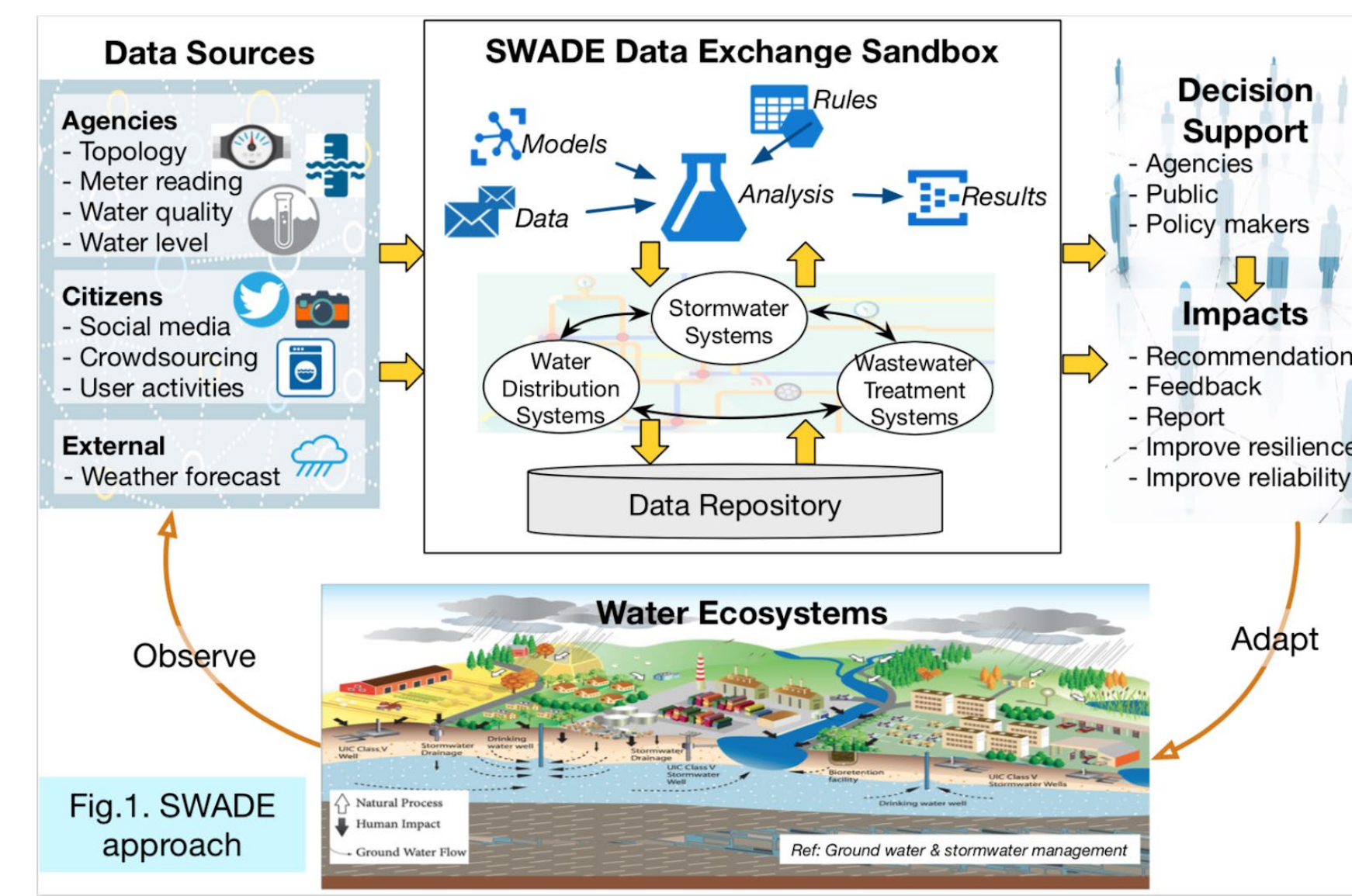
- Water infrastructure is aging and becoming increasingly complex: multiple infrastructures include stormwater, drinking water, and wastewater systems
- Agencies/utilities largely operate independently with specific regulatory compliance needs
- Data and structural information can help in planning and operation, but is often siloed within agencies and systems
- Decision-makers (agencies and policy makers) need tools to interpret the data, identify problem sources and translate it into actions

Key Premise: Water cycle data (historical and live) and its dependencies, a bulk of which resides within community agencies, if combined and enhanced with other geo-distributed data sources can enable new levels of efficiency and resilience.

Goals

- Structure the interconnection of different water infrastructures given an understanding of their reliance / impact on each other under normal operation and under extreme events
- Identify specific information required to enable a data-driven approach to extract and address gaps / opportunities at varying spatial and temporal granularities
- Determine barriers (societal, policy/regulatory) on gathering, sharing and usage of this data
- Identify challenges in translating (restricted) data to meaningful information for timely decision making, possibly under large disruptions (e.g., earthquakes)

- Data exchange architecture used to perform analysis tasks across different water organizations
- analysis performed automatically adapts to organization's data format



Challenges

IT Challenges

- Geo-distributed infrastructure, dynamically changing environment and physical (hydraulic) variables
- Heterogeneity of data sources at varying granularities
- Multiple Stakeholders requiring different combinations of information at different times
- Current approaches agnostic to these challenges: (1) requires significant effort to acquire and understand data; (2) delays in processing information; (3) high levels of data redundancy; (4) lack of infrastructure for data exchange

Interdisciplinary Challenges

- Modeling joint system structure and data interactions within and across systems
- Inter-agency and community data exchange constraints and barriers (data handling, privacy, security)
- Citizen/Community engagement for water-knowledge co-production

Community Engagement

- Stakeholder workshops and data challenge events
 - Workshop to survey stakeholders (agencies, policy makers, academics, industry groups, etc.)
 - What tools will be valuable to them? What are their concerns (privacy, data sharing, accessibility)? What tools can be built to facilitate cooperation?
- Engagement for platform creation and community data instantiation
 - Engagement from SWADE stakeholders to help design and develop platform
 - Leverage currently used tools, datasets, and data formats
- Validation Studies with Community Water Networks
 - Scenario Creation with Partner Agencies
 - Deployment of SWADE Tools for Analysis
- SWADE Internships
 - Embedding students as interns at various agencies
 - Undergraduate and High School Research Opportunities

Domain: Stormwater

Focus Problem: Source Identification for Dry Weather Monitoring



Network Structure and Time Series Analysis

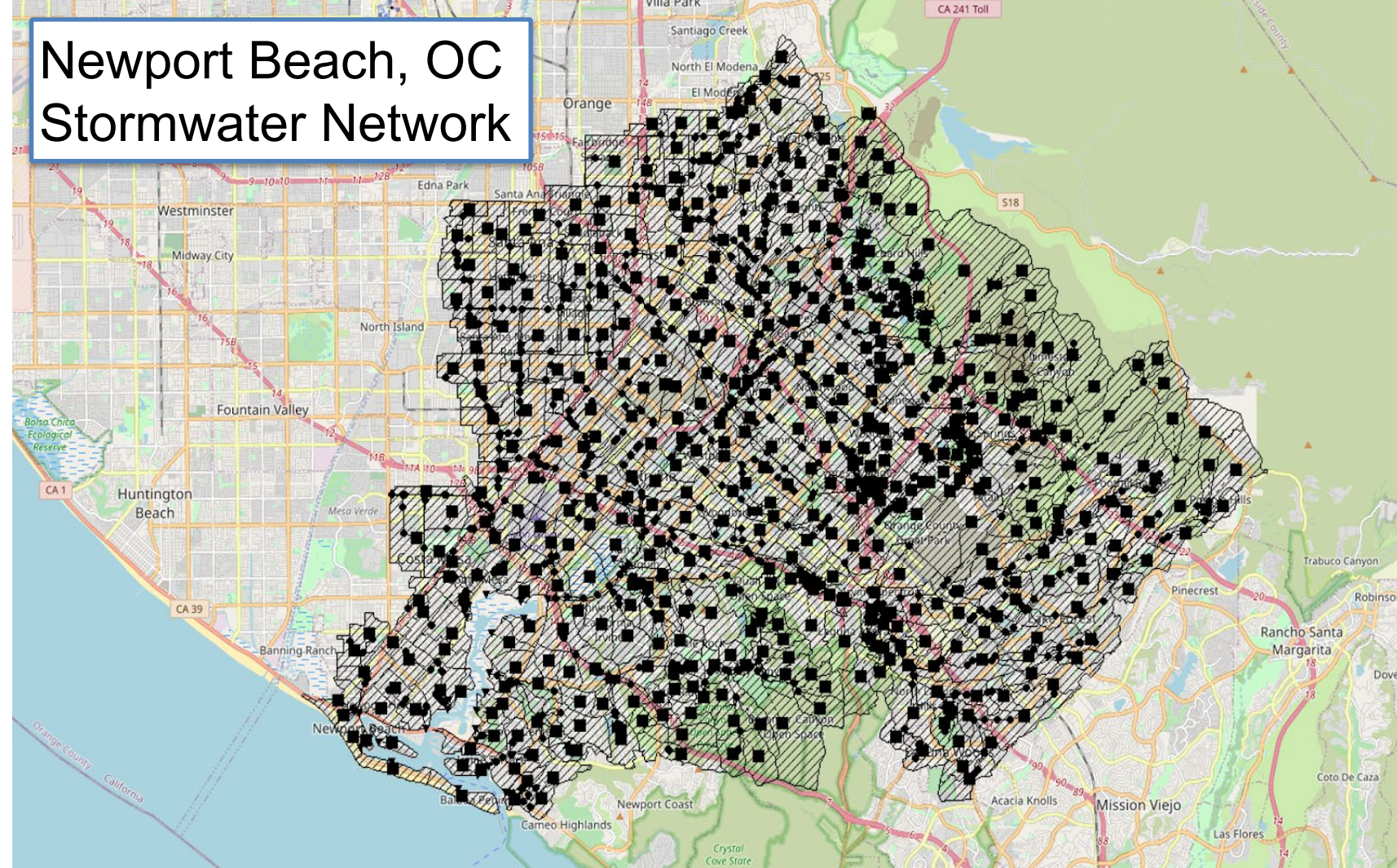
- Detect contamination events and sources rapidly
- Understand role of network structure and its relationship to data generated using time-series analytics

Sensor Placement

- Wet weather and Dry weather monitoring
- Consider tradeoff between Budget and Quality of Monitoring
- What kind of sensors should be considered? Where?

Resource-efficient monitoring

- Deployment & operational costs: underwater installation, human grab sampling.
- Efficiently switching between coarse-grained monitoring (efficient, but less accurate) and fine-grained monitoring (expensive, but high accuracy)



Generalizable data analytics

- Allow other agencies and communities to reuse successful models
- Train Machine Learning models that are robust to location-specific biases / patterns

Domain: Drinking Water

Focus Problem: Infrastructure Resilience to Extreme Events

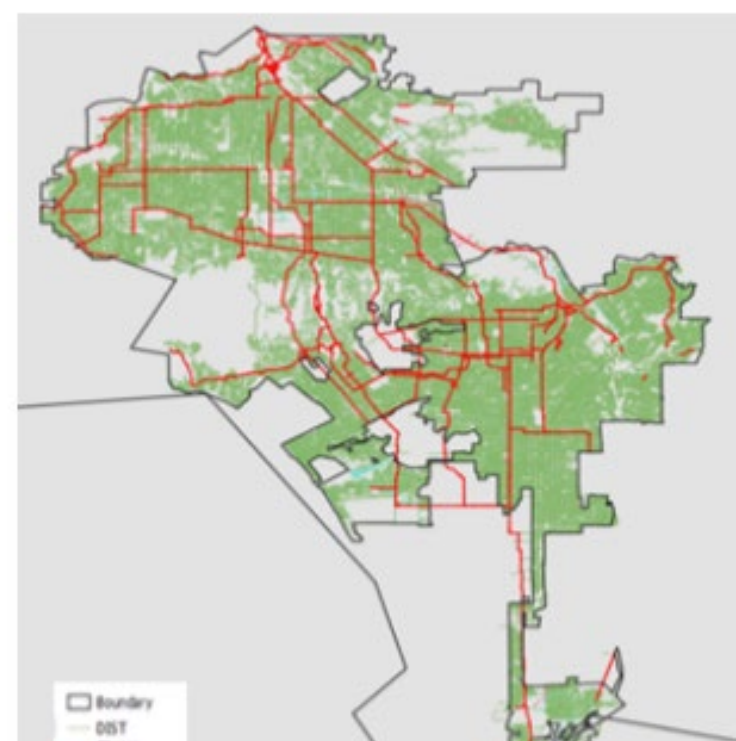
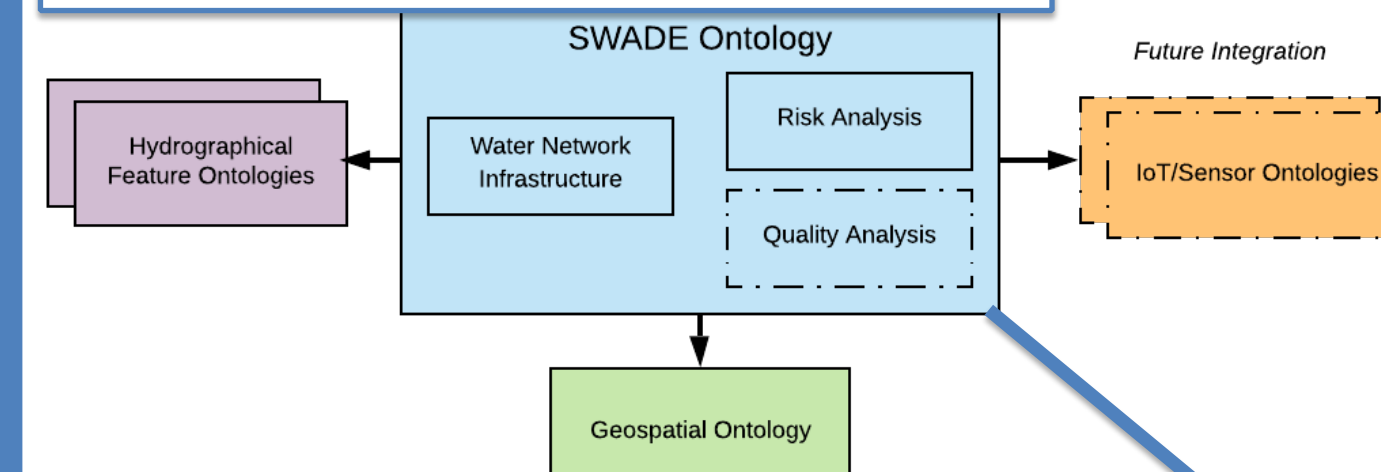


Figure 1: The water transmission and distribution pipeline systems for the City of Los Angeles.

Alignment with existing ontologies



SWADE Ontology

- Vocabulary to enable interoperability of water organizations
- Currently: Definition of water networks and relevant components, geographical elements, hazards, and risk analysis results
- Next: Create vocabulary to represent drinking water systems and water quality analysis

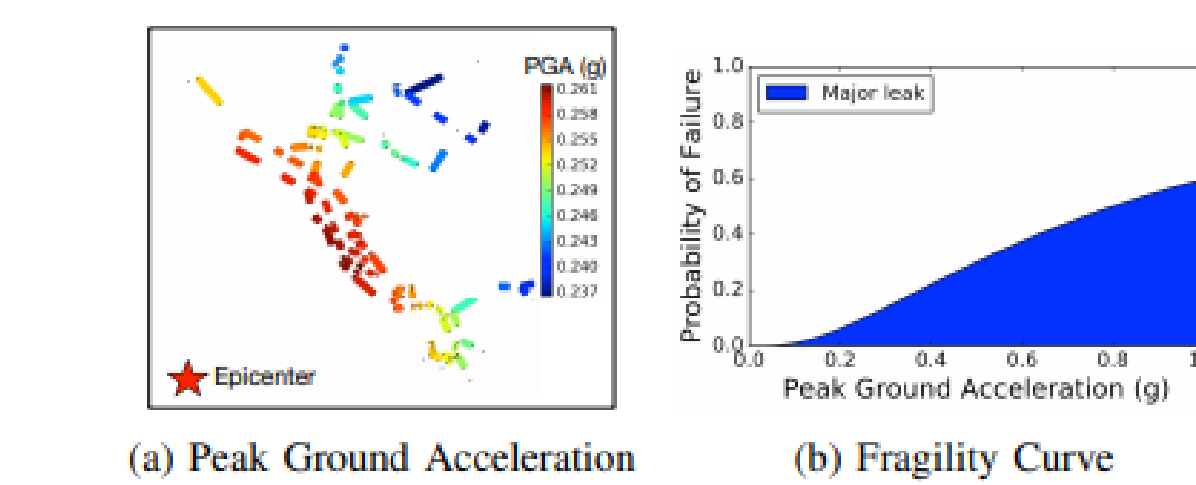


Fig. 7: (a) PGA of pipelines for a magnitude of 5.5 earthquake. (b) Fragility curve for pipe damage.

Exploiting robust simulations

- Identify seismic event scenarios with significant impacts on water distribution networks
- Enable additional instrumentation and retrofit to provide rapid predictions for community safety

Domain: Wastewater

Focus Problem: Process Mining for Wastewater Treatment Plants

Mining Implicit or Hidden Knowledge

- Extract implicit knowledge and experience from historical treatment event logs
- Model discovered knowledge or structural properties (e.g., interdependent flows, timing constraints)
- Improve the practical wastewater treatment workflow based on discovered knowledge and experience

North City Water Reclamation Plant Recycled Water Turbidity Report				
North City Water Reclamation Plant Recycled Water Coliform Report				
North City Water Reclamation Plant NS4-REC WATER: Recycled Water Chlorine Report				
NS4-REC WATER is compliance point for reclaimed water				
Month	Chlorine Residual (mg/L)	Chlorine Residual (mg/L)	Chlorine Residual (mg/L)	Time ¹
Jan	3.31	5.48	400 mg-mv/l	0
Feb	3.35	6.00	0	0
Mar	2.80	5.48	0	0
Apr	3.54	5.21	0	0
May	3.23	6.03	0	0
Jun	3.36	5.91	0	0
Jul	3.27	5.94	0	0
Aug	3.08	6.12	0	0
Sep	2.78	6.11	0	0
Oct	3.44	5.11	0	0
Nov	3.67	5.96	0	0
Dec	3.61	4.84	0	0
Average	3.29	6.42	Total: 0	0

