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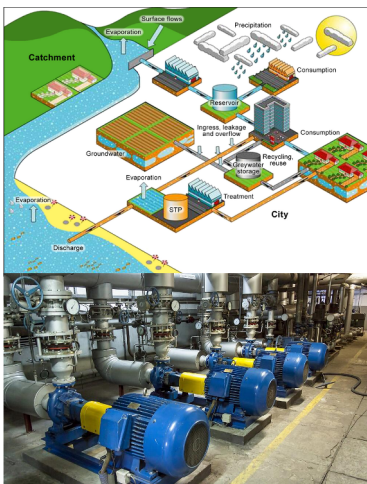
**SWADE Technology Partners & Community Advisory CAB**



# SWADE: Project Motivation

Water infrastructure is aging and becoming increasingly complex

- Multiple Infrastructures, and dedicated regulatory and compliance agencies  
Drinking water    Wastewater    Stormwater

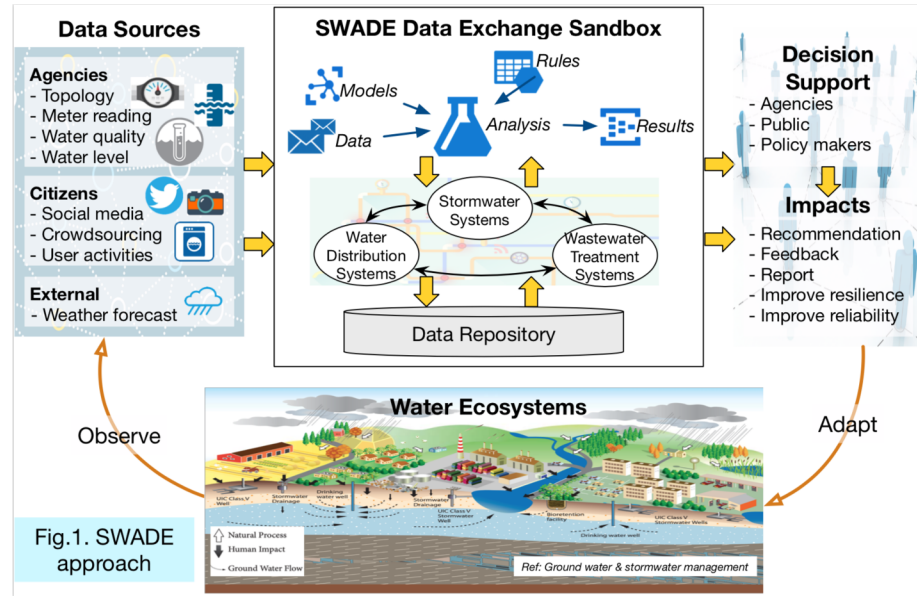


- Data and structural information siloed within agencies and systems
- Decision-makers (agencies, policy-makers) need tools to interpret data, identify problems and take actions

# SWADE: Project Overview

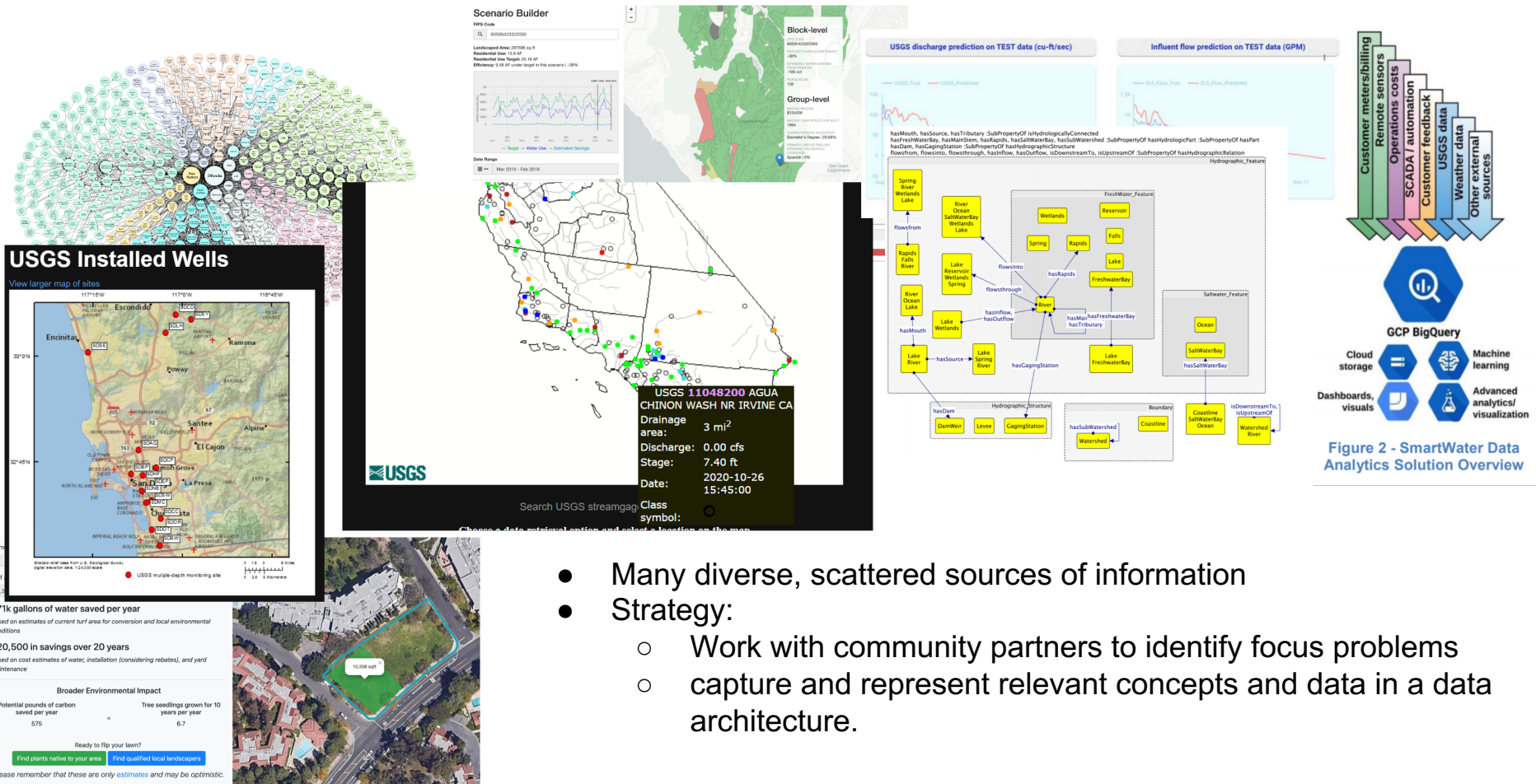
## Key Premise

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



- **Model the different water infrastructures** and interconnections under normal operation or extreme events
- Determine barriers (societal, policy/regulatory) on gathering, sharing and usage of data.
- Identify challenges in translating (restricted) data to meaningful information for timely decision making, especially under large disruptions
- Current approaches: (1) requires significant effort to acquire and understand data; (2) information processing delays; (3) high levels of data redundancy; (4) lack of infrastructure for data exchange

# SWADE: Project Evolution



- Many diverse, scattered sources of information
- Strategy:
  - Work with community partners to identify focus problems
  - capture and represent relevant concepts and data in a data architecture.

Figure 2 - SmartWater Data Analytics Solution Overview

# SWADE: Drinking Water Systems

## Focus Problem :

### Infrastructure Resilience to Extreme Events

- Exploit robust simulations to identify seismic event scenarios with significant impact on water distribution networks
- Enable additional instrumentation and retrofit to provide rapid predictions for community safety

**SWADE Ontology:** Vocabulary to enable interoperability of water organizations

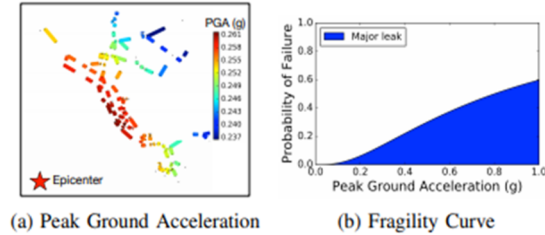
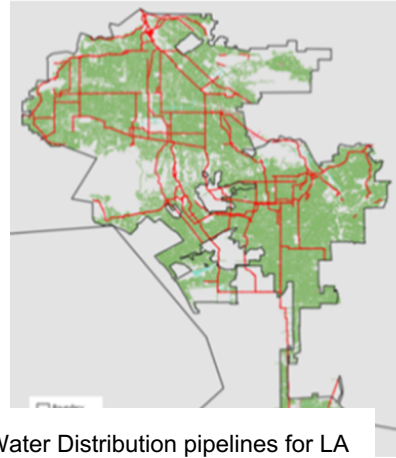
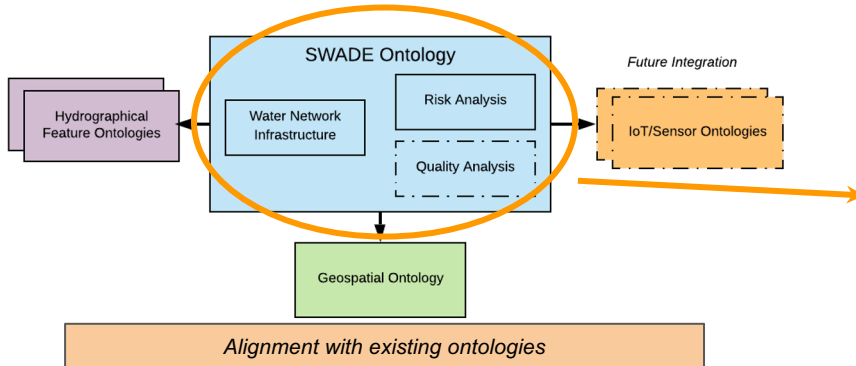
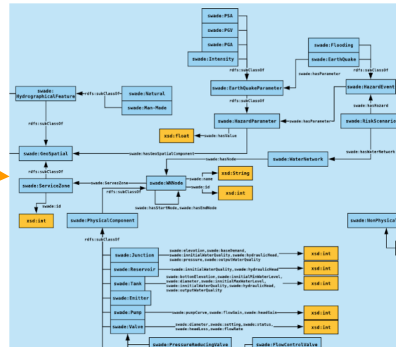
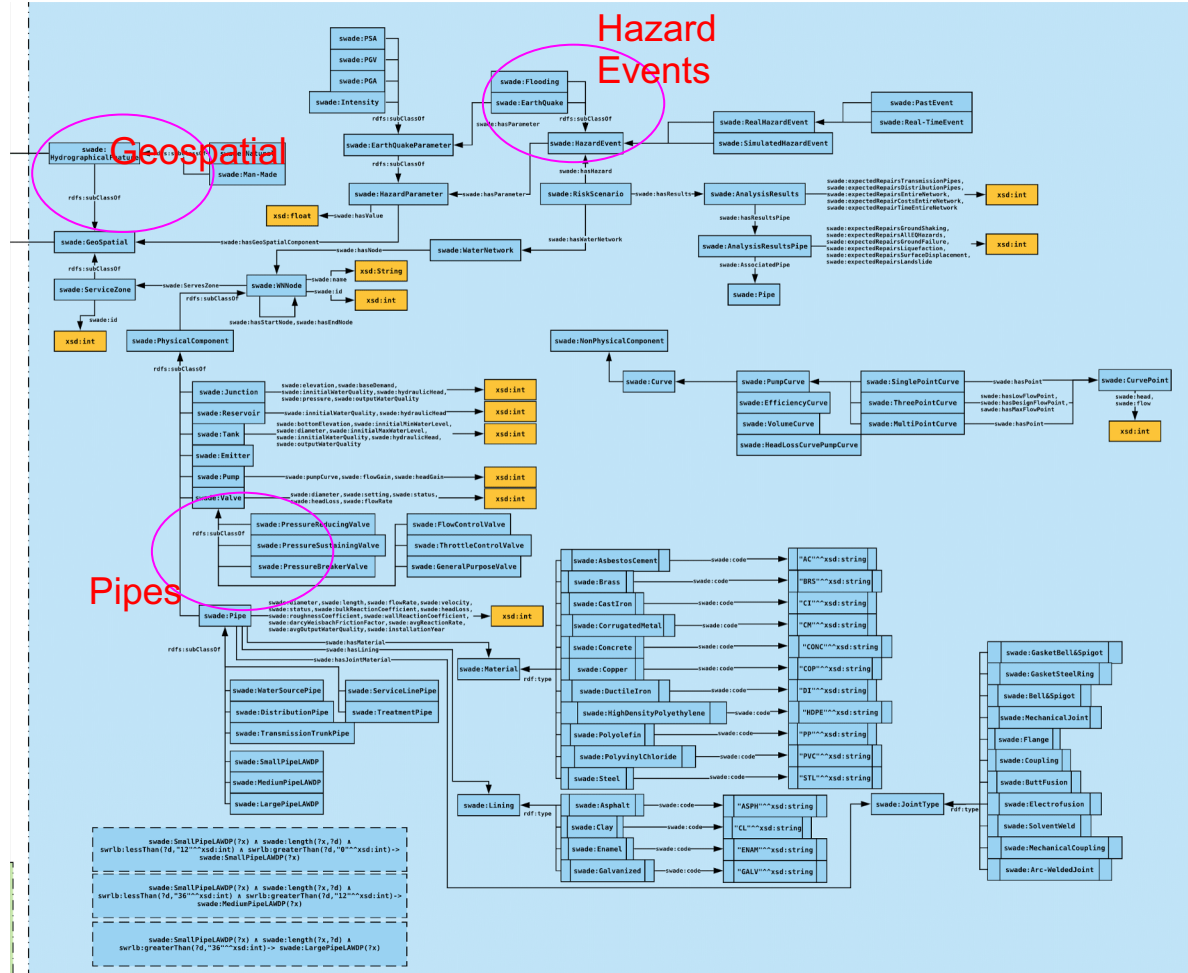


Fig. 7: (a) PGA of pipelines for a magnitude of 5.5 earthquake. (b) Fragility curve for pipe damage.

- **Current version:** Definition of water networks and components, geographical elements, hazards, and risk analysis results.



# SWADE: Ontology



# SWADE - Wastewater Systems

## Focus Problem :

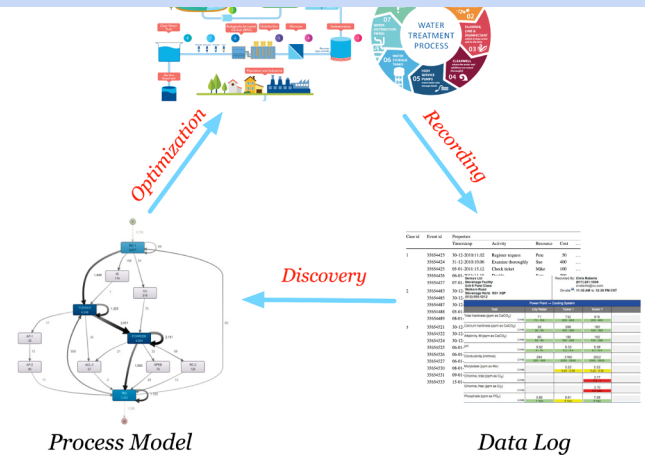
Process Mining for Wastewater Treatment Plants

## Mining Implicit or Hidden Knowledge

- Extract implicit knowledge and experience from historical treatment event logs
- Discover knowledge or structural properties (timing constraints, interdependent flows)
- Use to improve practical wastewater treatment workflow

## Improving operational efficiency with soft-sensors

- Reduce energy demands caused by excess aeration without sacrifice in effluent quality.
- Combine data from physical sensors with regression methods.



Wastewater Treatment Event Logs

North City Water Reclamation Plant				
(N34-REC WATER) Recycled Water Chlorine Report				
N34-REC WATER is compliance point for reclaimed water				
North City Water Reclamation Plant				
Recycled Water Coliform Report				
Operati	Data from in-plant meter *			
Date	Average Daily	Minimum Daily	Maximum Daily	Time Over
	Turbidity	Turbidity	Turbidity	5 WWS's
	(NTU)	(NTU)	(NTU)	(MAREFS)
Jan	0.19	0.16	0.20	0.00
Feb	0.30	0.26	0.50	0.00
Mar	0.31	0.26	0.72	0.00
Apr				
May				
Jun				
Jul				
Aug				
Sep				
Oct				
Nov				
Dec				
1 Minimum	0.16	0.16	0.16	0.00
2 Maximum	0.72	0.26	0.72	0.00
3 Total				

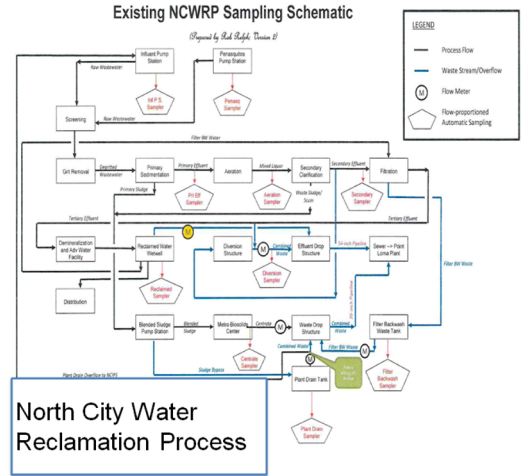
North City Water Reclamation Plant					
Recycled Water Turbidity Report					
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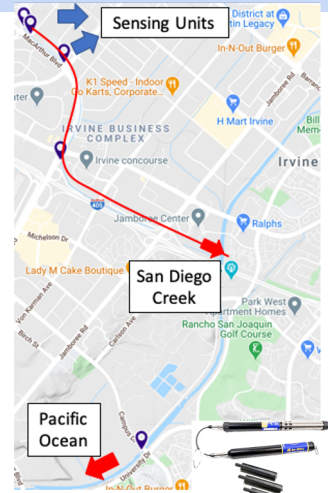
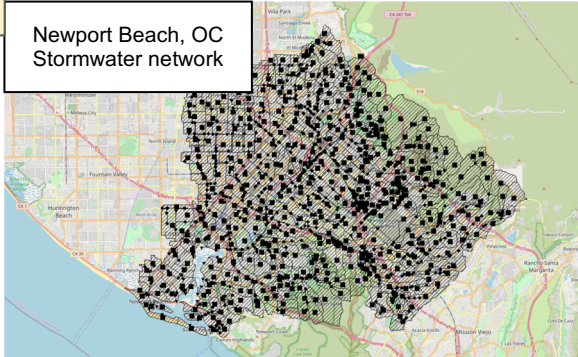


North City Water Reclamation Process

# SWADE - Stormwater Networks

## Focus Problem :

Source Identification for Dry weather monitoring



## Sensor Placement

- Tradeoff - Budget vs. Quality of monitoring
- Need to quickly detect high impact events

## Resource-efficient monitoring

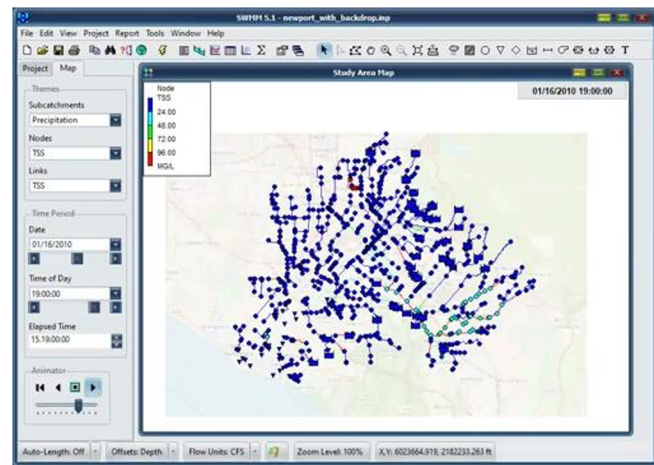
- High deployment & operational costs - underwater installation, human grab sampling.
- Coarse-grained monitoring (efficient, less accurate) vs. fine-grained monitoring (expensive, high accuracy)

## 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

## Generalizable Data Analytics

- Allow other agencies/communities to reuse successful models
- Train ML models that are robust to location-specific data biases





# 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 build to facilitate cooperation?

The logo for Water UCI, featuring the word "Water" in white and "UCI" in white on a dark blue rectangular background.

- **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 analytics

- **SWADE Internships**

- Embedding students as interns at various agencies.
- Undergraduate and High-school research opportunities

# SWADE: Ongoing work

- System-specific milestones
  - Drinking Water: Create vocabulary to represent drinking water systems and water quality analysis
  - Wastewater: Examine data representation needs for operational efficiency
  - Stormwater: Integrate time-series analytics with network structure analysis, explore generalization methods to improve robustness of ML techniques
- Identify relationships and interdependencies between different water infrastructure systems
- Expand data representation and ontology to capture multiple relationships