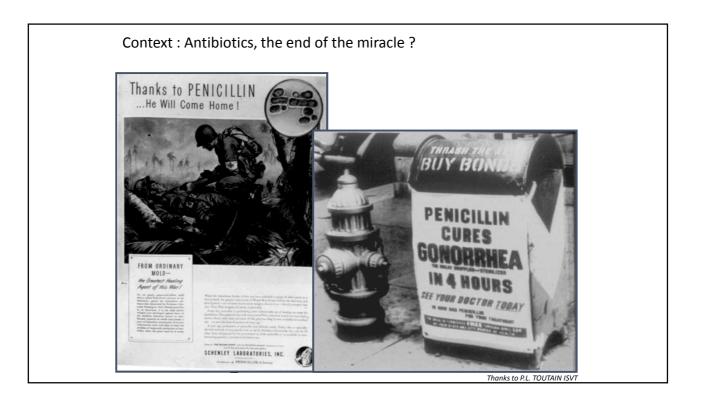
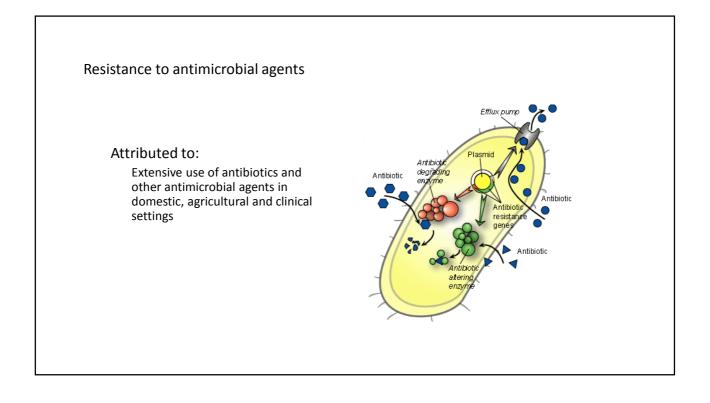


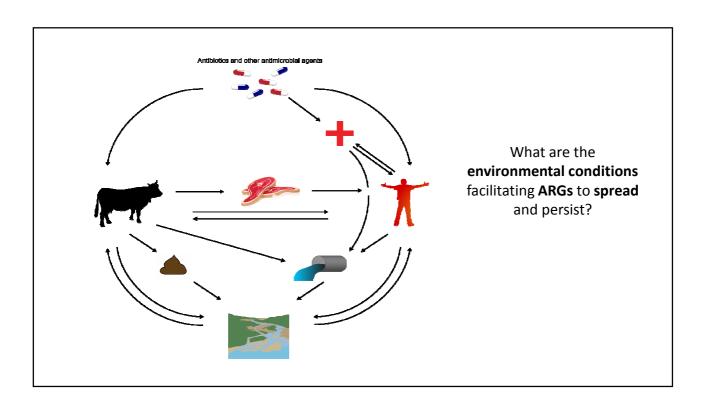
Robust discrimination of resistome and microbiome signatures in hospital and urban effluents

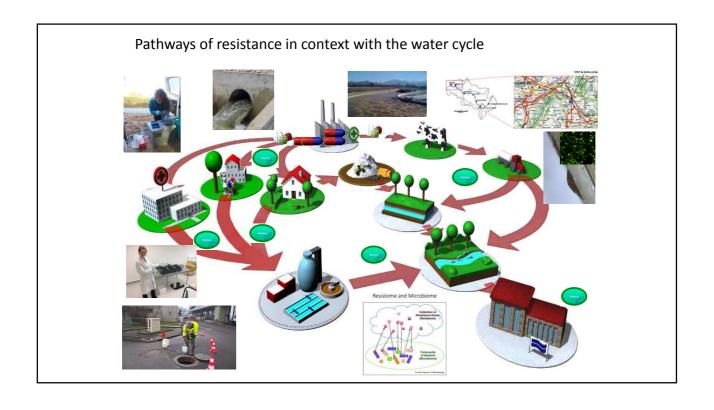
Elena Buelow

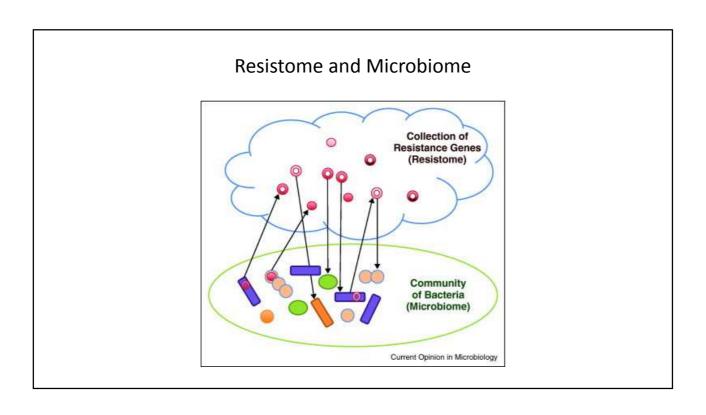


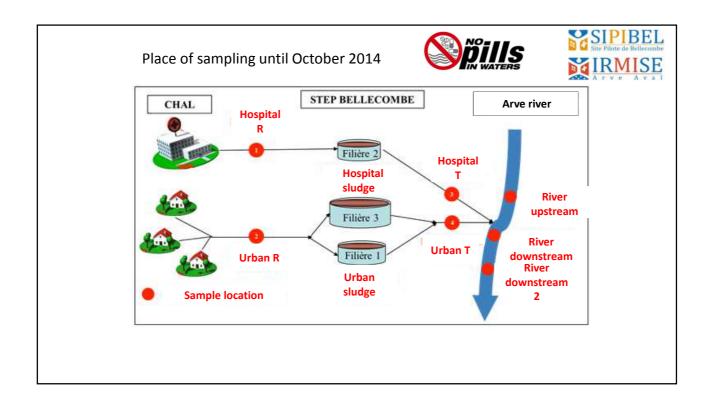
Resistance to antimicrobial agents – a major public health threat AMR in 2050 10 million Tetanus 60,000 Death attribuable to AMR every year Road traffic accidents 1.2 million Cancer compared to other 8.2 million AMR now major causes of death 700,000 (low estimate) Measles Cholera 130,000 100,000-120,000 J. O'Neil Review on Antimicrobial Resistance: Antimicrobial Resistance: Tackling a Crisis for the Health and Wealth of Nations. 2014 Diarrhoeal disease Diabetes 1.4 million 1.5 million

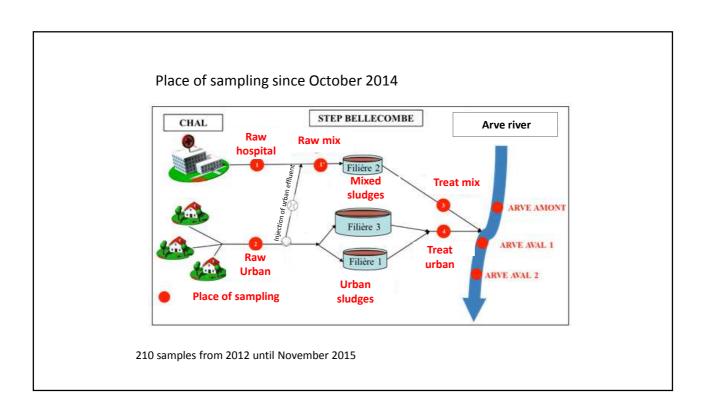


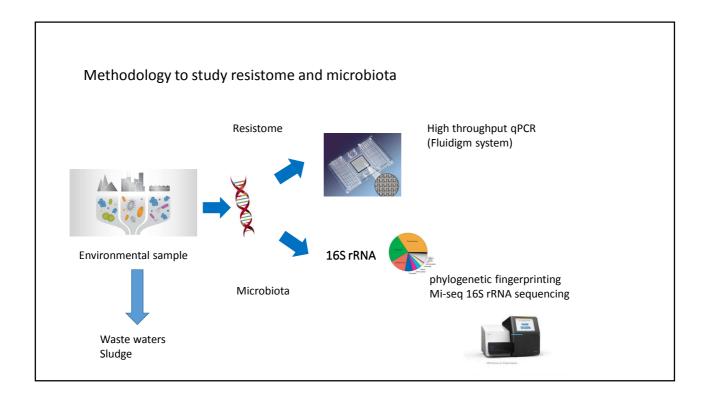


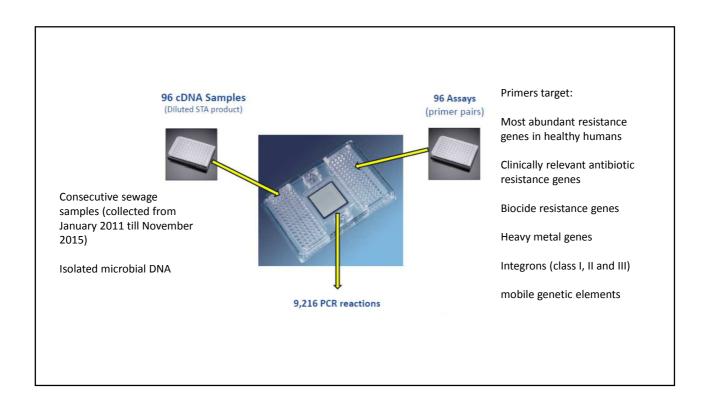


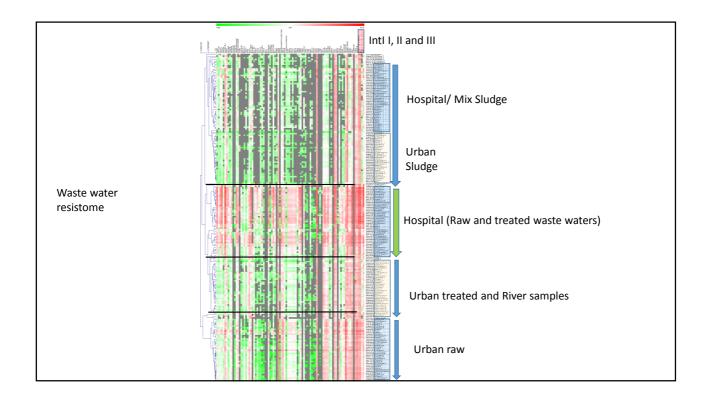








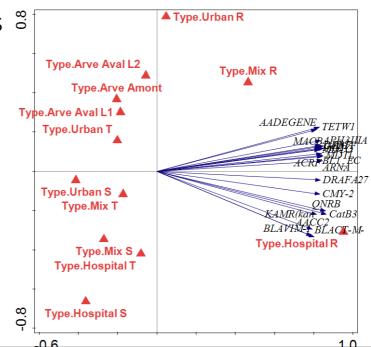




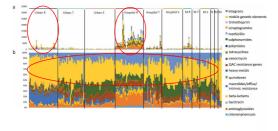
Multivariate analysis [∞]

Data visualization and statistical analysis

- RDA (constrained redundancy analysis)
 - Most significant explanatory variable is: TYPE of waste water
 - 69% of variation in resistiome explained by sample TYPE (p=0.002)



Relative and proportional abundance of ARGs and MGEs in hospital and urban wastewater over a period of 4 years

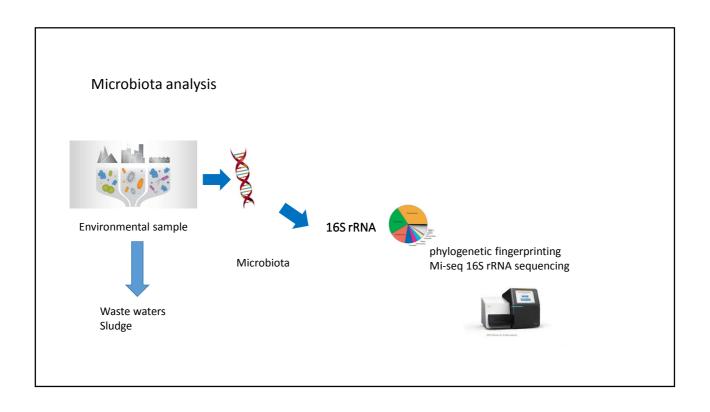


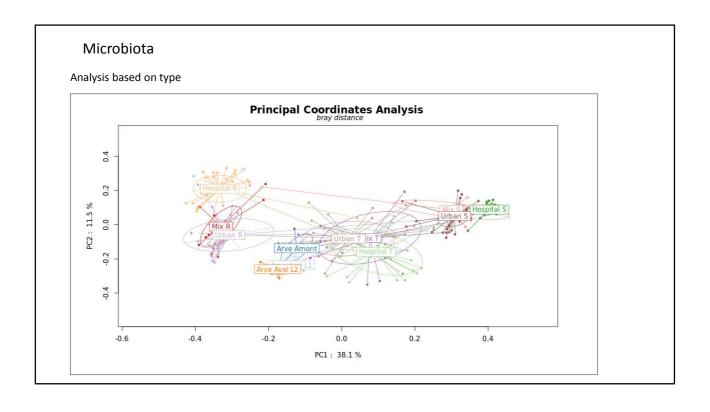
ARGs are significantly more abundant in hospital wastewater

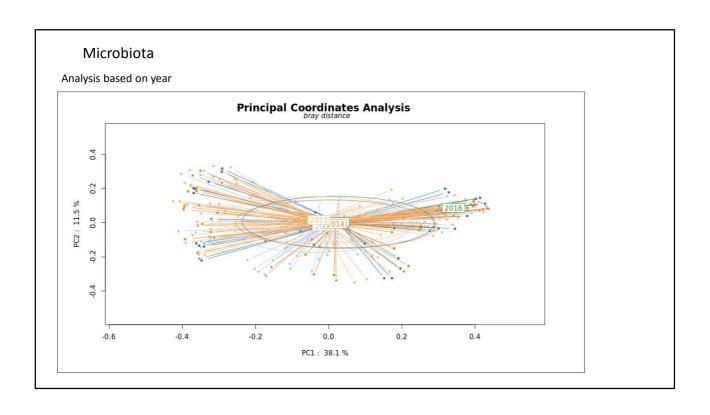
Relative and proportional abundance of ARGs does not change significantly over time (4 years)

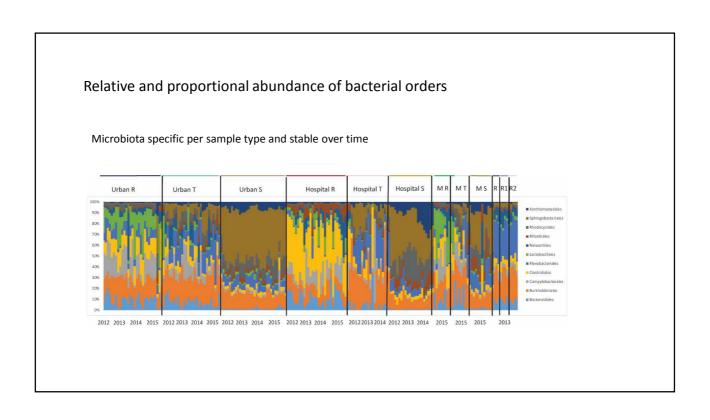
Distinct resistome signature

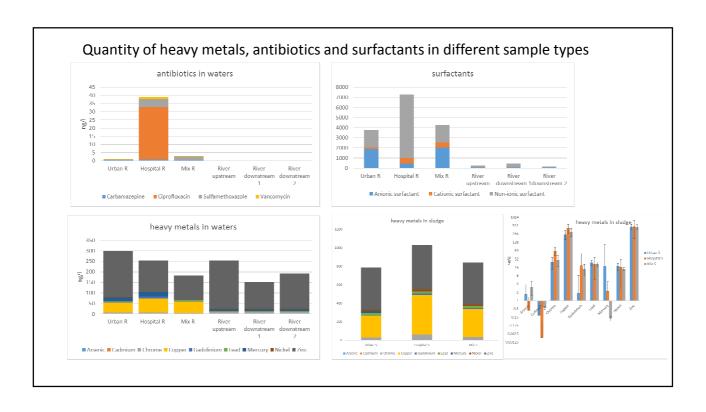
MGEs (transposases and integrons) are highly abundant in all waste and surface waters (represent up to 60% of detected genes in urban wastewaters)











Relationship between taxa, resistance genes and pharmaceuticals?

• Correlation analysis

chemical concentrations (p value is 0.02). Of

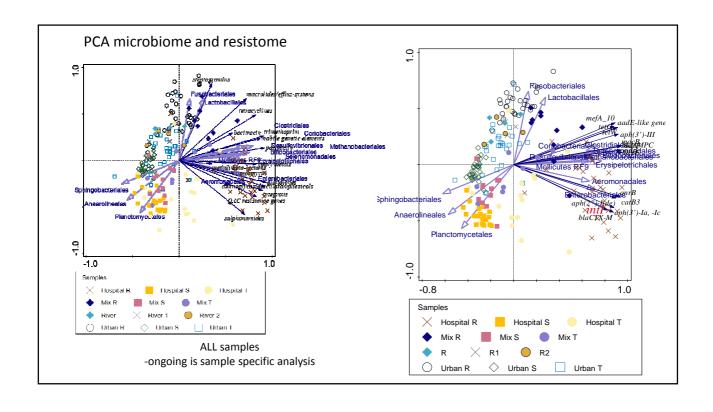
Dots represent samples.

this 61% is displayed on axis x and 8% in axis y.

concentrations (p value is 0.02). Of this 54% is displayed

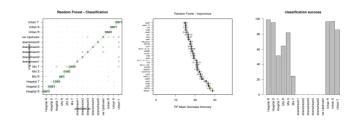
on axis x and 8% in axis y. Only the most relevant 15

genes are displayed. Dots represent samples.



Random Forrest Algorithm and machine based learning

- Does our data have predictive power?
- Putative implementations for risk assessment?
- Yes
- And probably yes



Random Forrest Algorithm and machine based learning Most important predictors (genes and taxa) for raw wastewaters Mixed wastewater (1Hospital:2Urban) carries the URBAN signature Mixed wastewater (1Hospital:2Urban) carries the URBAN signature

Summary

- Robust and distinct resistome and microbiota signature over time
- Hospital wastewaters contain high amounts resistance genes and mobile genetic elements (integrons)
- · Hospital wastewaters contain high amounts of pharmaceutical residues
- Mixed wastewater has Urban signature
 - resistome and microbiome composition is similar to Urban
 - relative abundance of ARGs decreases significantly to levels of ARGs in Urban wastewater when diluting hospital wastewater with Urban wastewater 1 in 2 (1 Hospital Wastewater, 2 Urban Wastewater).
- ARGs are correlated with specific taxa in each sample type
- · Pharmaceuticals and heavy metals significantly impact the resistome and specific mobile genetic elements

Possible implications / future outlook

- Causal relationship between the environment (e.g. chemical concentrations) and the waste- and riverwater resistome and microbiome – implications for public health?!
- Further need to invest in measures for selective removal of pharmaceuticals and chemicals from waste- and river waters -not wastewaters from wastewaters
- Selective removal/depletion of bacteria that carry and spread resistance genes (e.g. bacteria associated with MGEs, integrons) as a measure to reduce the dissemination and persistence of ARGs in the environment



- Christophe Dagot
- Andreu Rico (http://www.water.imdea.org/)
- Jose Laurenco
- Marie-Cecile Ploy
- Sandra Da Re
- Margaux Gaschet
- Sean Kennedy











