Statistical Eco(-toxico)logy

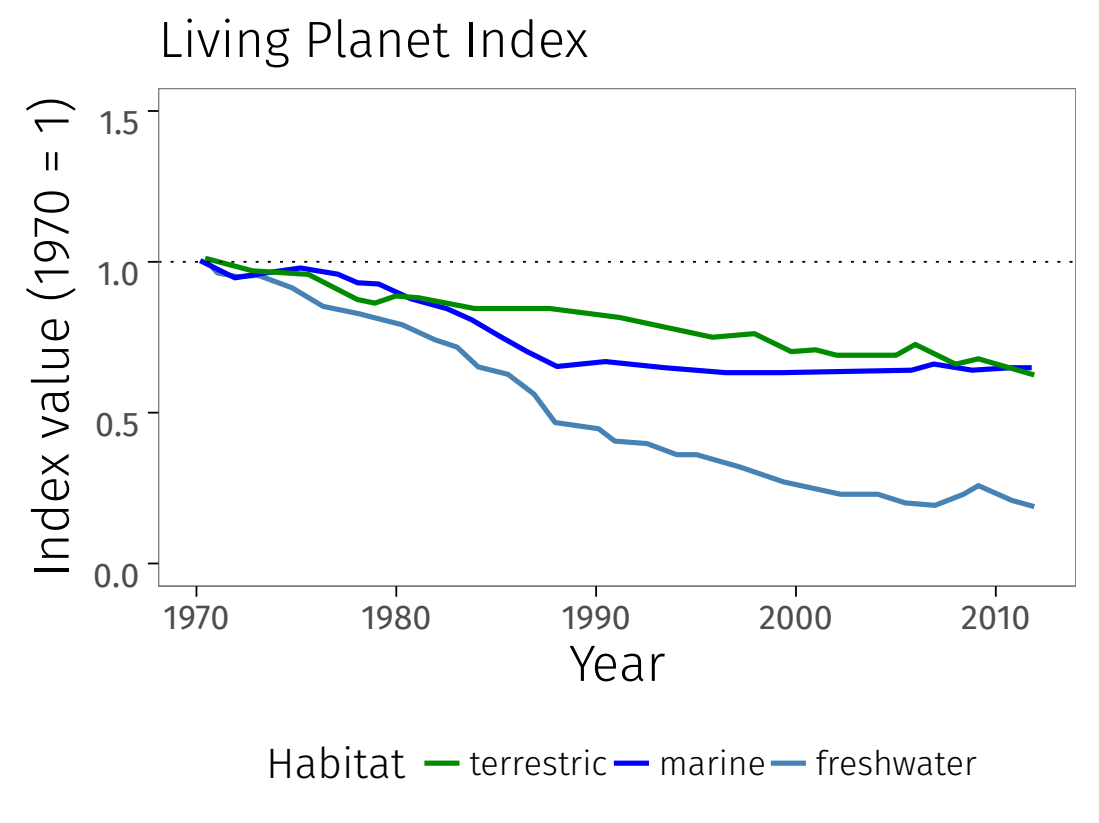
Improving the Utilisation of Data for Environmental Risk Assessment

Eduard Szöcs

25th January 2017

Table of contents

Freshwater biodiversity is strongly declining



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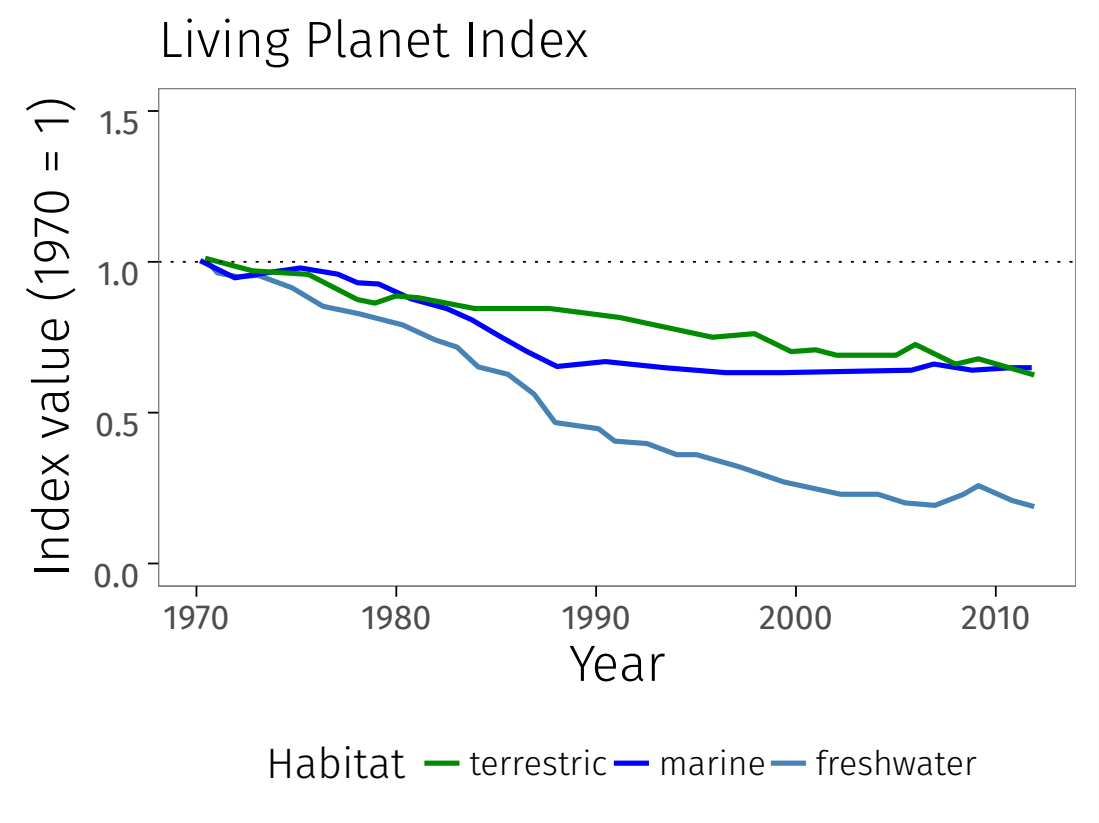

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- Pollution
- Invasive species

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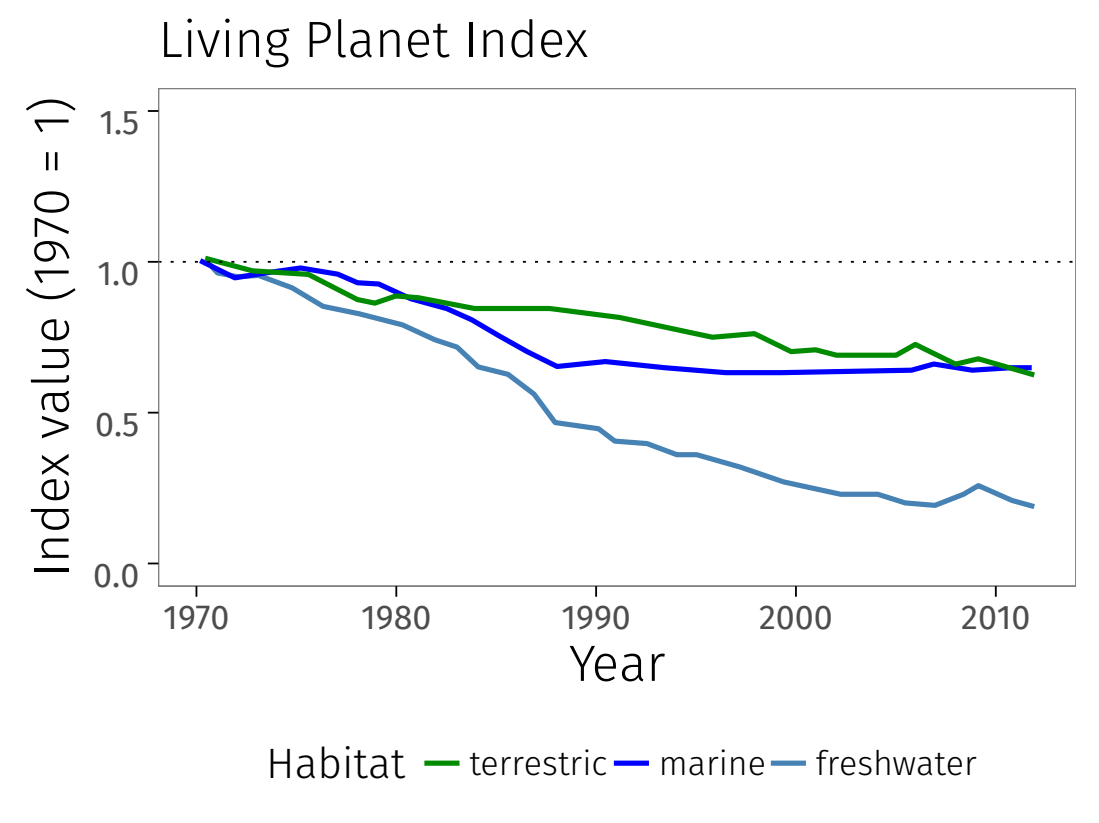

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Environmental Risk Assessment and Environmental Monitoring

Environmental Risk Assessment and Monitoring

Plant Protection Products
Regulation 1107/2009

Water Framework Directive
2000/60/EC

Environmental
Risk
Assessment

Environmental
Monitoring

Environmental Risk Assessment and Monitoring

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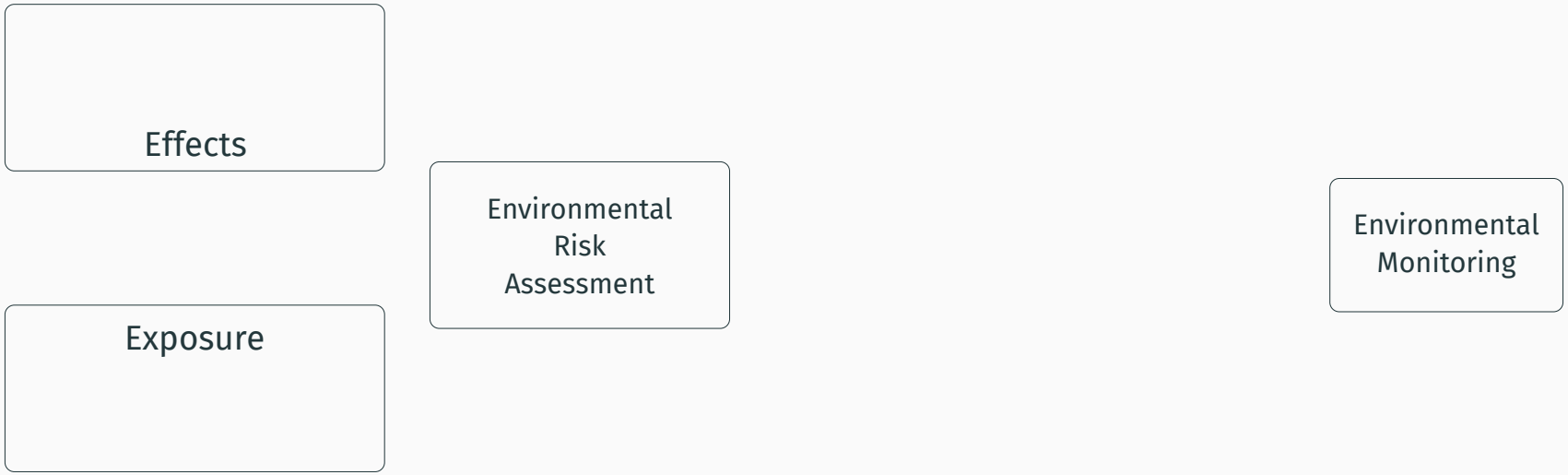

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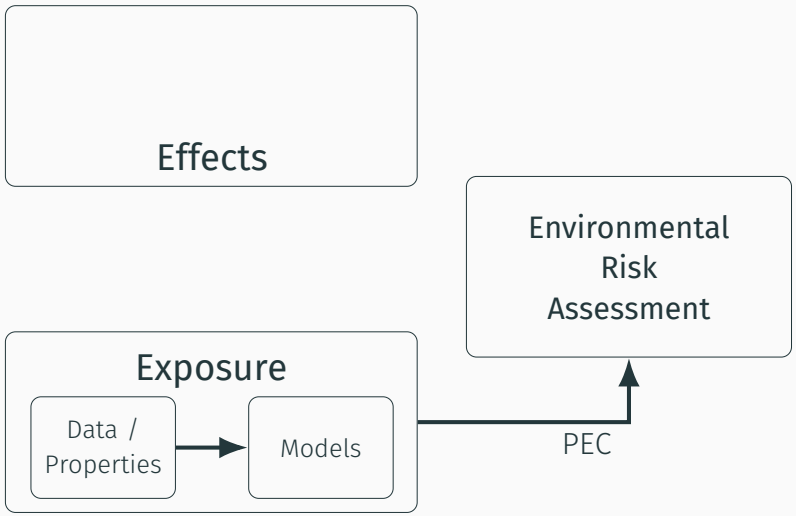

Environmental Monitoring

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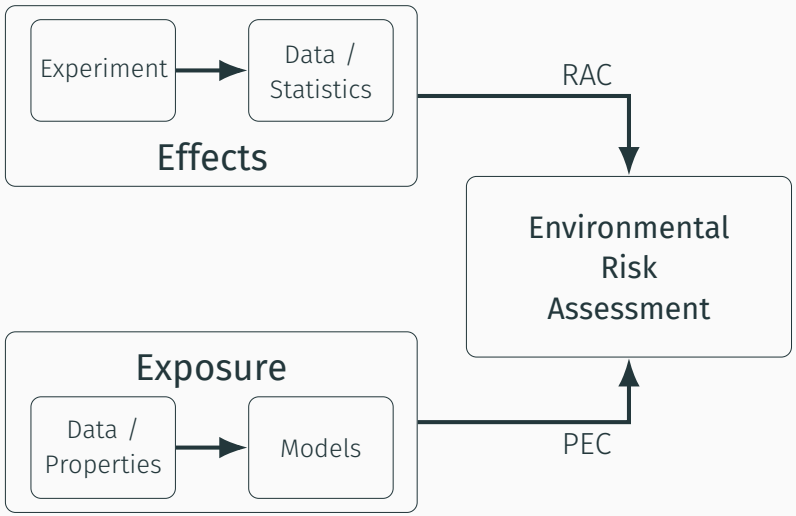

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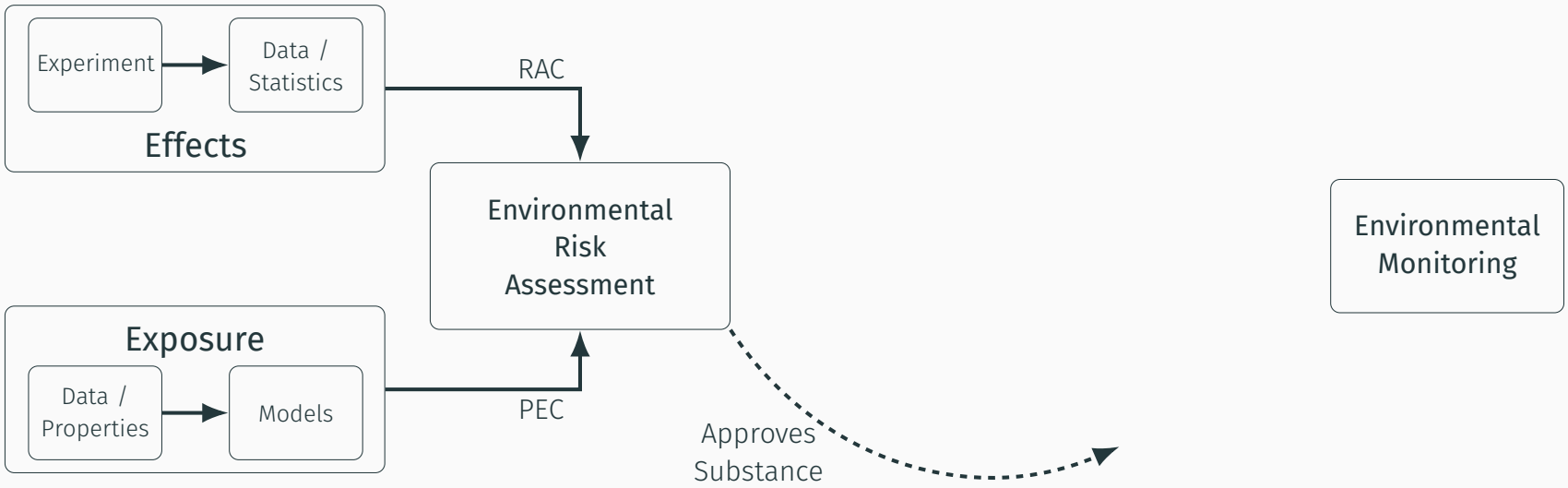

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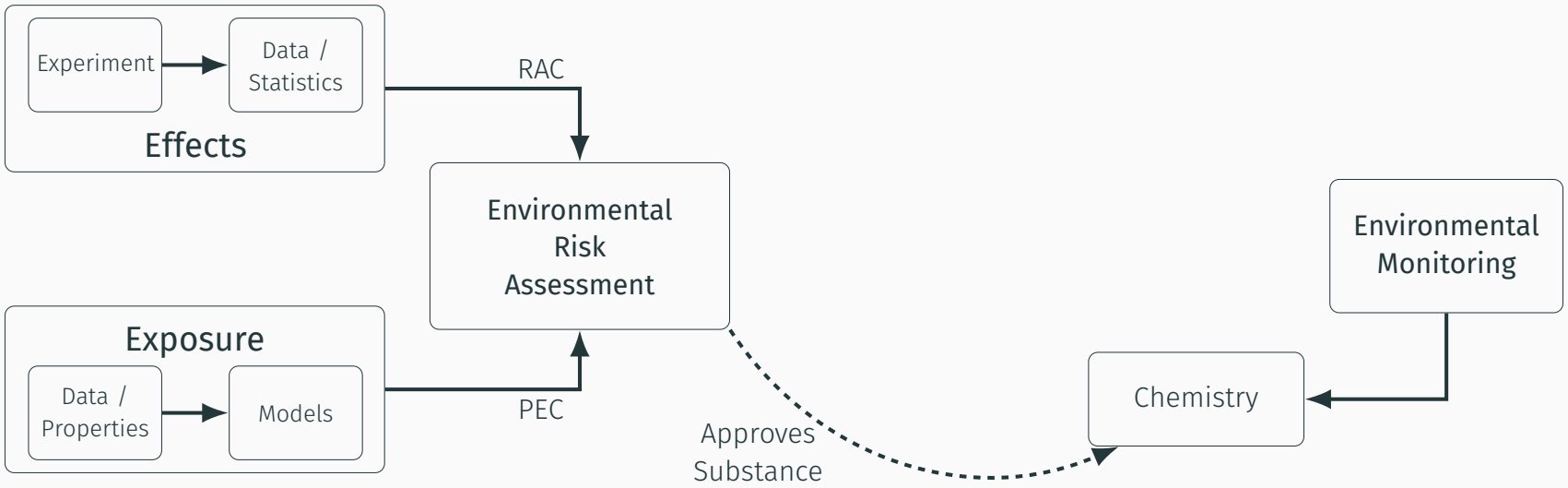

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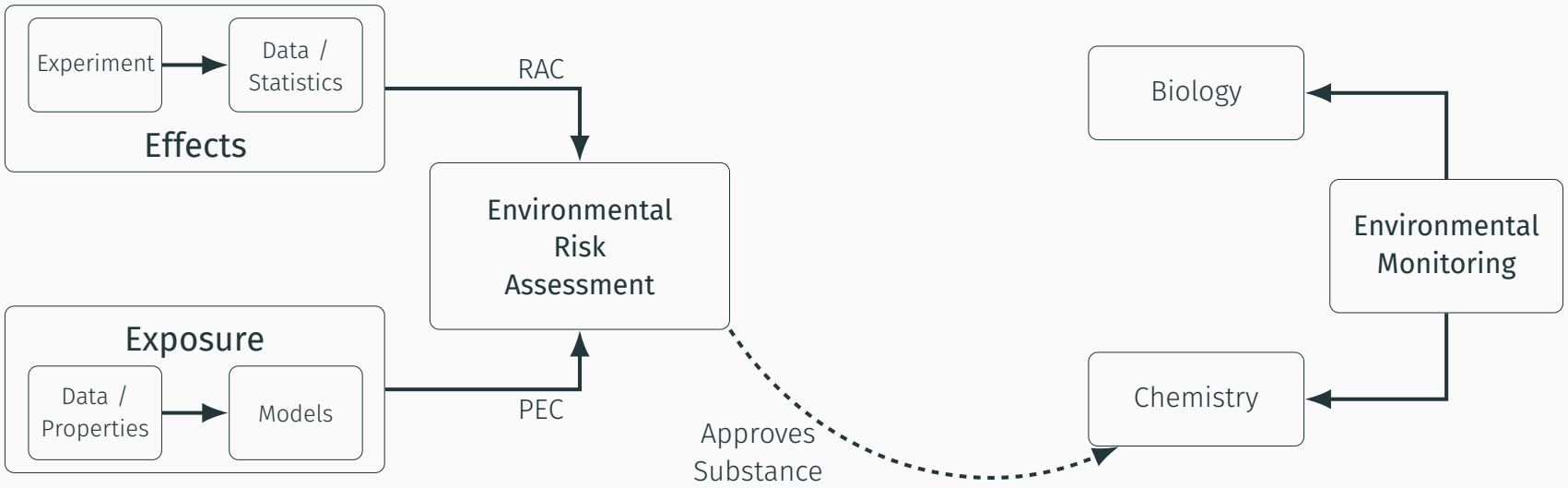

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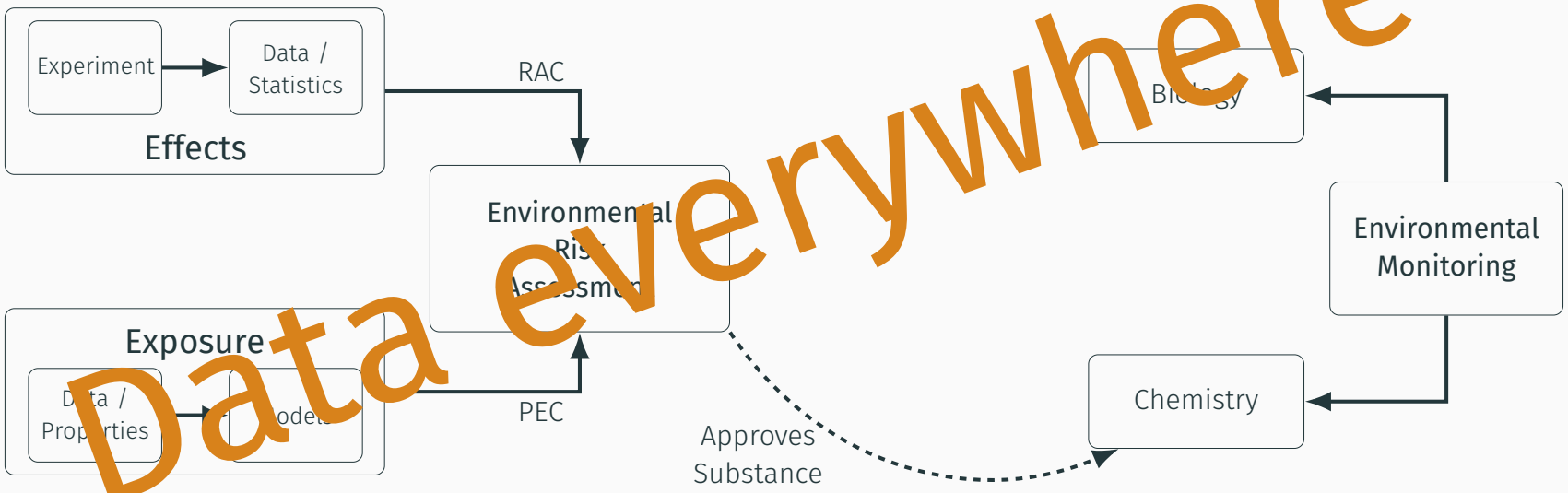
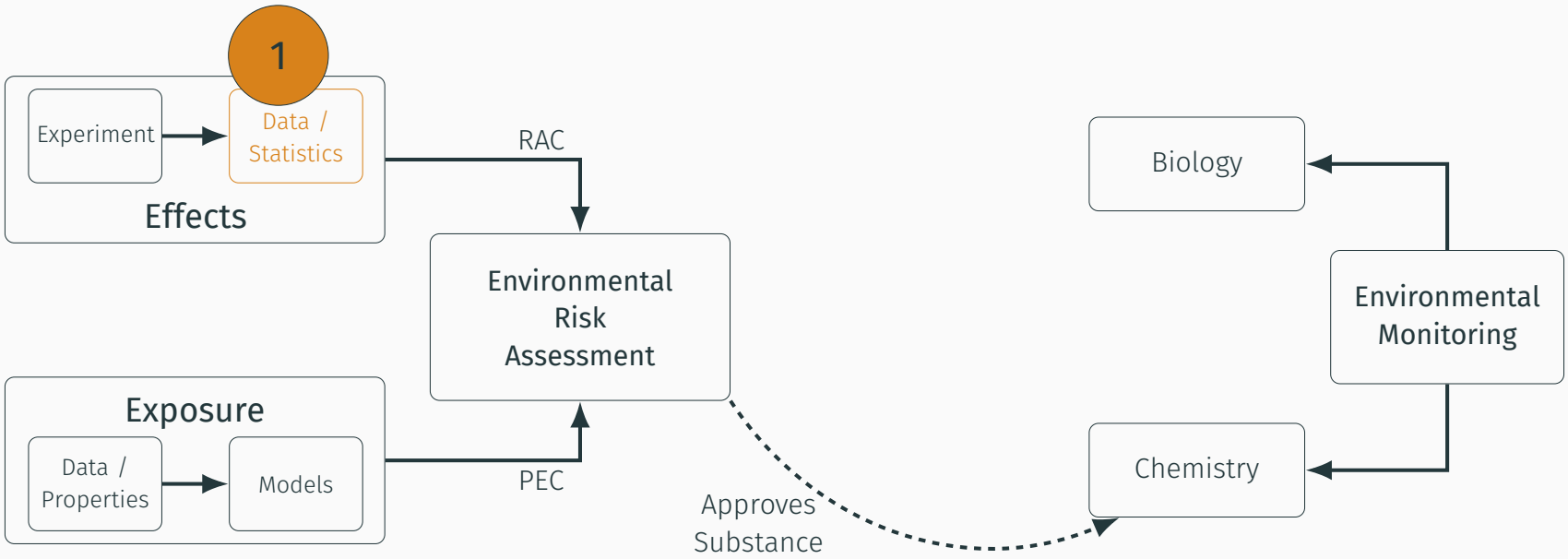


Data everywhere

Improving Statistics in ERA

Plant Protection Products Regulation 1107/2009

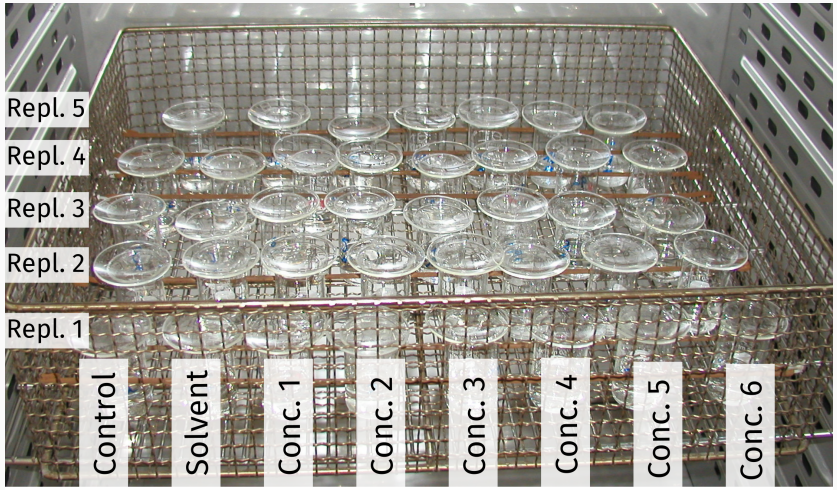
Water Framework Directive 2000/60/EC



1 Szöcs & Schäfer (2015). “Ecotoxicology is not normal”. ESPR 22(18), 13990–13999.

Improving Statistics in ERA

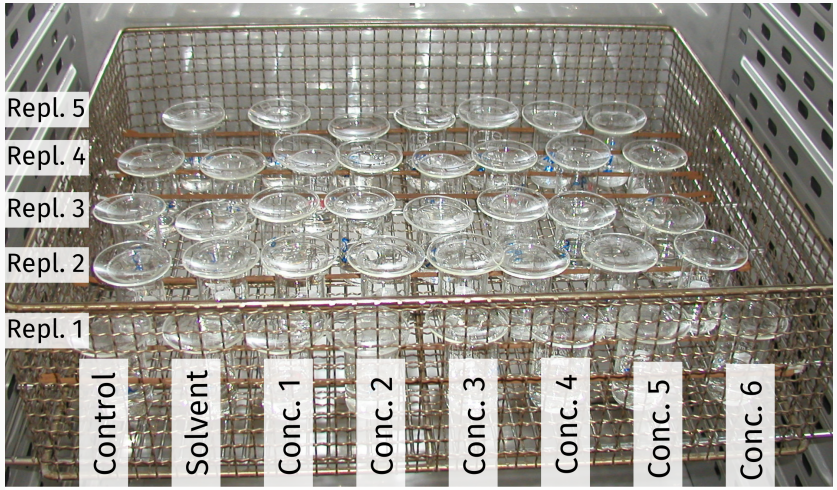
Experiments in Effect Assessment



- Daphnia Test
- *”x out of n survived”*

Source: `www.umweltbundesamt.de` (modified)

Experiments in Effect Assessment



- Daphnia Test
- *”x out of n survived”*

- Mesocosm
- *”number of animals”*

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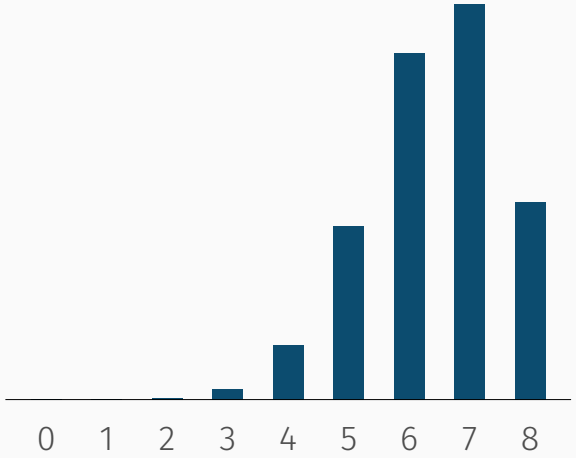
Ecotoxicology is not normal



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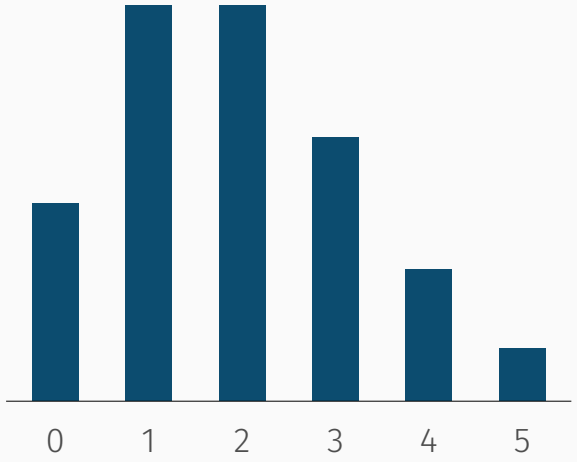

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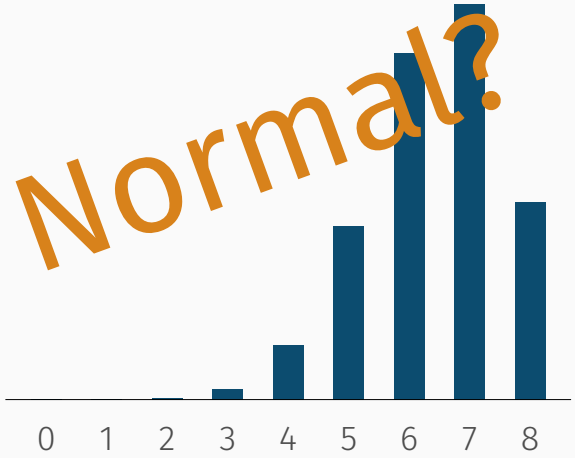

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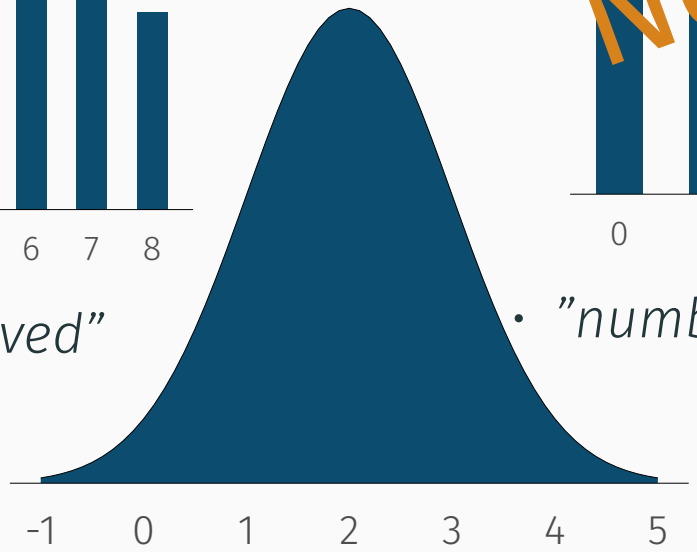

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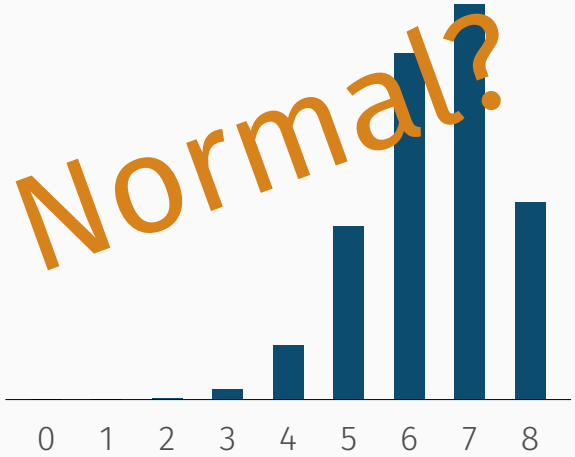

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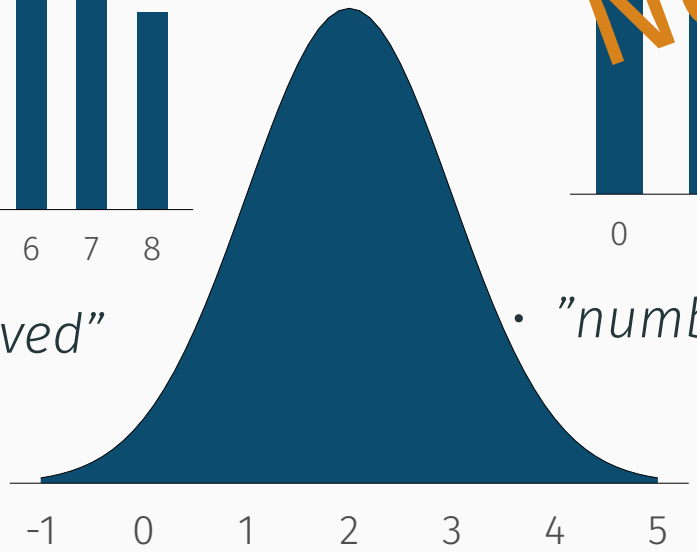

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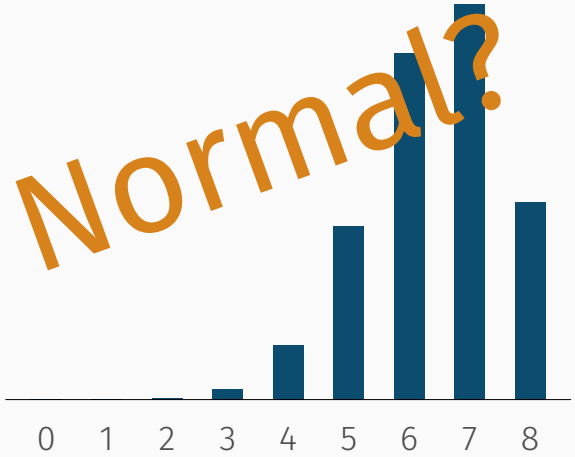

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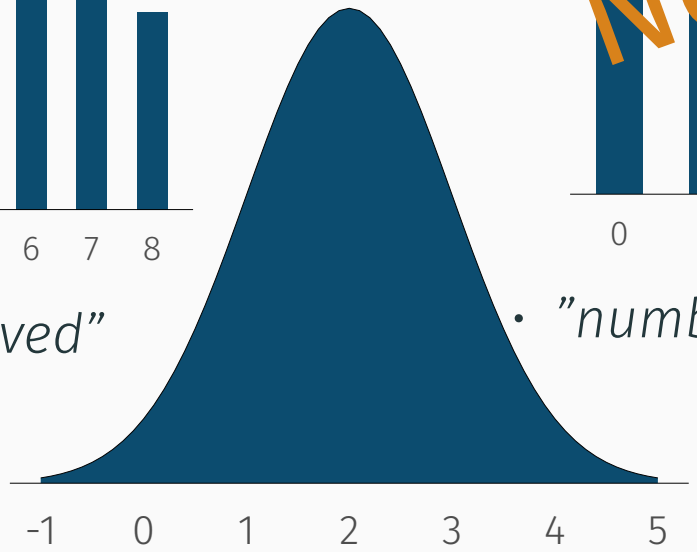

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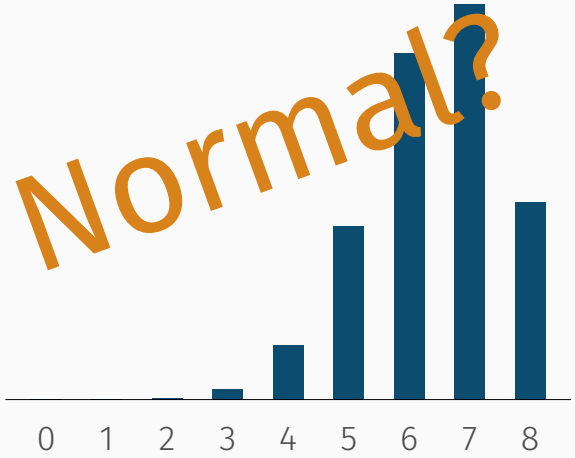

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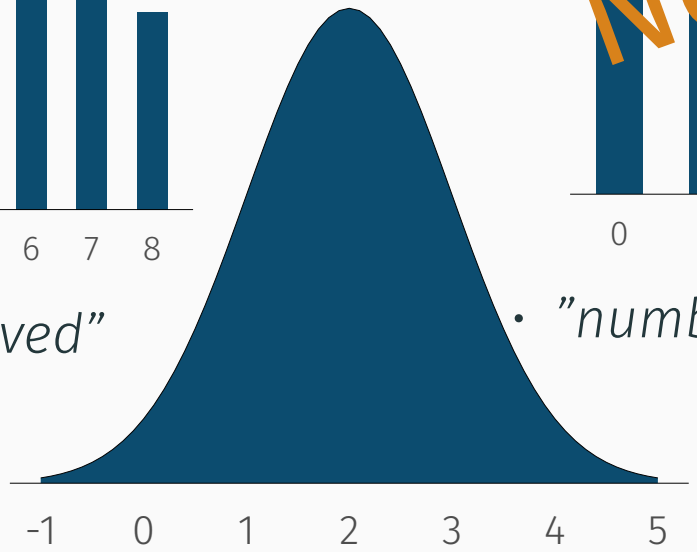

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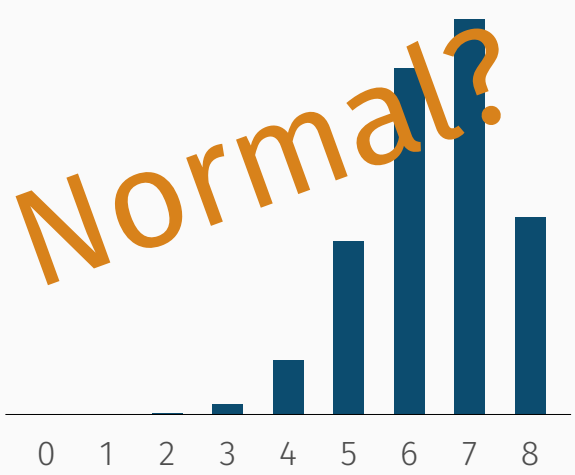

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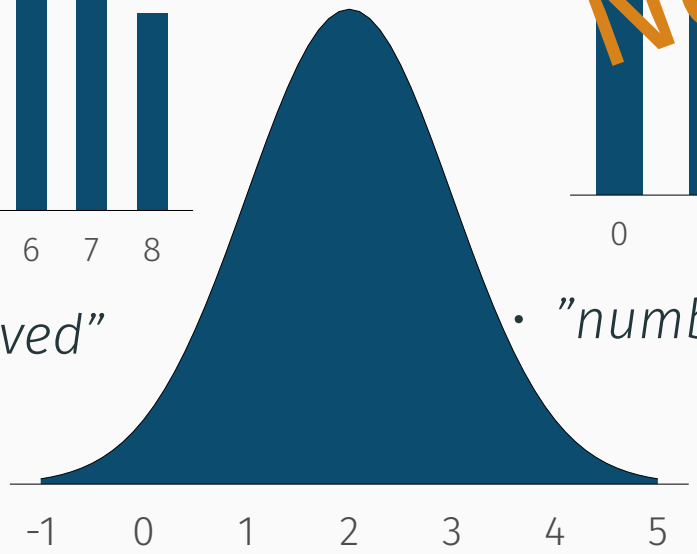

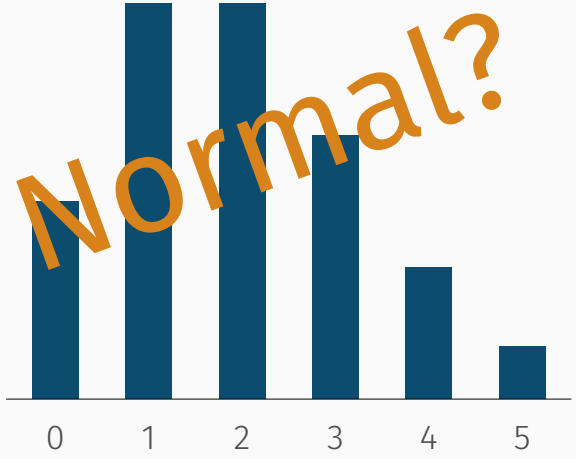

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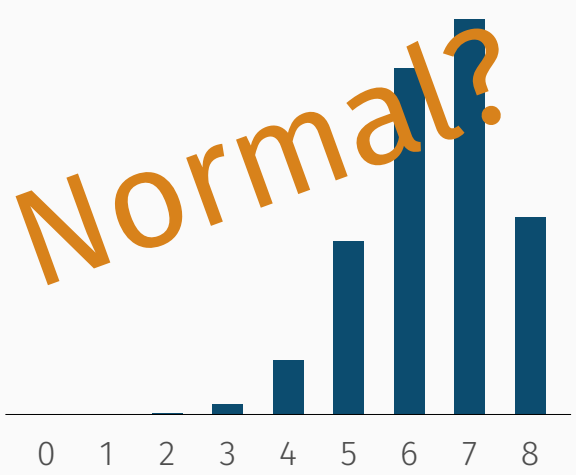

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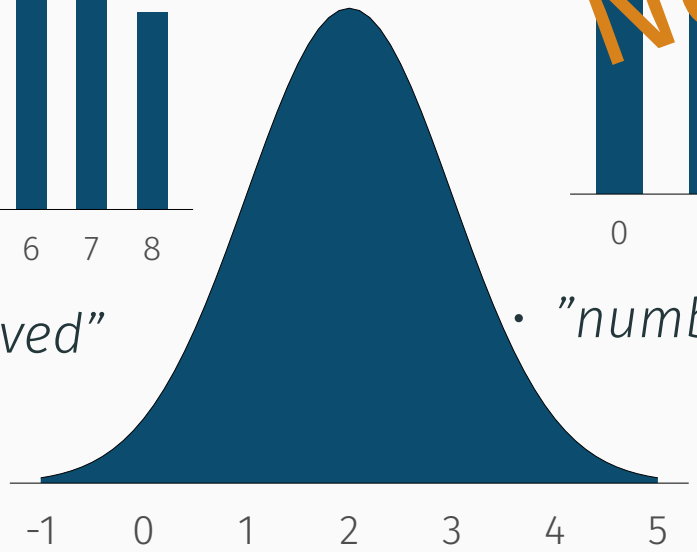

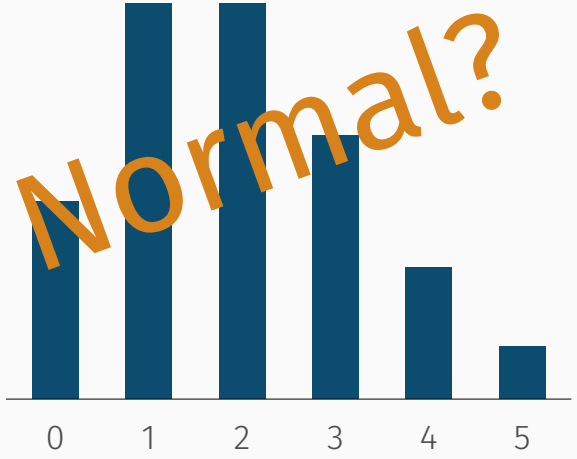

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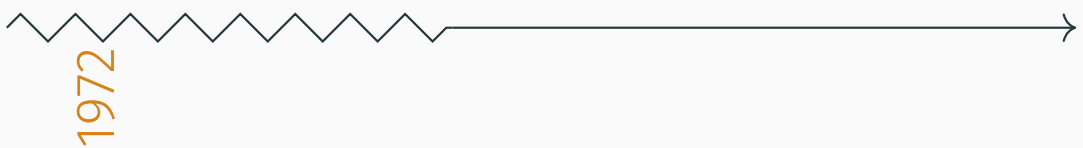
A recent history (uncomprehensive) of GLM in ecology

J. R. Statist. Soc. A,
(1972), **135**, *Part* 3, p. 370

Generalized Linear Models

By J. A. NELDER and R. W. M. WEDDERBURN

Rothamsted Experimental Station, Harpenden, Herts



A recent history (uncomprehensive) of GLM in ecology

Methods in Ecology and Evolution

British Ecological Society

Methods in Ecology & Evolution

doi: 10.1111/j.2041-210X.2010.00021.x

Do not log-transform count data

Robert B. O'Hara[1]* and D. Johan Kotze[2]

[1]*Biodiversity and Climate Research Centre, Senckenberganlage 25, D-60325 Frankfurt am Main, Germany and*
[2]*Department of Environmental Sciences, PO Box 65, University of Helsinki, Helsinki FI-00014, Finland*

A recent history (uncomprehensive) of GLM in ecology

Methods in Ecology and Evolution

British Ecological Society

Methods in Ecology and Evolution doi: 10.1111/j.2041-210X.2011.00127.x

Distance-based multivariate analyses confound location and dispersion effects

David I. Warton[1]*, Stephen T. Wright[1] and Yi Wang[1,2]

[1]*School of Mathematics and Statistics and Evolution & Ecology Research Centre; and* [2]*School of Computer Science and Engineering, The University of New South Wales, NSW 2052, Australia*

A recent history (uncomprehensive) of GLM in ecology

Ecotoxicology
DOI 10.1007/s10646-015-1421-0

Analysing chemical-induced changes in macroinvertebrate communities in aquatic mesocosm experiments: a comparison of methods

Eduard Szöcs · Paul J. Van den Brink · Laurent Lagadic · Thierry Caquet · Marc Roucaute · Arnaud Auber · Yannick Bayona · Matthias Liess · Peter Ebke · Alessio Ippolito · Cajo J. F. ter Braak · Theo C. M. Brock · Ralf B. Schäfer

A recent history (uncomprehensive) of GLM in ecology

Environ Sci Pollut Res
DOI 10.1007/s11356-015-4579-3

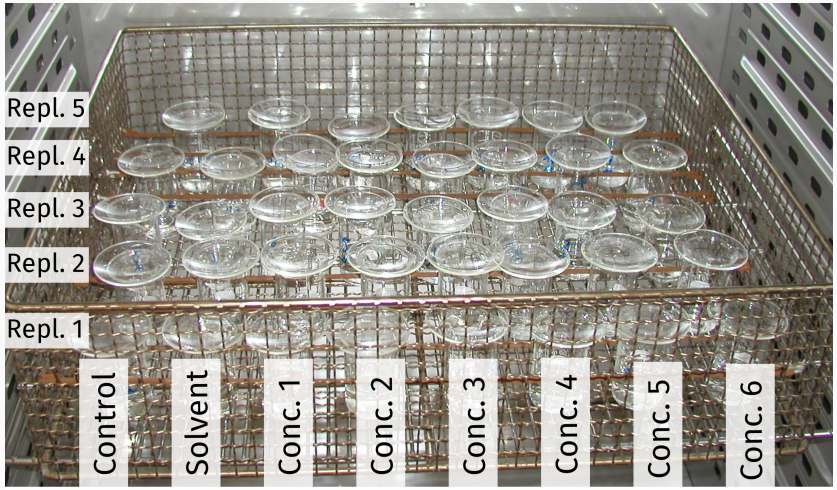
RESEARCH ARTICLE

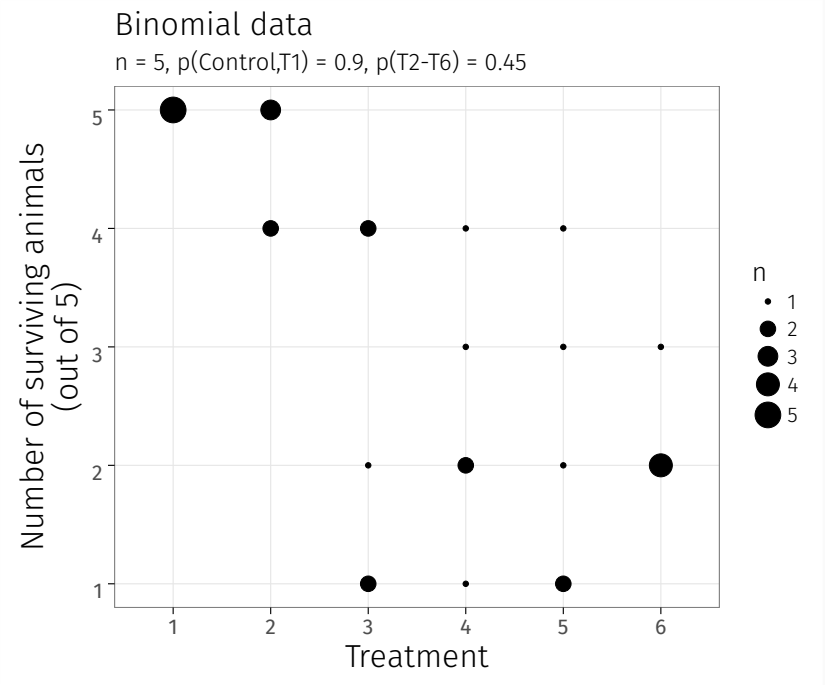
Ecotoxicology is not normal

A comparison of statistical approaches for analysis of count and proportion data in ecotoxicology

Eduard Szöcs[1] · Ralf B. Schäfer[1]

A simulation study

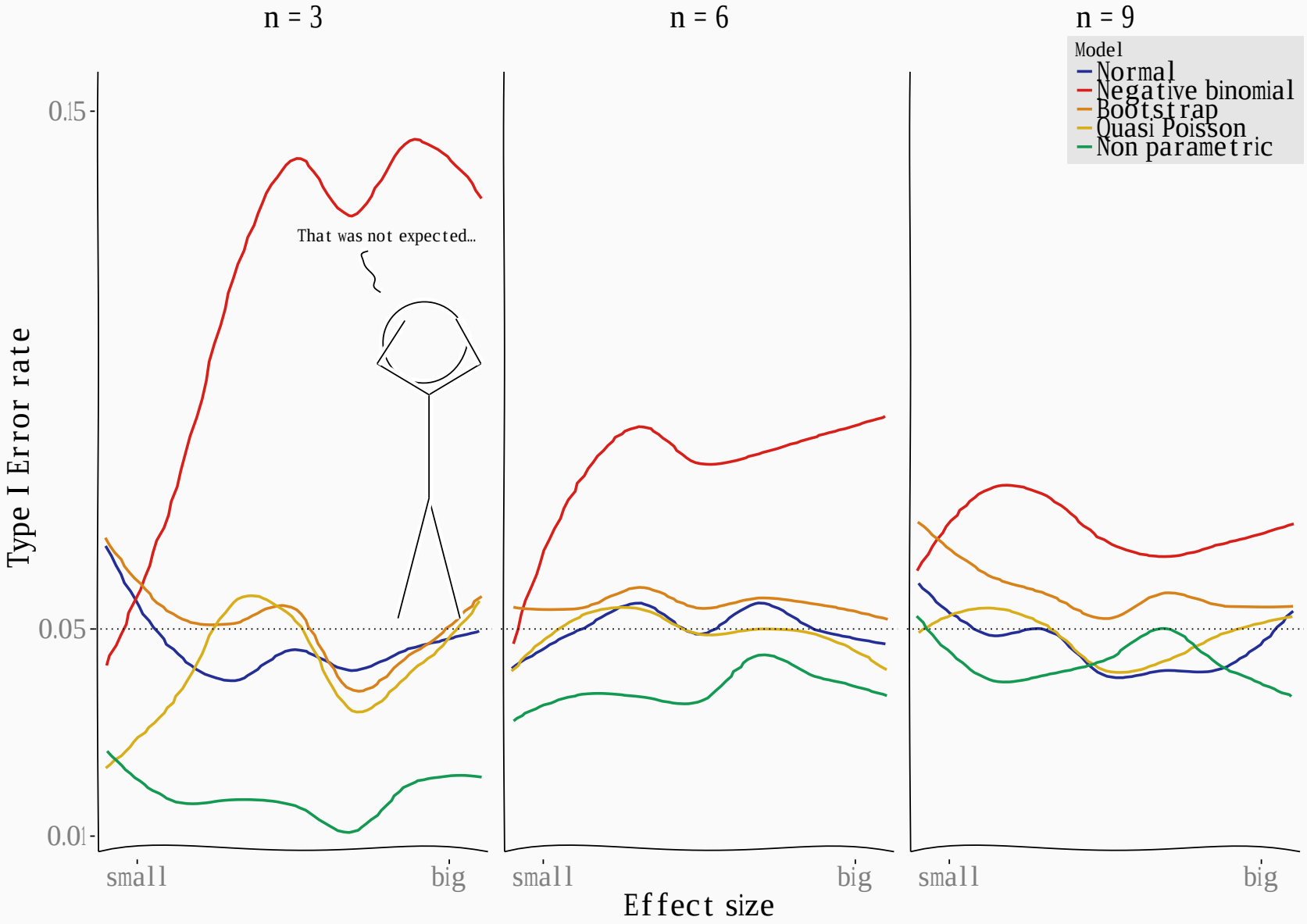




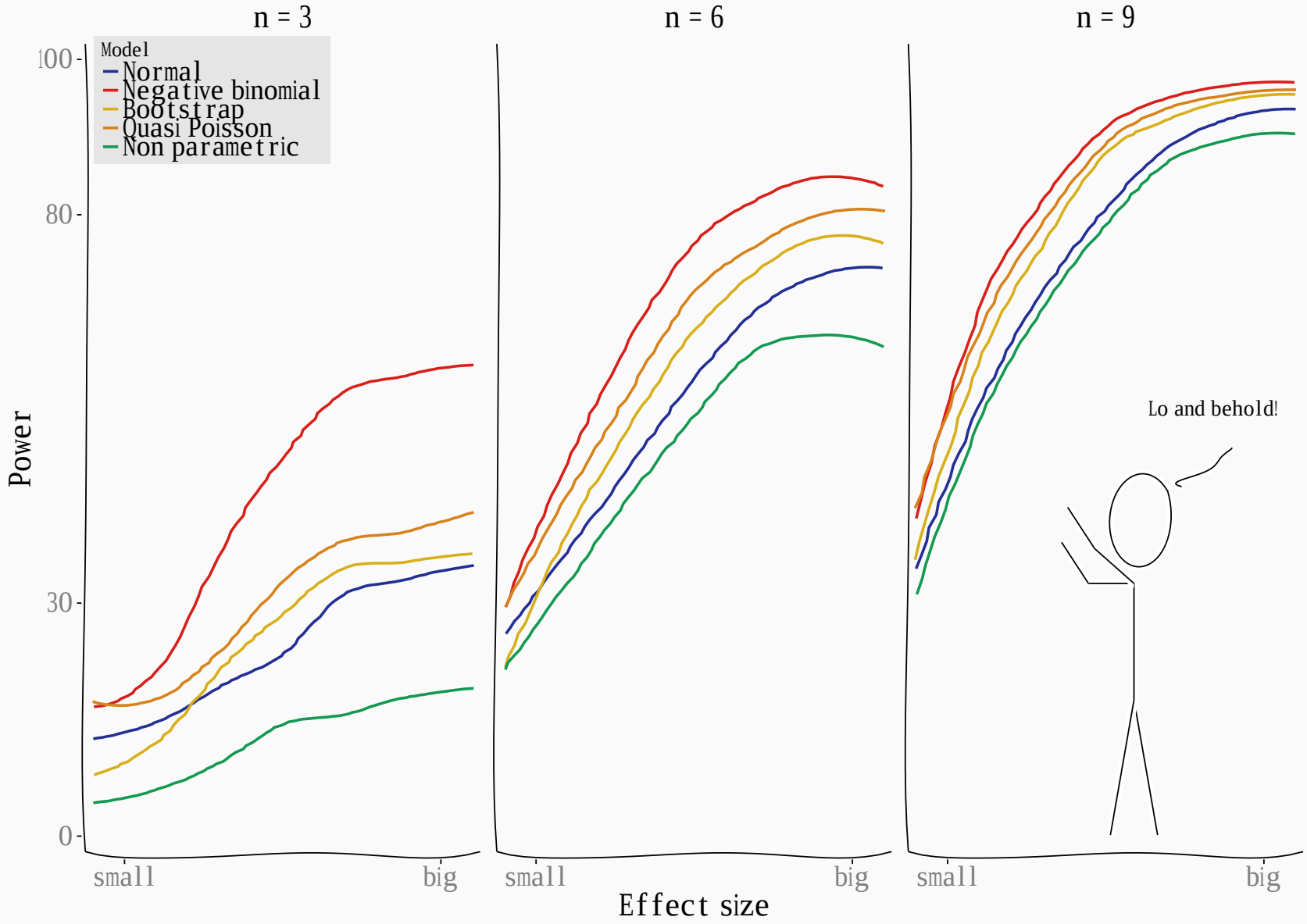
Simulation:

- Count & Binomial data
- Vary replicates & effect sizes
- LM, GLMs, Non-parametric
- Power & Type I Error

Type I Errors: GLMs can fail



Power: But GLMs can do also better



What we learned from this study

1. Negative-binomial GLM show increased Type I errors
2. Can be fixed via bootstrap
3. Ecotoxicological experiments commonly low power
4. GLMs can increase this power

Where are we today?

Three days earlier...

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Three days earlier...

Methods in Ecology and Evolution

British Ecological Society

Methods in Ecology and Evolution 2015, **6**, 828–835 doi: 10.1111/2041-210X.12386

For testing the significance of regression coefficients, go ahead and log-transform count data

Anthony R. Ives*

Department of Zoology, University of Wisconsin-Madison, Madison, WI 53706, USA

Where are we today?

Current state of knowledge:

Methods in Ecology and Evolution

British Ecological Society

Methods in Ecology and Evolution 2016, **7**, 882–890 doi: 10.1111/2041-210X.12552

FORUM

Three points to consider when choosing a LM or GLM test for count data

David I. Warton[1]*, Mitchell Lyons[2], Jakub Stoklosa[1] and Anthony R. Ives[3]

[1]*School of Mathematics and Statistics and Evolution & Ecology Research Centre, University of New South Wales, NSW 2052, Australia;* [2]*School of Biological, Earth and Environmental Sciences, University of New South Wales, NSW 2052, Australia; and* [3]*Department of Zoology, University of Wisconsin-Madison, Madison, WI 53706, USA*

Where are we today?

Three points to consider ...

1. Choose your model based on data properties
2. Fix Type I errors by resampling
3. GLMs have better power properties

Warton, D.I., Lyons, M., Stoklosa, J., Ives, A.R., 2016. Three points to consider when choosing a LM or GLM test for count data. Methods in Ecology and Evolution 7, 882–890.

Exploring Monitoring Data for ERA

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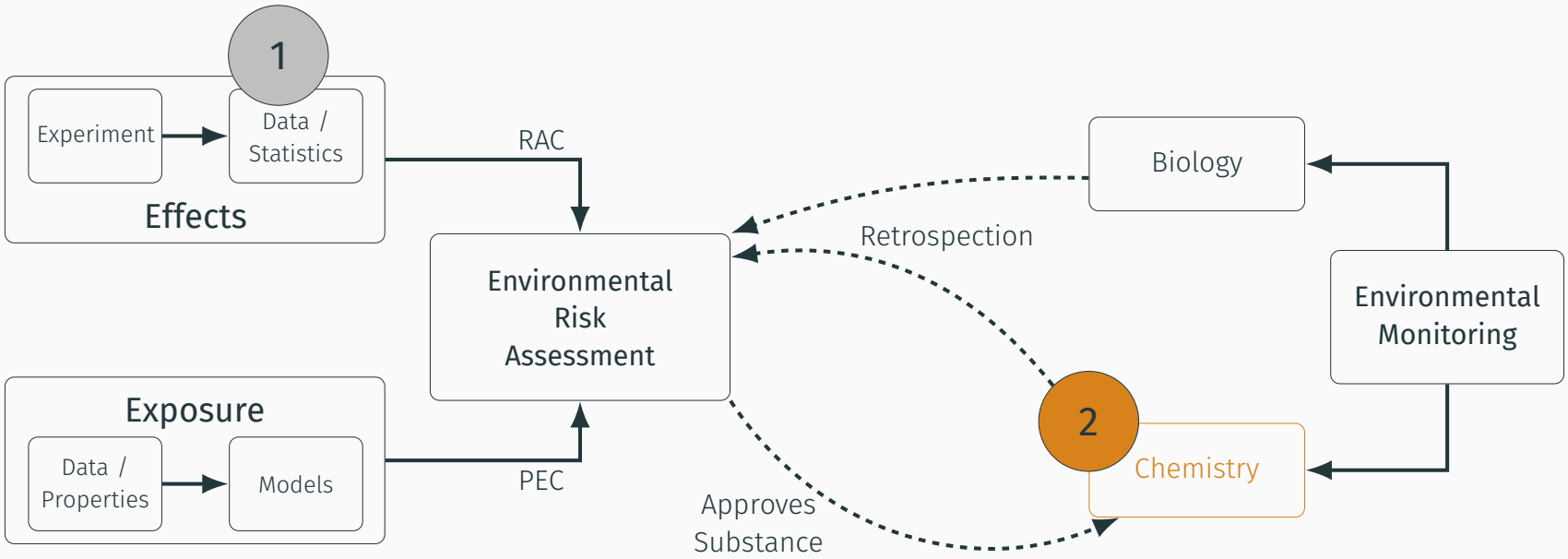
Water Framework Directive 2000/60/EC



2 Szöcs, Brinke, Karaoglan & Schäfer (submitted). “Large scale risks from pesticides in small streams”. Environmental Science & Technology.

Exploring Monitoring Data for ERA

Goals & Hypotheses

Goal: Combine monitoring and ERA

- Compile nation-wide monitoring data
- Focus on small streams
- Identify risks & influencing factors

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Hypotheses

1. Agriculture
2. Stream size
3. Precipitation
4. Annual dynamics

Analysing chemical concentrations

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- Concentrations < LOQ (96% of all measurements)

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$$y \sim ZAGA = \begin{cases} Binomial\ GLM & \text{if } y < LOQ \\ Gamma\ GLM & \text{if } y \geq LOQ \end{cases}$$

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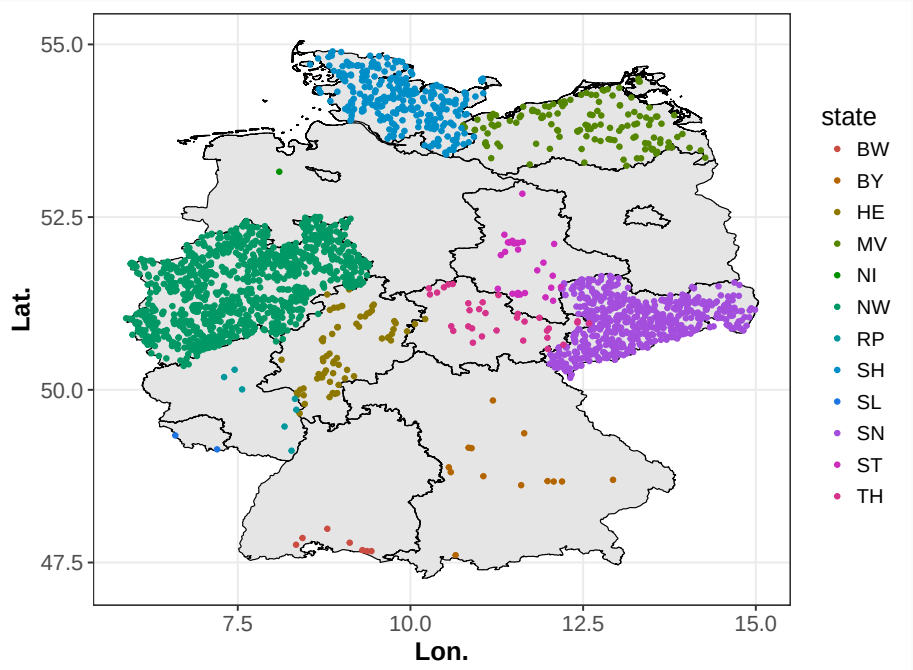
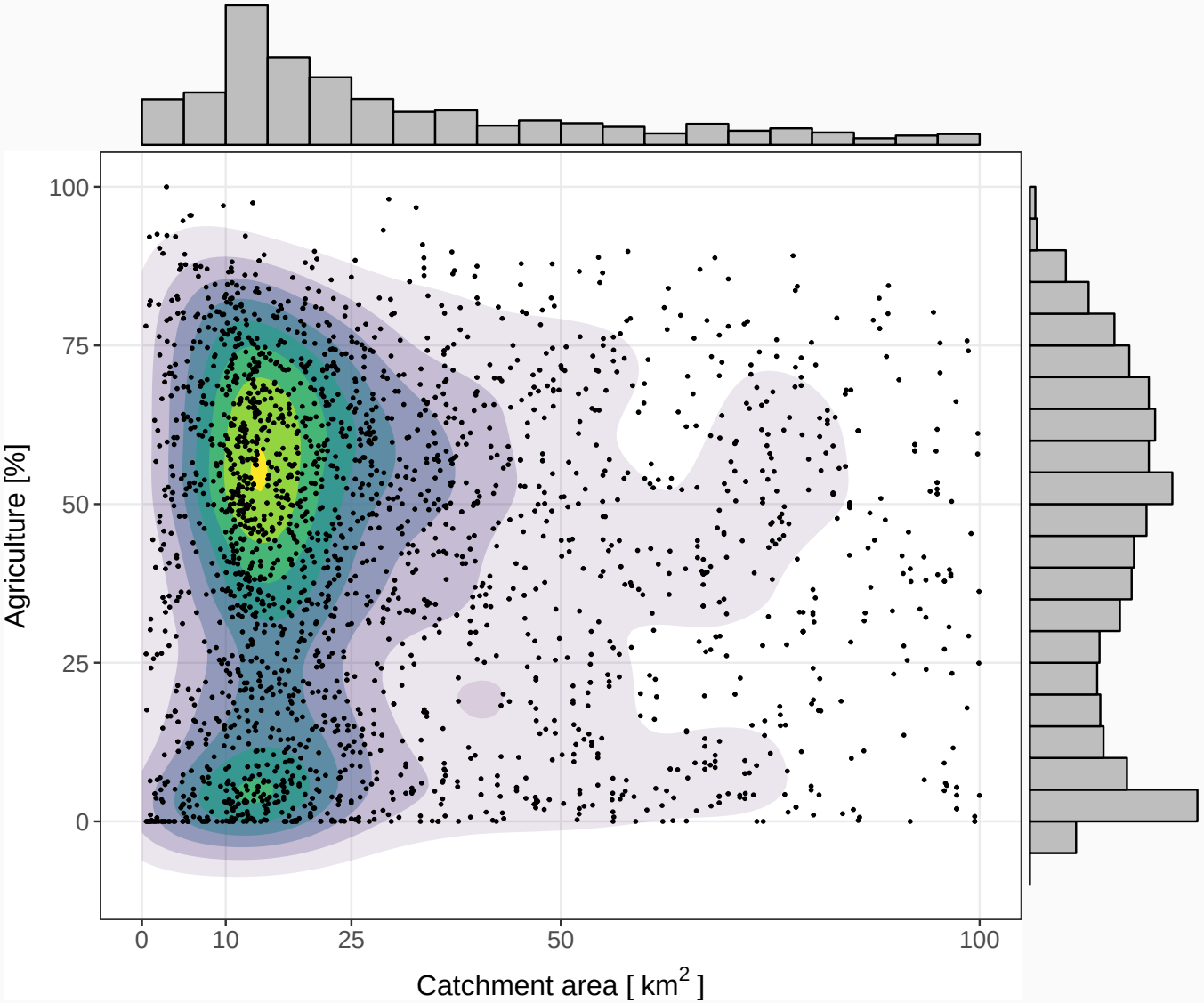
- Risk Quotient
  - $RQ = \frac{C}{RAC}$
- Predictors
  - Catchment Size
  - Agricultural land use
  - Precipitation

Compiled data: Big, but inhomogeneous

- 1,766,104 measurements
- 478 pesticides
- 24,743 samples
- 2,301 sites

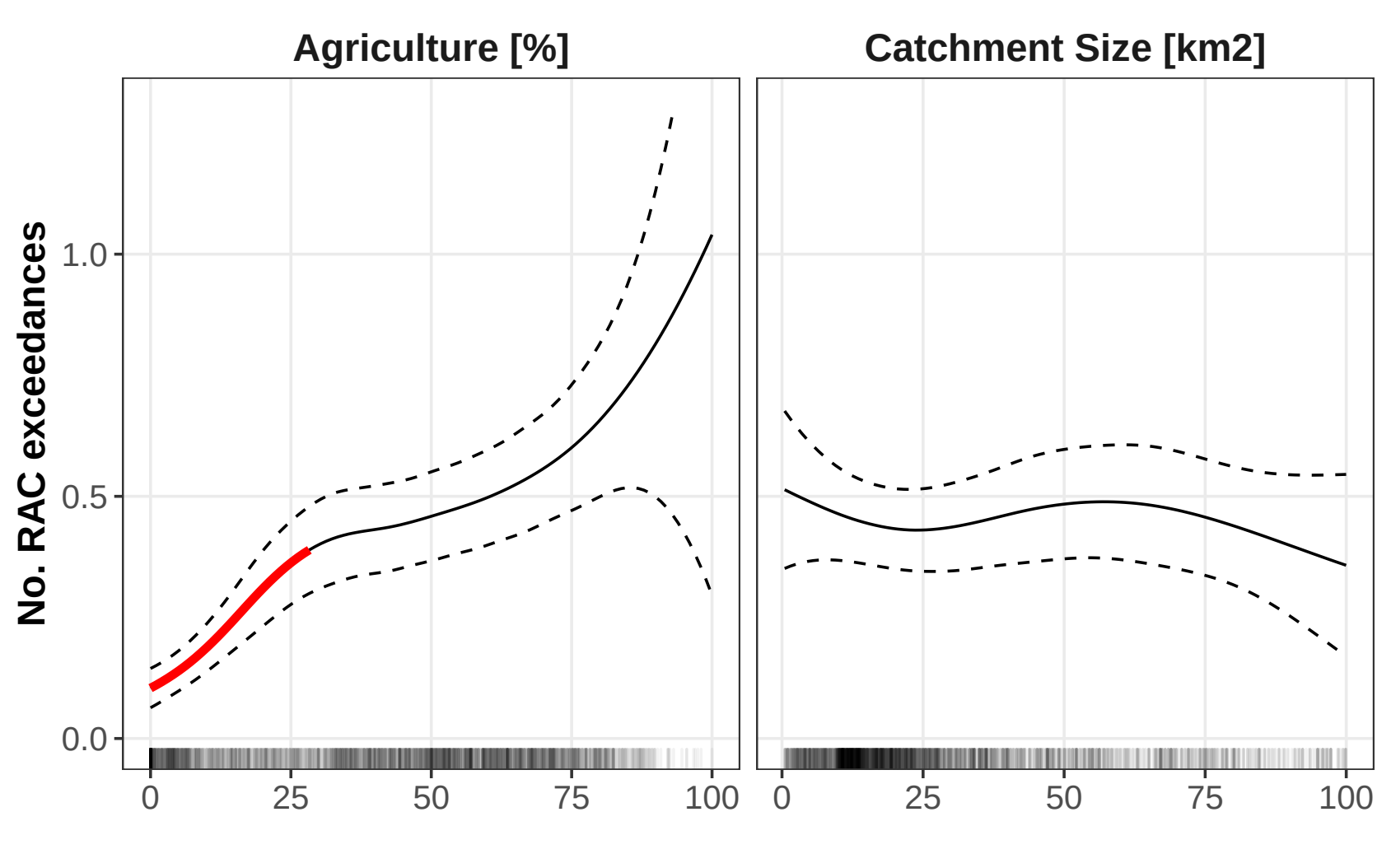


Monitoring: Small streams are underrepresented

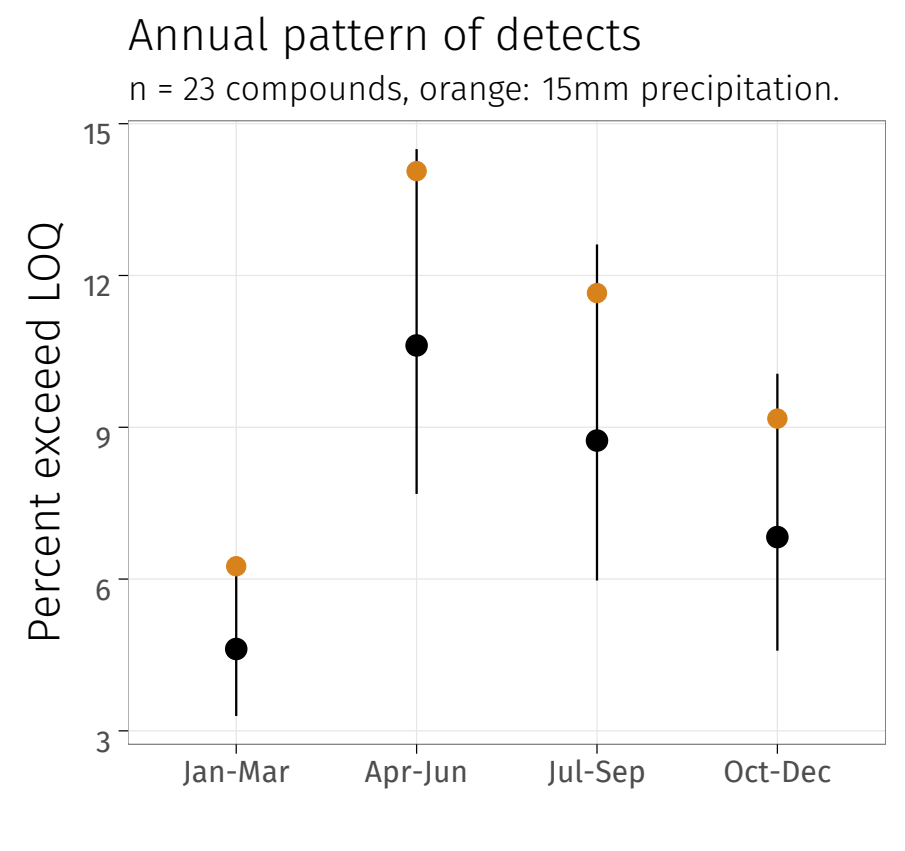


Landscape: Factors influencing risk

- 25% of sites with at least one concentration > RAC

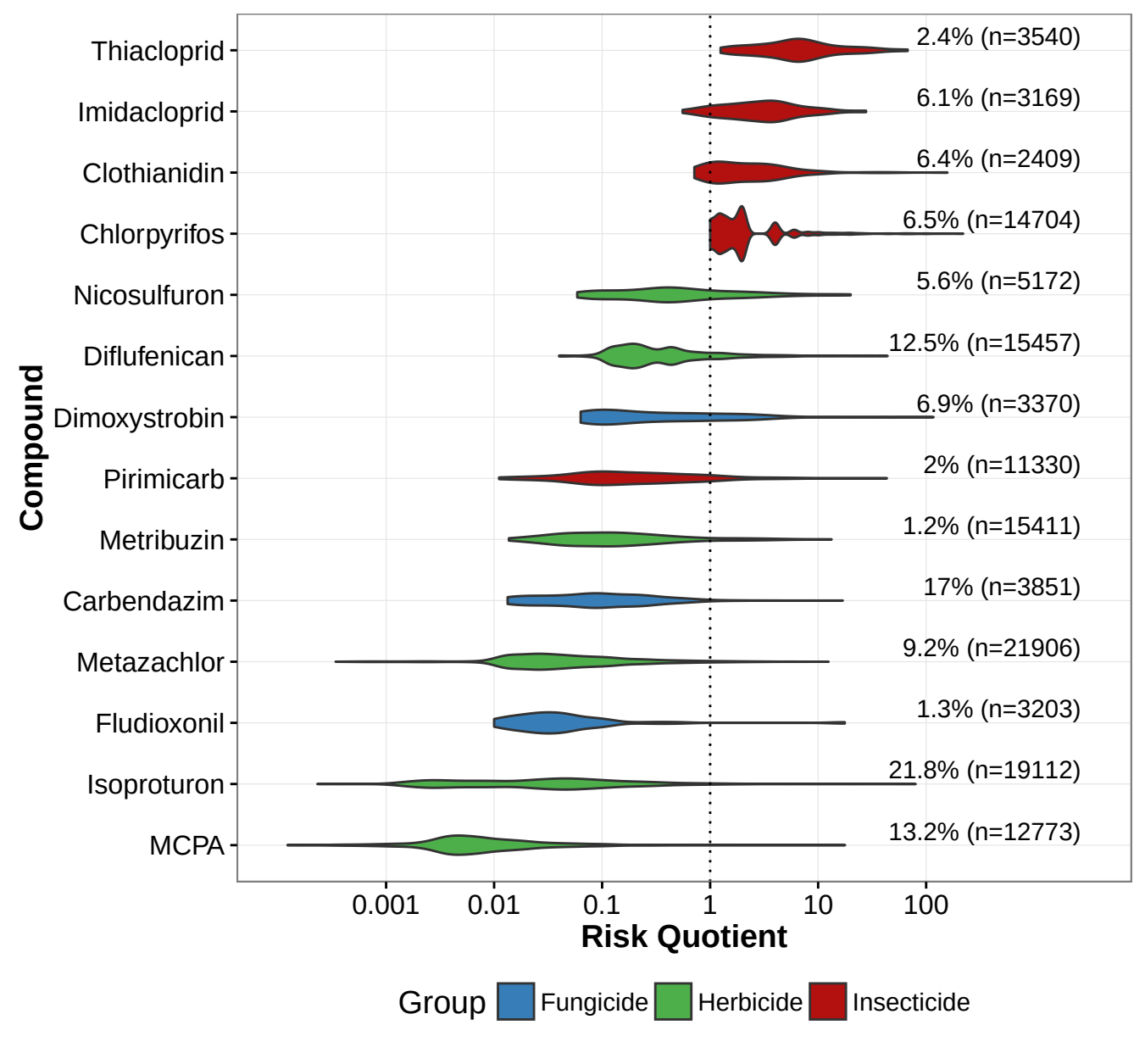


Sampling: Factors influencing risk



- Peak in summer
- Increase by precipitation
- absolute concentrations >>variability

Risks: Compounds exceeding risk thresholds

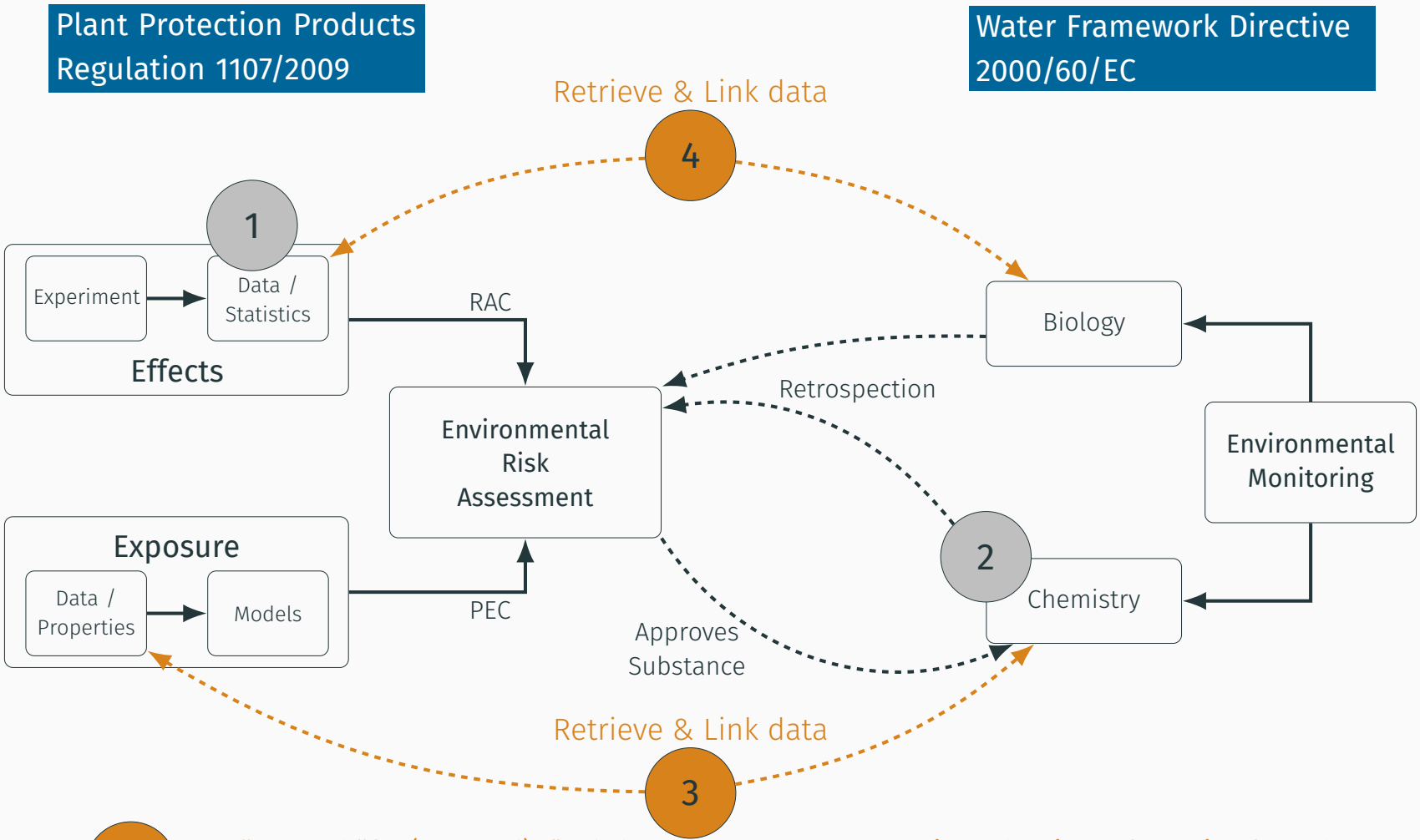


What we learned from this study

1. Differences between states
2. Small streams are underrepresented
3. Agricultural sources
4. LOQ gives additional insights
    - Annual dynamics
    - Precipitation increases concentrations
5. Neonicotinoids

Solutions for Data Handling in ERA

Solutions for Data Handling in ERA



3 Szöcs & Schäfer (accepted). "webchem: An R Package to Retrieve Chemical Information from the Web". Journal of Statistical Software.

4 Chamberlain & Szöcs (2013). "taxize: taxonomic search and retrieval in R". F1000Research 2(191).

Biologists & Chemists face the same problems

Biologists & Chemists face the same problems

Names

Osmia rufa, Osmia bicornis, Osmia ruffa, Osmia unilandauis

Chlorpyrifos, Chlorpyriphos, Chlorphyrifos, Chlorpypifot

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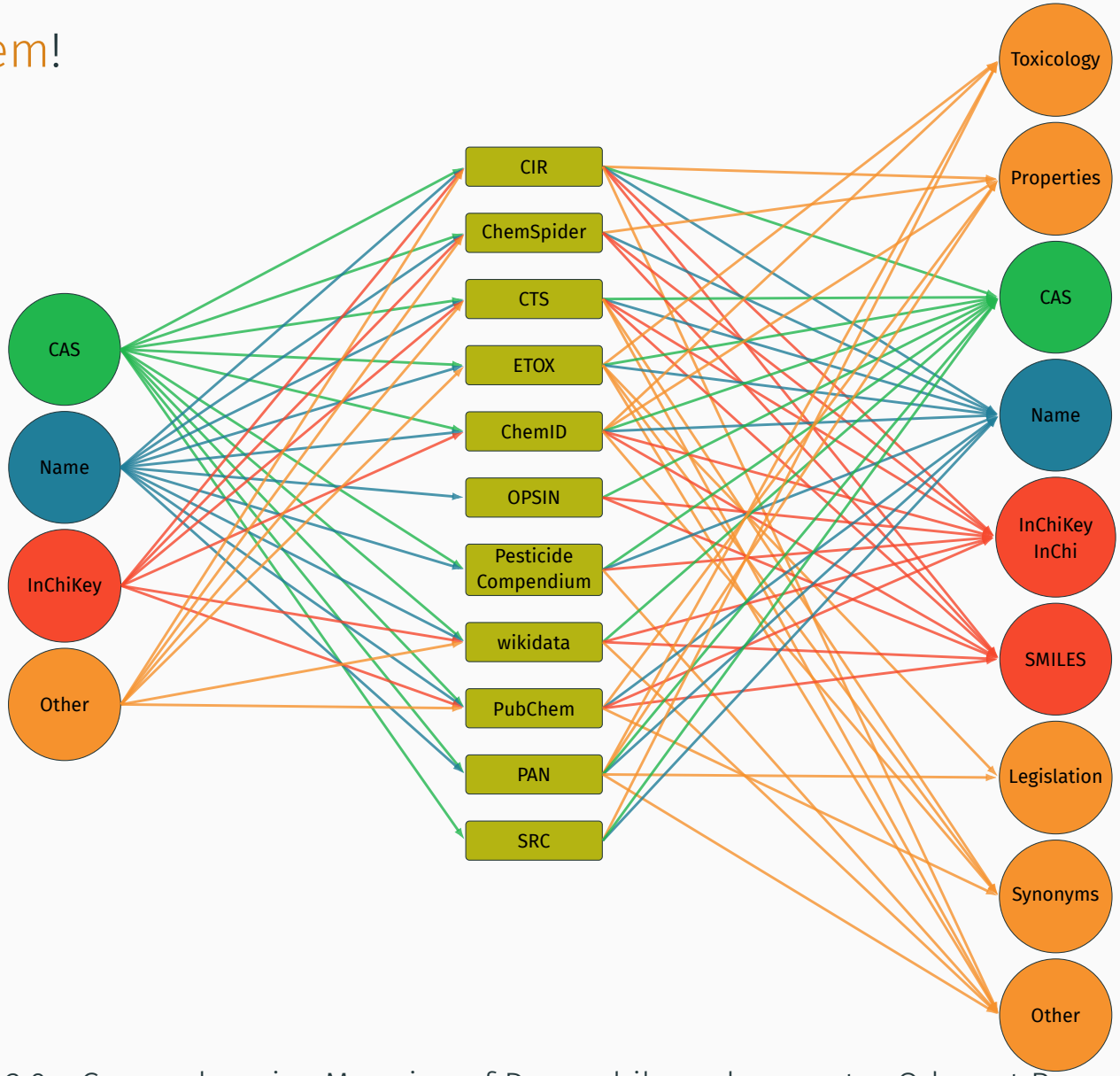
Amount of data

2993 taxa

478 pesticides

Instead of wasting time...

... use webchem!



Münch et al. (2016). DoOR 2.0 - Comprehensive Mapping of Drosophila melanogaster Odorant Responses. Scientific Reports 6, 21841

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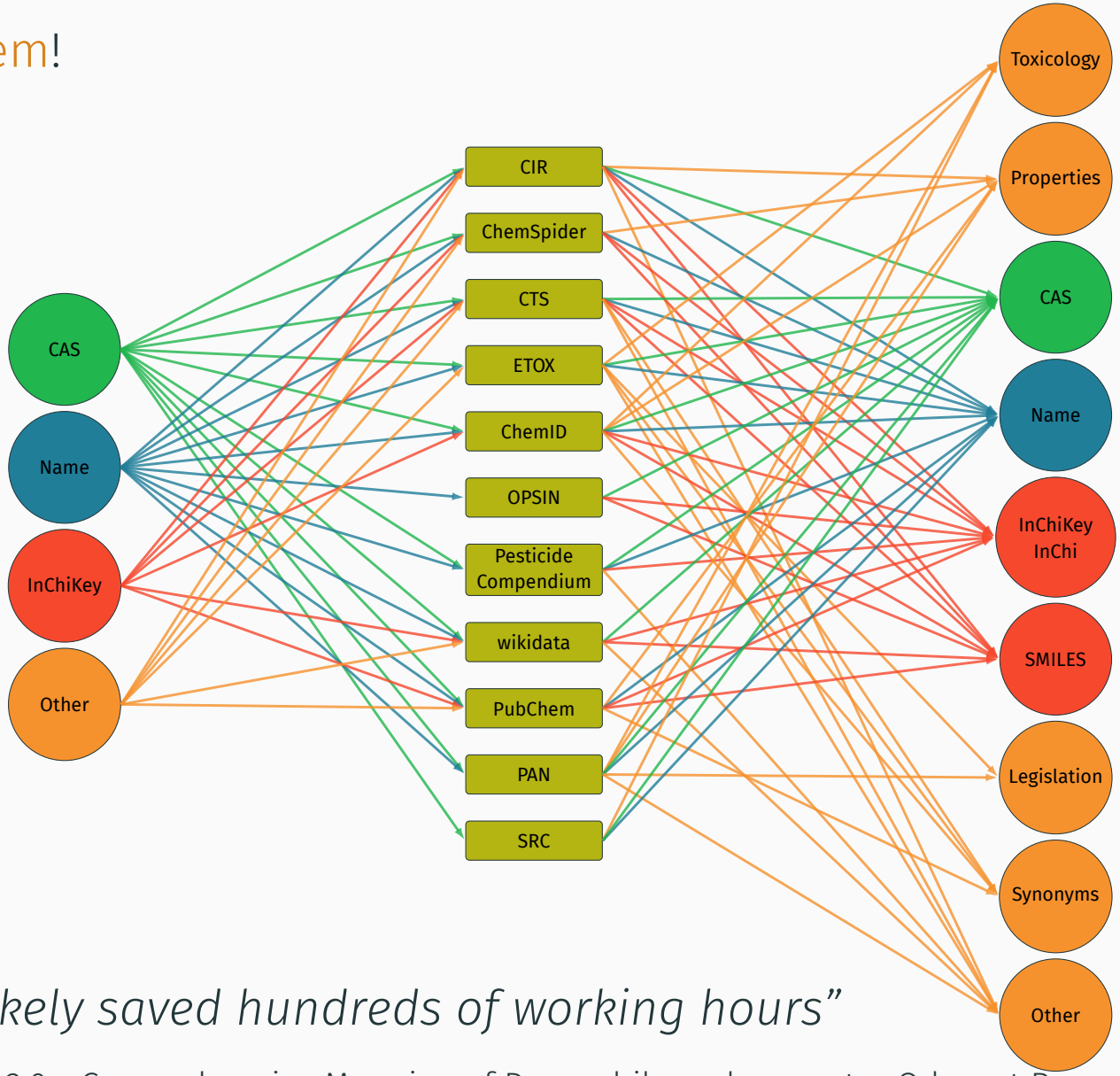

"webchem ...likely saved hundreds of working hours"

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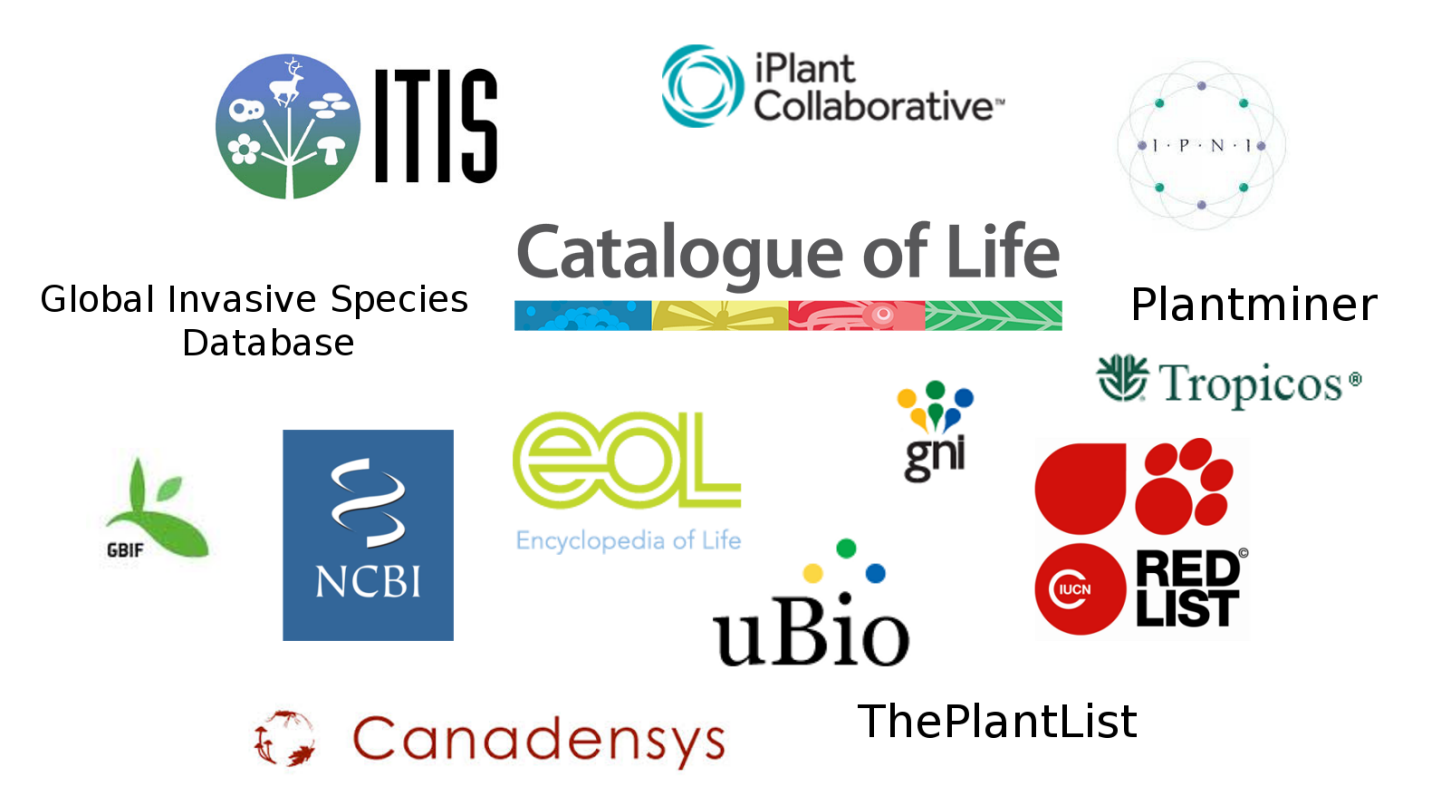

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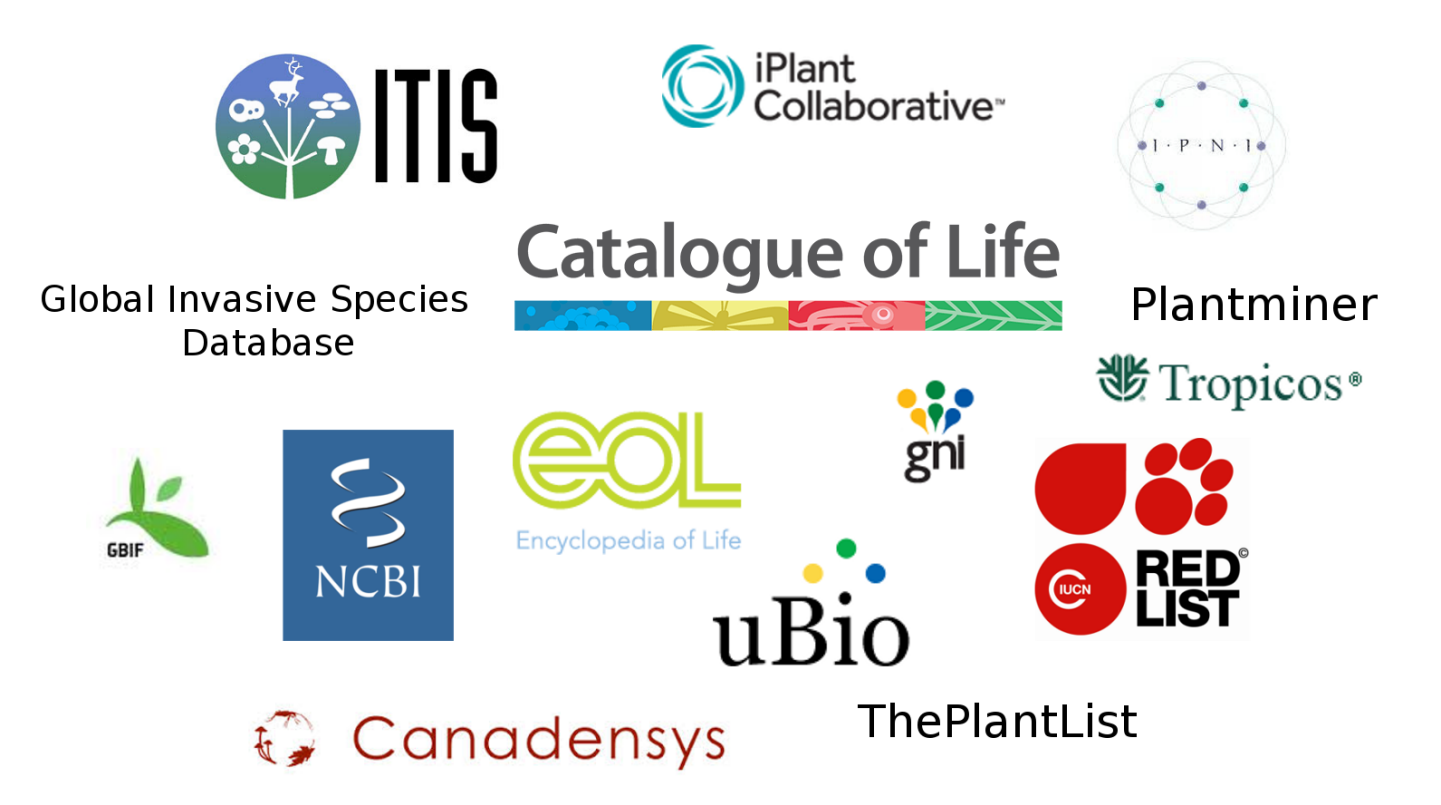
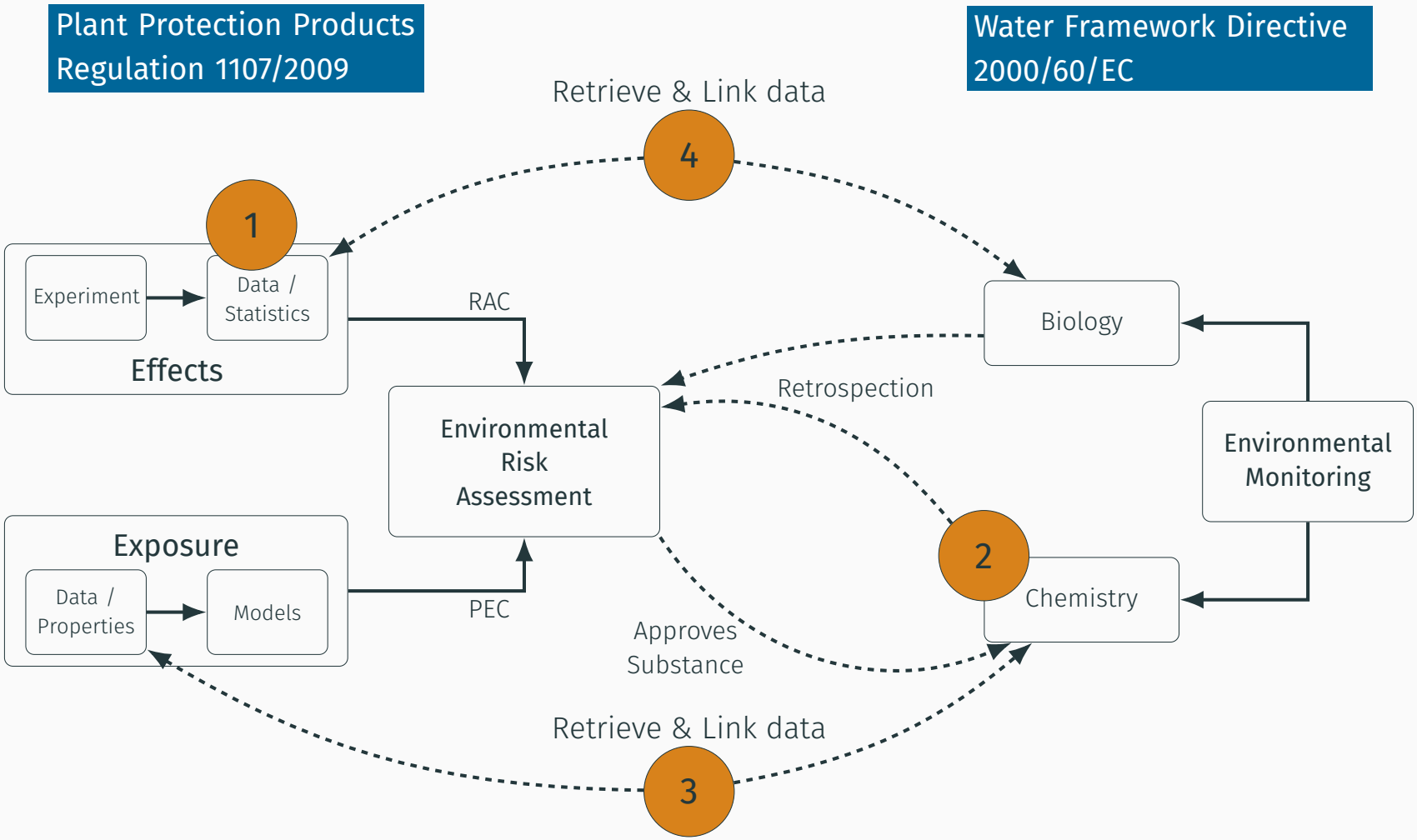


"Days of searching done during my morning coffee. Amazing. taxize."

Dr. Susan E. Johnston, University of Edinburgh. On twitter.

Recap

Recap: What did I look at?



Recap: What we learned from my PhD Thesis

✓Improving Statistics in ERA

- Change your model, not your data
- Take LOQ into account

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- Risk drivers and dynamics
- Agricultural small streams neglected & at risk
- Neonicotinoids

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✓Solutions for Linking Data in ERA

- Handling big eco(toxico-)logical data not easy
- Now easier

Statistical Ecotoxicology

Improving the Utilisation of Data for Environmental Risk Assessment

Eduard Szöcs

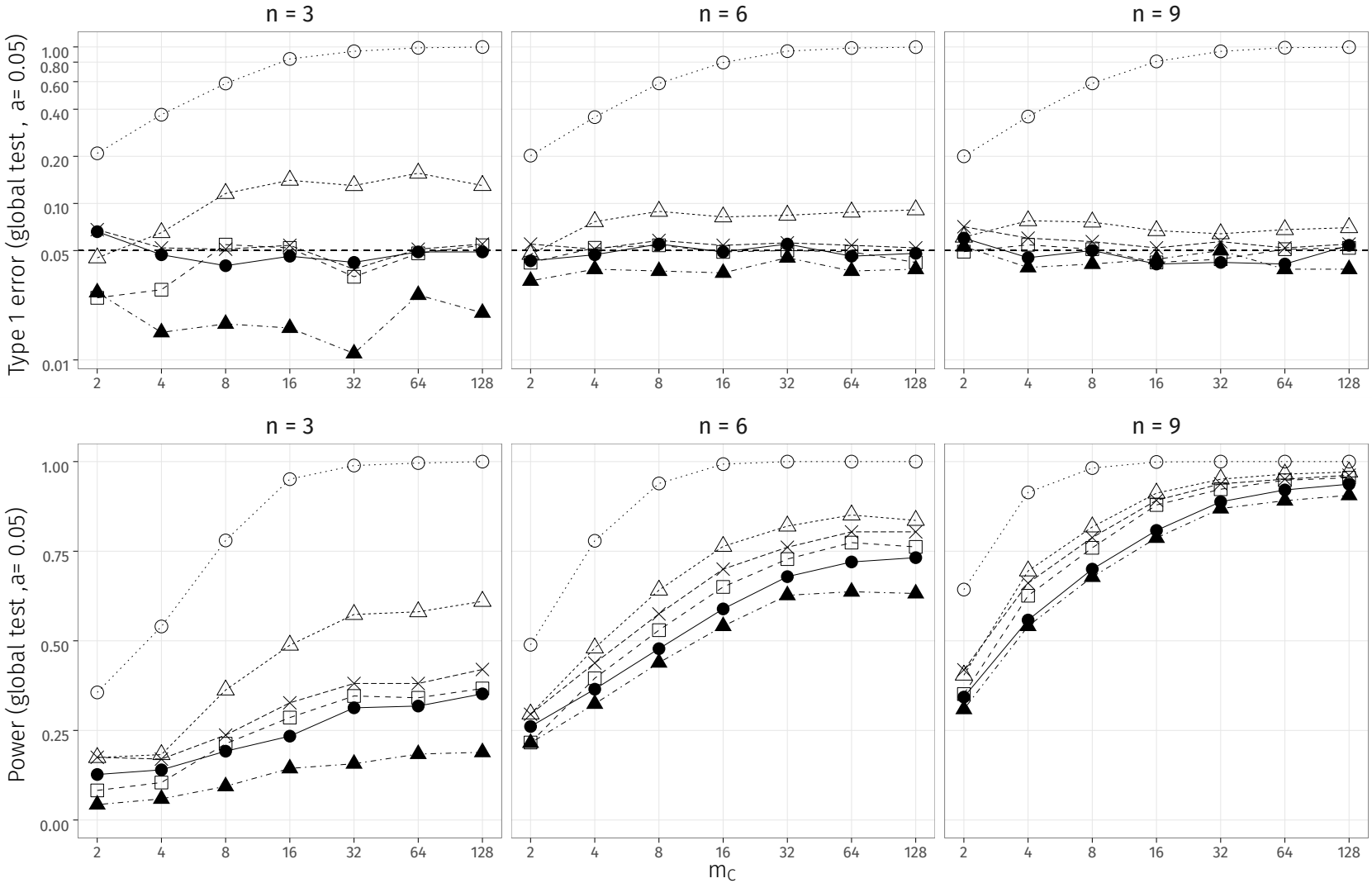
http://edild.github.io/

@EduardSzoecs

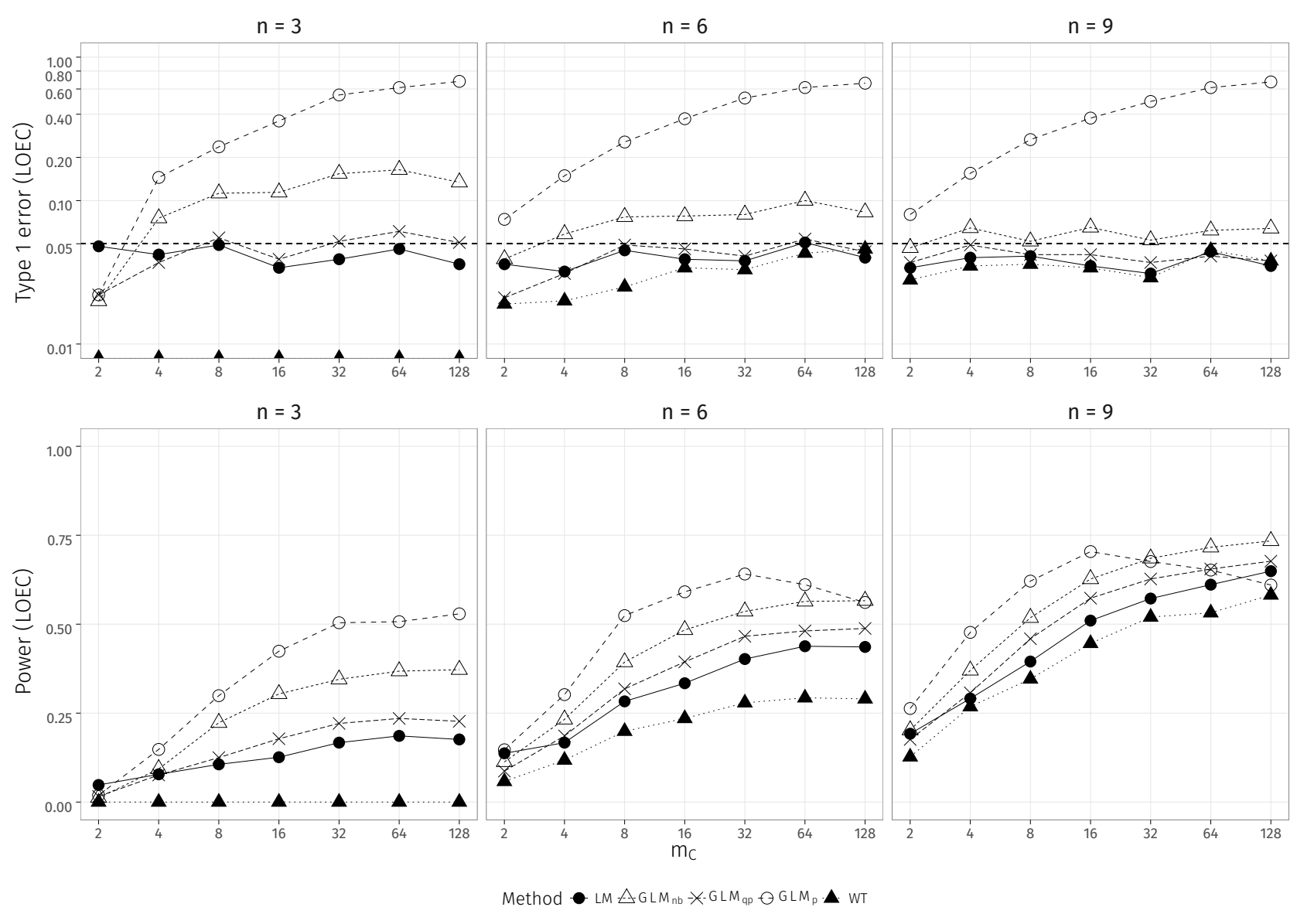
https://github.com/edild/phd_defense

https://github.com/edild/phd_thesis

Power en detail



For LOEC it is even worse



Comparison with Ives...

Szöcs (2015)

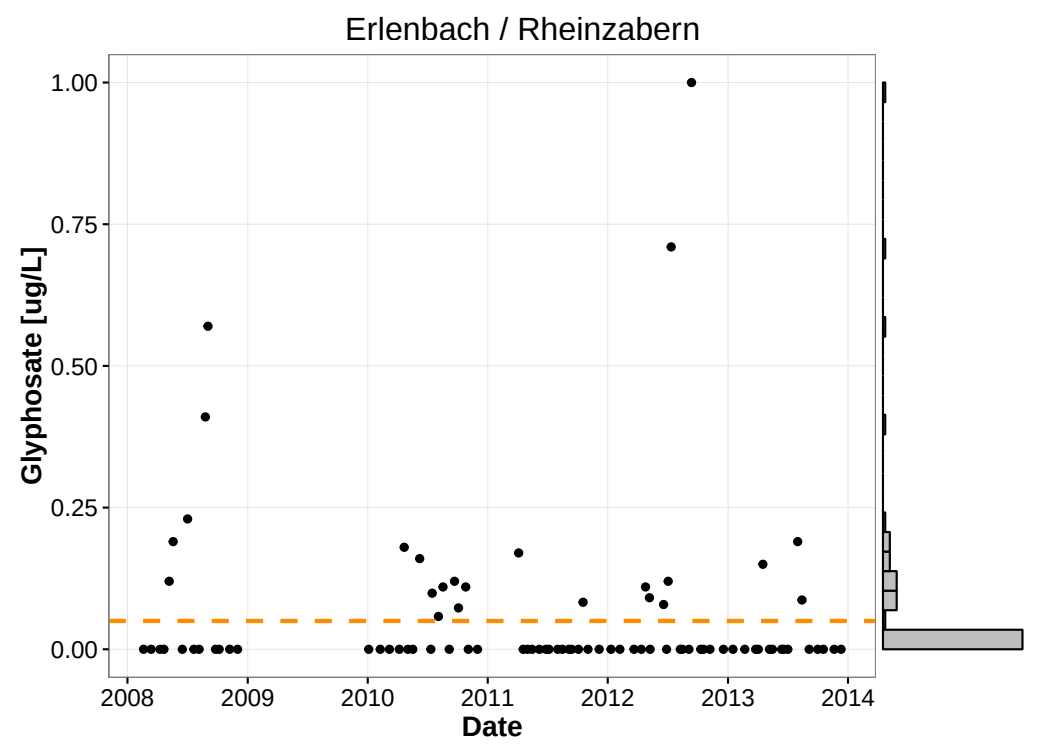
- factorial design
- one predictor
- low replicated
- LM, GLM, bootstrap
- High T1 error of NB
- Quasi-Poisson worked well
- Bootstrap fixes the problems

Ives (2015)

- continuous design
- two predictors
- well replicated
- LM, GLM
- High T1 error of NB
- Quasi-Poisson has problems with multiple predictors
-

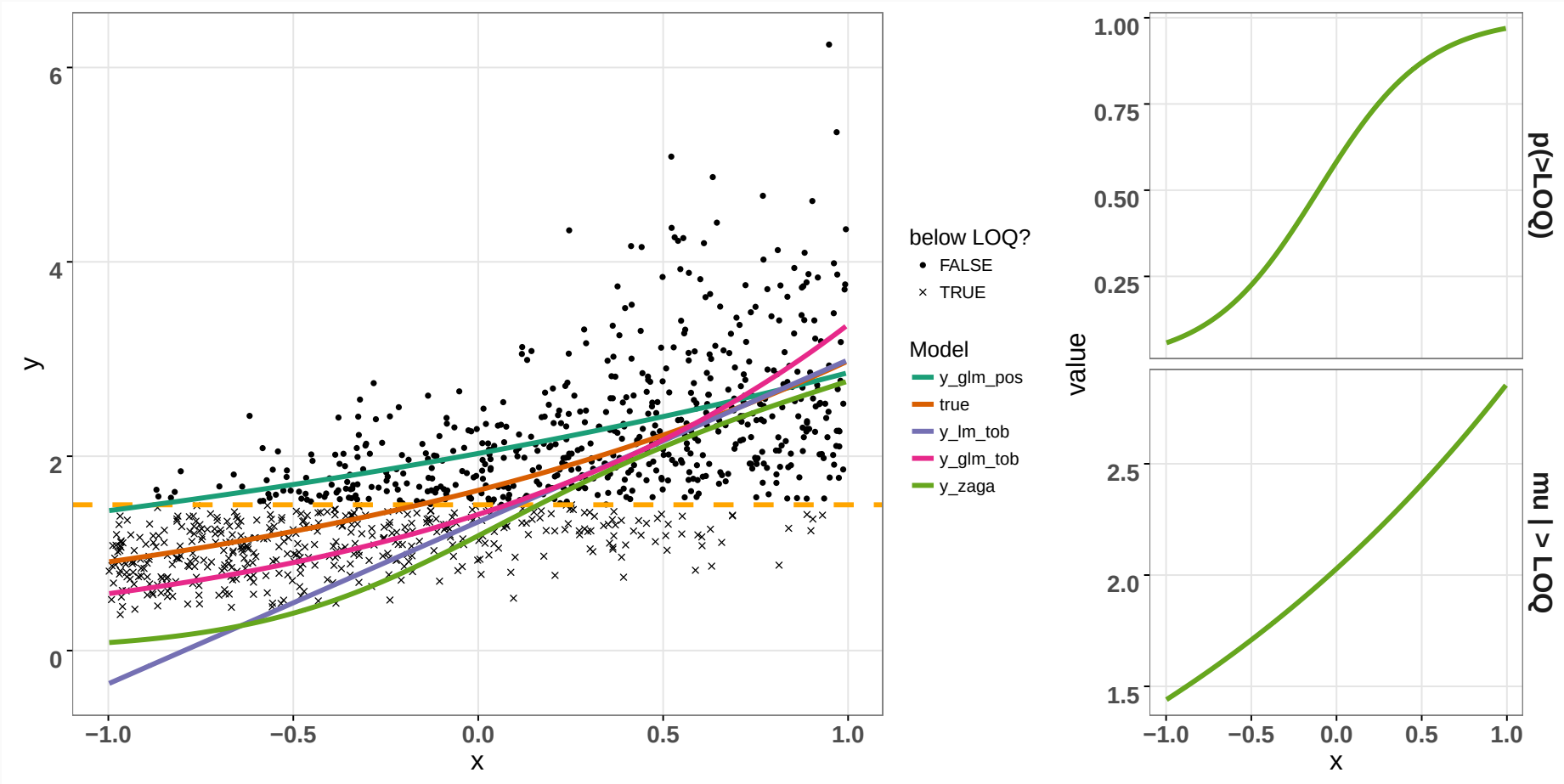
Idiosyncrasies of chemical concentrations

- continuous distribution in $\mathbb{R}_0^+$
- censoring (x <LOQ)
- non-linearity (season, trends)
- dependency (spatial, temporal)
- missing data

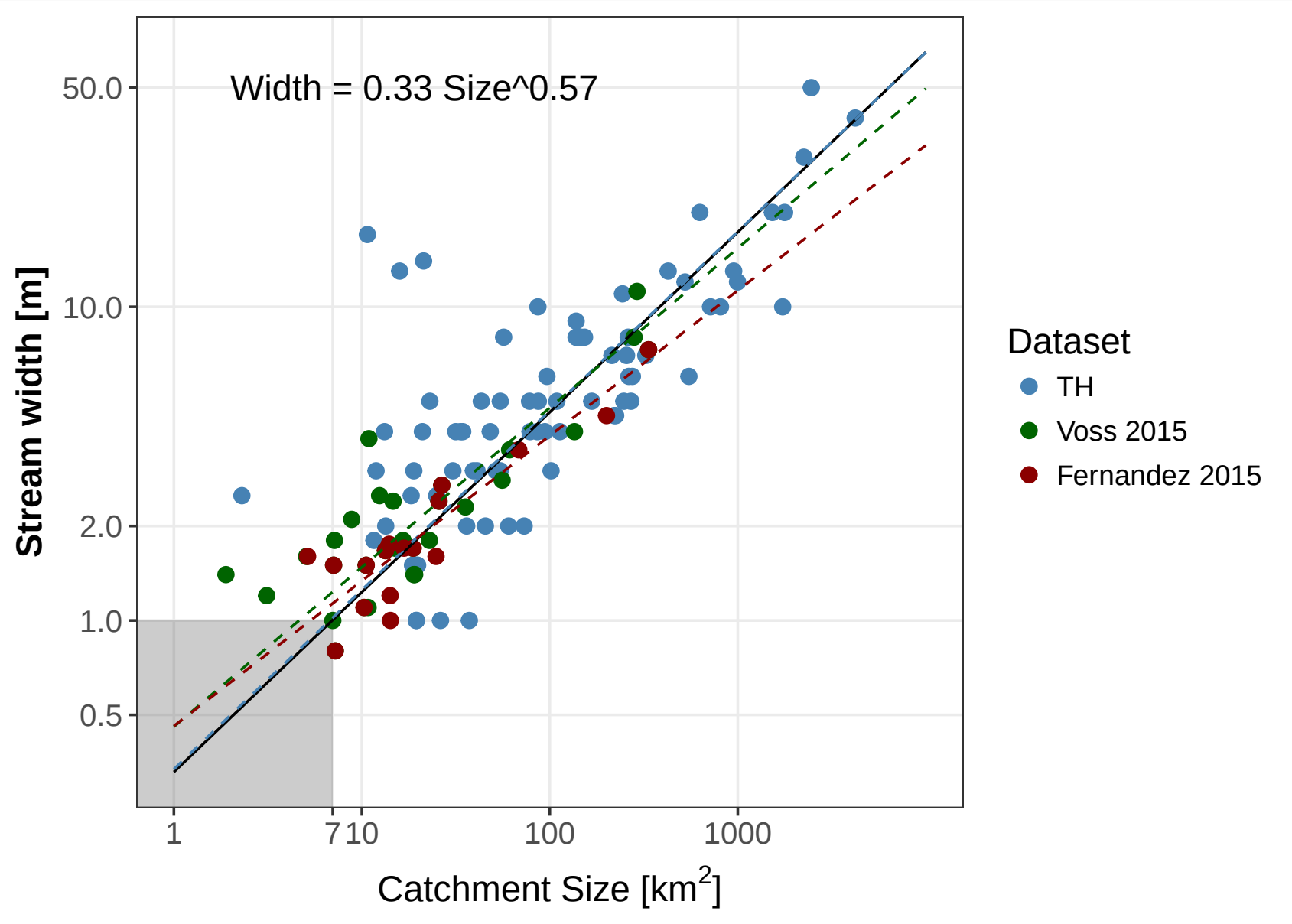


ZAGA what...?

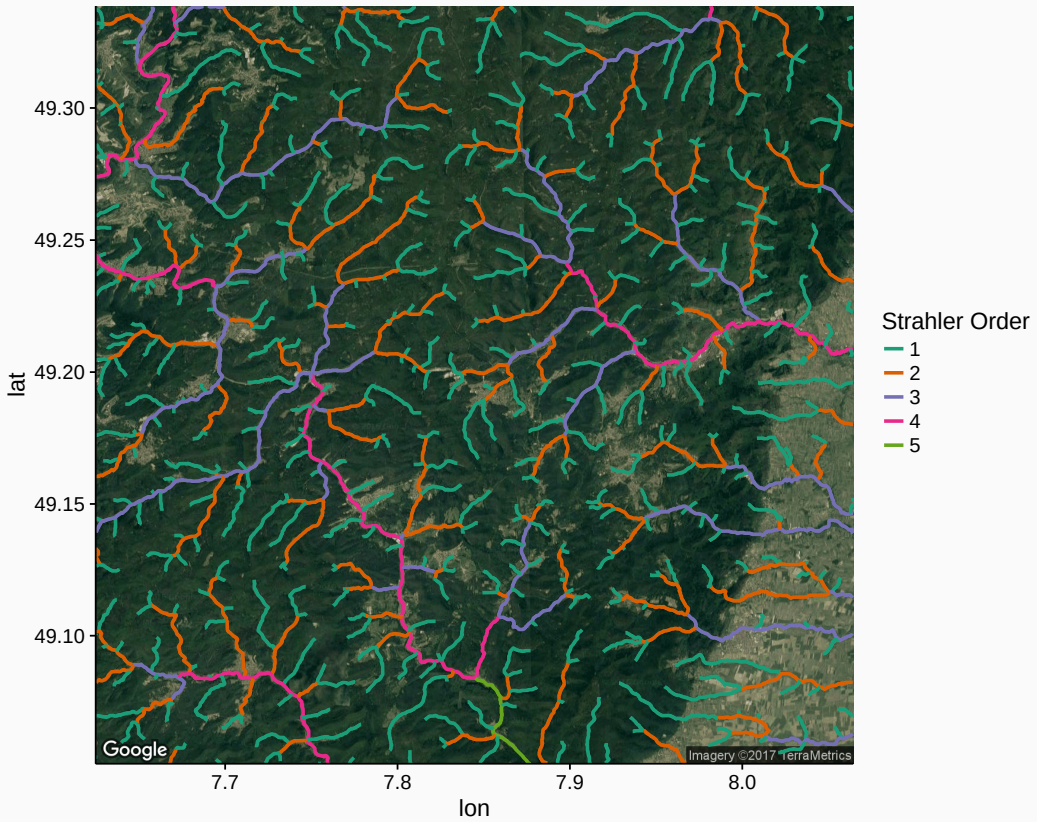
shiny app: `http://uni-ko-ld.de/g4`

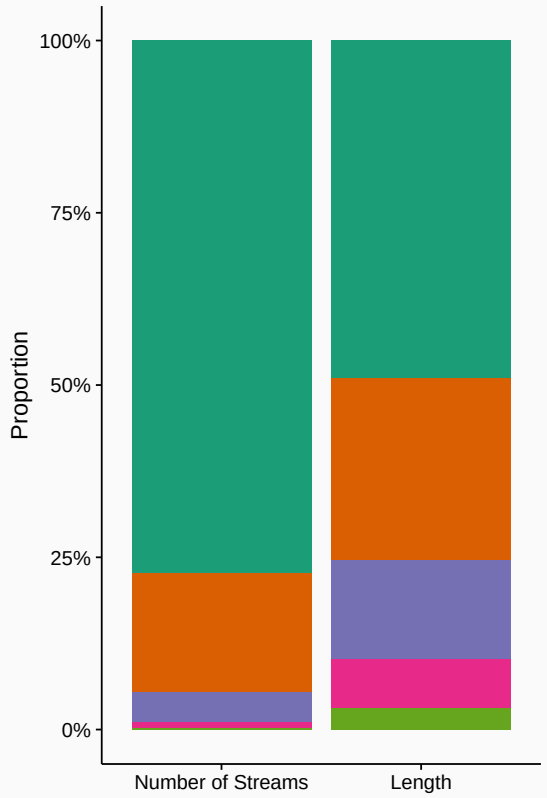


Stream size / width



Importance of small streams

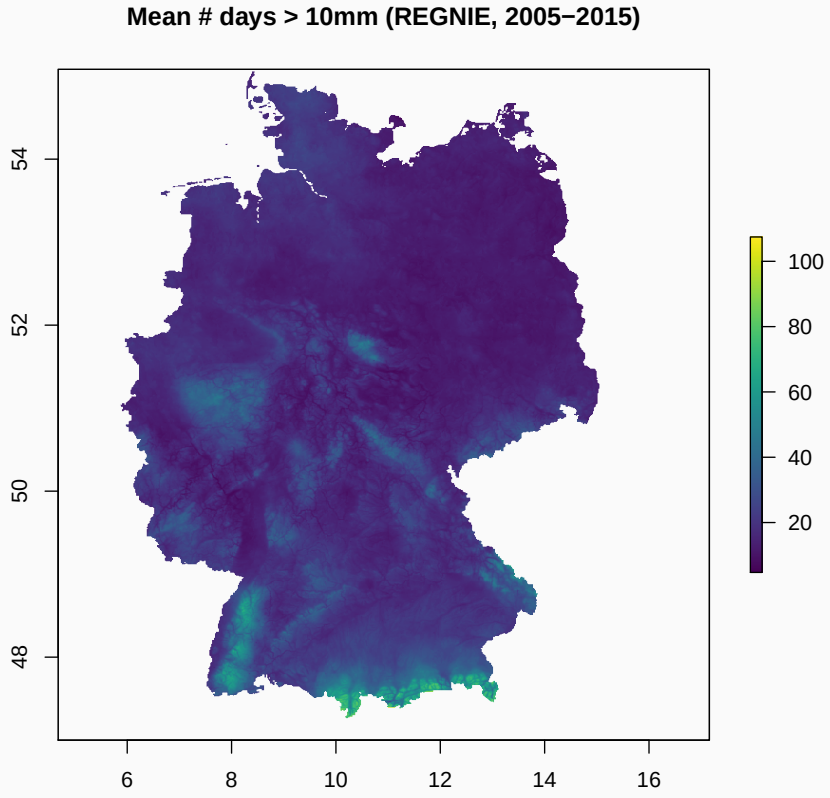


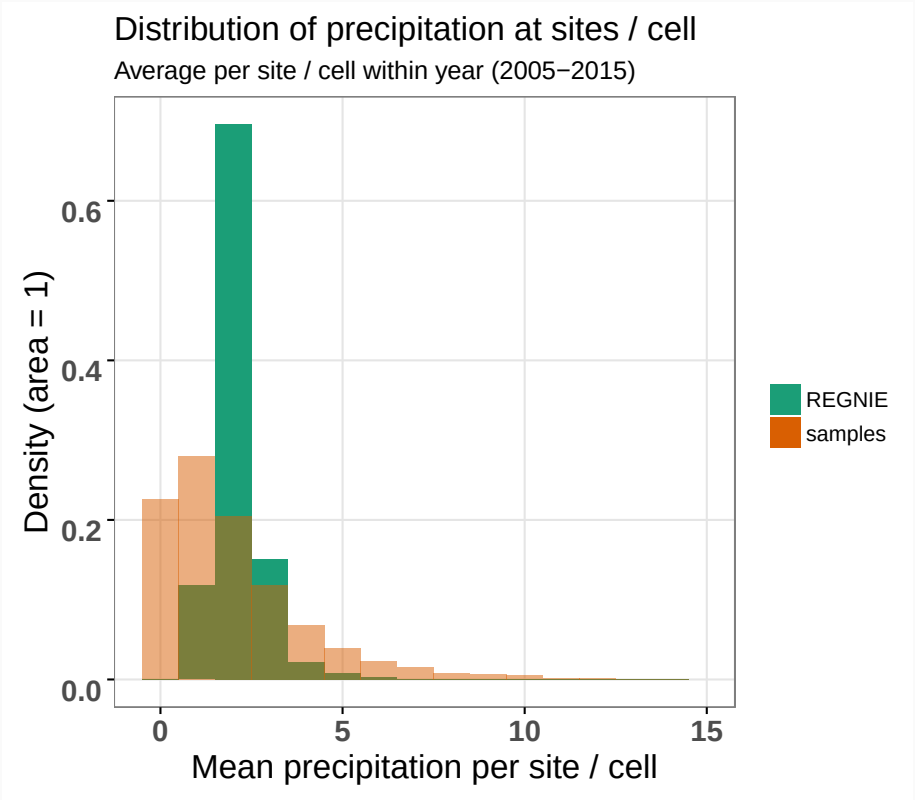


- Biodiversity
- Refuge for re-colonisation

Biggs et al. (2016). The importance of small waterbodies for biodiversity and ecosystem services: implications for policy makers. Hydrobiologia. DOI: 10.1007/s10750-016-3007-0.

Precipitation in Germany and the samples





Comparison with other studies

Szöcs (2016)

- Germany
- Monitoring
- Grab sampling
- Pesticides
- Neonics + Chlorpyrifos (OP)
- ZAGA (<LOQ)

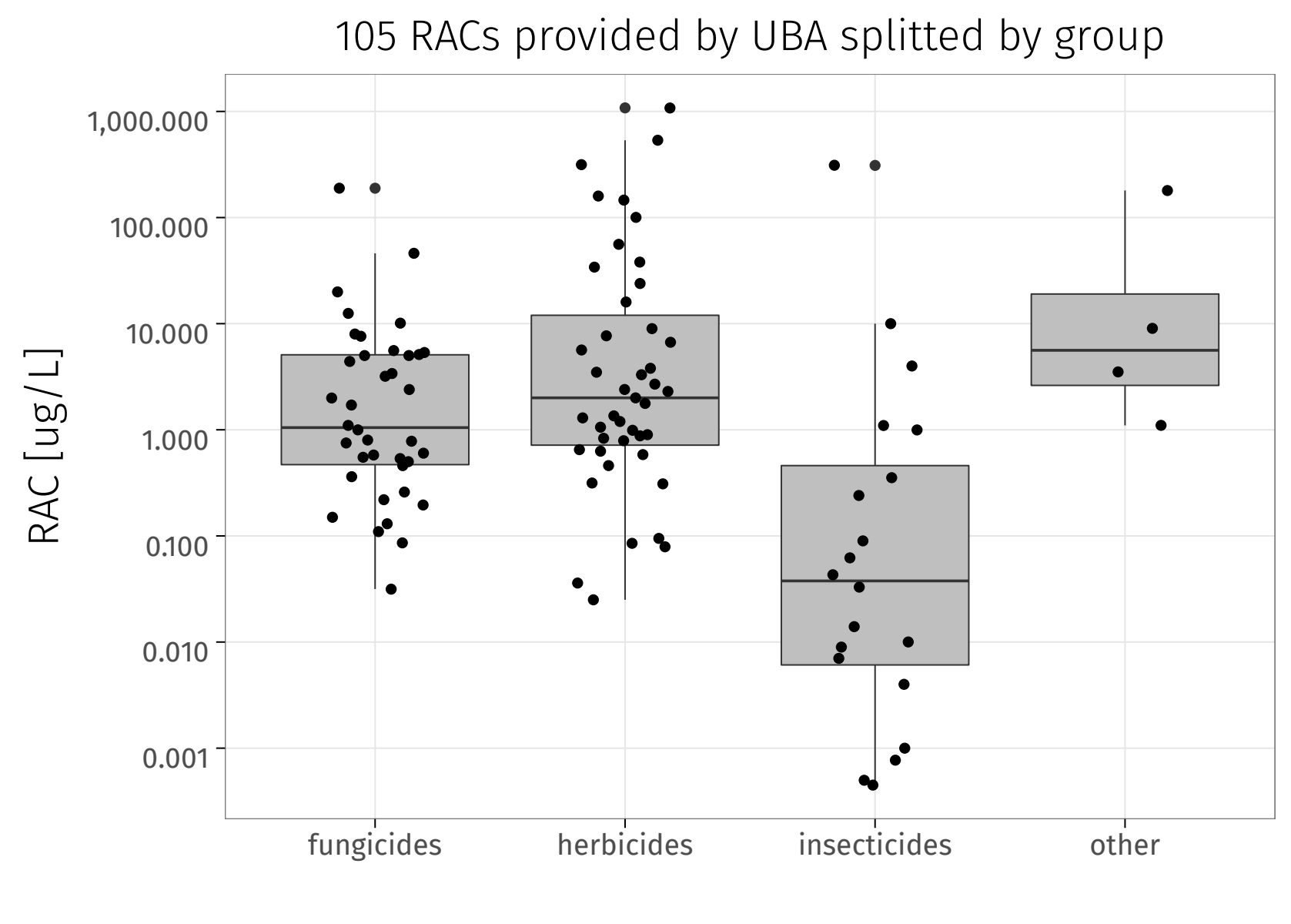
Stehle (2015)

- Europe / Global
- Publications
- Grab & Event driven sampling
- Insecticides
- Organophos.+ Pyrethroids
- LM for >LOQ

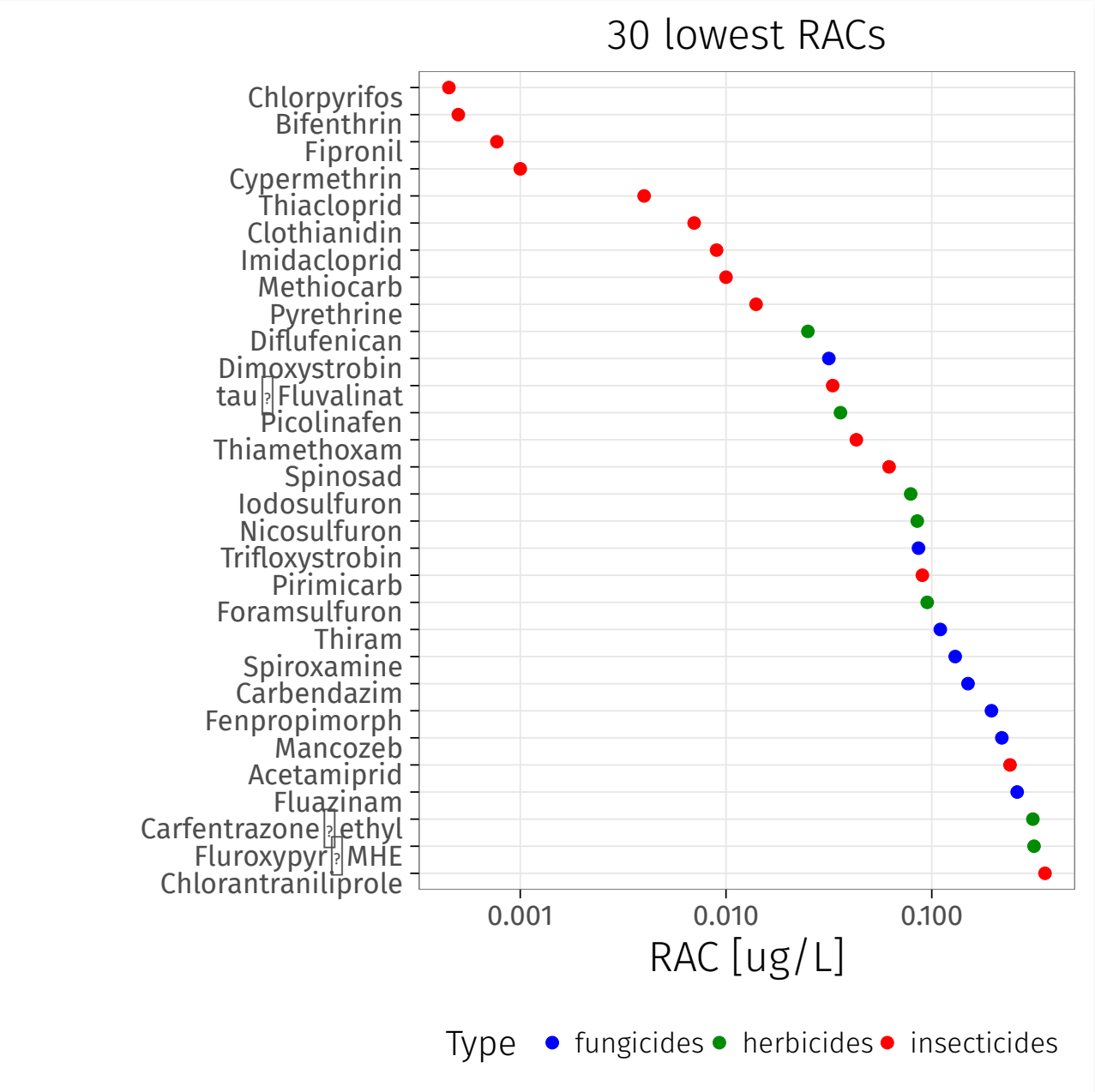
Knauer (2016)

- Switzerland
- Monitoring
- Grab sampling
- Pesticides
- Chlorpyrifos + Herb + Fung
- no model

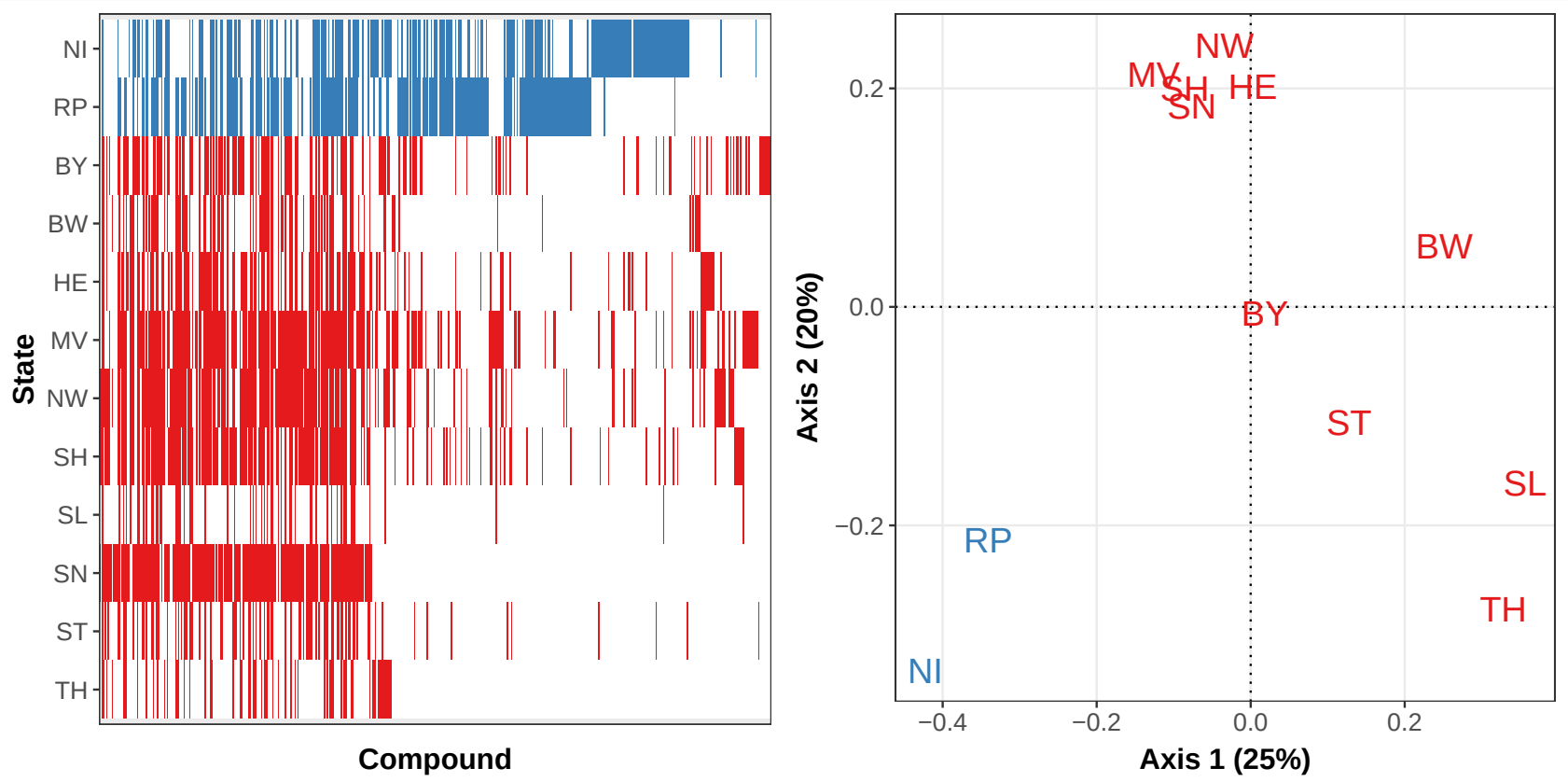
RACs by Type


105 RACs provided by UBA splitted by group
RAC [ug/L]
1,000.000
100.000
10.000
1.000
0.100
0.010
0.001
fungicides
herbicides
insecticides
other

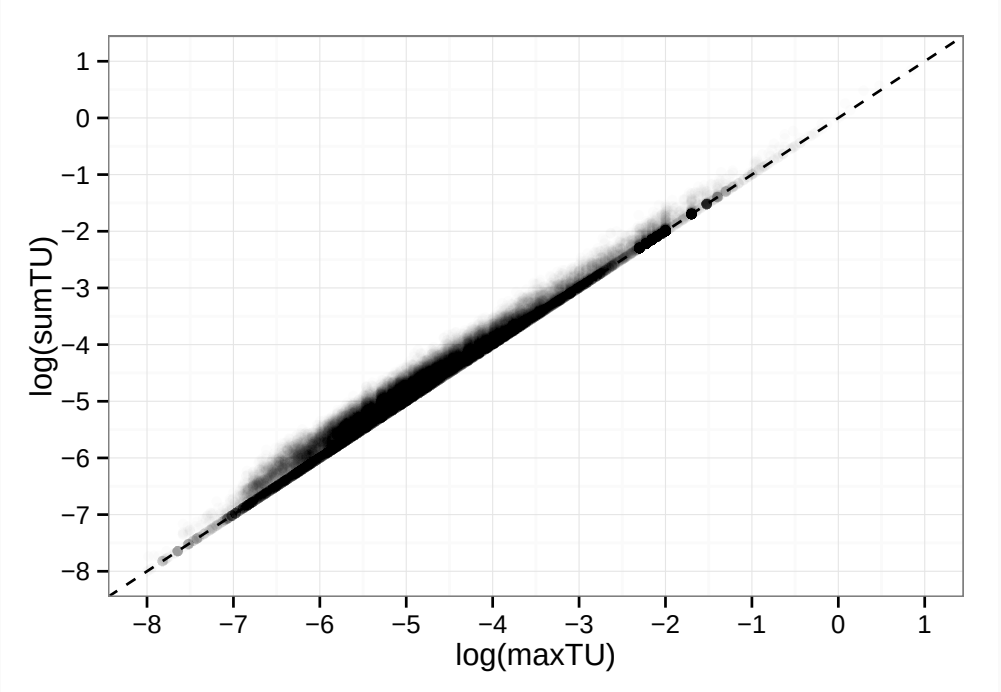
RACs by Compound



Analysed compound spectra by state



Mixtures are common, but one compound dominates the risk

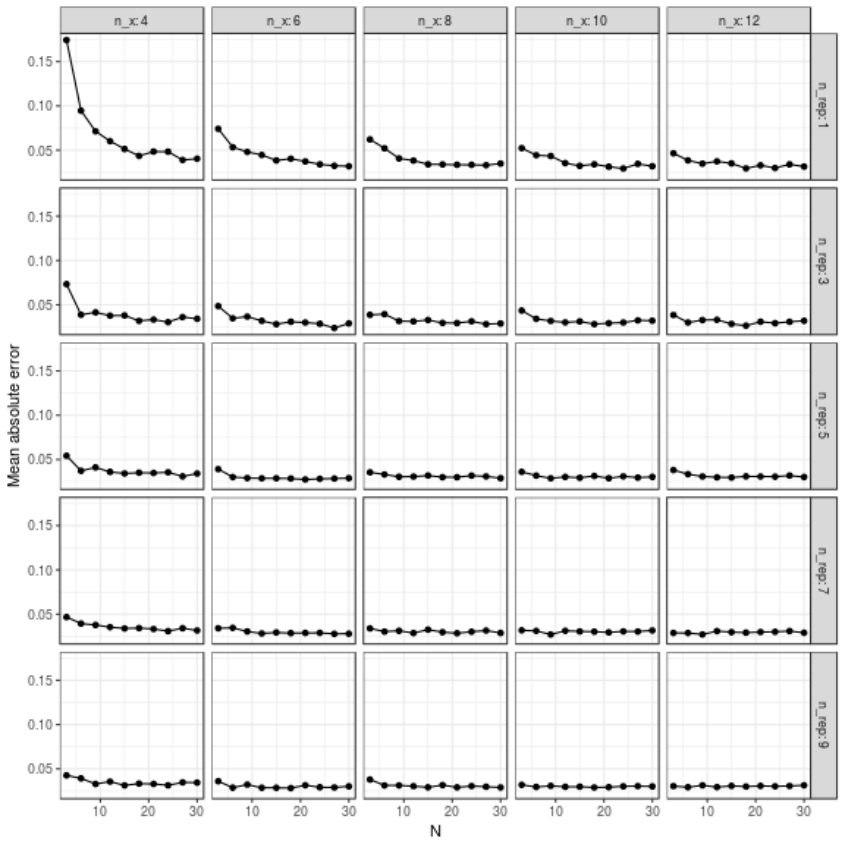


- up to 50 compounds in one sample
- high correlation
- $\sim$ 0.5 TU increase
- mainly one compound responsible for risk

Simulations are worth their work, use them *a priori*!

Experimental design for dose-response experiments - a simulation

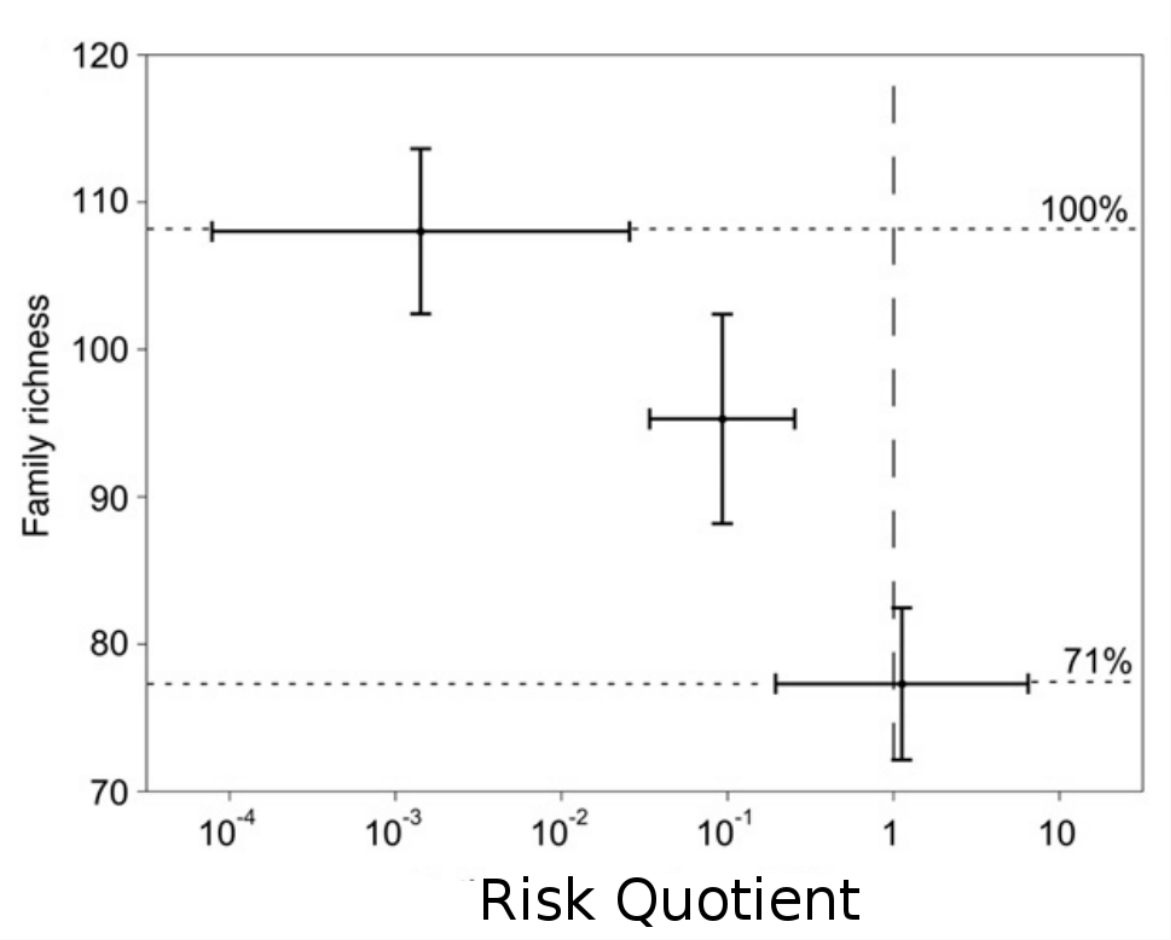
`http://edild.github.io/lc50_bias_sim/`



GLM-Explorer: `http://uni-ko-ld.de/g3`

Effects of RAC exceedances

- RACs should never be exceeded (=protection goal)
- If so, biological effects likely



Modified from: Stehle, Schulz (2015). Agricultural insecticides threaten surface waters at the global scale. PNAS 112, 5750–5755.

Reasons for observed RAC exceedances

- Risk Mitigation fails
  - Risk mitigation measures (erosion rills, wind)
  - Farmer do not adhere (GAP, no spray zones)
- Risk Assessment fails
  - Exposure Assessment
    - Models not working (Knäbel et al.)
  - Effect Assessment
    - Missed sensitive species
    - New document asks also for insects

Risk Mitigation Measures

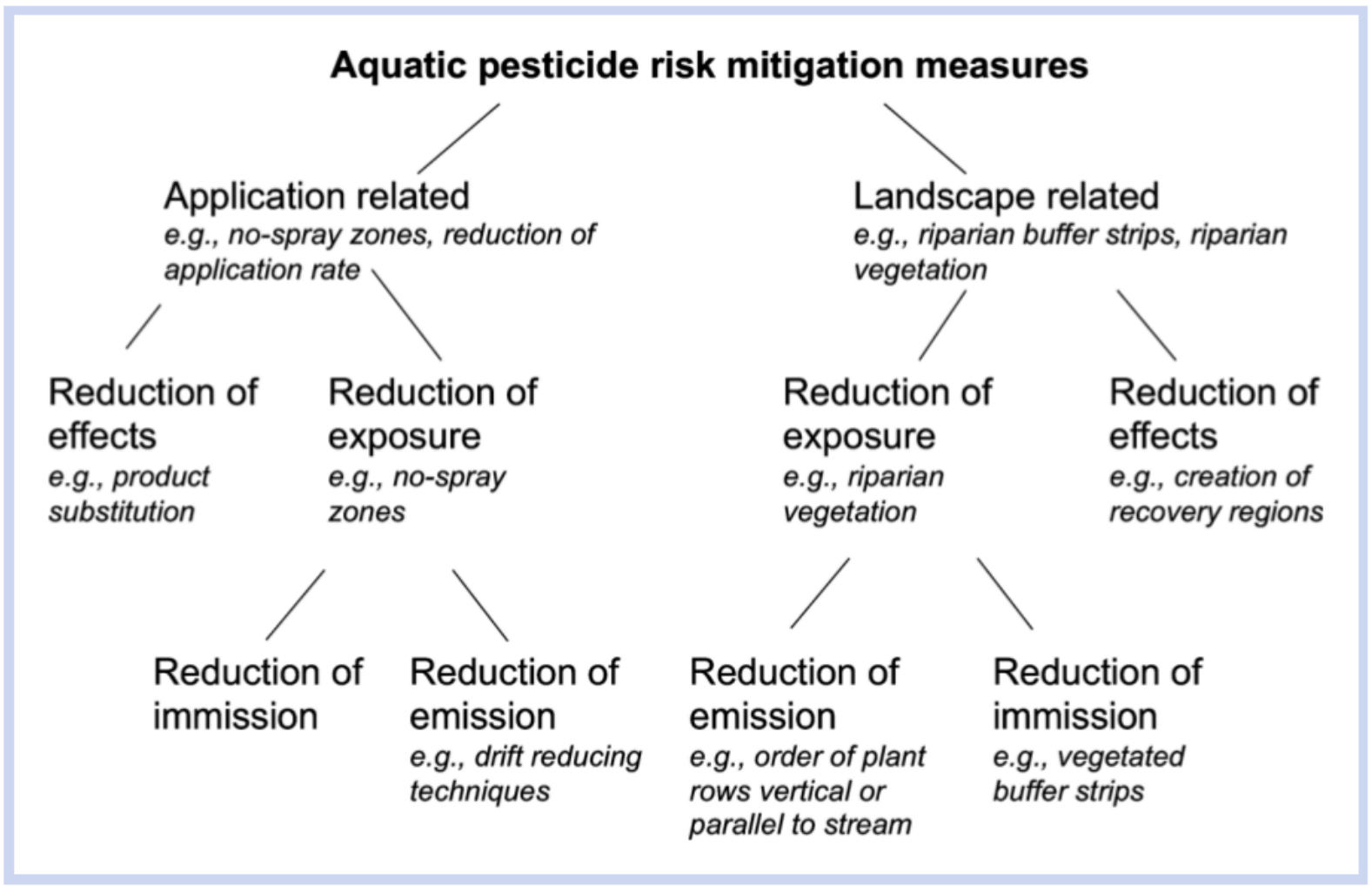
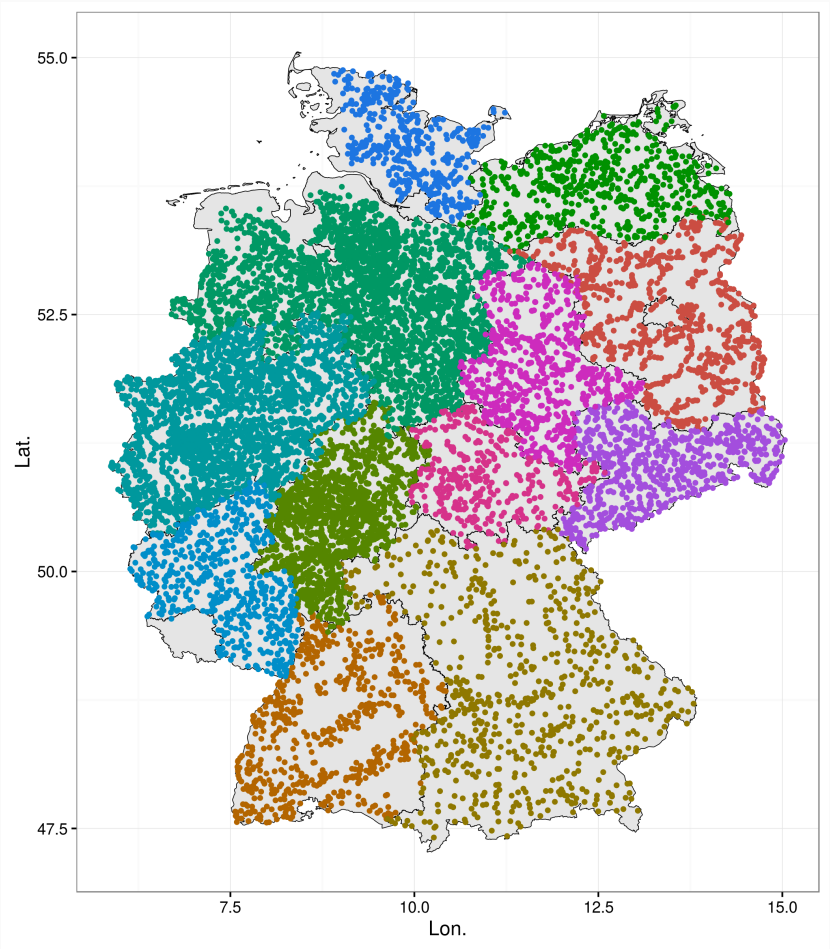


Figure 1. Classification of pesticide risk mitigation measures.

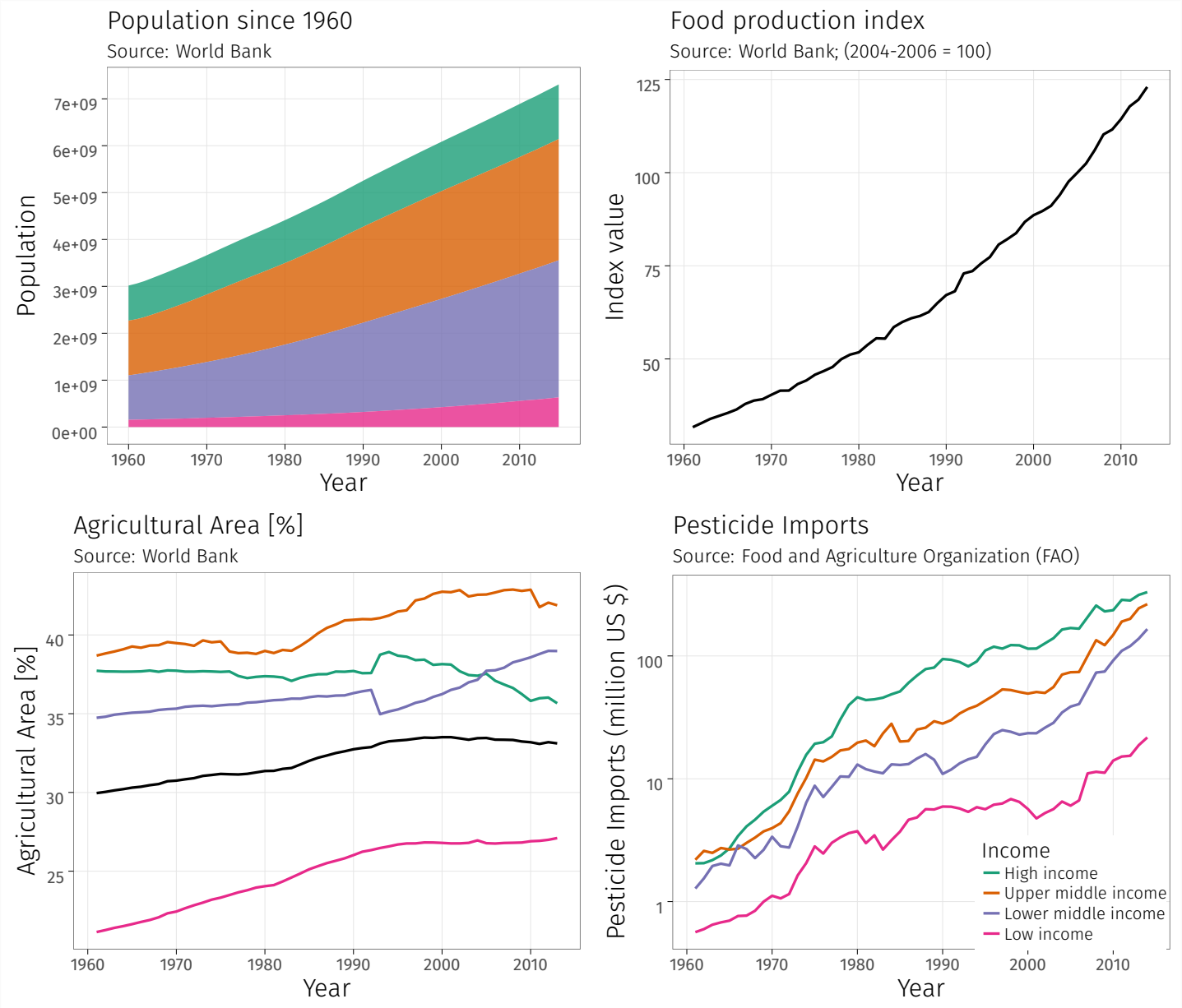
Bereswill, Streloke, Schulz (2014). Risk mitigation measures for diffuse pesticide entry into aquatic ecosystems: Proposal of a guide to identify appropriate measures on a catchment scale: Guide to Identify Pesticide Risk Mitigation Measures. IEAM 10, 286–298.

Biotic field effects



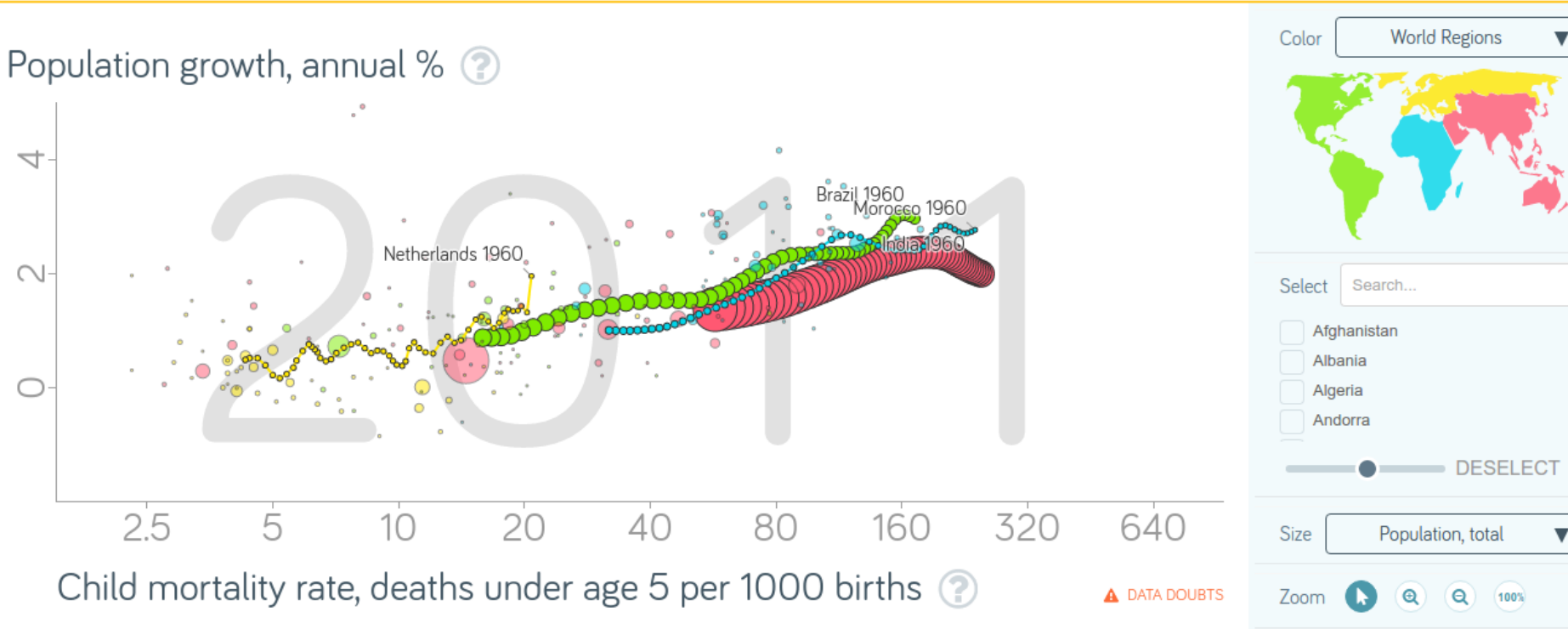
- biological data with good spatial coverage
- 60% of spatial congruence
- Large scale effects largely unknown.
- Some work left...
- Future....

A global perspective (I)



Source code to retrieve data and reproduce results: `http://uni-ko-ld.de/g7`

A global perspective (II)



Source: `https://www.gapminder.org/tools`

Software availability

Stable versions on CRAN, dev versions on github.

```
webchem github.com/ropensci/webchem
   taxize github.com/ropensci/taxize
```

Best practices for Software:

- open source (permissive MIT License)
- version control (git)
- automated tests (Travis-CI)
- in source documentation (roxygen)

Poisot, T., 2015. Best publishing practices to improve user confidence in scientific software. Ideas in Ecology and Evolution 8.

Many Thanks To

- My supervisor Prof. Dr. Ralf. B. Schäfer (for his support, openness, opportunities & discussions)
- My colleagues & collaborators (to many to list here)
- German Environment Agency (for funding & collab)
- My parents Anca & Helmut (for their support)
- My girlfriend Anja (for everything)