



Combatting seafood fraud: Application of machine learning to develop chemometric differentiation tools using metal isotope ratios and chemical contaminants in eastern oysters

Elizabeth Crowther, Steven Christopher, Michael Winchester,
Nathan Mahynski, Debra Ellisor



Collaboration Among NIST Groups

- Inorganic Chemical Metrology Group → Elizabeth, Steve, Mike
- Chemical Informatics Group → Nate
- Biospecimen Science Group → Deb











Seafood Fraud

- Examples include mislabeling of species, weight, geographic origin, wild-caught vs. farm-raised sourcing
- 36% of ~9,000 seafood products across 30 countries were found to be mislabeled [Leahy 2021]
- Loss of ~\$26-50 billion per year from the global food industry [Sumaila et al. 2020]
- Interferes with consumers' decisions, poses health risks, undermines environmental conservation and management, negatively affect marine ecosystems

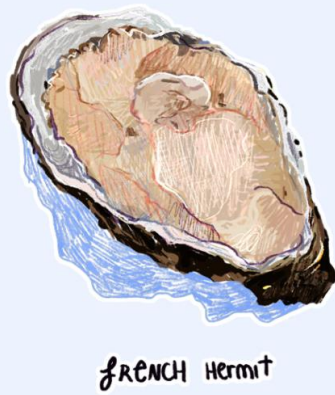
THE COST OF SEAFOOD FRAUD

Example Price Differences for Commonly Swapped Species (8 oz filets)

| AT THE GROCERY STORE | IN A RESTAURANT |
|--|--|
| <p>IF YOU MEAN TO BUY:</p> <p>GROUPEUR \$7.00</p>  <p>BUT YOU GET:</p> <p>TILAPIA \$2.99</p>  <p>YOU LOSE: \$4.01</p> | <p>IF YOU MEAN TO BUY:</p> <p>GROUPEUR \$27.00</p>  <p>BUT YOU GET:</p> <p>TILAPIA \$15.00</p>  <p>YOU LOSE: \$12.00</p> |
| <p>IF YOU MEAN TO BUY:</p> <p>WILD CHINOOK SALMON \$7.50</p>  <p>BUT YOU GET:</p> <p>ATLANTIC FARMED SALMON \$3.99</p>  <p>YOU LOSE: \$3.51</p> | <p>IF YOU MEAN TO BUY:</p> <p>WILD CHINOOK SALMON \$26.30</p>  <p>BUT YOU GET:</p> <p>ATLANTIC FARMED SALMON \$20.70</p>  <p>YOU LOSE: \$5.60</p> |

Goal – Chemometric Modeling

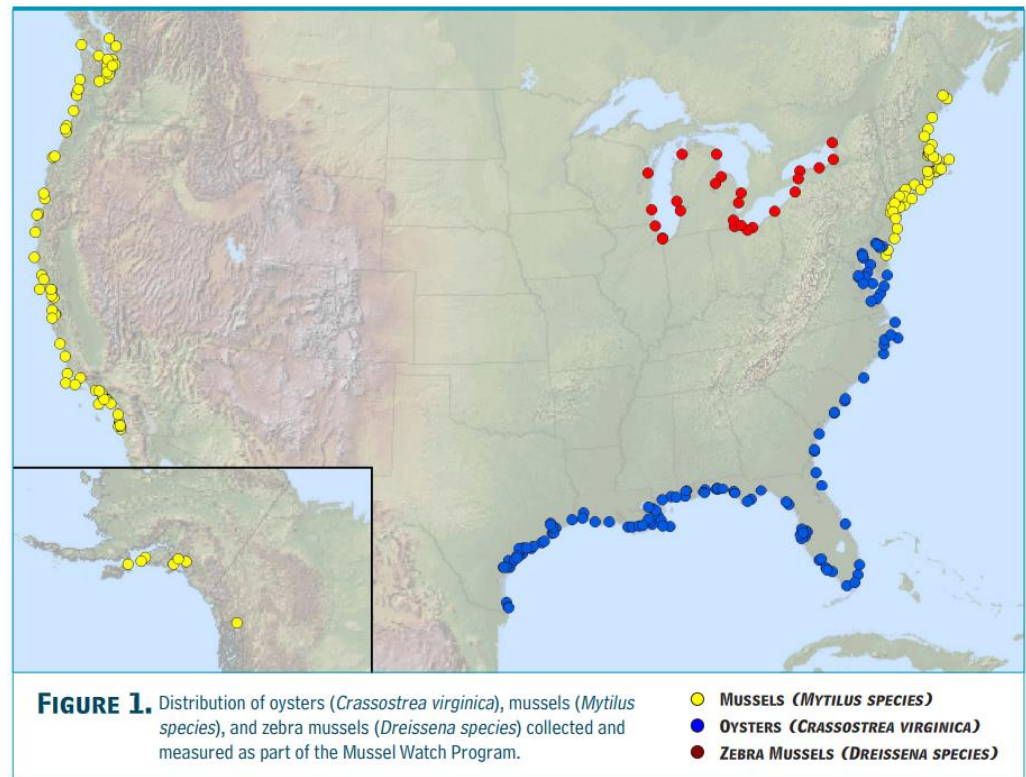
- Eastern oysters have distinct flavors, shell shapes, & textures depending on their geographic origin – influenced by salinity and temperature
- Goal is to determine which types of chemical analyses (metals, organics, isotopes) could be used to determine the geographic origin of oysters
- Develop chemometric models using previous measurements of metal & organic contaminant concentrations, plus new measurements of metal isotope ratios within oyster tissue collected from known locations





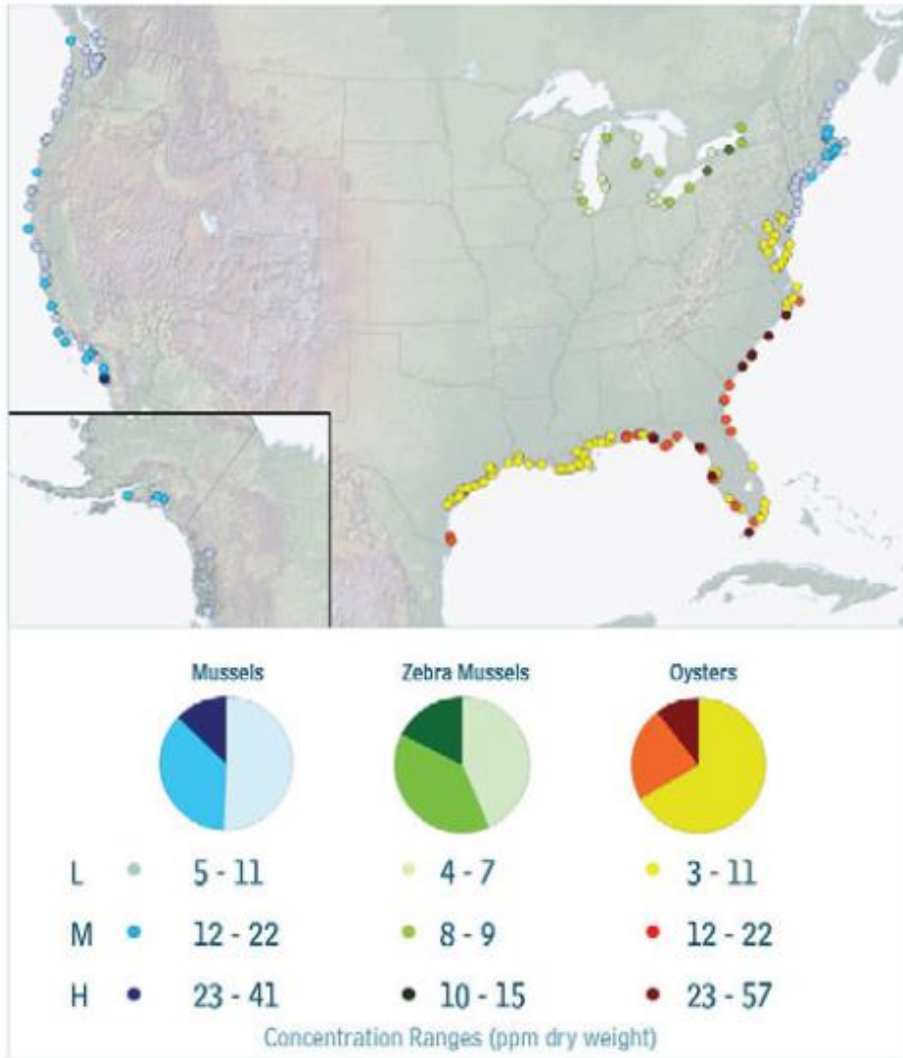
Mussel Watch Program

- Part of NOAA's National Status and Trends (NS&T) Program
- 1986-2012, over 36,000 eastern oysters collected from 123 sites across 61 waterbodies
- ~1,700 oyster tissue composites, each analyzed for ~140 contaminants to assess long-term temporal & spatial trends
- ~174 samples frozen at the NIST Biorepository



Kimbrough et al. 2008

Mussel Watch Program



- Variation in contaminant concentrations in oyster tissue along Gulf & eastern coasts of the US
- Spatial variation is necessary for chemical fingerprinting
- But so is temporal stability...

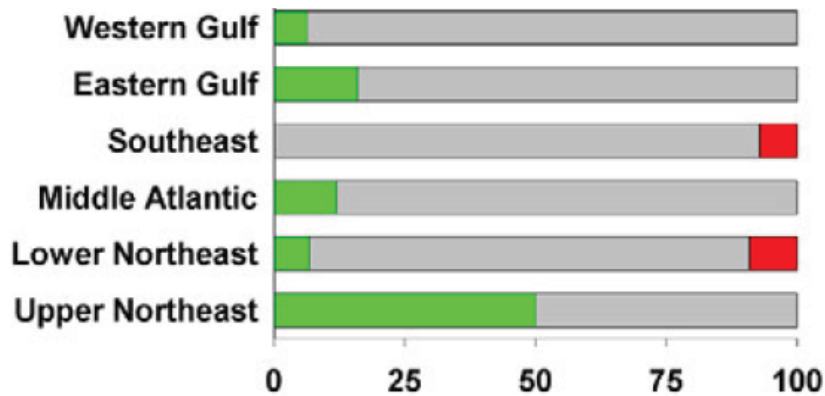
Kimbrough et al. 2008

Mussel Watch Program

- Metal concentrations in oyster tissue were generally stable, except in localized areas where they either increased or decreased
- Organic contaminant concentrations in oyster tissue tended to decrease over time

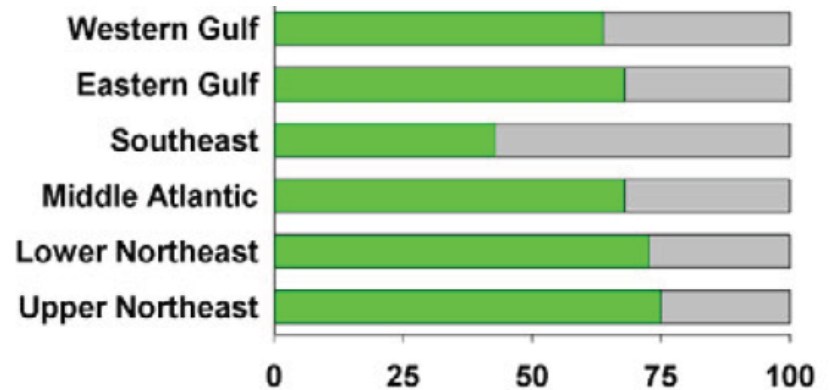
Gray = stable
Green = decreased
Red = increased
(1986-2005)
Kimbrough et al. 2008

Changes in metal concentrations



Regional Comparison of Trends (%)

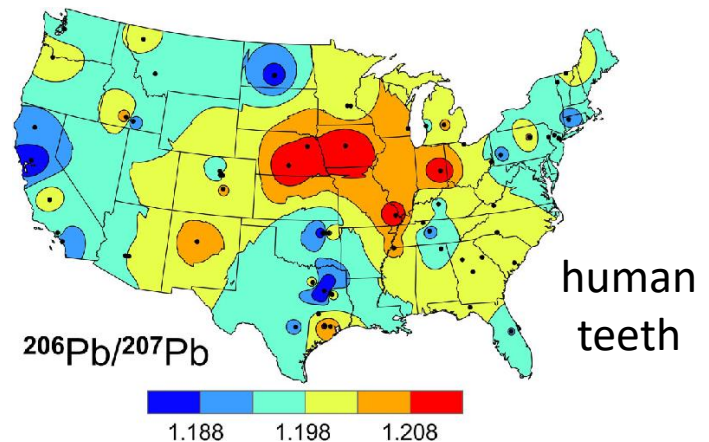
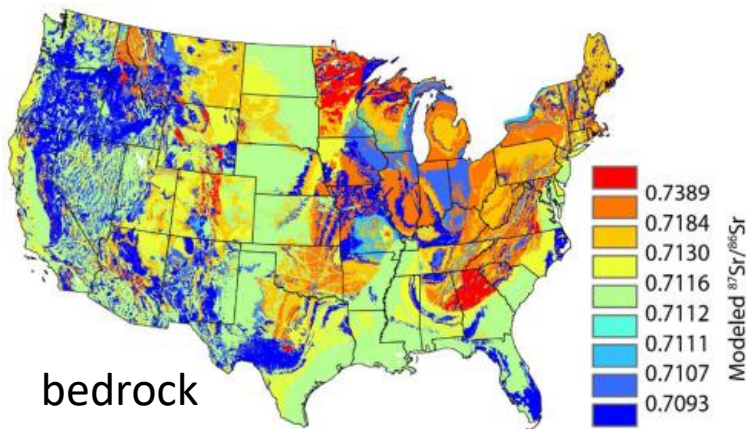
Changes in organic contaminant concentrations



Regional Comparison of Trends (%)

Isotopic Measurements

- Metal isotope ratios may be both spatially variable & stable over time
- Strontium (Sr) → although ocean water has a uniform Sr isotopic composition, stream water along the coast is isotopically variable due to weathering of local bedrock (Bataille & Bowen 2012)
- Lead (Pb), zinc (Zn), & copper (Cu) → isotope ratios in oysters would reflect local anthropogenic contamination, with Zn concentrations in oyster tissue being especially high due to bioaccumulation



Fresh & Archived Samples

Fresh oysters collected in Feb 2023 by:

- Texas Parks & Wildlife Dept
- Louisiana Dept of Wildlife & Fisheries
- Choctawhatchee Basin Alliance (FL)
- Florida Fish & Wildlife Conservation Commission
- South Carolina Dept of Natural Resources
- Chincoteague Bay Field Station (VA)



Sampling sites and dates of oyster composite samples for Sr, Pb, Zn, & Cu isotope analysis.

| State | General location | Specific location | NS&T site code | Oyster sample collection dates |
|-------|--------------------|---------------------|----------------|------------------------------------|
| TX | Galveston Bay | Confederate Reef | GBCR | 1986, 1987, 1992, 2007, 2023 |
| LA | Breton Sound | Bay Gardene | BSBG | 1986, 1991, 2005, 2006, 2008, 2023 |
| FL | Choctawhatchee Bay | Off Santa Rosa | CBSR | 1986, 1991, 1992, 2006, 2008, 2023 |
| FL | Matanzas River | Crescent Beach | MRCB | 1988, 2005, 2007, 2023 |
| SC | Charleston Harbor | Fort Johnson | CHFJ | 1986, 2008, 2023 |
| SC | Charleston Harbor | Shutes Folly Island | CHSF | 1987, 2023 |
| SC | Winyah Bay | Lower Bay | WBLB | 1990, 2006, 2008, 2023 |
| VA | Chincoteague Bay | Chincoteague Inlet | CBCI | 1988, 2006, 2008, 2023 |

Lab Methods

- Oysters were shucked, and tissue was cryogenically homogenized
- Next steps: acid digestion → metal concentration analysis (triple quad ICP-MS) → resin column separation → recovery check (triple quad ICP-MS) → isotopic analysis (MC-ICP-MS)



Chemometric Modeling

- Filter data to eliminate “holes” → choose thresholds for number of measurements per site & per year; fraction of data below detection limit; number of times particular measurements were not performed, etc.
- Option to calculate Jensen-Shannon divergence (JSD) for each analyte at each location → Higher JSD values signify analyte concentrations or isotope ratios that stand out at a particular sampling site compared to other sites (e.g., due to environmental contamination)
- Potential models to try [Hastie et al. 2009; Kamath and Liu 2021]
 - partial least-squares discriminant analysis (PLS-DA)
 - soft independent modelling by class analogy (SIMCA)
 - random forest
 - gradient-boosted tree
 - explainable boosting machine
 - support vector classifier
 - k-nearest-neighbor

Application & Significance

- Combining data from long-term monitoring efforts with new measurements on both archived and freshly collected samples
- Leverages government infrastructure and long-term investments in sample collection, analysis, and archival storage
- Applicable to the U.S. oyster industry, and would provide a framework for developing models applicable to other types of seafood
- Combatting seafood fraud and illegal, unreported, & unregulated (IUU) fishing is a top priority for the U.S. within the 2020 Maritime SAFE Act
- This Act outlines ways in which federal agencies will “expand the role of technology for combatting IUU fishing and seafood fraud” and develop methods of sharing information and data on IUU fishing and seafood fraud using “big data analytics and machine learning”.

Questions?

