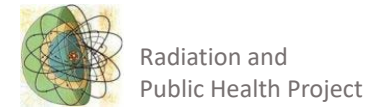


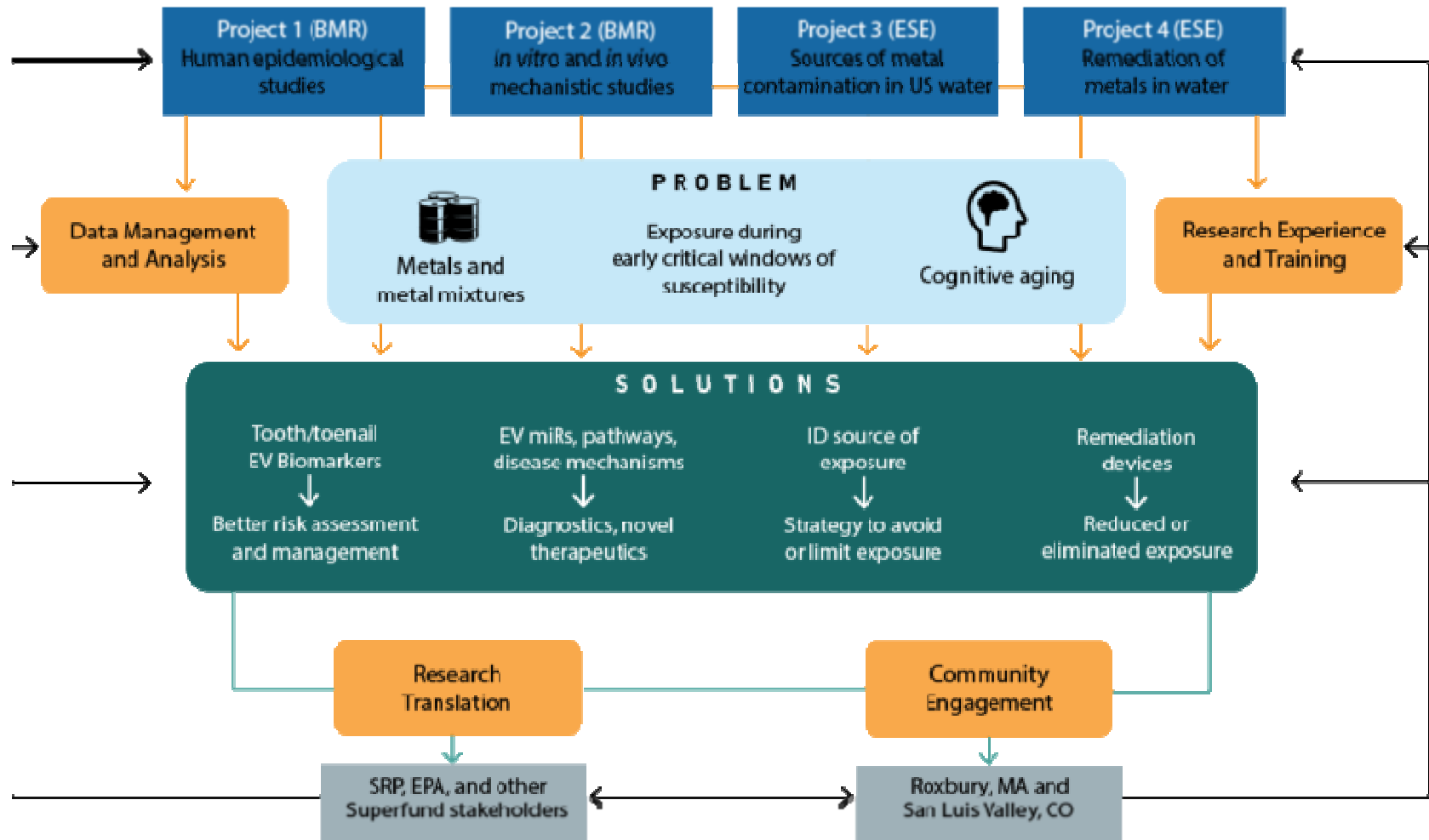
# MEMCARE (Metals and Metal Mixtures: Cognitive Aging, Remediation, and Exposure Sources)

To understand and mitigate the effects of exposure, particularly early life exposure, to metals and metal mixtures on late life cognitive health.





# MEMCARE SUPERFUND RESEARCH CENTER



# Project 1: Early life metals exposures and late life cognitive function

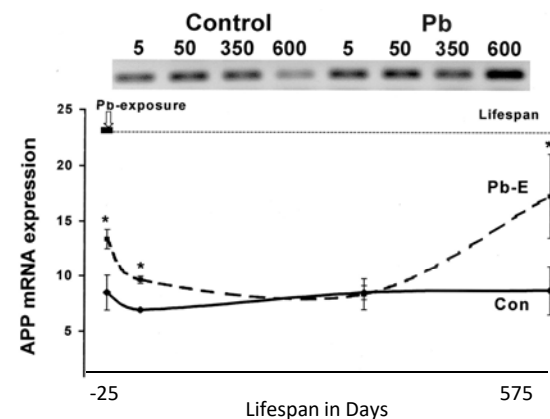
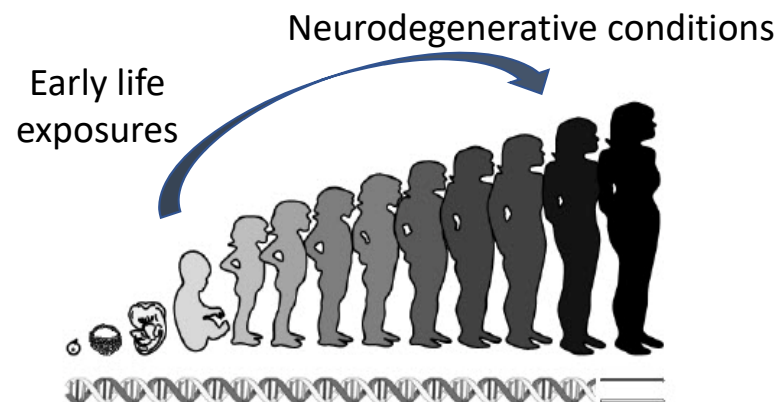
Co-Leaders: Marc Weisskopf and David Christiani  
Harvard T.H. Chan School of Public Health

## Current knowledge

- In utero/early life metals exposures associated with impaired neurodevelopment
- Later life metals exposure associated with worse cognitive function

## Developmental Origins of Health and Disease (DOHAD)

- In utero/early neo-natal lead (Pb) exposure related to late life Alzheimer's like neuropathology in animal models
- No human data



Rat data

*Basha et al., 2005*

# Project 1: Early life metals exposures and late life cognitive function

## Aims:

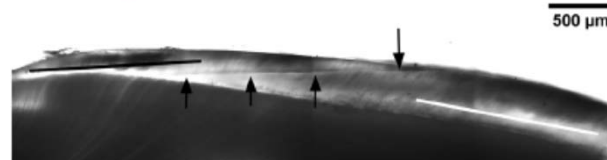
- Assess association of in utero/early life exposure to metals with worse cognitive function in later life
- Assess adult EV miRNA for associations with metals and cognition

## Teeth as a biomarker of early life metals exposure

- Baby teeth develop both pre and postnatally with identifiable timing
- Laser ablation ICP-MS to measure metals

## St. Louis Baby Teeth (SLBT) study

- Children born in the 1950s donated baby teeth
- ~70 years old now
- Re-contacting to assess cognitive function and other aspects of life and health



Part of a lower incisor of a SLBT participant

Enamel (lighter) on top; dentin (darker) below.

Black arrows: neonatal line visible as a darker line in the lighter enamel.

White bar (lower right): Prenatal region (roughly second trimester)

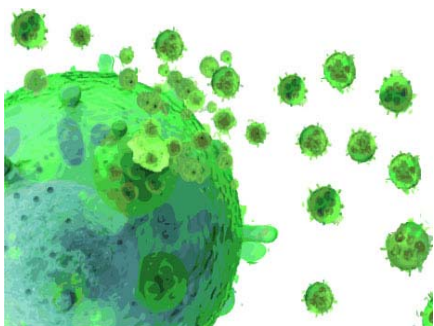
Black bar (top left): Postnatal region (roughly first few months post pregnancy)



## Project 2: Extracellular Vesicle (EV) miRNAs in cognitive function decline associated with early-life metal exposure

Co-Leaders: Quan Lu, Harvard T.H. Chan School of Public Health and  
Takao Hensch, Boston Children's Hospital

### Extracellular vesicles



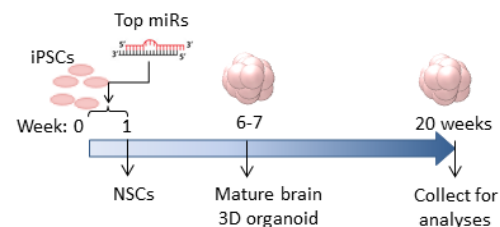
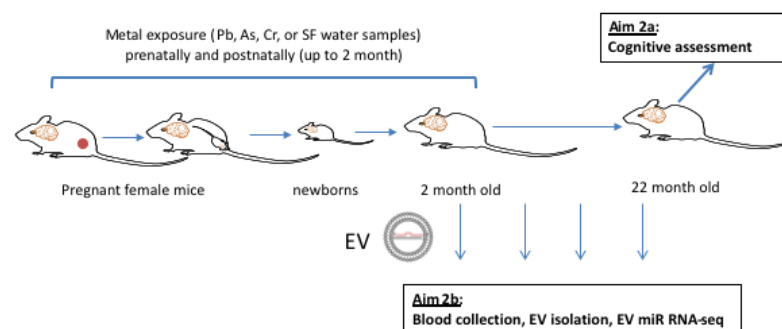
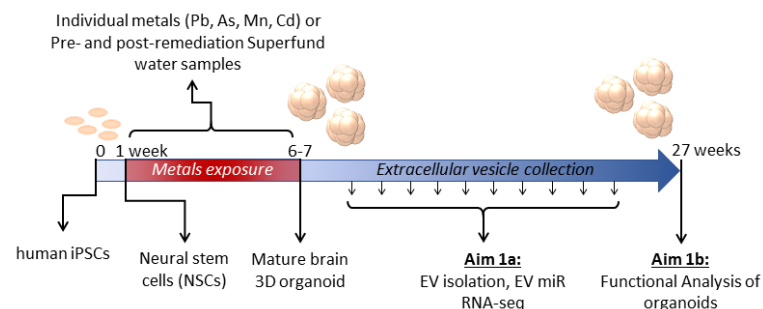
- Cell communication
- Biomarkers

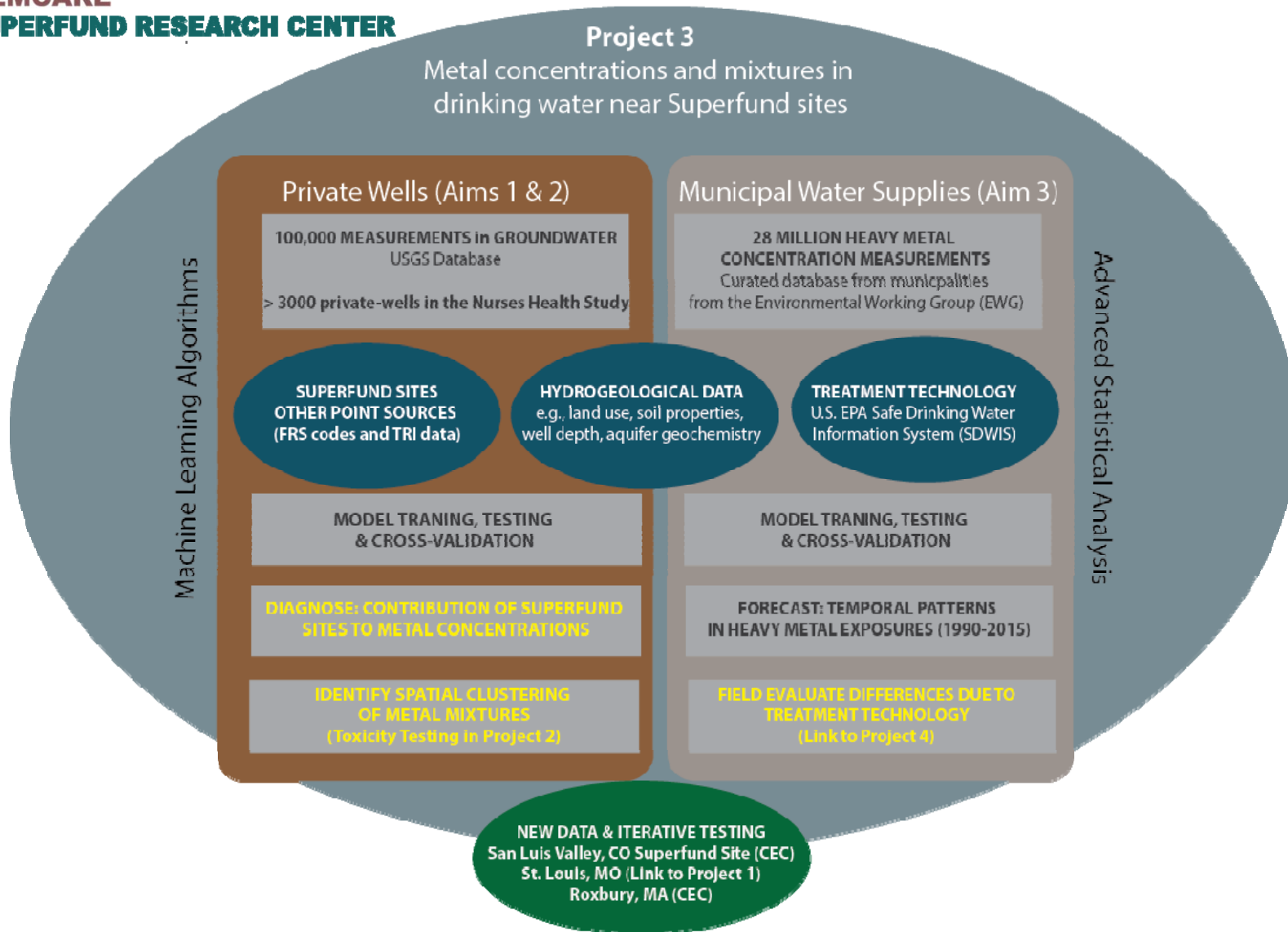
Our goal is to establish EV miRNAs not only as novel biomarkers for metal exposure-related cognitive function, but also as a likely mechanistic basis for metal-induced neurotoxicity and cognitive impairment.

We hypothesize that metal exposures in early life alter EV miRNAs in the brain and that these changes in EV miRNAs affect the function of neurons and neural stem cells to accelerate cognitive aging.

## Specific Aims of Project 2

- to determine the effects of metal exposures on EV miRNAs and neural cell functions using 3D brain organoids
- to determine the effects of early-life metal exposures on cognitive function and EV miRNAs in late-life mice
- to determine the functional role of EV miRNAs in modulating functions of neural cells and cognitive function in mice and in brain organoids





Co Leaders:  
Elsie Sunderland  
Francine Laden



## Motivation

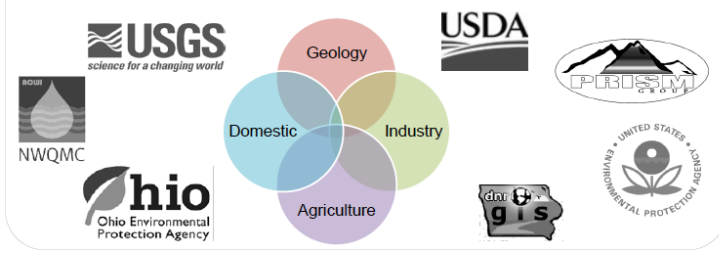
Contamination of drinking water is a significant public health concern

- Private wells serve **15%** of US population and lack routine monitoring
- USGS in 2009 found **23%** contaminated at a level of health concern

Direct measurements of water quality is:



Models with readily available data can help



## Research objectives

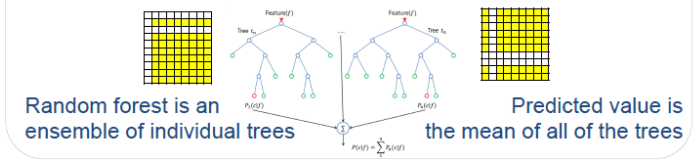
- Assemble a **comprehensive dataset** of measurement data for private wells, and predictor variables representing the **sources, transport and fate** of heavy metals in soils and groundwater
- Develop and test a **random forest model** to estimate groundwater exposure to **arsenic, lead and cadmium** in private wells
- Assess the **predictive performance** of the model at **levels of exposure** that are relevant for **epidemiologic studies**.

## Data & Method

- National heavy metals occurrence data from the Water Quality Portal for well water quality collected during 1989-2017
- Candidate predictors are informed by ground water contamination processes



- Random forest model was trained to predict As, Pb, Cd levels and the probability of exceeding 10 µg/L





## Results

72,976 wells measured for As  
 91,600 wells measured for Pb  
 99,055 wells measured for Cd (1989-2017)

Using a hybrid mechanistic-empirical approach, the model captured the variability in measurements of As, Pb, and Cd levels in ground water.

Table 1. Evaluation of cross-validated models for predicting concentrations and probability of exceeding 10 µg/L

	$R^2_{cont}$	$MSE_{cont}$	$R^2_{dicho}$	$MSE_{dicho}$
As	0.78	234.5	0.78	0.035
Pb	0.80	221.8	0.83	0.013
Cd	0.51	1498.0	0.74	0.024

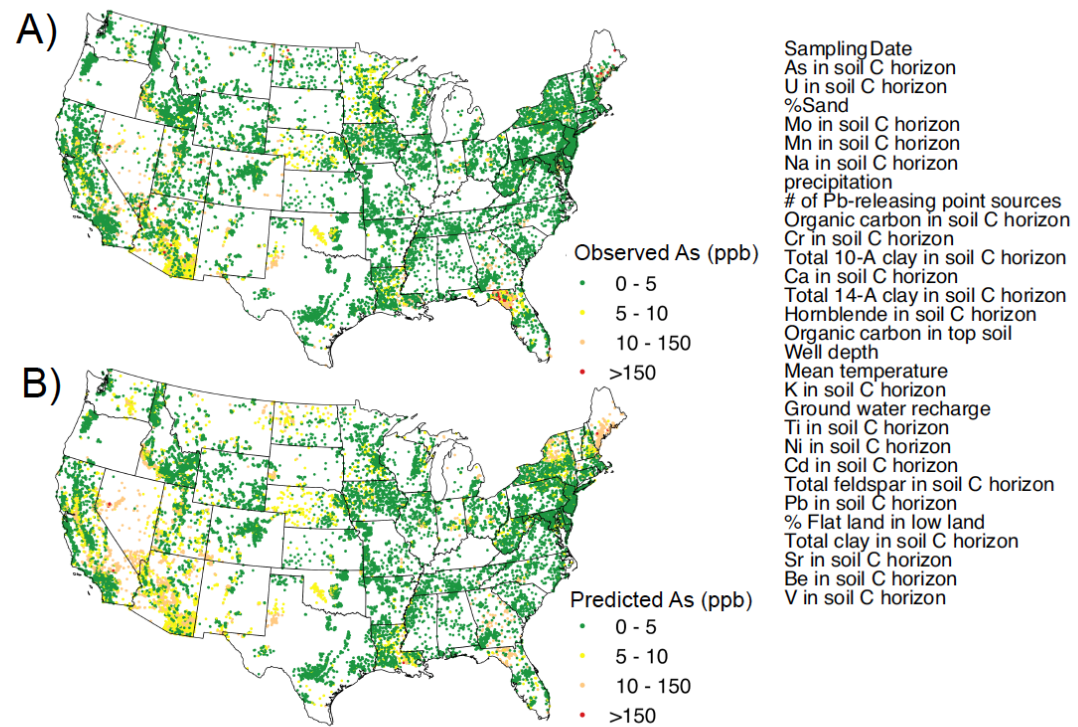


Figure 2. Spatial distribution of A) observed and B) predicted concentration of As in wells.

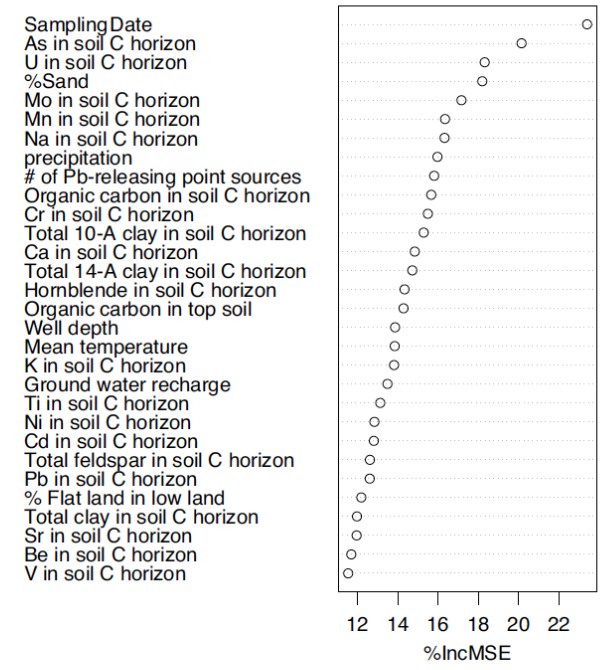


Figure 3. Top 30 variables predicting As levels in random forest model.

## Results (cont.)

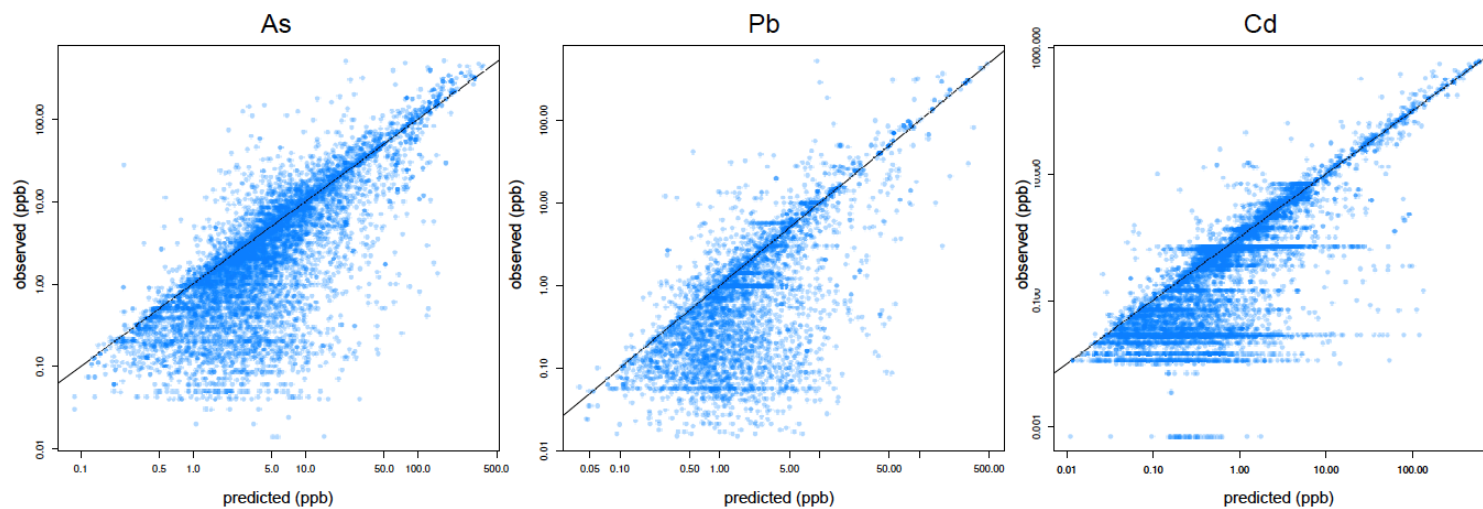


Figure 4. Observed vs predicted groundwater concentrations of As, Pb and Cd for private wells in the testing set across the continental US.

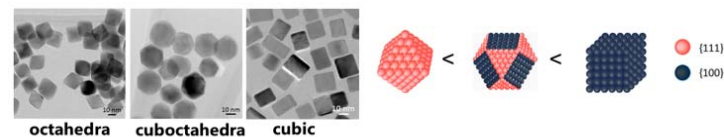
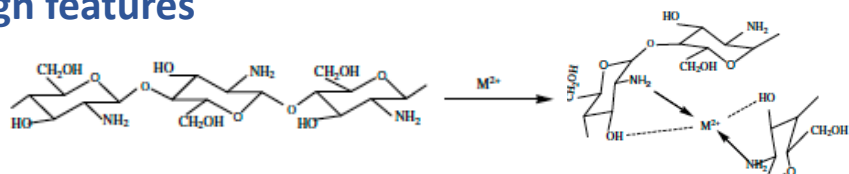
## Project 4: Design and optimization of advanced selective sorbent materials for metal remediation of drinking water

Julie Zimmerman, Yale University

Christopher Muhich and Paul Westerhoff, Arizona State University

The goal of this project is to **design and develop advanced selective adsorbent materials** informed by empirical observations on capacity and selectivity to evaluate functional performance; fine and near edge x-ray spectroscopy to elucidate mechanism; and novel computational approaches from the molecular to system scale to inform sorbent optimization.

# Aim 1: Optimizing synthesis and systematic characterization of two sorbent platforms with novel bottom-up design features

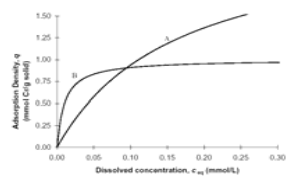


## Platform 1: Transition metal crosslinked biopolymers (TMC)

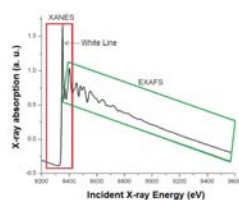
## Platform 2: Crystal facet engineering of nano metal oxides (NMOs)

### Aim 2: Design of optimized Platform 1 for selective removal of target metal oxoanions in competitive environments through small-scale packed bed water treatment systems

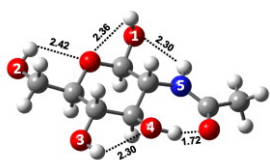
1. batch sorption experiments



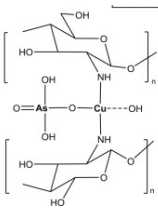
2. characterization of sorption mechanism



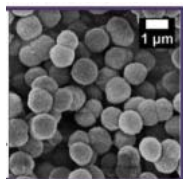
3. DFT models + machine learning



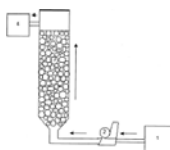
4. Surface complexation and mass transport modeling



5. *a priori* design of optimized Platform 1 sorbents

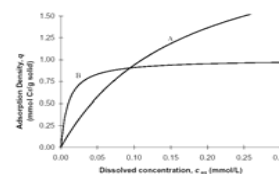


6. scale up and pilot testing in packed beds

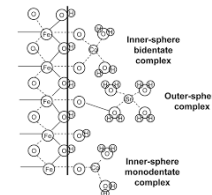


### Aim 3: Design of optimized Platform 2 and incorporation in macroporous electrospun polymers for selective removal of target metals in competitive environments through small-scale fiber membrane-based water treatment systems

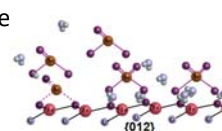
1. batch sorption experiments



2. characterization of sorption mechanism



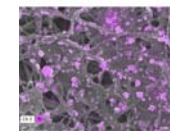
3. DFT models + machine learning



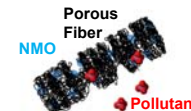
4. *a priori* design of optimized Platform 2 sorbents



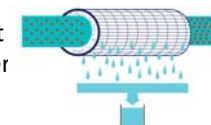
5. Incorporating optimized Platform 2 NMO into porous, electrospun fibers



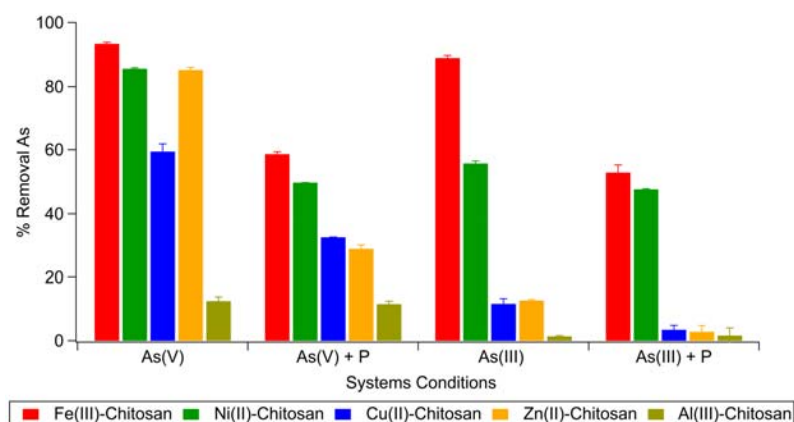
6. optimization of NMO-fibers through mass transport modeling



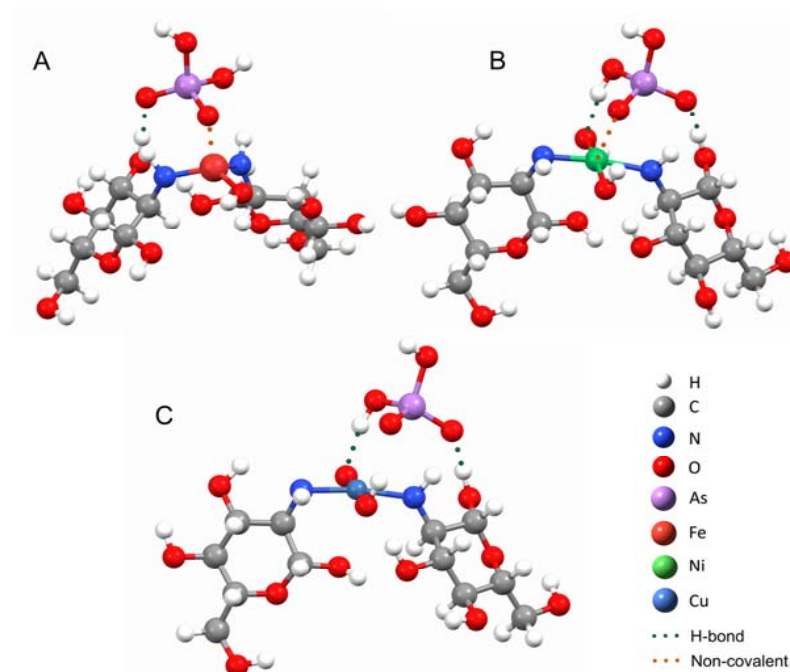
7. scale up and pilot testing in NMO-fiber membrane



# Platform 1

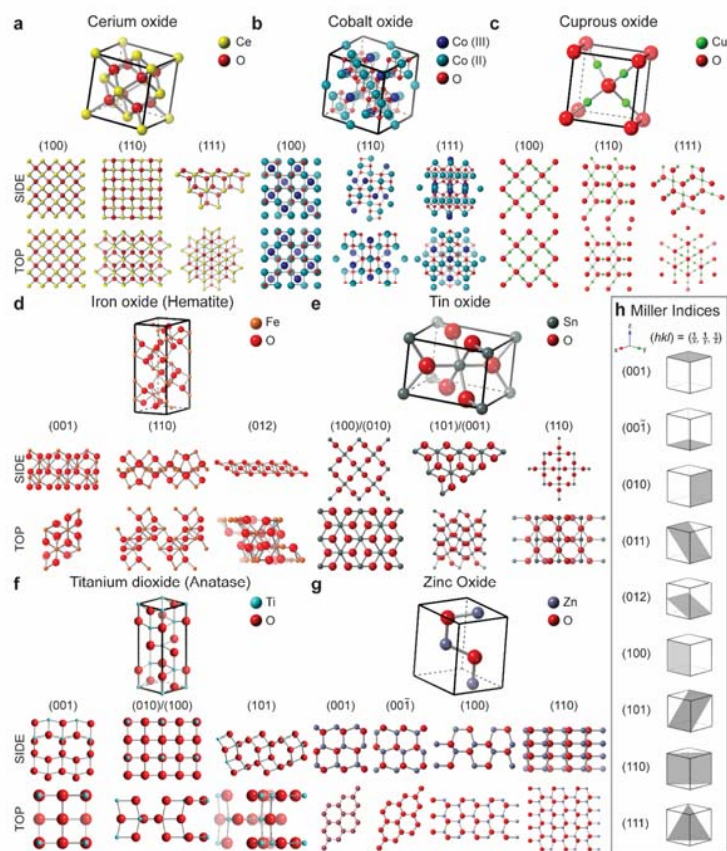


**Figure 1.** Comparison of arsenic removal performance by Fe(III)-chitosan, Ni(II)-chitosan, Cu(II)-chitosan, Zn(II)-chitosan, and Al(III)-chitosan in various systems conditions. Starting concentrations were 4 ppm As and 25 mM acetate buffer pH 6, and 16 ppm P when present.



**Figure 2.** DFT optimized geometries of As(V) adsorption by Fe(III)-chitosan (Figure 4.6a), Ni(II)-chitosan (Figure 4.6b) and Cu(II)-chitosan (Figure 4.6c). Each element in the structure is represented with a different color with white for H, gray for C, blue for N, red for O, purple for As, light red for Fe, green for Ni, and light blue for Cu. Green dots depict hydrogen bonding interactions between As(V) and the metal-chitosan complex. Orange dots indicate inner-sphere non-covalent interactions between As(V) and the metal-chitosan complex.

# Platform 2



**Figure 3.** (Top) Atomic crystal structure with unit cell and respective facet surface structures (middle: side view; bottom: top view) of nanoscale metal oxides: (a) Cerium oxide , (b) Cobalt oxide, (c) Cuprous oxide, (d) Iron oxide (hematite), (e) Tin oxide, (f) Titanium dioxide (anatase), and (g) Zinc oxide. (h) Common low-index crystal planes or facets identified by Miller indices.

# MEMCARE Cores

## **Administrative Core**

Director: Quan Lu

Deputy Director: Julie Zimmerman

Research Translation Coordinator: Trina von Stackelberg

Administrative Coordinator: Julie Goodman

## **Research Experience and Training Coordination Core (RETCC)**

Susan Korrick

Elsie Sunderland

## **Data Management and Analysis Core (DMAC)**

Brent Coull

Xihong Lin

## **Community Engagement Core (CEC)**

Kathy James

Tamarra James-Todd

## Data Management and Analysis Core (DMAC)

- **DMAC Leads:**

- Brent Coull, [bcoull@hsph.harvard.edu](mailto:bcoull@hsph.harvard.edu), 617-432-2376
- Xihong Lin, [xlin@hsph.harvard.edu](mailto:xlin@hsph.harvard.edu), 617-432-2914

- **Aims:**

- State-of-the-art data management
- Ensure sound statistical principles for center design and analysis
- Provide support in geographic information systems (GIS)
- Conduct mission related research.
- Disseminate methodological developments via articles, case studies, web-based software, and short courses.
- Provide education and training for Center students and researchers
- Ensure all projects use state-of-the-art approaches to statistical computing.



## Data and DMAC Activities

- **Data:** 'omics data; imaging data; neurological phenotypes; metal biomarker data; water metal concentrations; residential locations, rich point and areal spatial data; simulation output of molecular geometries, energies, and charges.
- **Data Sharing Strategies:**
  - Submission of 'omics data to dbGAP.
  - Submission of data to NIH Data Commons when appropriate.
  - Use of an open science web portal (OSF) in conjunction with the Harvard Dataverse
  - Free and open-source software packages (DEGAUSS) that are based on containerization, meaning these executable programs contain all code and data when appropriate.
- **Also Using:**
  - Bioinformatics
  - Geographical Information Sciences (GIS) technologies
  - Computational modeling



## Community Engagement Core

Kathy James, University of Colorado

Tamara James-Todd, Harvard T.H. Chan School of Public Health

### The Dimock Center (Roxbury, MA)

*Harvard University*

- Urban partner
- Contamination of neighborhood soil—air pollution
- *Possible* water contaminants due to older housing stock/pipes

### San Luis Valley (Colorado)

*University of Colorado-AMC*

- Rural partner and mining community
- Long history of water contamination with metals
- Soil contamination with metals

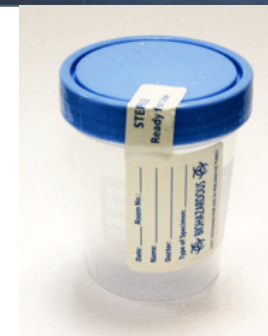
## Objective 1: Community Activities—Online Presence

- Social media— community partners, health clinics, public health centers
- Website advertising: community partners
- Local newspaper and radio
- Hard copy: community flyers
- Facebook, YouTube channel, and Twitter
- Set up bi-annual webinars for mass viewing and recorded
- Set up a community Vlog and Q&A location through YouTube and Facebook
- Online focus groups targeting pregnant women



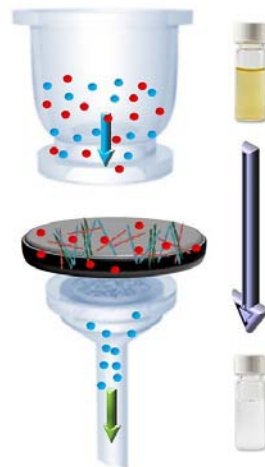
## Objective 2: Participant Recruitment and Engagement in Citizen Science

- Recruitment of Pregnant Women and Children
  - Clinic nurse/midwife will recruit and provide study information
  - Study personnel will contact participants via email or phone
  - Online/telephone consent process (Cisco Jabber)
- Data Collection (4 collection points across 1 year of follow up)
  - Demographic and exposure survey completed online
  - Citizen Science
    - Water, urine, soil sample collection kit mailed to participant
    - Sample collection supplies and instructions (8<sup>th</sup> grade level with pictures)
    - Electronic gift cards to local grocery store with fresh fruit and veggies



## Objective 3: Evaluating interventions for mitigation

- Install water treatment system in qualified participant homes (metals levels > EPA maximum contaminant levels)
  - Monitor water and urine levels after installation
- Improve environmental health literacy within the community related to metals, neurological outcomes, and reducing exposure to vulnerable populations





THANKS to

Dr. Bill Suk, Dr.  
Danielle Carlin and  
many others at the  
Superfund Research  
Program (SRP) for  
guidance and advice

## Questions / Contacts

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Julie Goodman

Center Coordinator

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Visit our Website for more information:

<https://www.hsph.harvard.edu/memcare/>