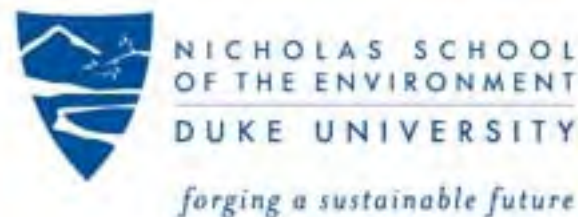


Helping contaminants emerge: The role of non-targeted analysis in monitoring contaminants of emerging concern

P. Lee Ferguson,

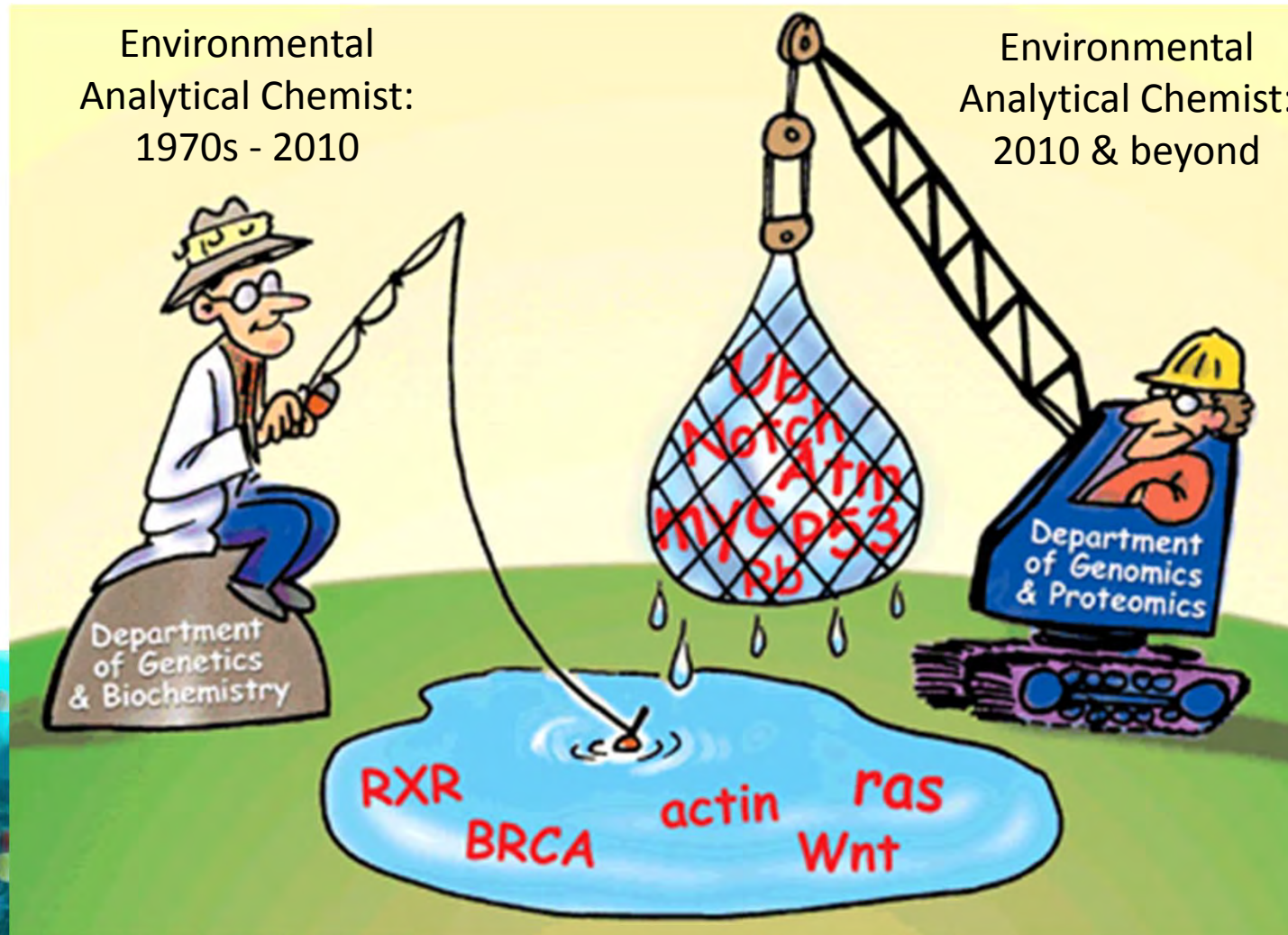
Department of Civil & Environmental Engineering and Nicholas School of the Environment, Duke University, Durham, NC

lee.ferguson@duke.edu



What are the next emerging contaminants and how can we find them in the environment?

LF5



Science 16 February 2001: vol. 291 no. 5507 1221-1224

Slide 2

LF5

Probably expand this a bit to make it more relevant to wastewater... tie to treatment processes, etc. etc.

Lee Ferguson, 6/7/2013

LC-HRMS: An emerging technique for “helping contaminants emerge”

LC-MS strategies for characterization of organic contaminants			
Screening technique:	<i>Targeted</i>	<i>Suspect</i>	<i>Non-target</i>
Question:	<i>Are compounds x, y, & z present in this sample?</i>	<i>Which compounds of a defined list are present in this sample?</i>	<i>Which compounds are present in this sample?</i>
Compound Types:	<i>Known-knowns</i>	<i>Known-unknowns</i>	<i>Known-unknowns & unknown-unknowns</i>



“There are known knowns; there are things we know we know. We also know there are known unknowns; that is to say we know there are some things we do not know. But there are also unknown unknowns – the ones we don't know we don't know.”

In context of water reuse, it is critical to identify the most relevant CECs for monitoring

- The major challenge in non-target analysis of CECs is compound ***prioritization*** for data reduction.
- CECs may have process-dependent fate within water/wastewater treatment.
- Needed: methods to allow focus on recalcitrant, high-abundance, and/or particularly toxic CECs.

Two Approaches:

- Fate-dependent analysis of CECs in water/wastewater treatment
- Activity-directed analysis of CECs that may have particular biological activities

Fate-dependent analysis of CECs in wastewater^{LF6}



Slide 5

LF6

Update and modify to reflect the new study

Lee Ferguson, 6/7/2013

Sample Preparation and Instrumental Analysis

Sampling:

- Daily grab samples (Tue-Fri)
- Triplicate sampling on one day



Sample enrichment:

- 500 mL sample (primary effluent diluted 1:5 (v/v))
- Spiked with stable isotope labeled standards (19)
- Automated SPE, 500 mg Oasis HLB



UHPLC

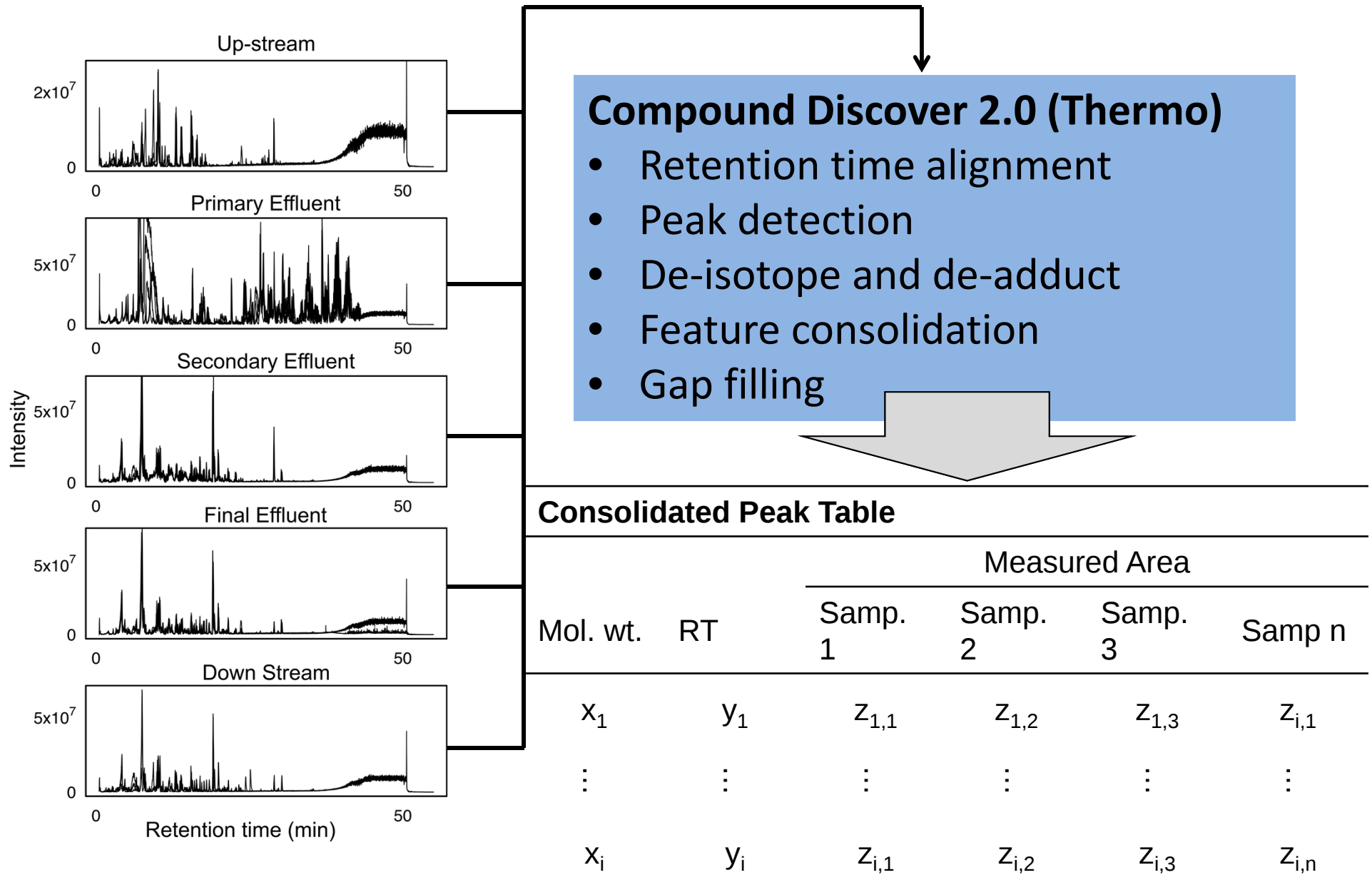
- Dionex Ultimate 3000, 100x2.1 Thermo Hypersil Gold aQ
- H₂O/ACN gradient, 95% to 1% H₂O in 55 min, 0.5 mL min⁻¹

High-resolution mass spectrometry

- Thermo LTQ Orbitrap Velos, ESI(+)
- Full-scan (m/z 100-2000), accurate mass, R=60k FWHM
- Top-4 data-dependent accurate mass MS², R=7500 FWHM



Data analysis workflow

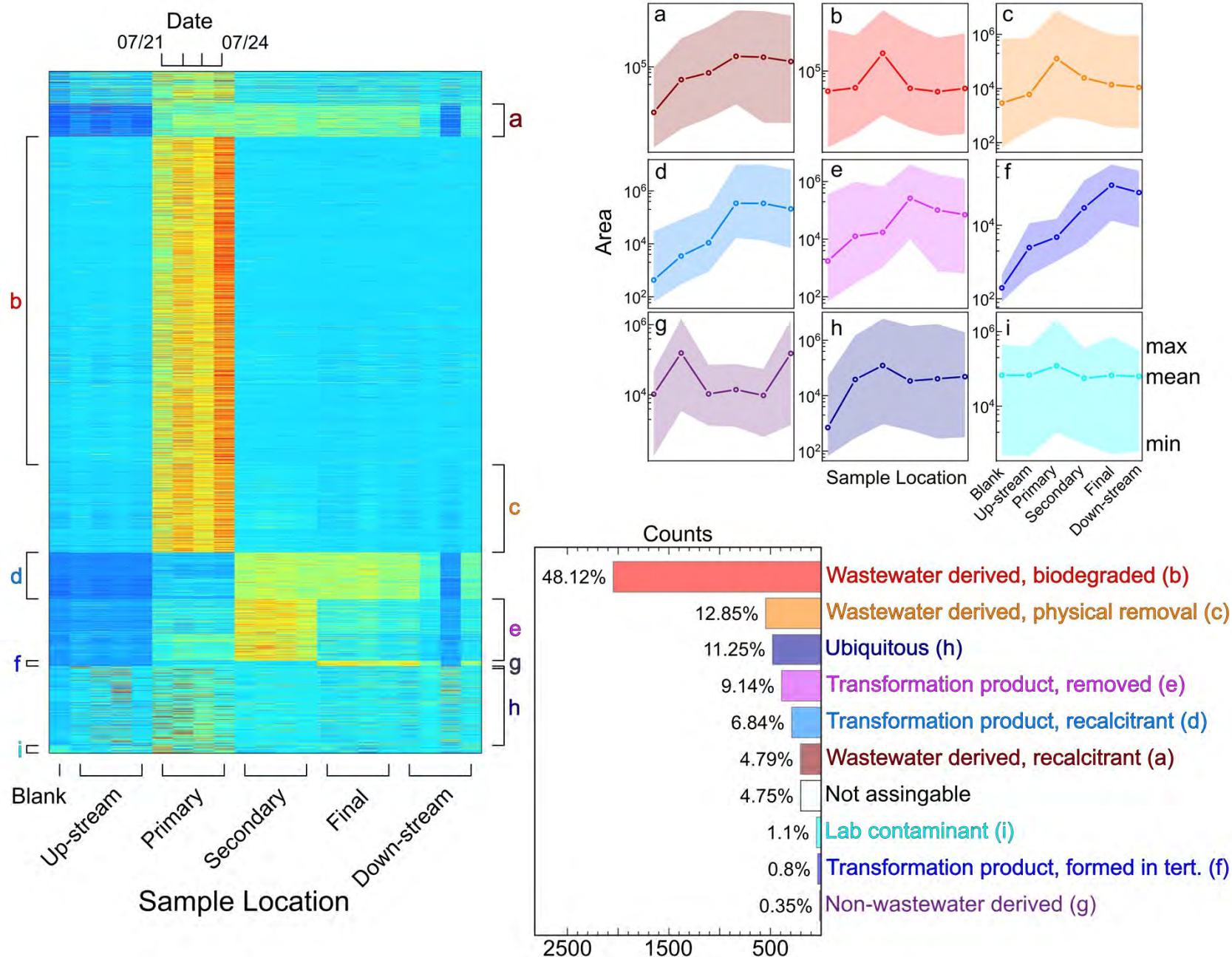


Differential analysis of non-target compounds in wastewater to reveal emerging contaminants

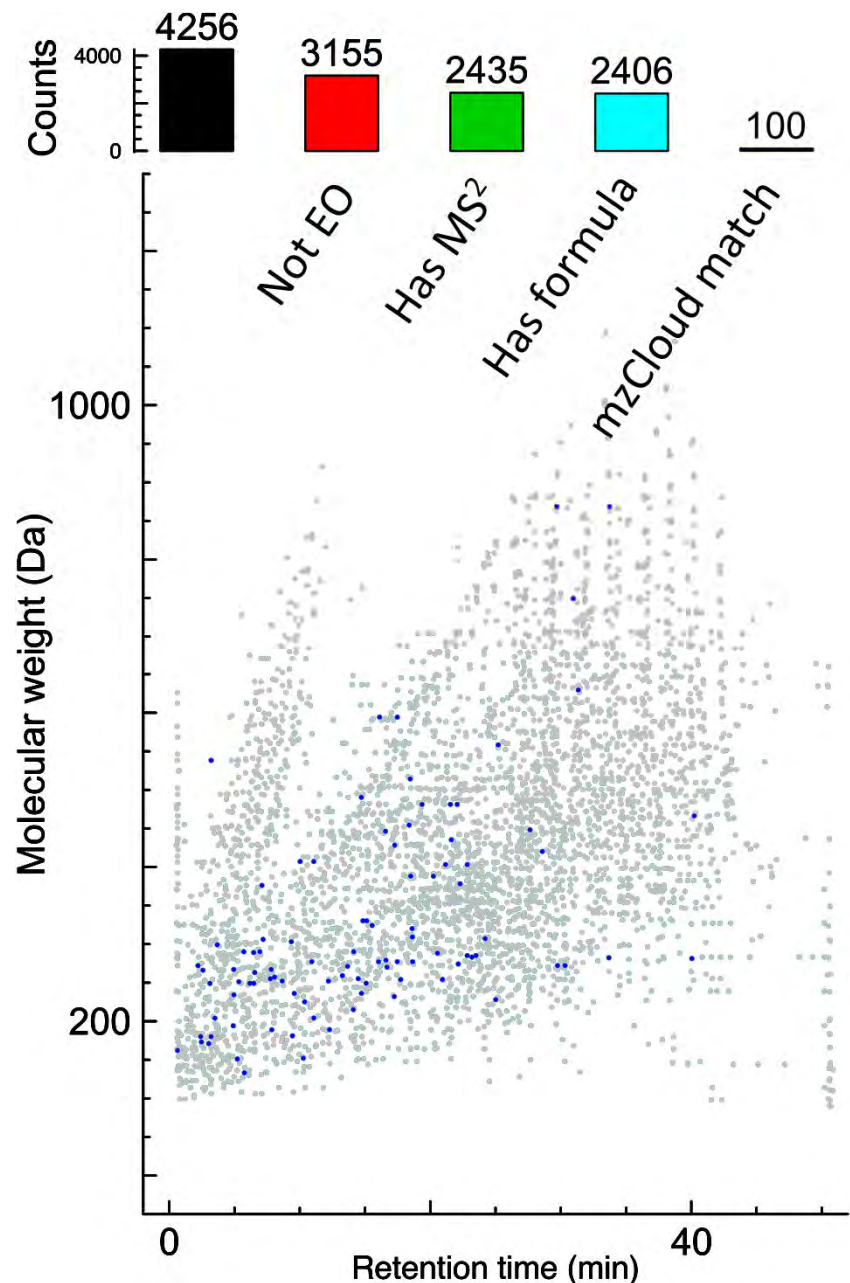
- Screening results were used to generate target lists for differential screening
- ID in screening = mass error $< \pm 2.5$ ppm and $\Delta RT < 0.5$ min

Category	Blank	Upstream	Primary eff.	Secondary eff.	Final eff.	Downstream
Wastewater-derived, biodegradable	○	○	↑	↓	→	→
Wastewater-derived, physical removal	○	○	↑	↘	↘	→
Wastewater-derived, recalcitrant	○	○	↑	→	→	→
Transformation product, removed	○	○	○	↑	↓	→
Transformation product, recalcitrant	○	○	○	↑	→	→
Transformation product, produced in tertiary treatment	○	○	○	↗	↑	→
Non wastewater-derived	○	↑	○	○	○	↑
Ubiquitous	○	→	→	→	→	→
Laboratory contaminant	→	→	→	→	→	→

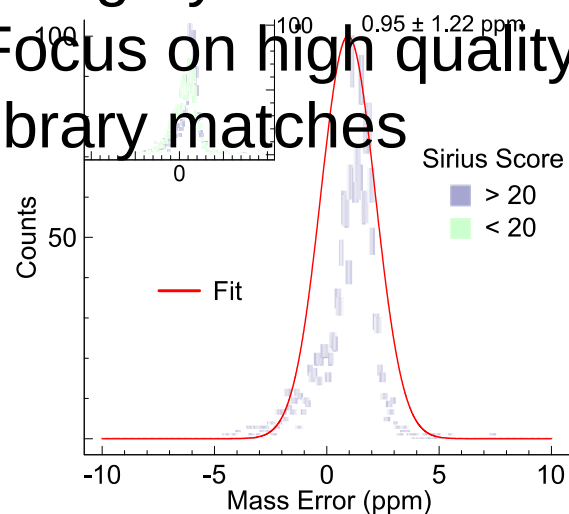
Fate-dependent feature prioritization



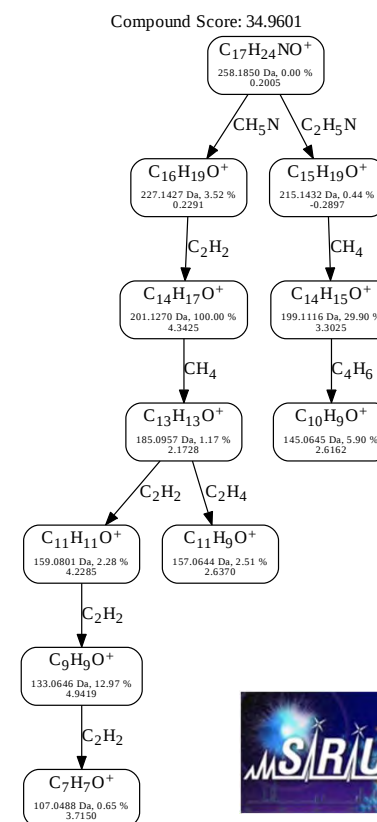
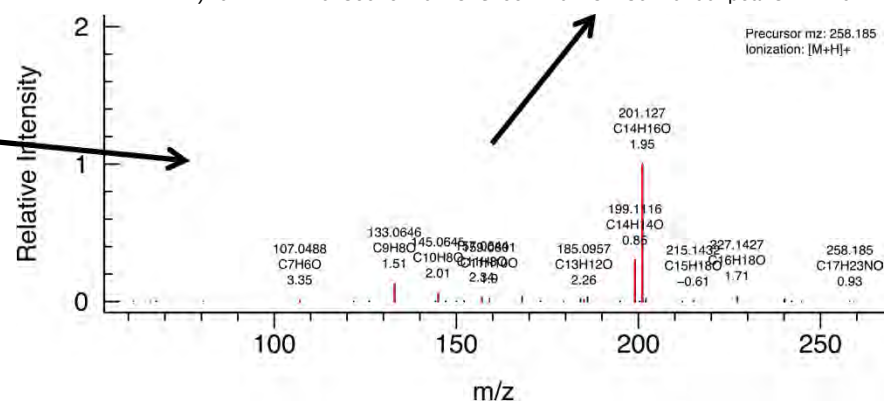
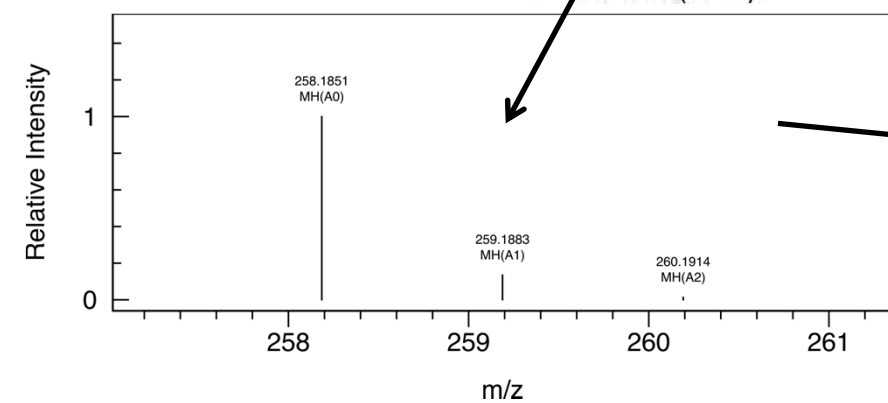
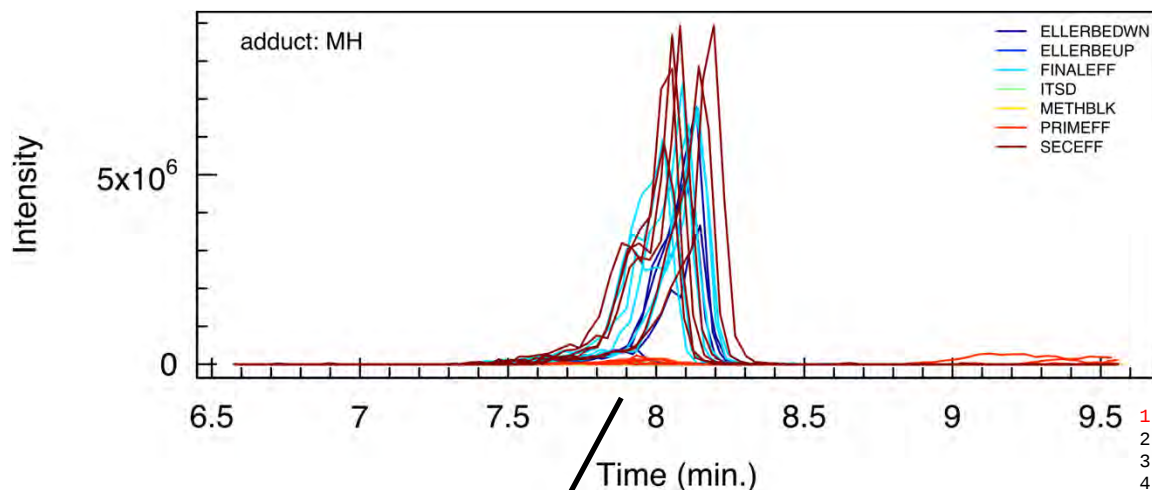
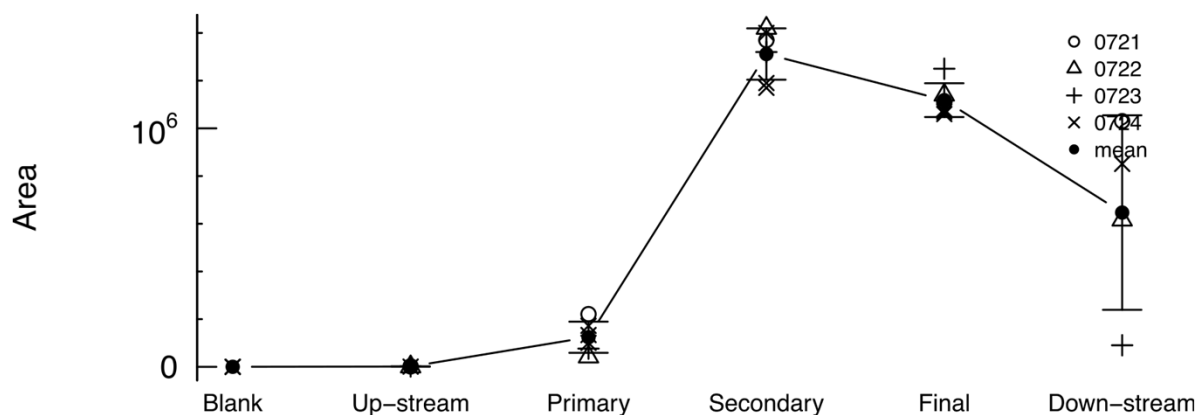
Data filtering



- Remove features corresponding to ethoxylated oligomers
- Filter based on MS² spectra
- Retain features with molecular formula assignment
- Prioritize based on differential category
- Focus on high quality mzCloud library matches



257.1777 Da feature follows a recalcitrant transformation product profile

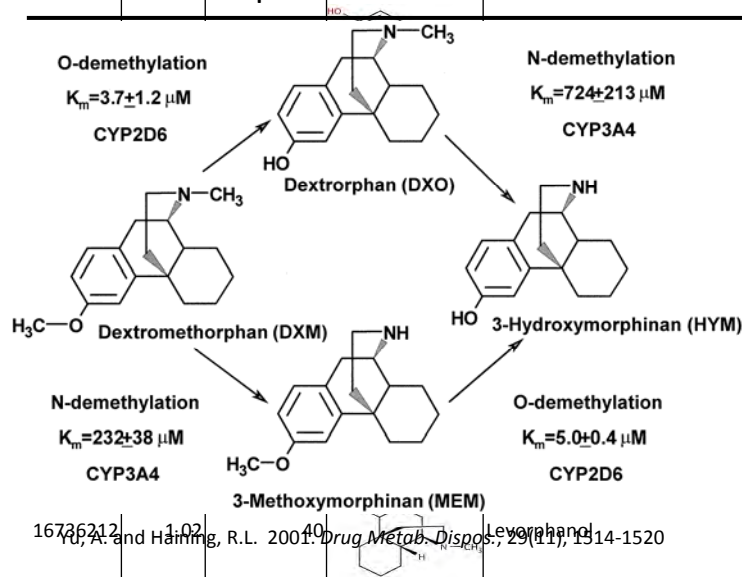


- 1.) C₁₇H₂₃NO score: 34.96 tree: +28.10 iso: 6.86 peaks: 11 91.33 %
- 2.) C₁₃H₂₆N₂O score: 23.35 tree: +15.03 iso: 8.33 peaks: 13 93.83 %
- 3.) C₁₂H₂₂FN₄O score: 21.51 tree: +12.90 iso: 8.61 peaks: 11 88.13 %
- 4.) C₇H₂₁F₂N₇O score: 6.13 tree: +6.13 iso: 0.00 peaks: 11 91.26 %

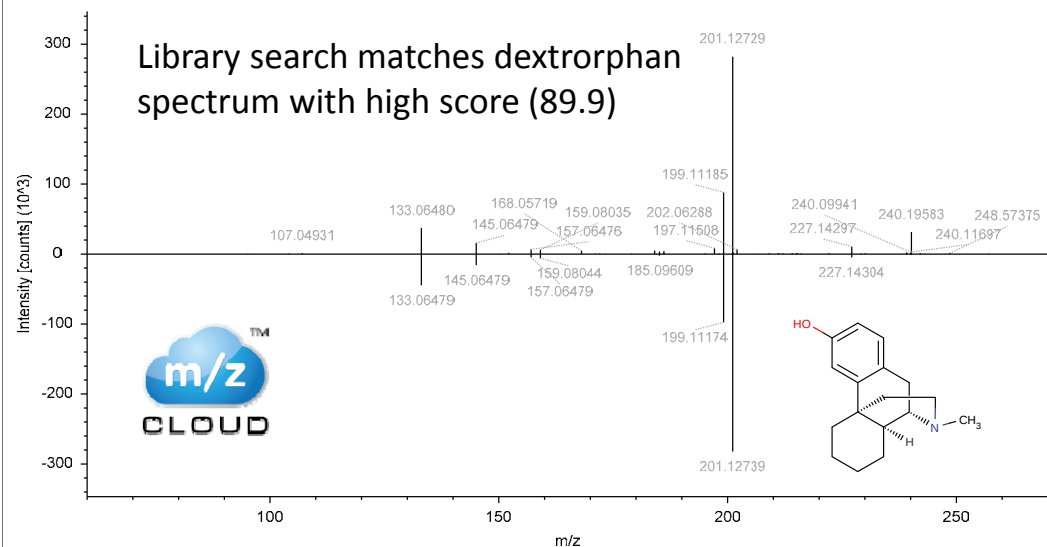
Dextrorphan tentatively identified by library match and *in silico* fragments

Chemspider matches for C₁₇H₂₃NO

CSID	MW	# References	Structure	Name
1723	270.34	1		dextromethorphan

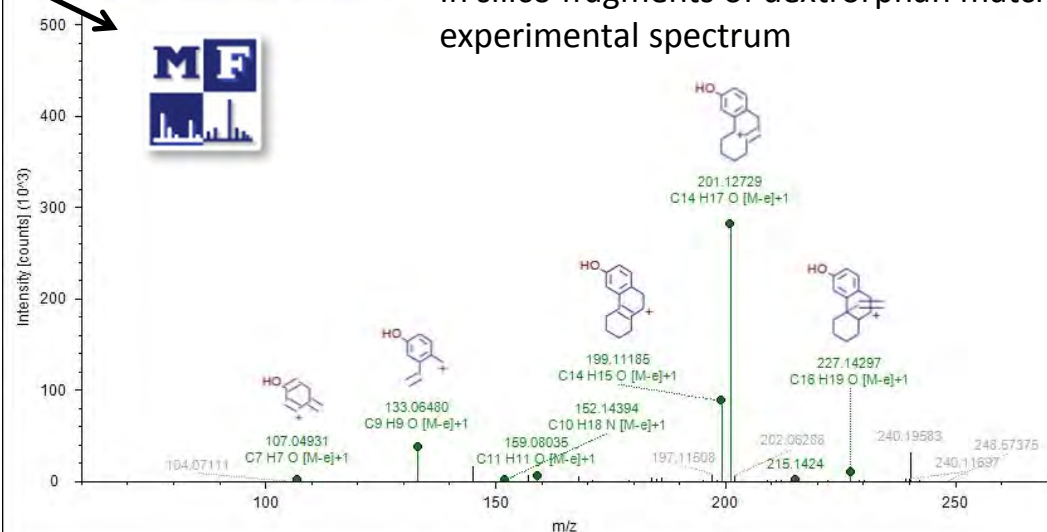


RAWFILE: PRIMEFF_0721_GRAB-1, #956, RT=7.877 min, FTMS (+), MS2 (CID, DDF, m/z=258.18, z=+1)
REFERENCE: MzCloud library C17 H23 N O Dextrophan

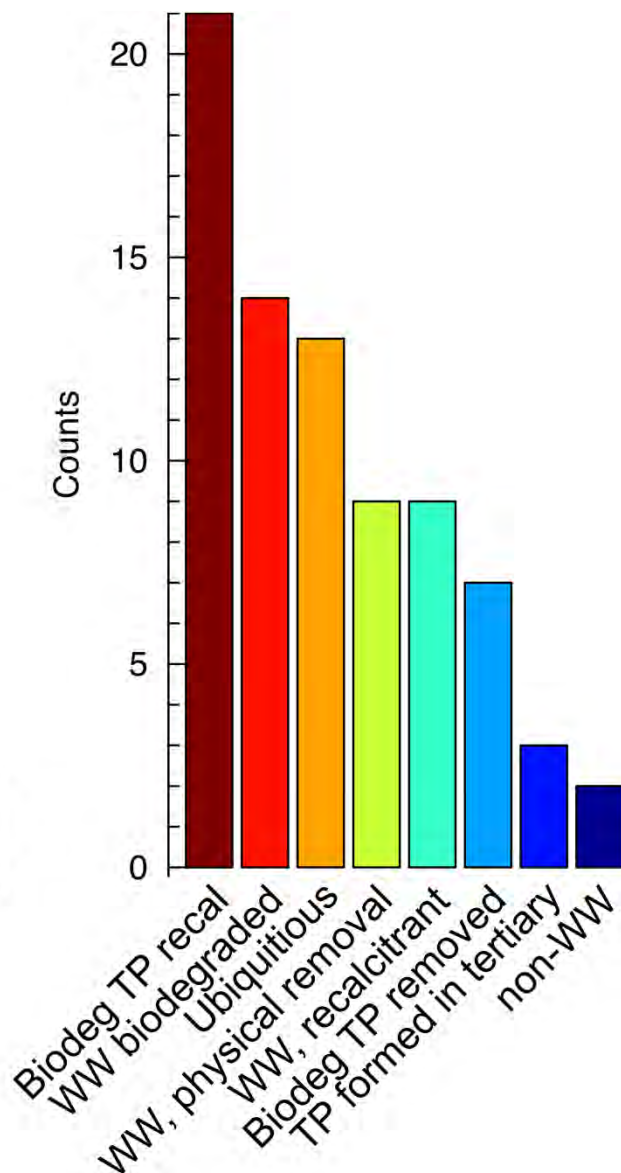
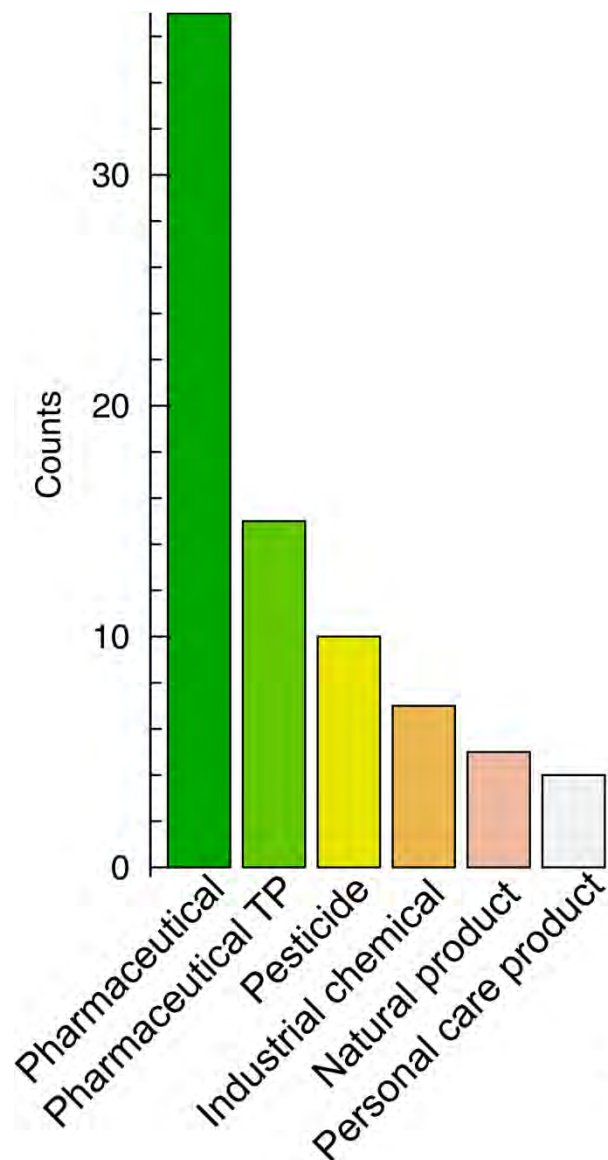


PRIMEFF_0721_GRAB-1, #956, RT=7.877 min, FTMS (+), MS2 (CID, DDF, m/z=258.18, z=+1)
Dextrophan C17 H23 N O, MW: 257.17796, Area: 44246
FISH Coverage: 8 Direct, 32 Unmatched, 23 Skipped

In silico fragments of dextrorphan match experimental spectrum



Results: Suspect compounds tentatively identified



- 78 compounds from six classes were tentatively identified in wastewater/surface water
- 20 of these were confirmed with standards (100% correct assignment)
- A further 1,101 features were annotated as polyethoxylated surfactants (comprising 39% of features identified as wastewater-derived, biodegradable)
- The largest fraction of ID'd compounds was classified as recalcitrant transformation products.

Finding the “causative stressors” within a water reuse scenario using non-targeted analysis coupled with bioanalytical tools



Bioanalytical pollutant monitoring on Kiawah Island, SC

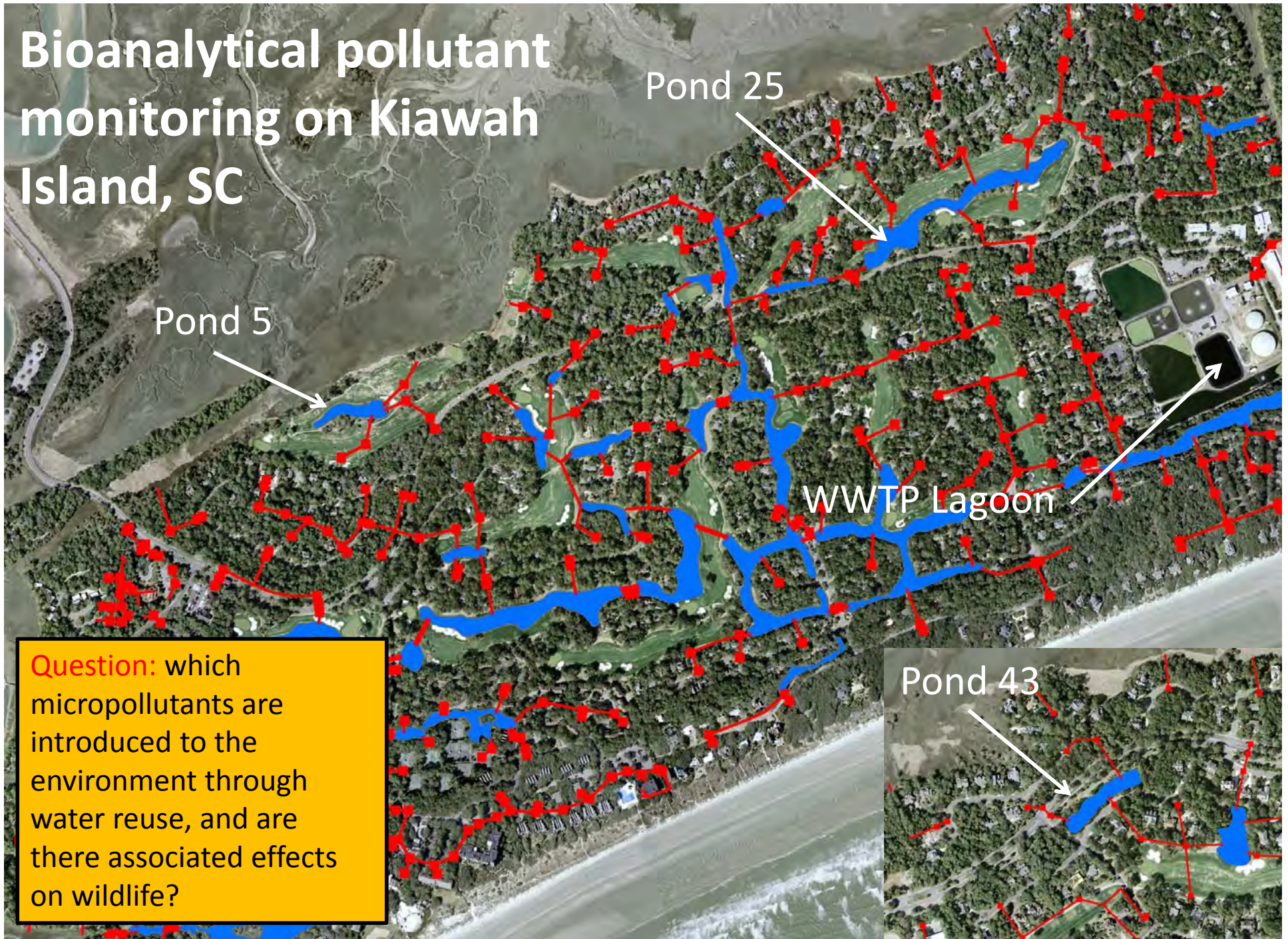
Pond 25

Pond 5

WWTP Lagoon

Question: which micropollutants are introduced to the environment through water reuse, and are there associated effects on wildlife?

Pond 43



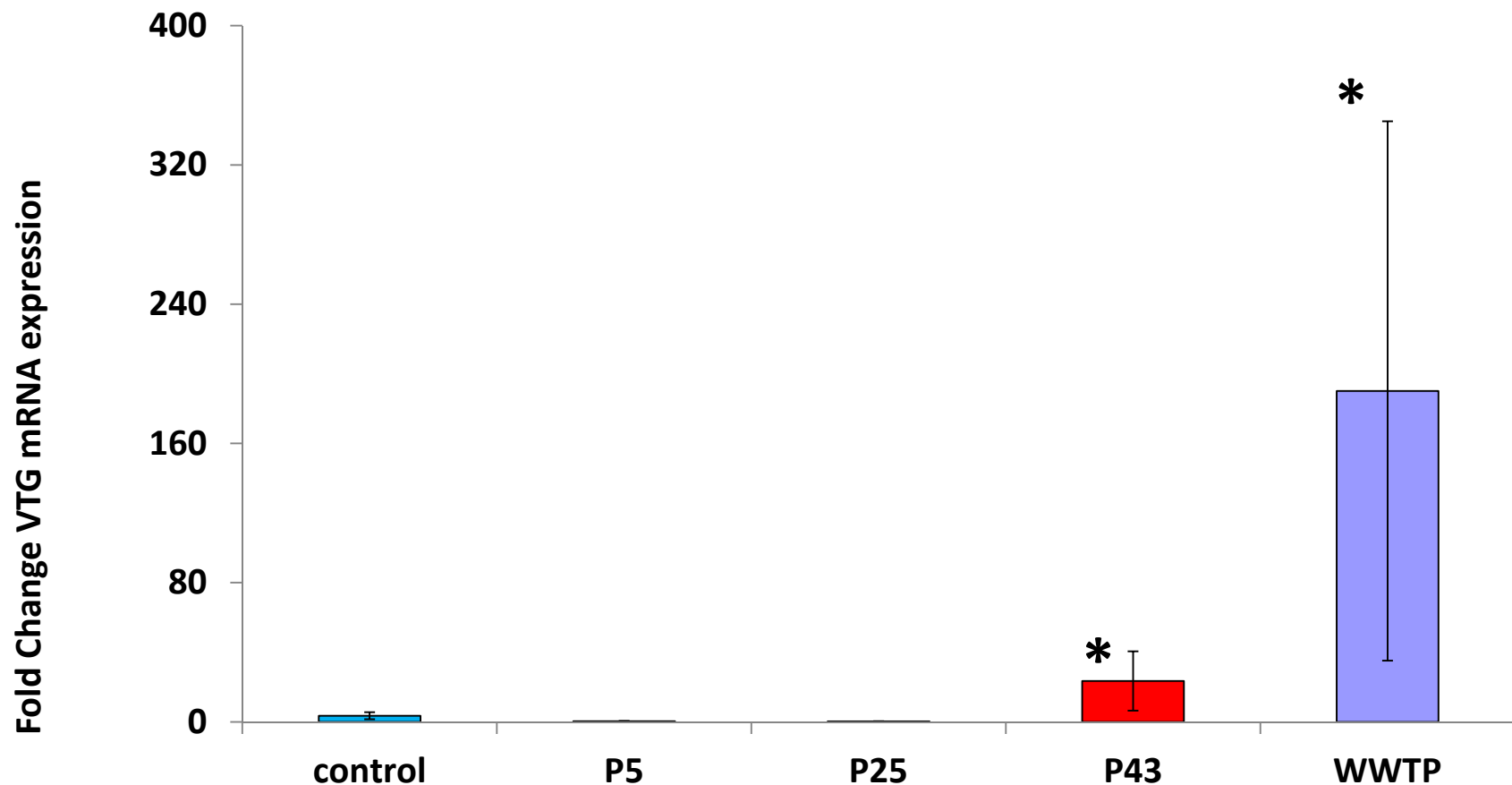
Male Fathead Minnow Exposures



Mini Mobile Units for
“safer” in-situ fish
exposure (Alan
Kolok)

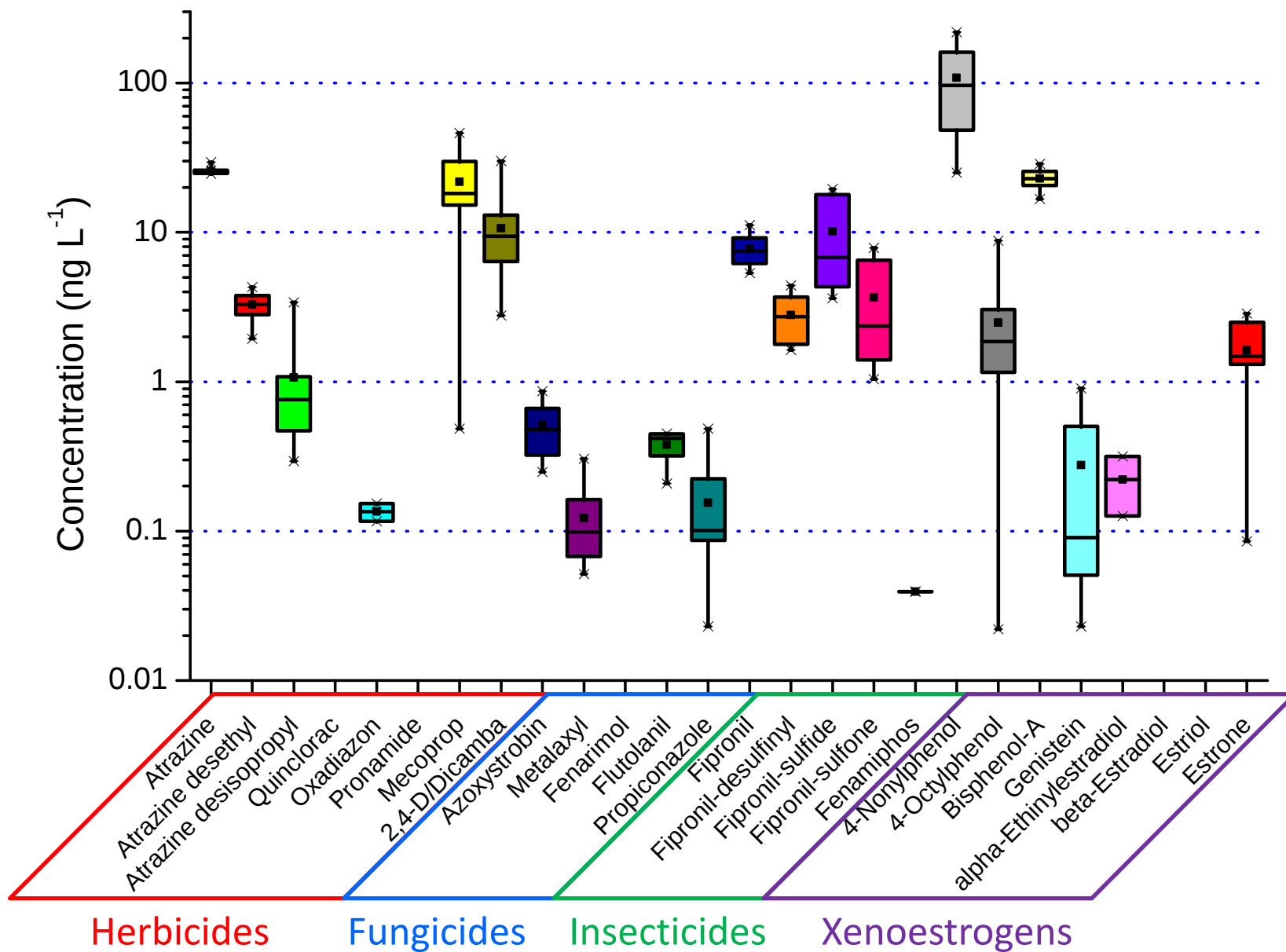


Hepatic Vitellogenin mRNA Expression in Male Fathead Minnows After 1 Week Exposure



* Indicate statistically significant values compared to control using one-way ANOVA ($p < 0.05$)

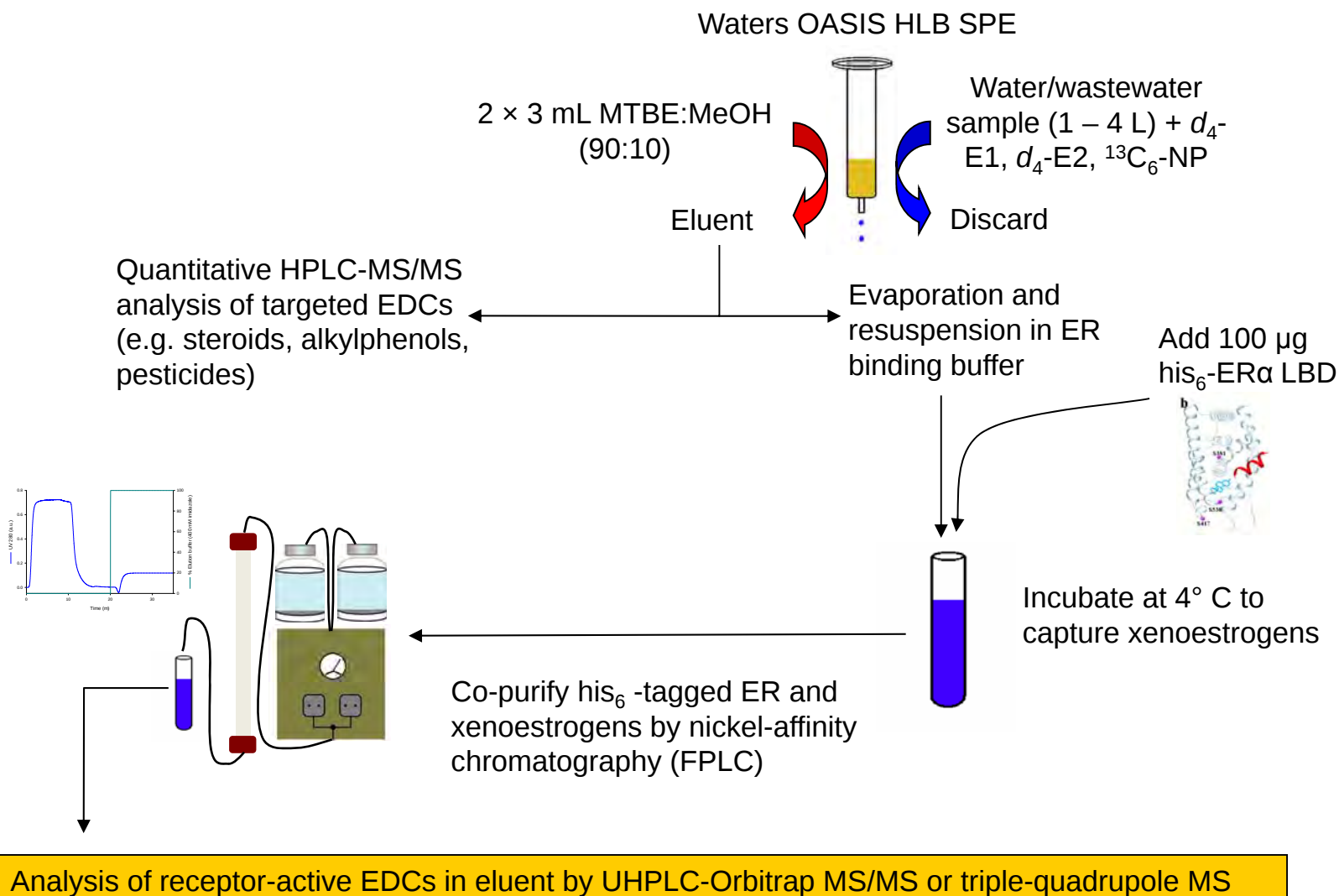
Target Micropollutants in WWTP Lagoon



Yes, but what ELSE might be contributing to estrogenicity in the reclaimed water??

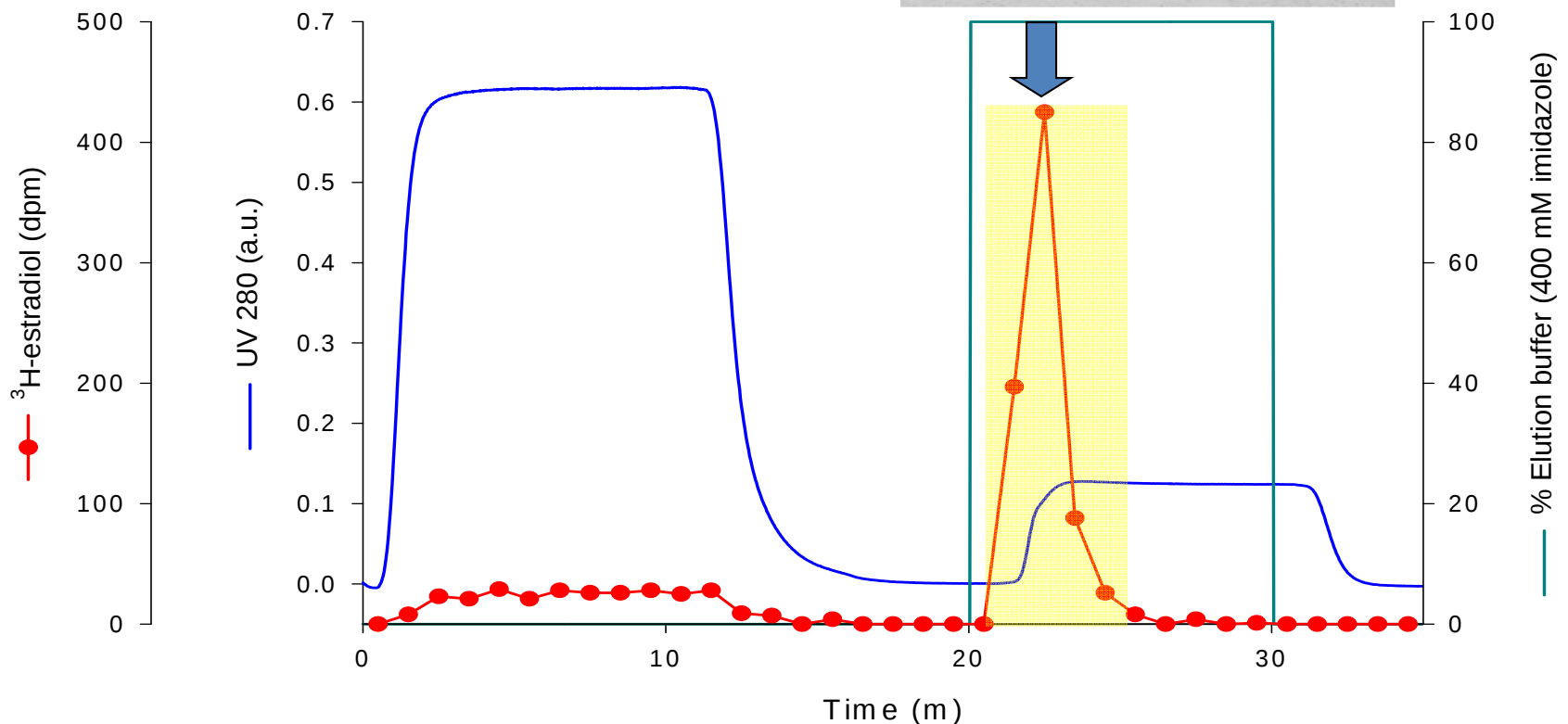
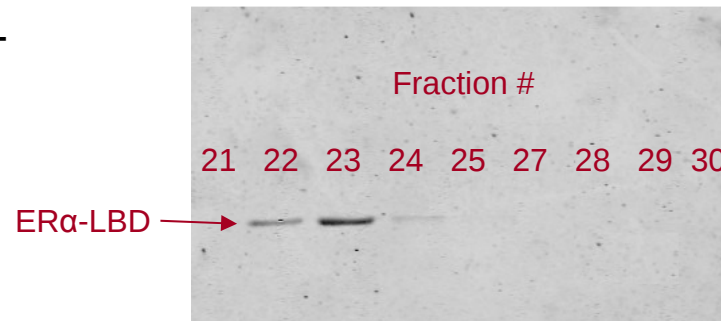
- **Receptor affinity extraction:** Similar in concept to immunoaffinity chromatography – relies on high specificity/selectivity molecular interaction to isolate target analytes from a mixture prior to analysis
- Recombinant protein engineering:
 - ER α ligand binding domain triple Cys $\rightarrow$ Ser mutant
 - Fusion of thioredoxin to ER enhances solubility
 - His₆ tag allows subsequent purification
- Proteins are cloned, expressed in bacterial vectors, and purified chromatographically

Estrogen receptor-affinity isolation for activity-directed analysis

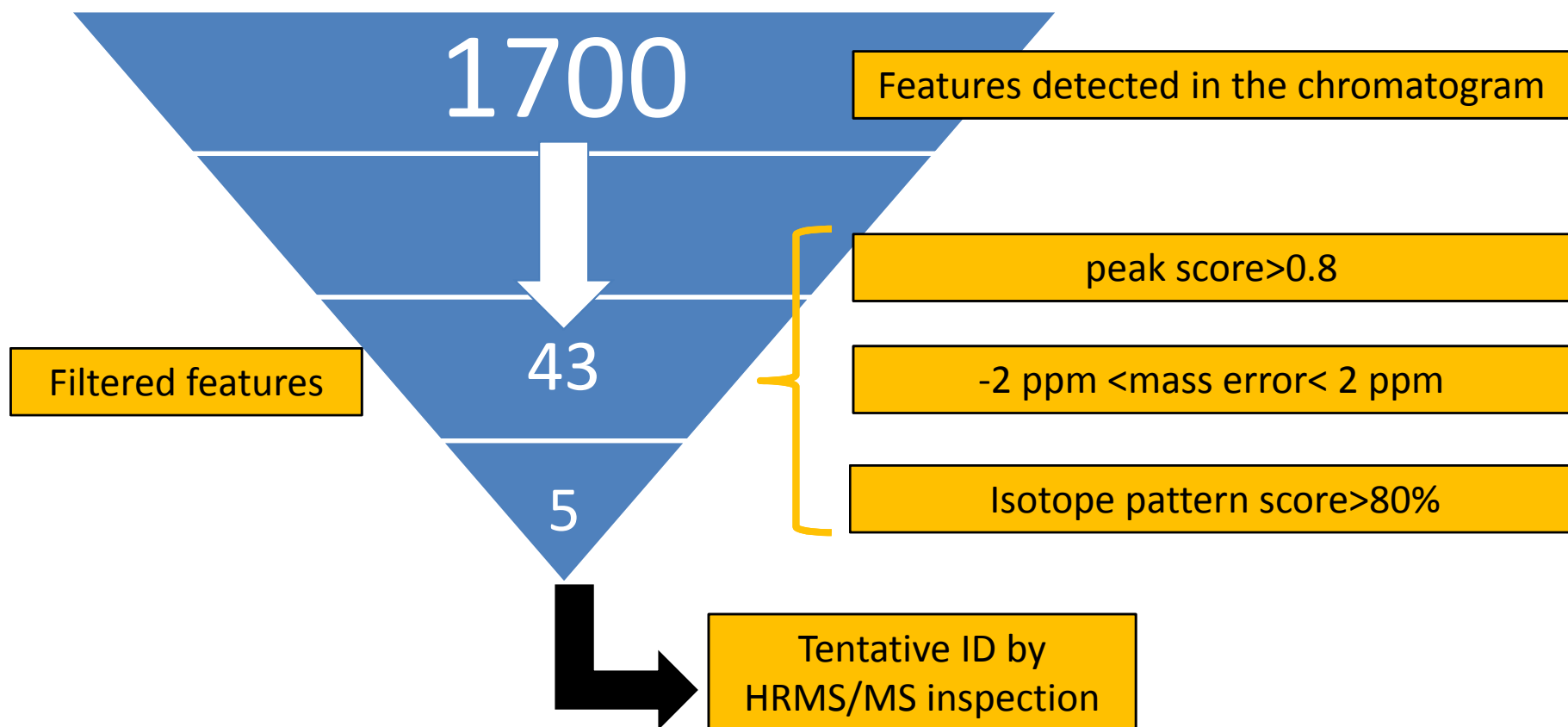


Purification of ER-bound xenoestrogens from wastewater

ER α -LBD and radioligand tracer (^3H -17 β -estradiol) are selectively retained from wastewater extract by Ni^{2+} agarose, and can be co-eluted with 400 mM imidazole



Data filtering and tentative compound identification



Xenoestrogen extracts

Name	Class
17β-estradiol	Endogenous steroid
Estrone	Endogenous steroid
Celecoxib	NSAID Pharmaceutical
PFOA	Fluorinated surfactant
Efavirenz	Antiviral pharmaceutical

Efavirenz directly modulates the oestrogen receptor and induces breast cancer cell growth

MJ Sikora,¹ JM Rae,^{1,2} MD Johnson³ and Z Desta⁴
¹Department of Pharmacology, ²Department of Internal Medicine, University of Michigan Medical Center, Ann Arbor, MI, USA, ³Lombardi Cancer Center and Department of Oncology, Georgetown University Medical Center, Washington, DC, USA and ⁴Division of Clinical Pharmacology, Department of Medicine, Indiana University, Indianapolis, IN, USA

Objectives
Efavirenz-based HIV therapy is associated with breast hypertrophy and gynaecomastia. Here, we tested the hypothesis that efavirenz induces gynaecomastia through direct binding and modulation of the oestrogen receptor (ER).

Methods
To determine the effect of efavirenz on growth, the oestrogen-dependent, ER-positive breast cancer cell lines MCF-7, T47D and ZR-75-1 were treated with efavirenz under oestrogen-free conditions in the presence or absence of the anti-oestrogen ICI 182,780. Cells treated with 17β-oestradiol in the absence or presence of ICI 182,780 served as positive and negative controls, respectively. Cellular growth was assayed using the crystal violet staining method and an *in vitro* receptor binding assay was used to measure the ER binding affinity of efavirenz.

Results
Efavirenz induced growth in MCF-7 cells with an estimated effective concentration for half-maximal growth (EC₅₀) of 15.7 μM. This growth was reversed by ICI 182,780. Further, efavirenz binds directly to the ER [inhibitory concentration for half maximal binding (IC₅₀) of ~ 52 μM] at a roughly 1000-fold higher concentration than observed with 17β-oestradiol.

Conclusions
Our data suggest that efavirenz-induced gynaecomastia may be caused, at least in part, by drug-induced ER activation in breast tissues.

Keywords: efavirenz, gynaecomastia, highly active antiretroviral therapy, oestrogen receptor, oestrogens

Accepted 12 January 2010

Introduction
The introduction of highly active antiretroviral therapy (HAART) multi-drug combination regimens has considerably improved the prognosis of patients infected with HIV by reducing AIDS-related morbidity and mortality [1]. However, chronic treatment with these regimens is associated with multiple adverse effects, nonadherence and eventually therapy failure [2]. Treatment regimens containing the nonnucleoside reverse transcriptase inhibitor efavirenz are preferred in treatment-naïve patients and are widely used in other settings [3]. While efavirenz is generally well tolerated, concentration-dependent side effects that impact drug adherence and promote resistance have been documented [4]. Common adverse effects of efavirenz include central nervous system symptoms, occurring in up to 50% of patients [5], but other less common adverse effects have also been reported. An increasing number of reports suggest that the use of HAART, in particular efavirenz-based therapy, is associated with breast hypertrophy or gynaecomastia [6–11]. While mechanisms underlying efavirenz-induced gynaecomastia are not well understood, a number of hypotheses exist, including a direct oestrogenic effect, induction of an immune response, or altered steroid hormone metabolism by cytochrome P450 enzymes. To our knowledge, none of

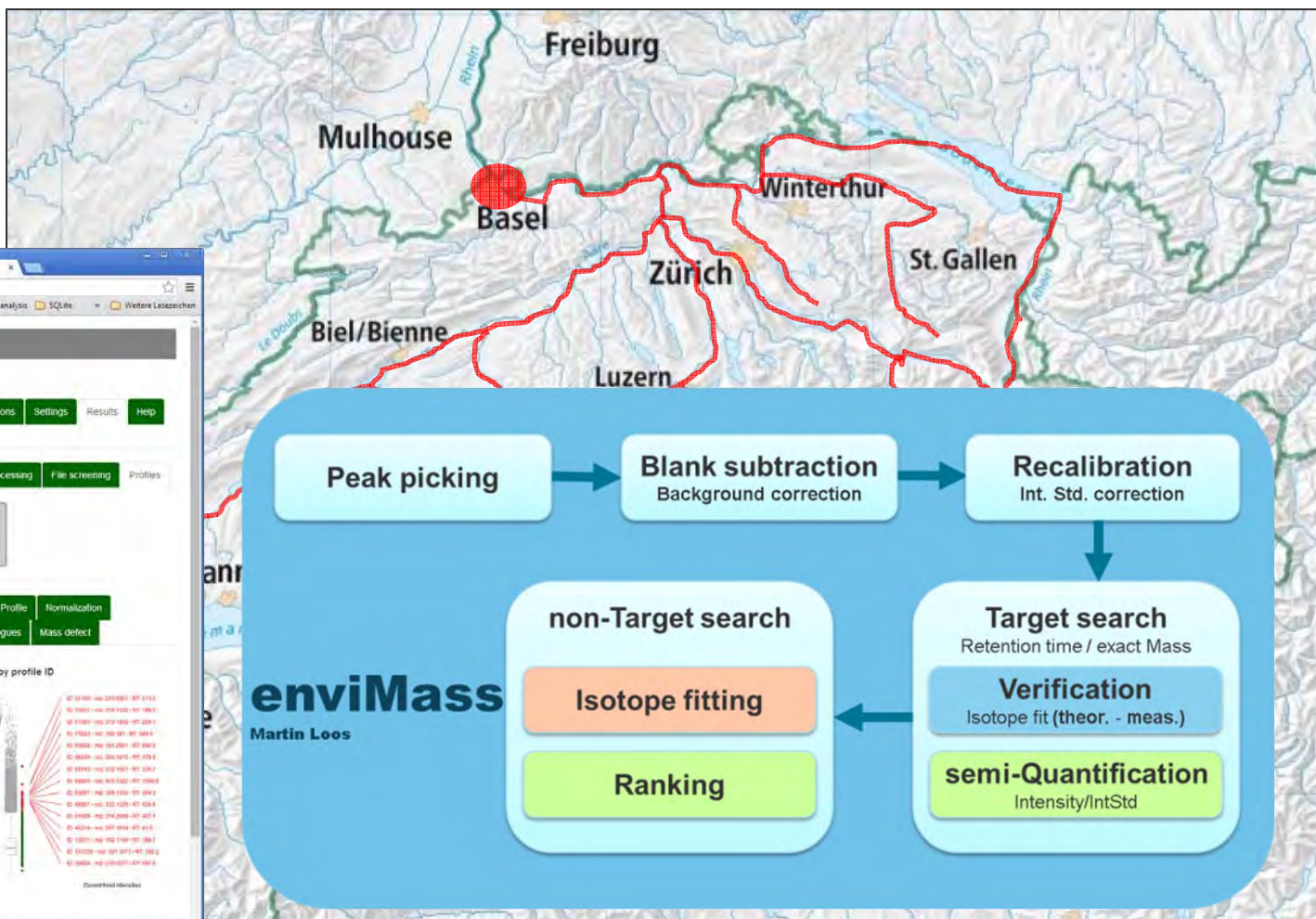
tor affinity screening

Pattern ore	Confirmed & Quantified?
00	0.70 ng/L
05	11.5 ng/L
00	no
00	no
00	no

Case study: Daily routine monitoring of CECs in the Rhine river at the Swiss-German border

(Matthias Ruff, Heinz Singer, and Juliane Hollender)

enviMass

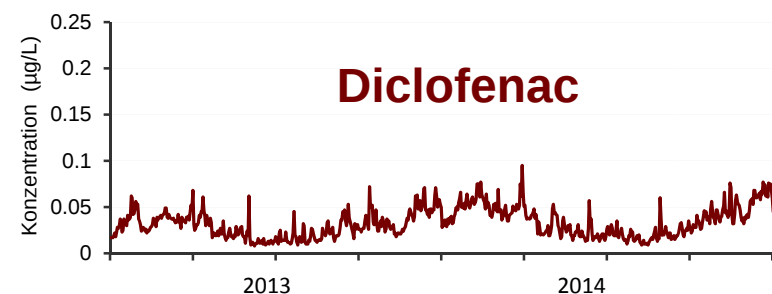


Target/non-target CEC monitoring in the Rhine

spill



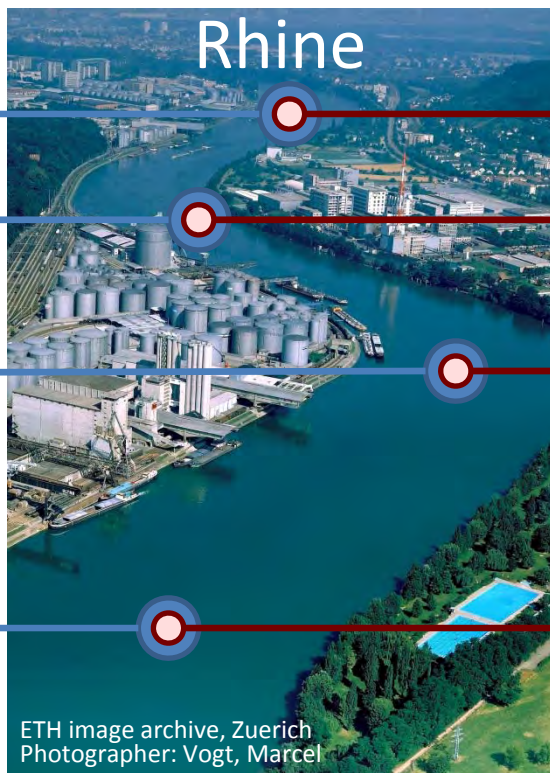
trend



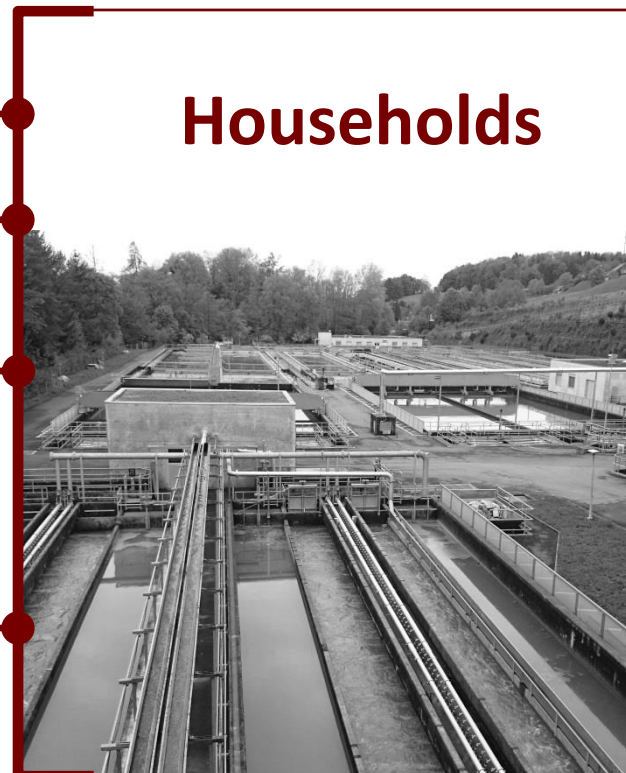
Industry

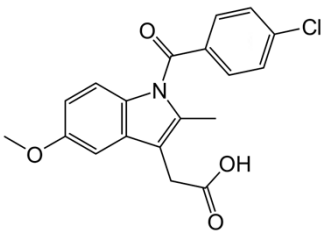


Rhine



Households





March 2014	1	2	3	4	5	6	7	8		25
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Ruff et al., Water Research, 2015 (87)

Results

- Comprehensive and real time trend and spill detection
- Implementation of non-target/suspect screening into routine monitoring

Results:

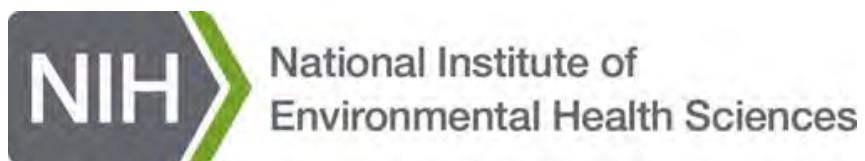
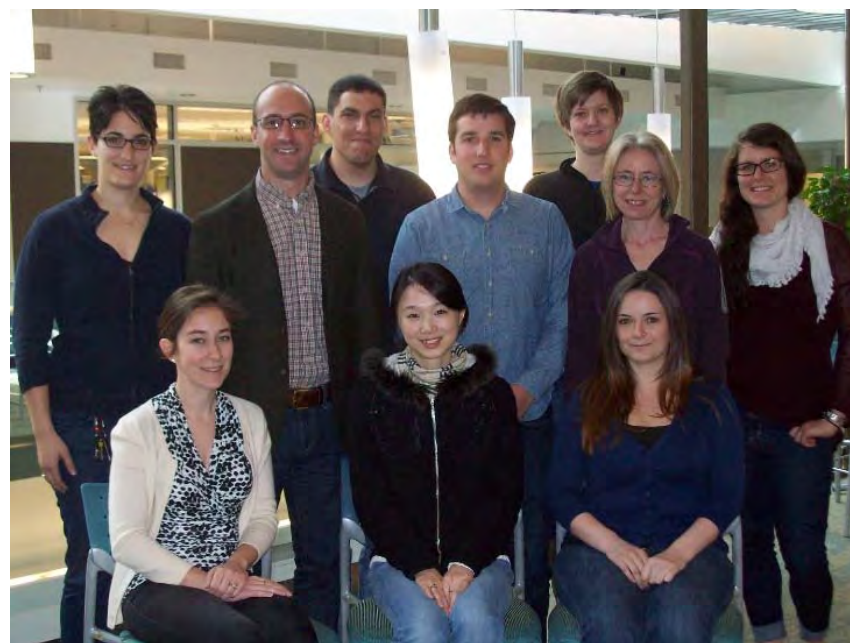


Conclusions and perspectives for future work

- LC-HRMS coupled with optimized non-targeted screening workflows provide essential tools for conducting “fate-directed analysis” and prioritization of CECs in water/wastewater.
- Coupling non-targeted analysis with mechanism-based bioanalytical tools provides a means of identifying “causative stressors”.
- Non-targeted analysis is complementary to targeted CEC monitoring in recycled water, and may best be applied to supply “leads” for CEC inclusion on monitoring lists.
- Significant challenges remain for robust compound identification tools (chemoinformatics), harmonization of methods, and “routine” application of non-targeted analysis.

Acknowledgement

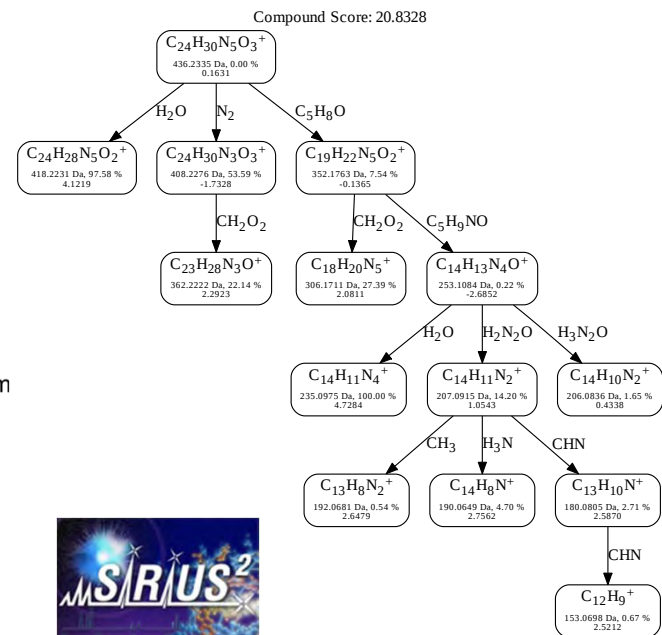
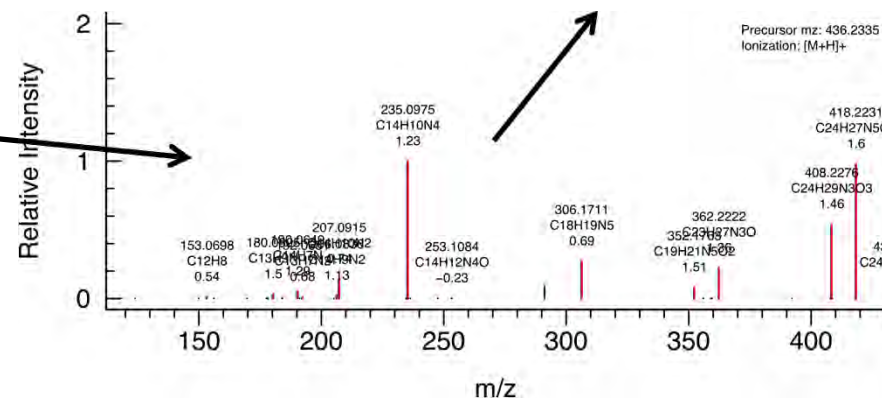
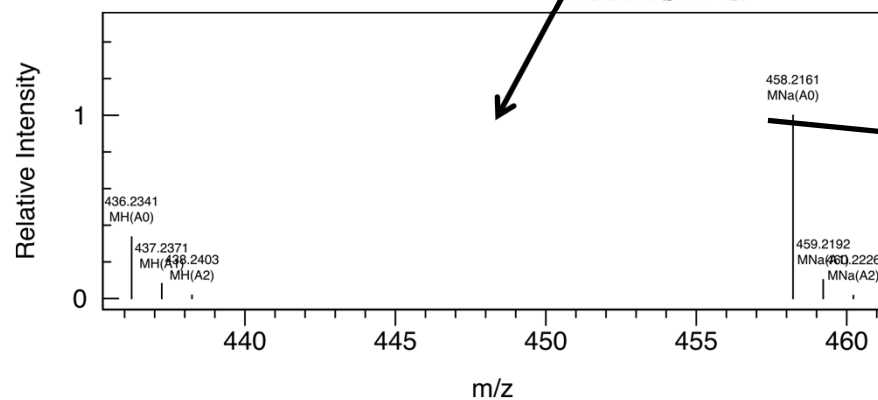
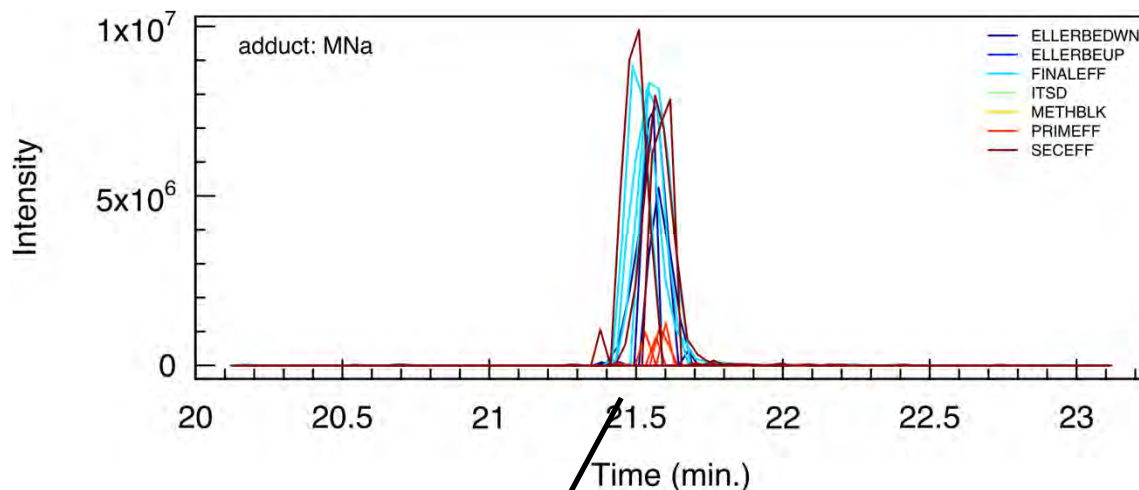
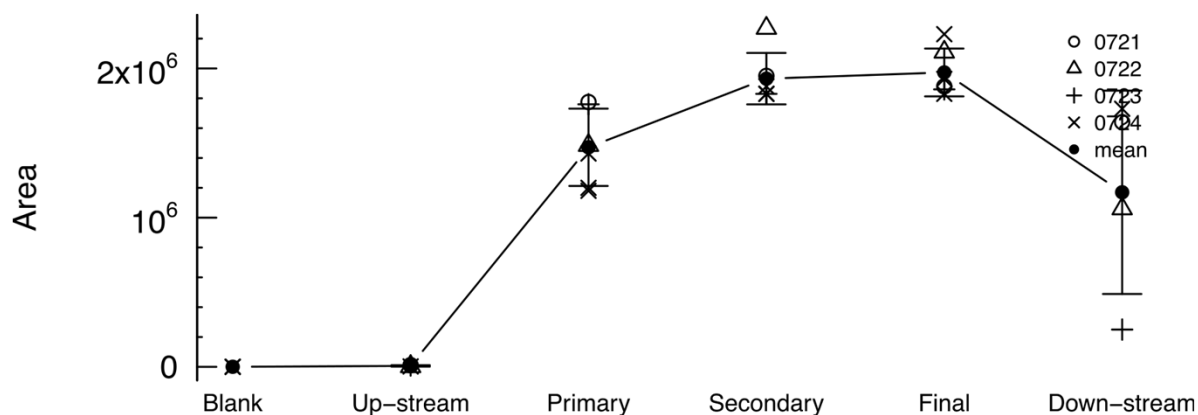
- **Ferguson Lab Group**
- **Heinz Singer, Juliane Hollender (EAWAG)**



ThermoFisher
SCIENTIFIC

Richard Jack and Dipankar Ghosh

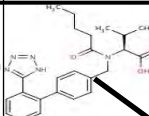
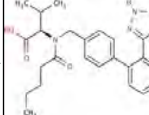
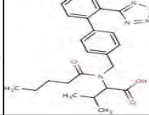
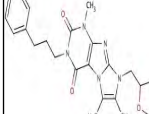
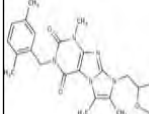
A feature at 435.2244 Da is wastewater-derived and recalcitrant



- 1.) C₂₄H₂₉N₅O₃ score: 20.83 tree: +20.83 iso: 0.00 peaks: 14 95.67 %
- 2.) C₂₂H₂₇N₈O₂ score: 15.72 tree: +15.72 iso: 0.00 peaks: 15 95.87 %
- 3.) C₁₉H₂₈F₈O₃ score: 13.65 tree: +13.65 iso: 0.00 peaks: 14 95.67 %
- 4.) C₁₆H₃₃N₇O₅ score: 13.08 tree: +13.08 iso: 0.00 peaks: 12 85.63 %
- 5.) C₁₉H₂₉F₄N₅O₂ score: 12.88 tree: +12.88 iso: 0.00 peaks: 13 74.11 %
- 6.) C₁₈H₃₀N₉O₂P score: 12.26 tree: +12.26 iso: 0.00 peaks: 11 64.07 %
- 7.) C₂₁H₃₁F₄N₂O₃ score: 12.00 tree: +12.00 iso: 0.00 peaks: 14 95.40 %
- 8.) C₁₉H₃₀C₁N₉O score: 11.25 tree: +11.25 iso: 0.00 peaks: 10 63.87 %
- 9.) C₁₅H₃₃F₂N₄O₈ score: 10.95 tree: +10.95 iso: 0.00 peaks: 14 95.67 %
- 10.) C₁₅H₃₁F₉N₃O₃P score: 9.57 tree: +9.57 iso: 0.00 peaks: 13 80.27 %

Valsartan tentatively identified by library match and *in silico* fragments

Chemspider matches for $C_{24}H_{29}N_5O_3$

CSID	Δ Mass [ppm]	# References	Structure	Name
54833	1.4	1050		Valsartan
4447678	1.4	34		N-pentanoyl-N-([2'-(1H-tetrazol-5-yl)biphenyl-4-yl]methyl)-D-valine
5448	1.4	26		N-Pentanoyl-N-([2'-(1H-tetrazol-5-yl)-4-biphenyl]methyl)valine
2414607	1.4	15		1,6,7-Trimethyl-3-(3-phenylpropyl)-8-(tetrahydro-2-furanylmethyl)-1H-imidazo[2,1-f]purine-2,4(3H,8H)-dione
2414601	1.4	13		3-(2,5-Dimethylbenzyl)-1,6,7-trimethyl-8-(tetrahydro-2-furanylmethyl)-1H-imidazo[2,1-f]purine-2,4(3H,8H)-dione

Valsartan is a highly prescribed angiotensin II receptor antagonist and has previously been shown to be recalcitrant to biodegradation in wastewater treatment.

Bergheim, M. et al. 2014. *Environ. Chem.*, 11, 431-444

