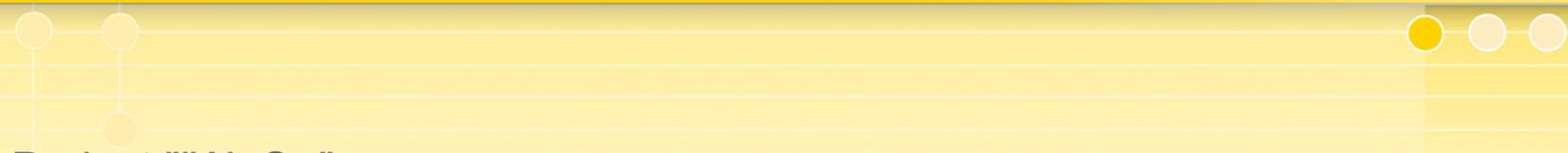


Prediction, valuation and ecological effects of future stream water quality based on socio-economic scenarios and climate change predictions for 2050.

Nele Schuwirth, Mark Honti, Anne Dietzel, Peter Reichert, Christian Stamm

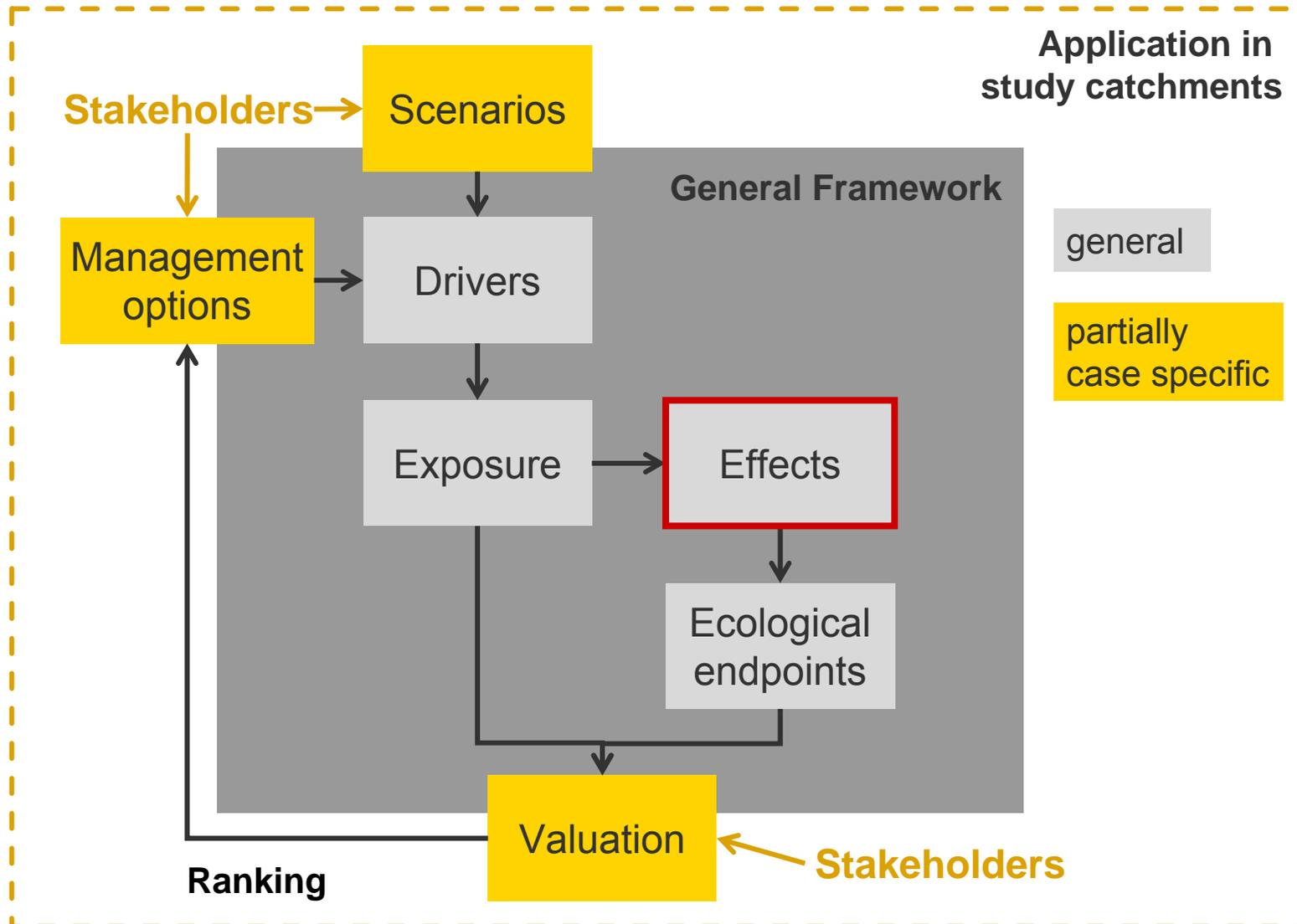


Project "iWaQa":

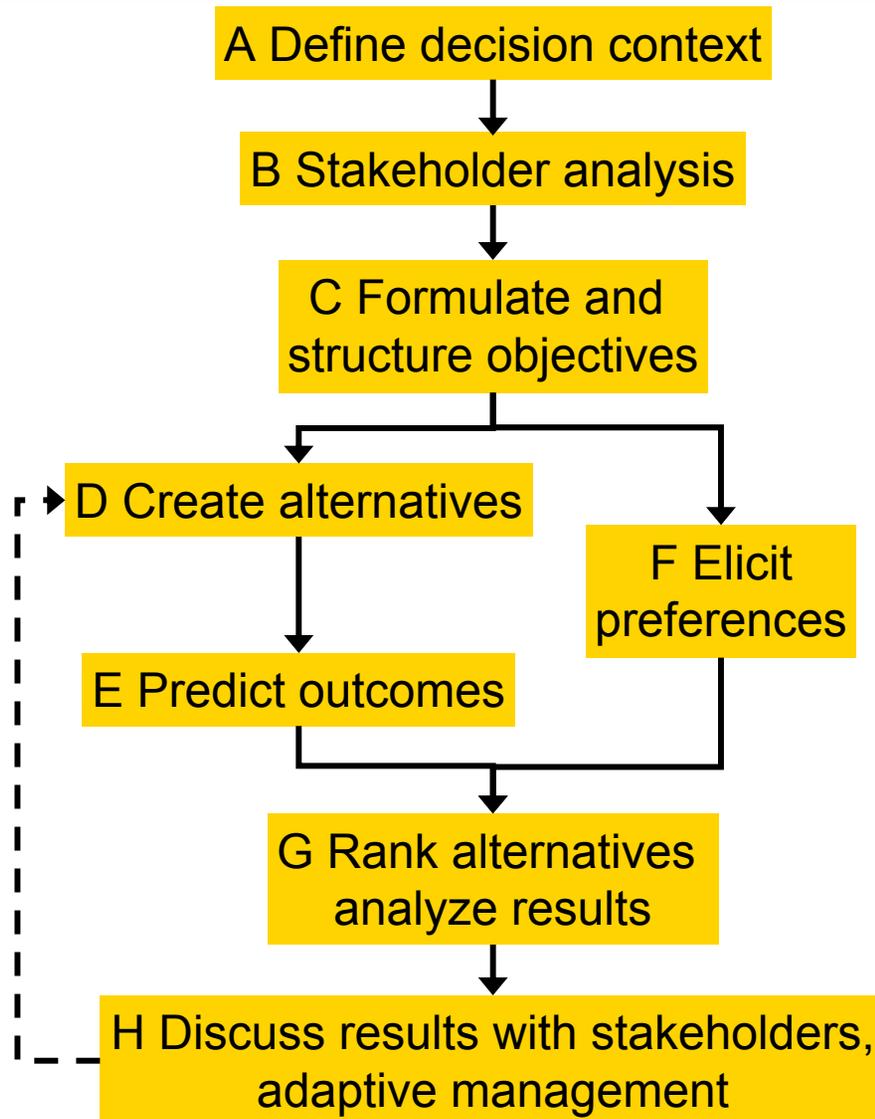
Integrated management of river water quality within the National Research Program 61 on Sustainable Water Management

Conceptual framework - project iWaQa

"Integrated management of river water quality"



Decision Analytic Framework

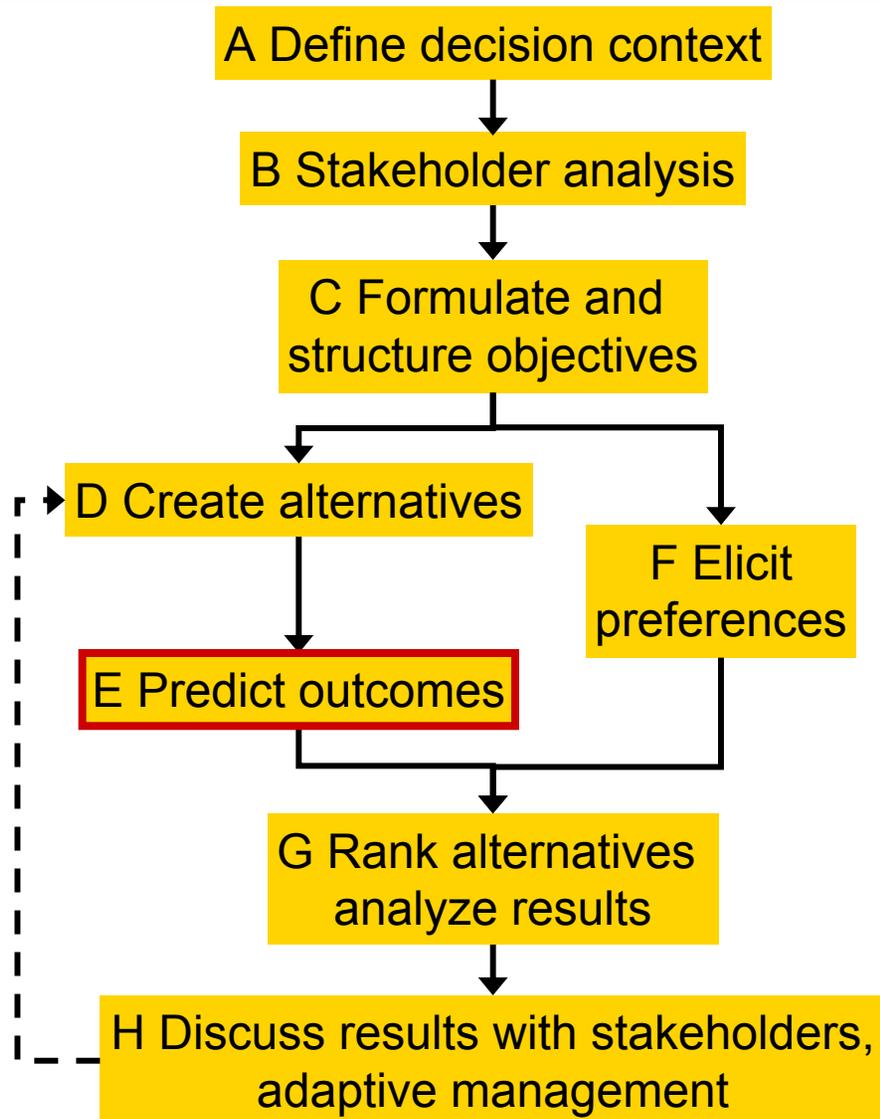


Aims:

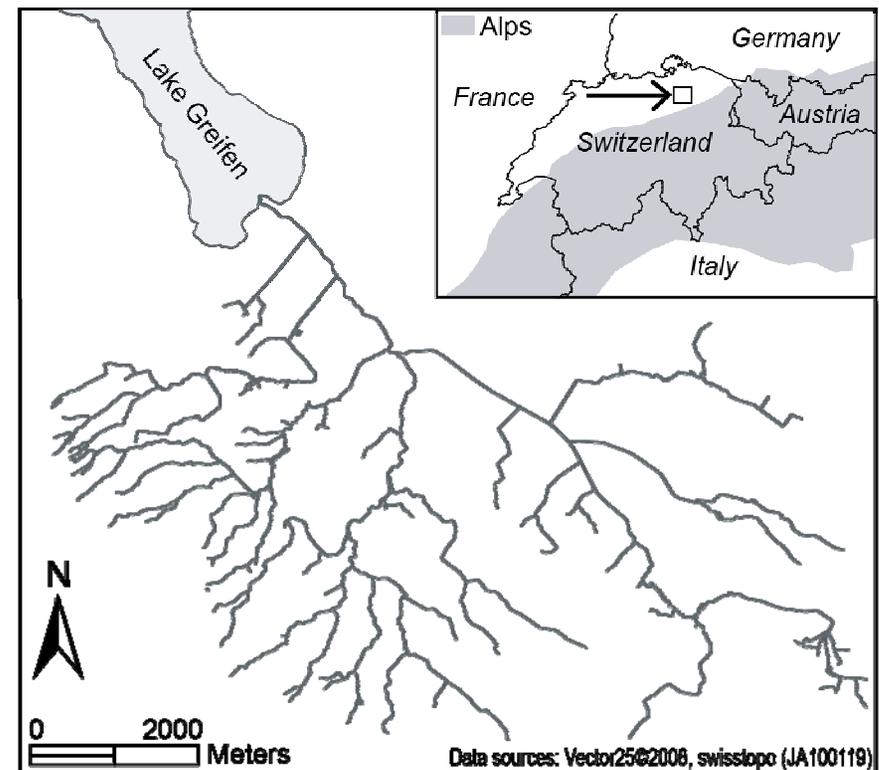
- Learning about management alternatives
 - Learning about stakeholder opinions
 - Learning about influence of uncertain future
 - Identification of potential consensus-solutions
- separating prediction of outcomes (experts) from evaluation (subjective opinion)

Schuwirth et al. 2012 EJOR: Methodological aspects of multi-criteria decision analysis for policy support: A case study on pharmaceutical removal from hospital wastewater

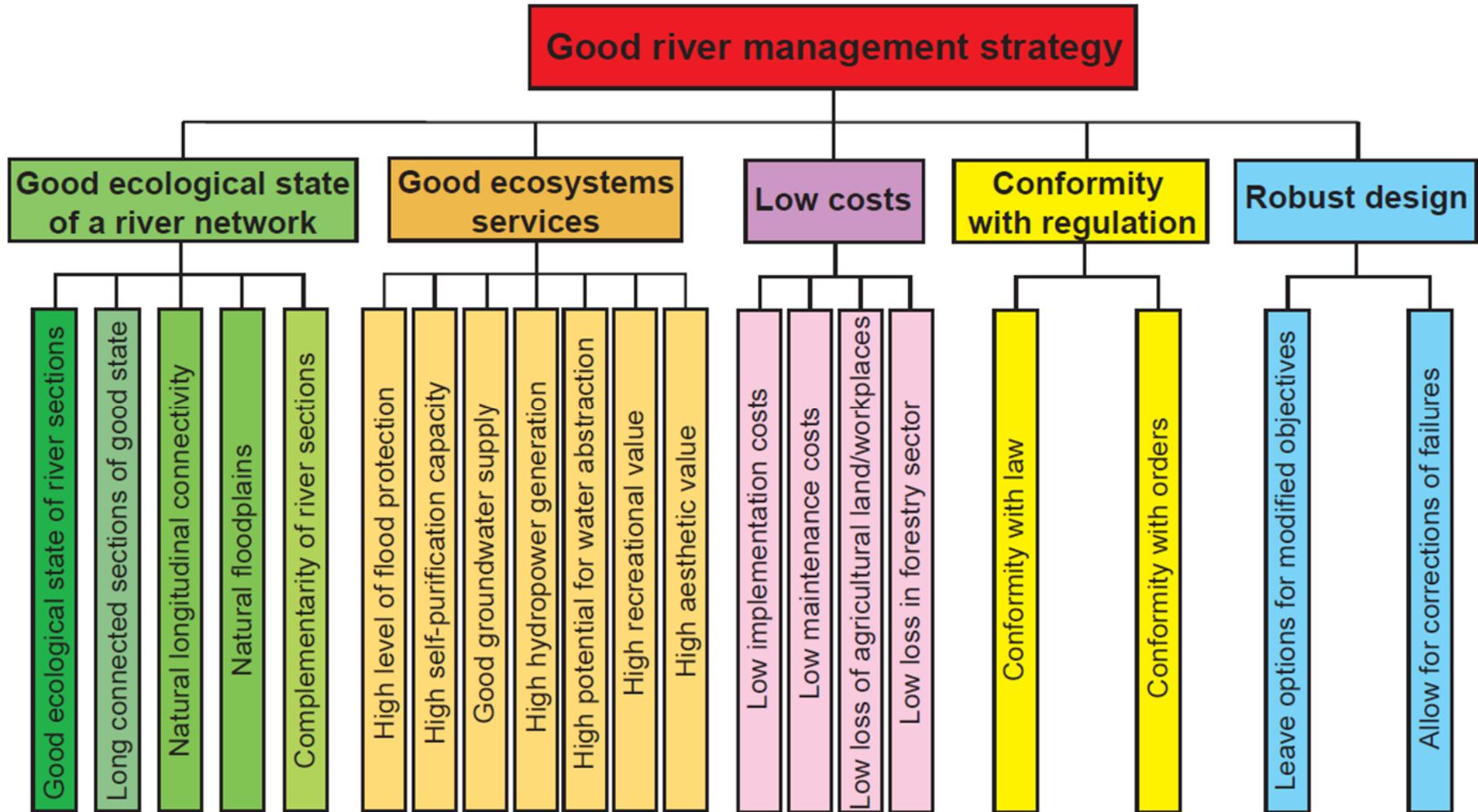
Decision Analytic Framework



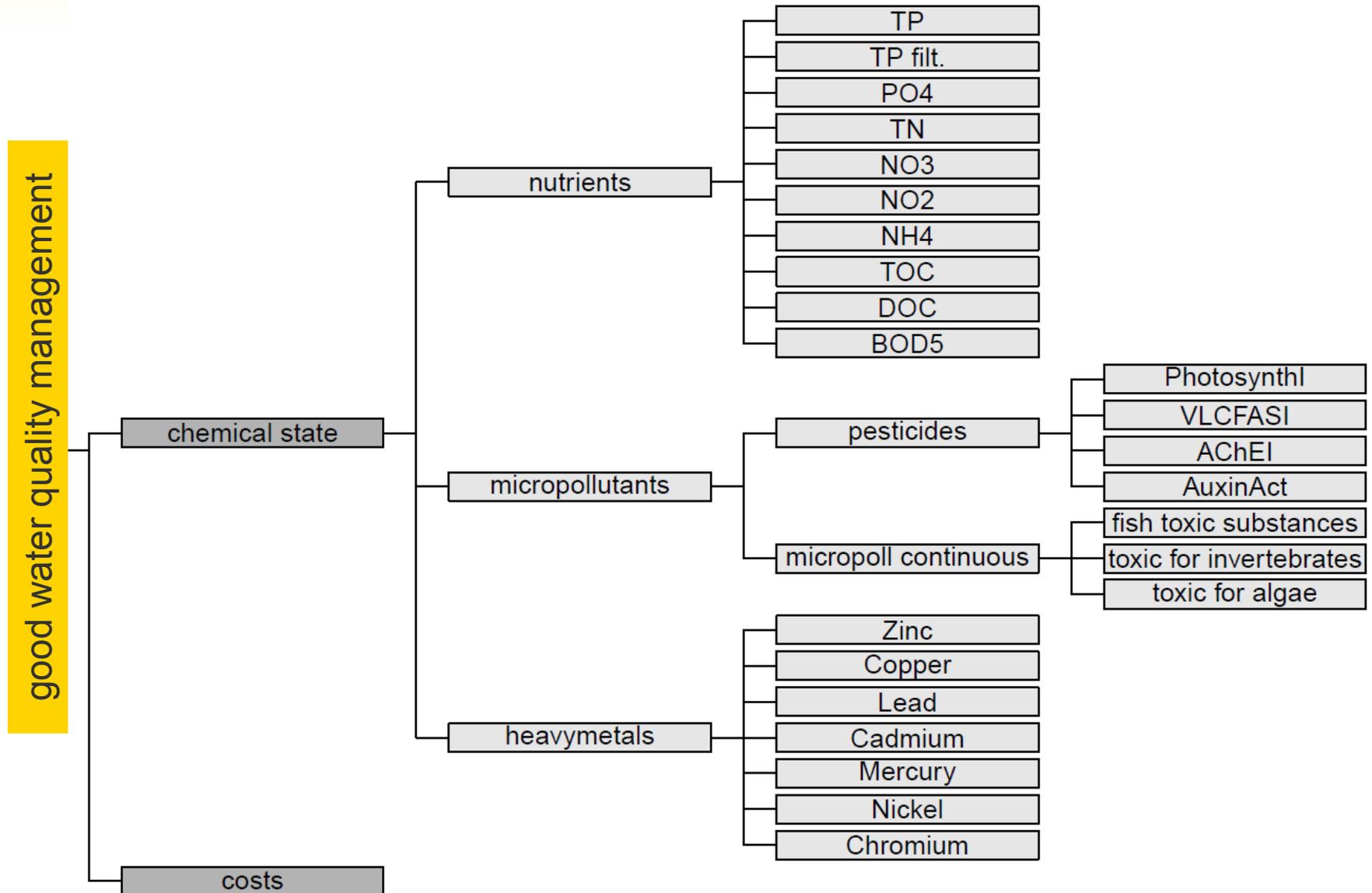
How to improve river water quality in a catchment of the Swiss Plateau?



C Objectives River Management

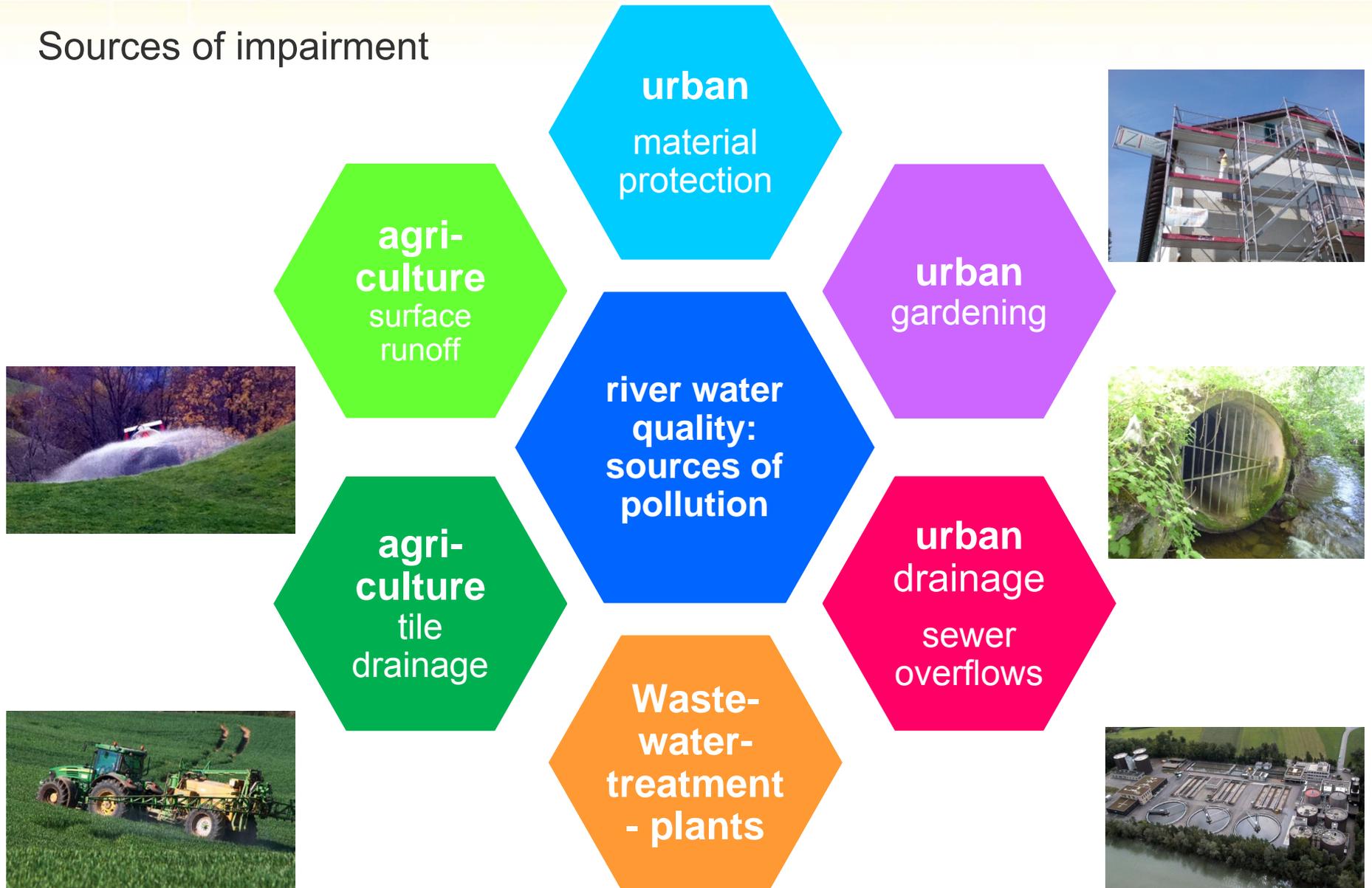


C Objectives Water quality management



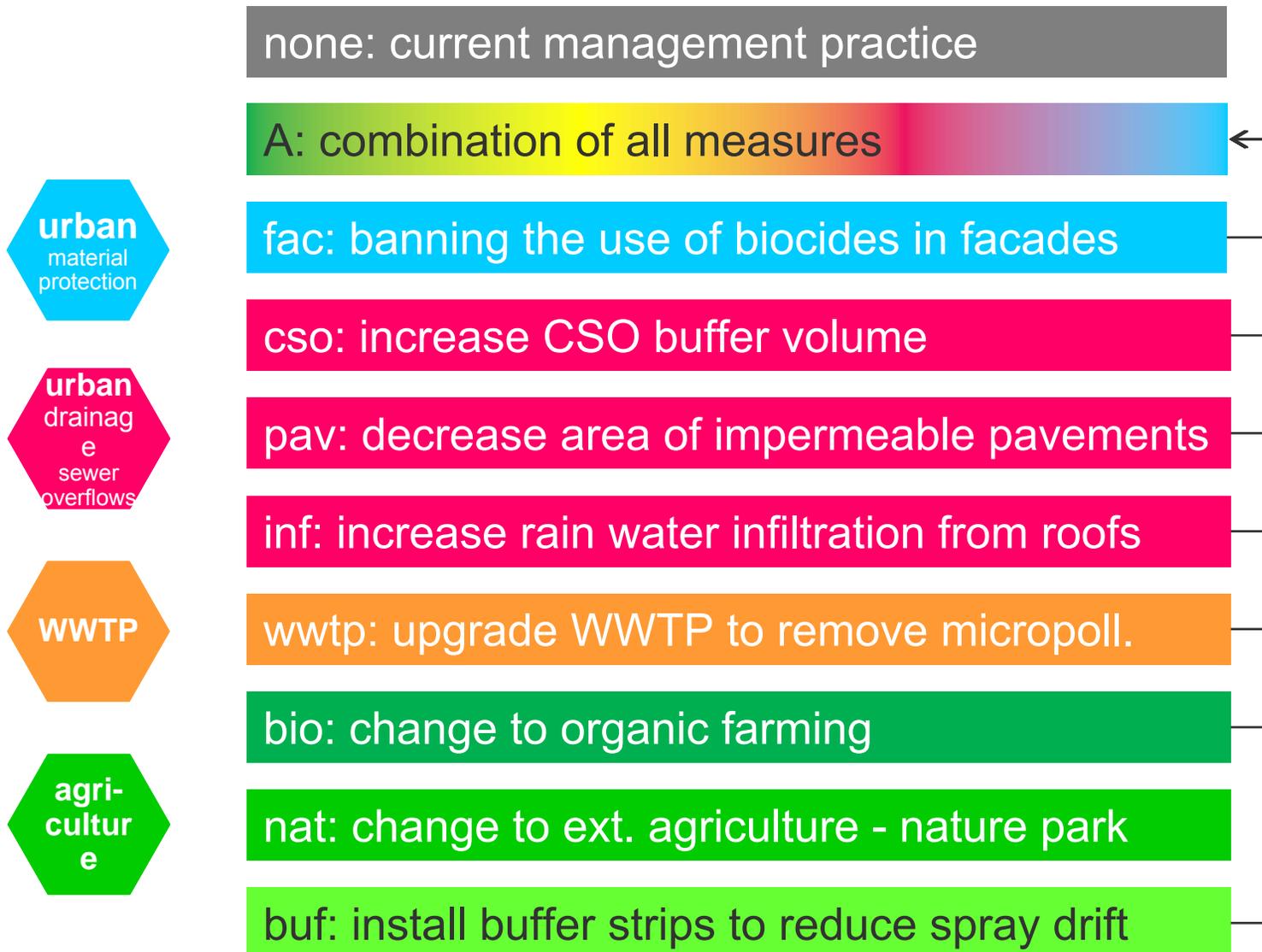
D Create alternatives

Sources of impairment



D Create alternatives

Management alternatives to improve water quality



E Socio-economic scenarios horizon 2050

Boundary conditions that cannot be influenced by WQ management alternatives

	mean income	population	urban area
Status quo	+0.4 %/year	as today	as today
Boom	+4	+730 %	+300 %
Doom	-1.5	-20 %	-
Quality of life	+2	+20 %	+5 %

Boom



Doom



Quality of life



Developed in a stakeholder workshop, Lienert et al. (submitted)

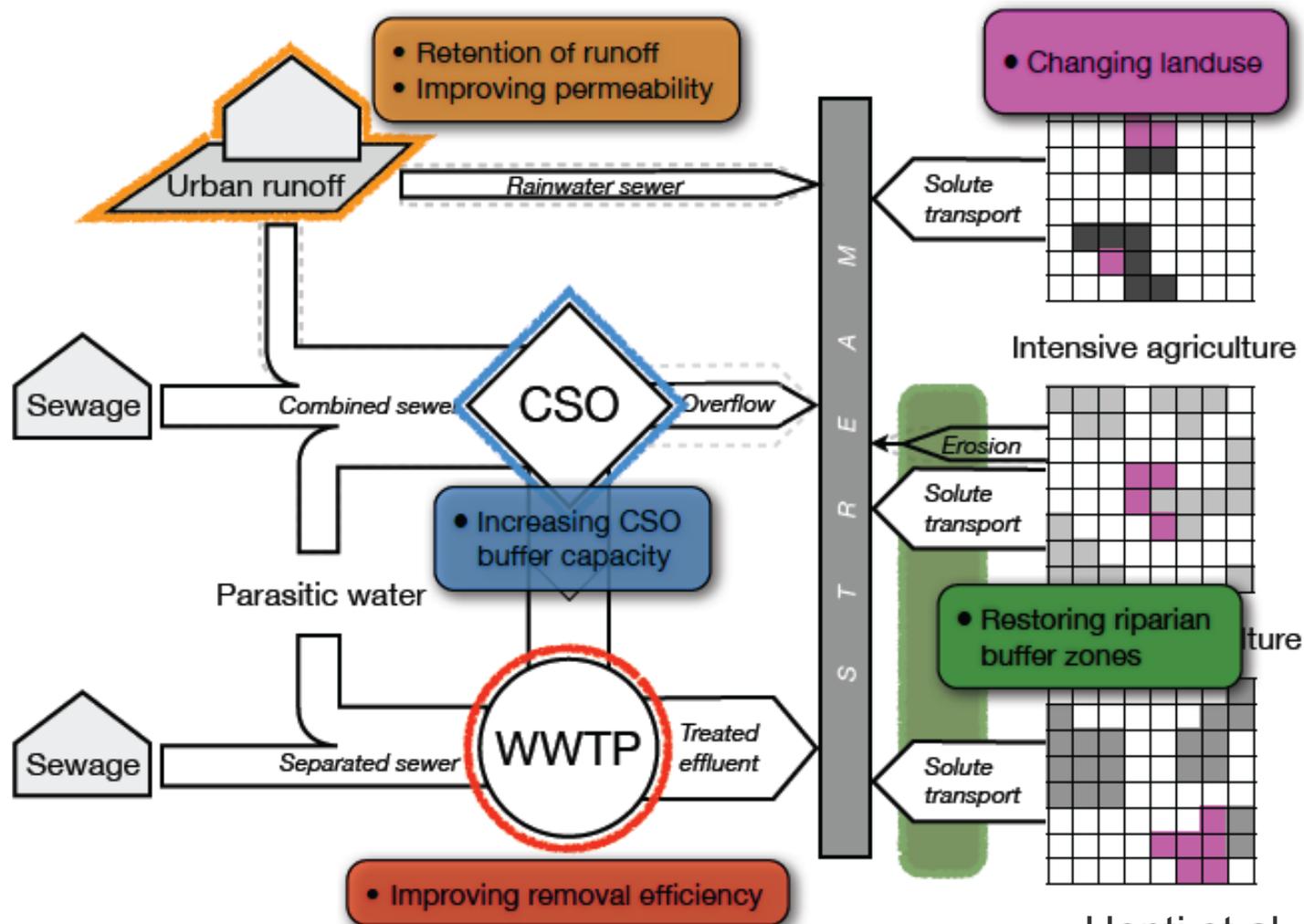
E Climate change - horizon 2050

- IPCC A1B emission scenario (scenarios hardly differ until 2050)
- 10 ENSEMBLES GCM-RCM model chains, daily data for precipitation & air temperature
- Neyman-Scott Rectangular Pulses (NSRP) model to generate hourly precipitation
- UKCP09 weather generator (which is based on the NSRP model) for daily values of other weather parameters

Honti, M., Scheidegger, A., Stamm, C. (2014): Importance of hydrological uncertainty assessment methods in climate change impact studies. Hydrol. Earth Syst. Sci. Discuss., 11, 501-553

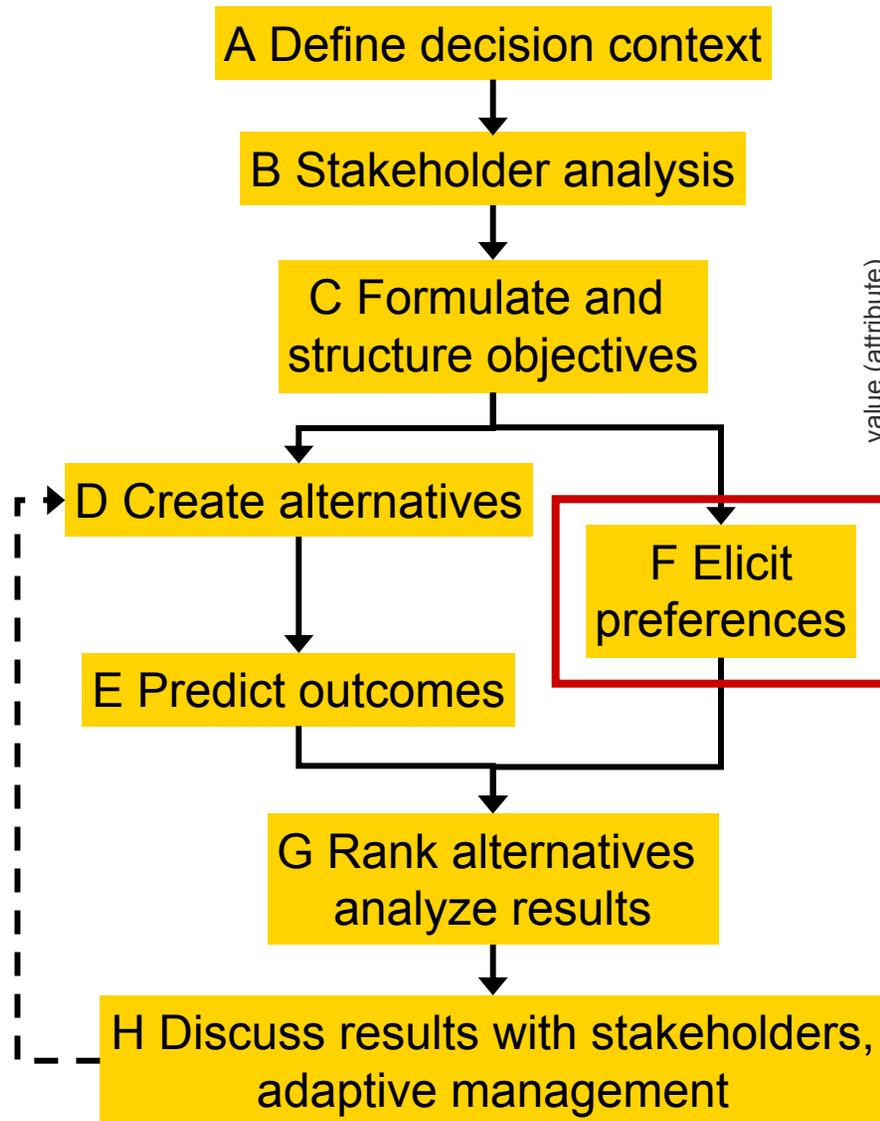
E Water quality model

Pollutant transport management



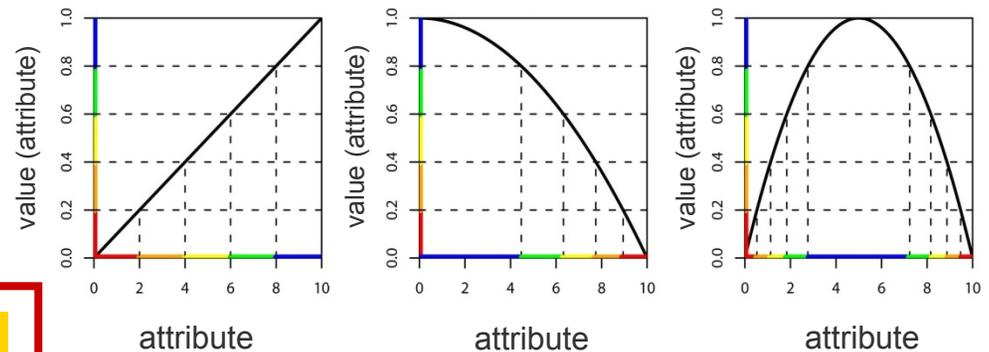
Honti et al. (in prep.)

F Preferences



Multi-attribute value theory

1. value functions for lowest level sub objectives

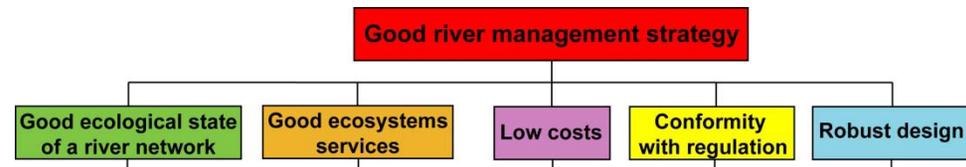


2. hierarchical aggregation:
trade-offs between objectives

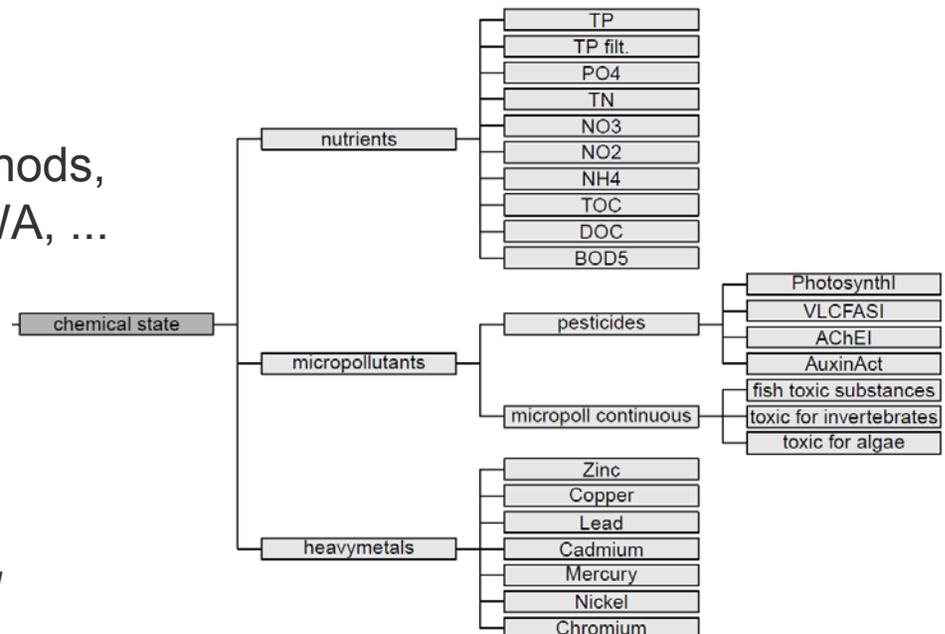
Schuwirth et al. 2012 EJOR: Methodological aspects of multi-criteria decision analysis for policy support: A case study on pharmaceutical removal from hospital wastewater

F Elicitation of Preferences

- Higher levels: societal values elicited from stakeholders case specific

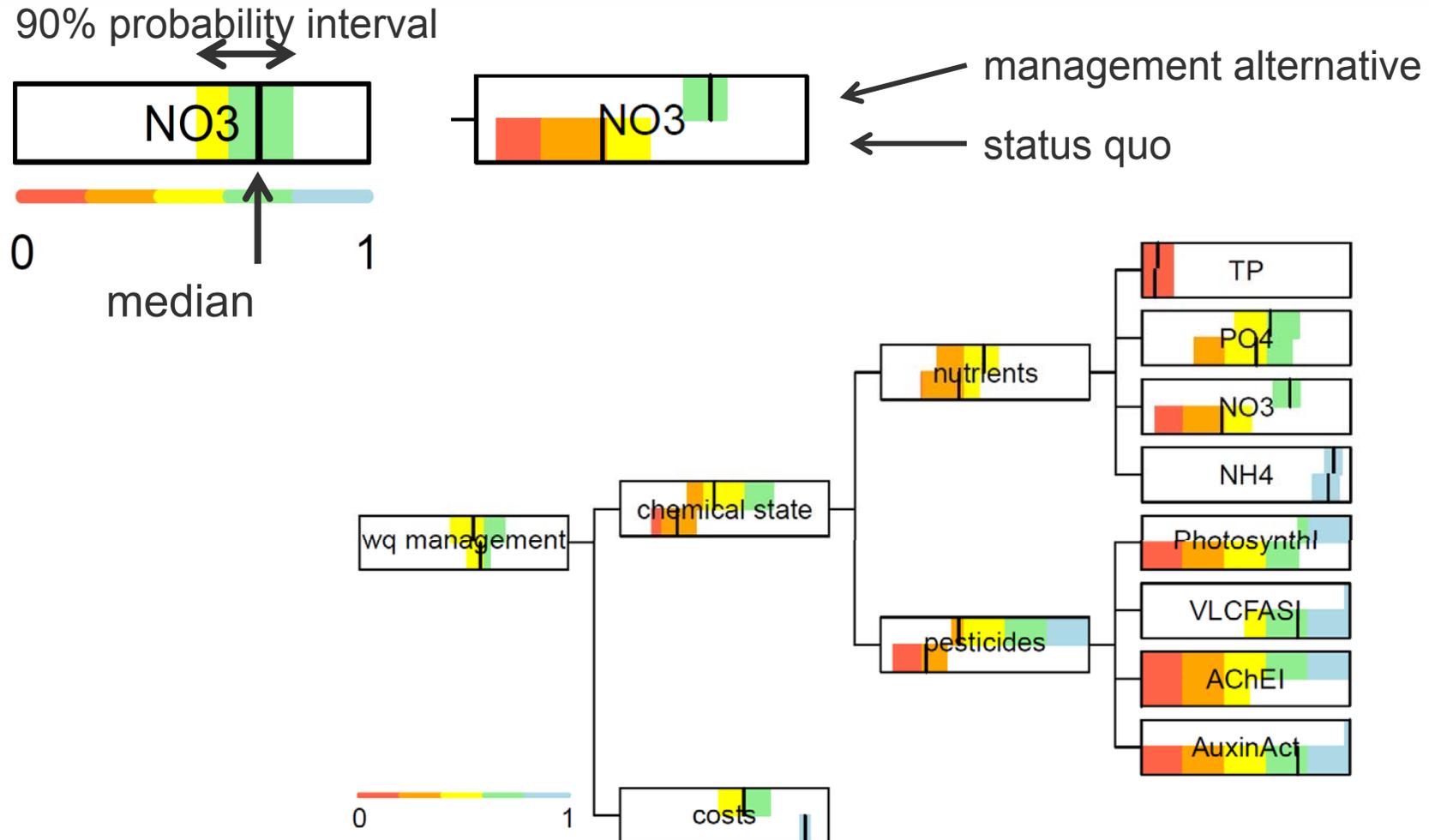


- Lower levels of some branches: require expert knowledge should be generic translation of existing assessment methods, e.g. www.modulstufenkonzept.ch, LAWA, ...



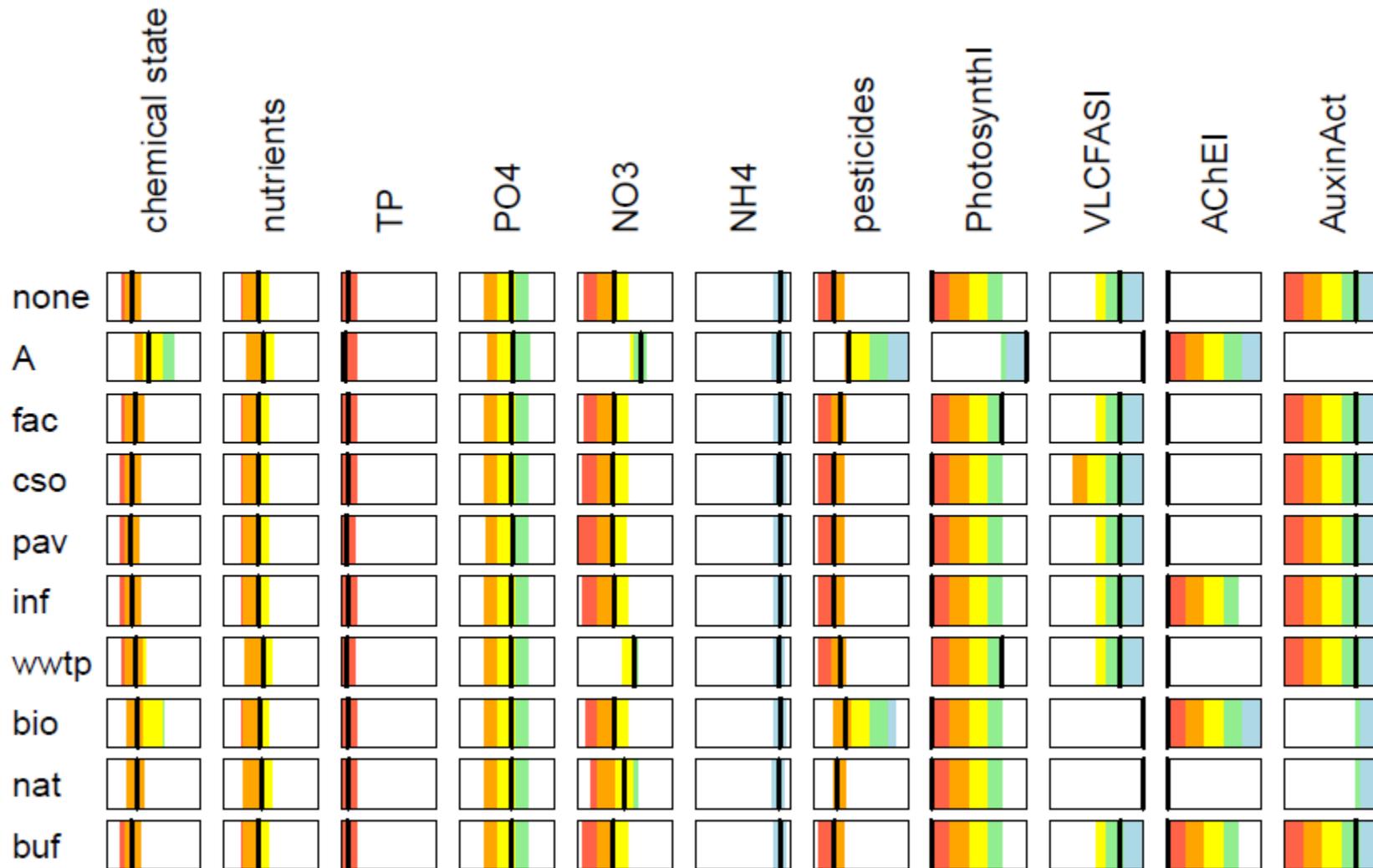
Langhans, Lienert, Schuwirth, Reichert (2013): How to make river assessments comparable: a demonstration for hydromorphology. *Ecological Indicators* 32, 264–275

G Visualization of results - hierarchy



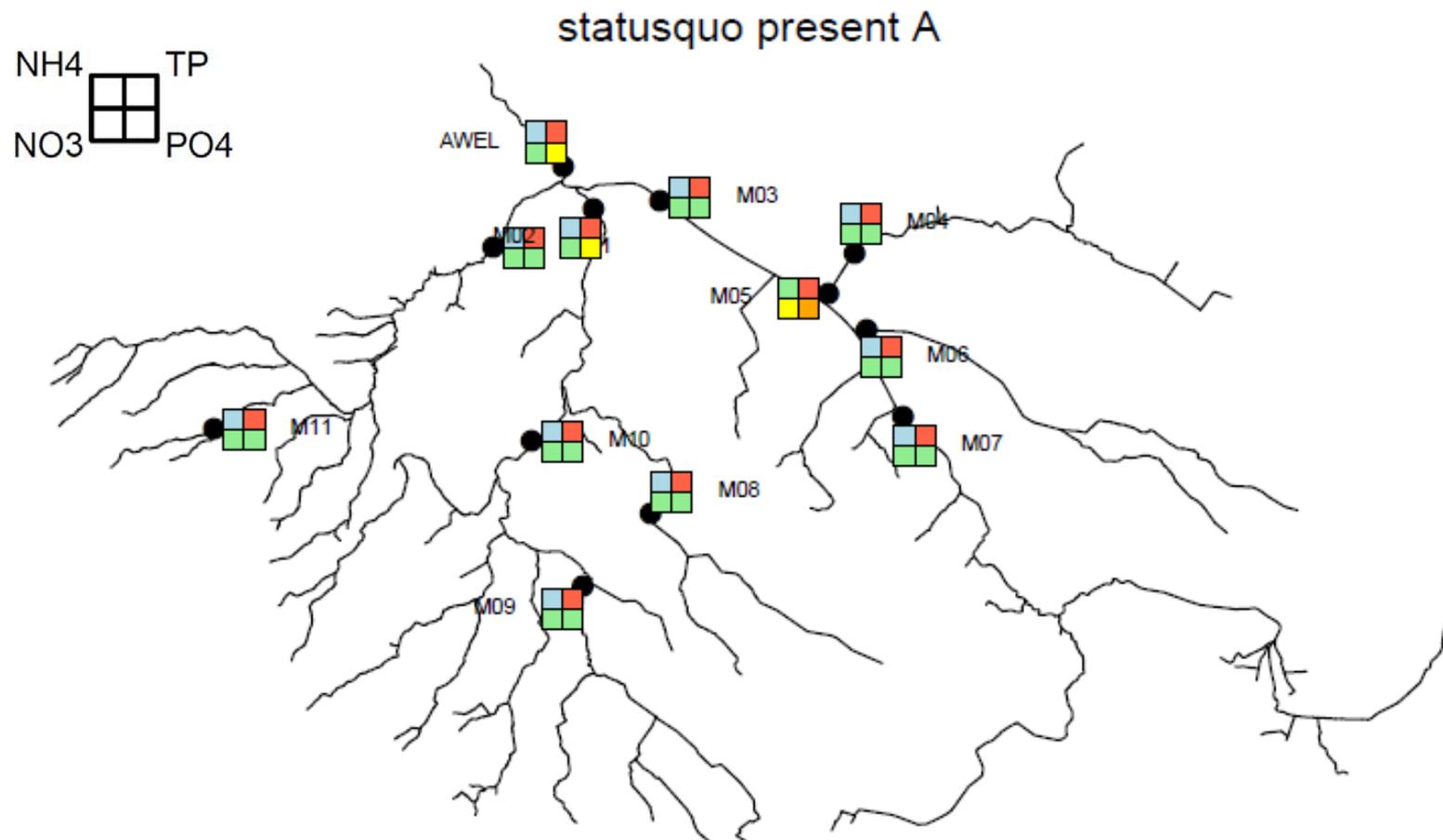
Reichert, Schuwirth, Langhans (2013): Constructing, evaluating and visualizing value and utility functions for decision support, *Environmental Modelling & Software* 46, 283-291. [see also R-package "utility".]

G Visualization of results - table



Reichert, Schuwirth, Langhans (2013): Constructing, evaluating and visualizing value and utility functions for decision support, *Environmental Modelling & Software* 46, 283-291. [see also R-package "utility".]

G Visualization of results - catchment



Reichert, Schuwirth, Langhans (2013): Constructing, evaluating and visualizing value and utility functions for decision support, *Environmental Modelling & Software* 46, 283-291. [see also R-package "utility".]

Prediction of ecological effects: Streambugs



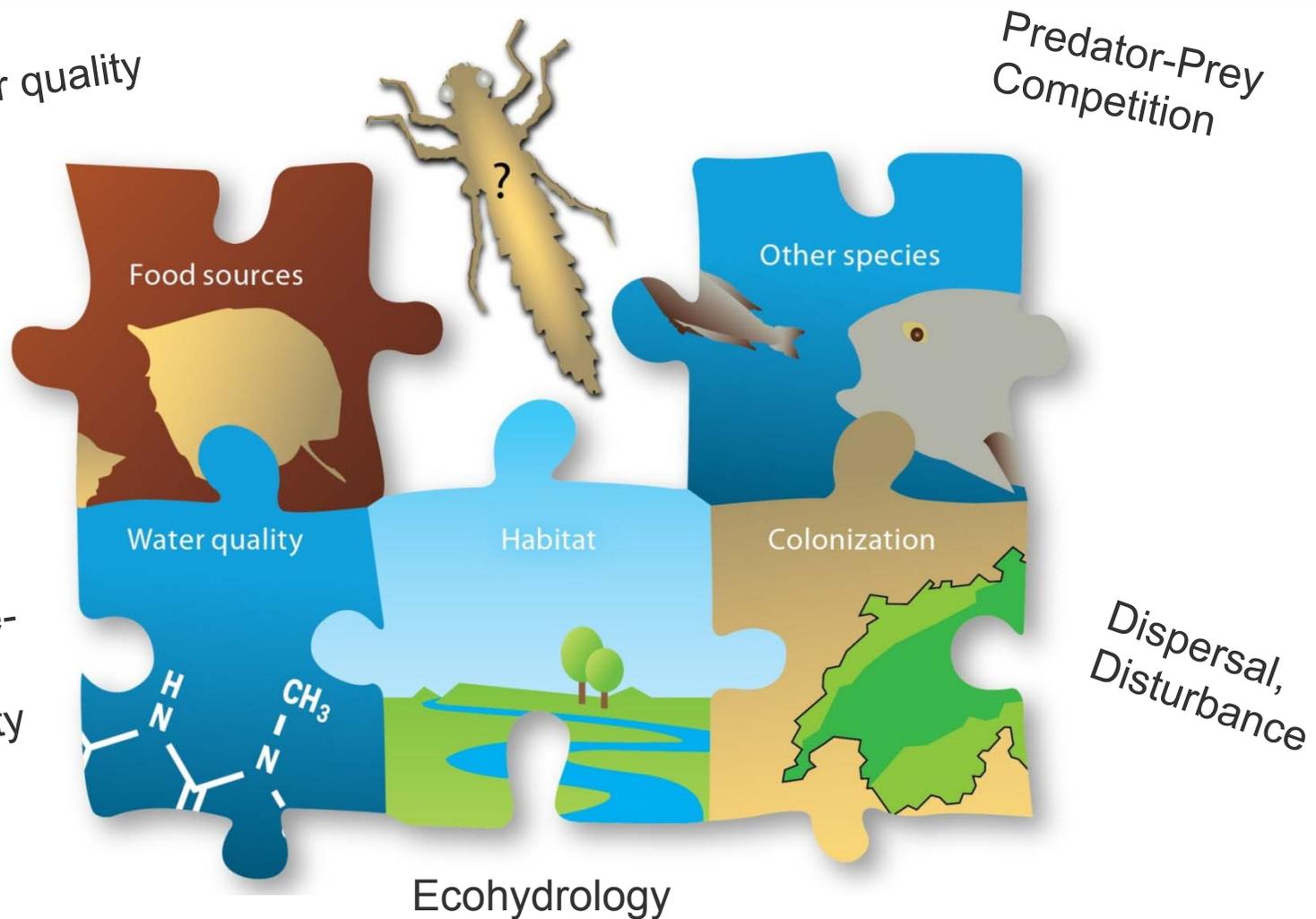
Motivation - predicting the community composition of macroinvertebrates in (wadeable) streams to:

- test **understanding** of **ecosystem functioning**
- test hypotheses on **food-web stability**
- identify influential sources of **human disturbance**
- predict consequences of **management** options
(river restoration, upgrade of waste water treatment plants)
- assess impact of **future development** (climate change, land use)



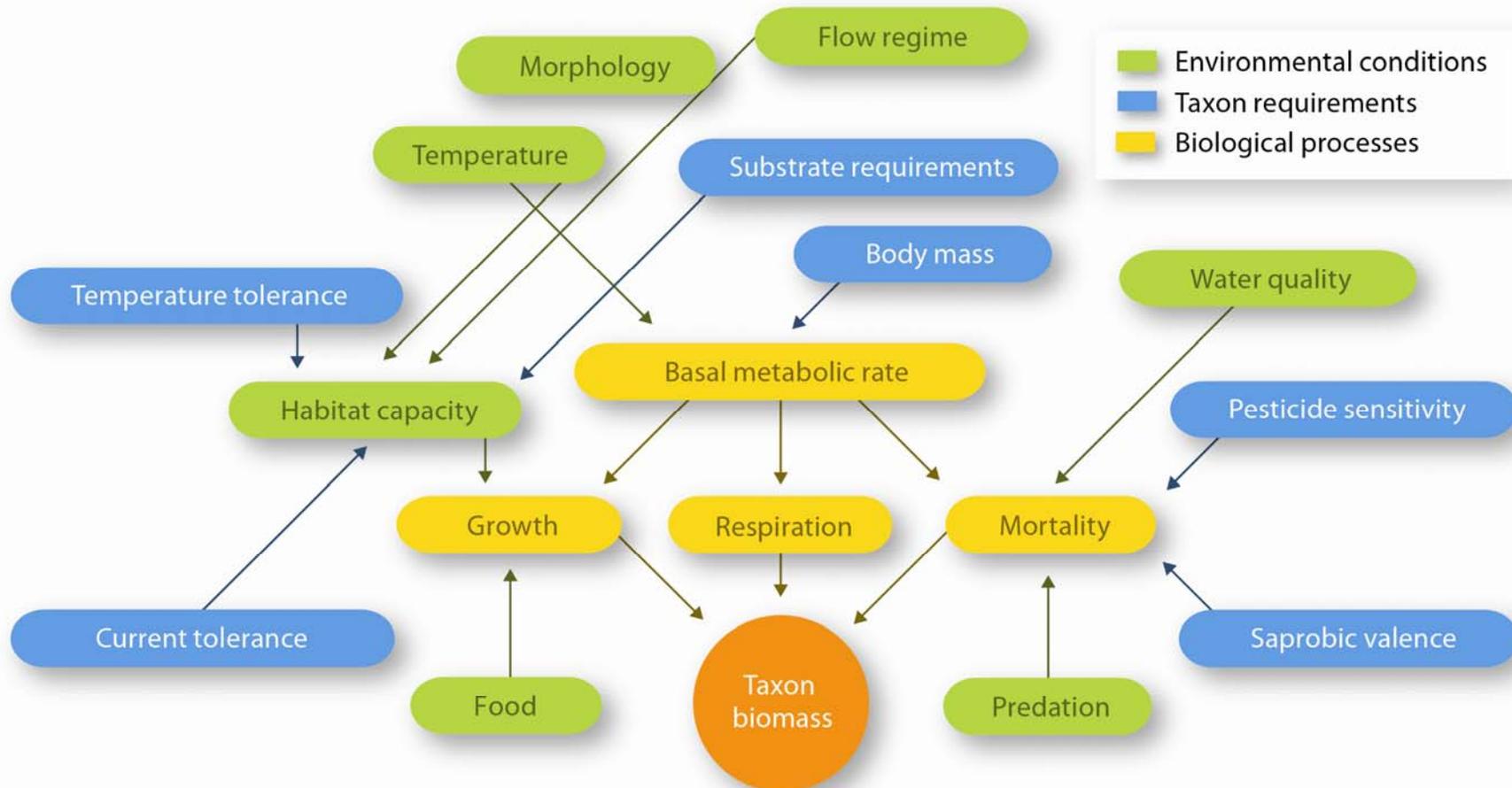
Prediction of ecological effects: Streambugs

Leaf litter quality
Shading



Schuwirth & Reichert (2013): Bridging the gap between theoretical ecology and real ecosystems: modeling invertebrate community composition in streams. *Ecology* 94, 368–379.

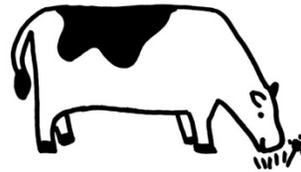
Streambugs 1.0



Schuwirth & Reichert (2013): Bridging the gap between theoretical ecology and real ecosystems: modeling invertebrate community composition in streams. *Ecology* 94, 368–379.

Ecological traits from databases

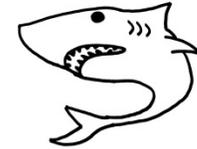
- feeding types
- body mass M
- energy content
- habitat tolerances (current, temperature, substrate, pH, salinity)
- sensitivity towards pollutants (pesticides, saprobic conditions)
- mobility, emergence,



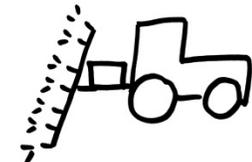
grazing



shredding



predation



e.g. www.freshwaterecology.info

SPEAR ("Species at risk", Liess et al 2008)

Tachet (Tachet et al. 2000)

Site specific model input

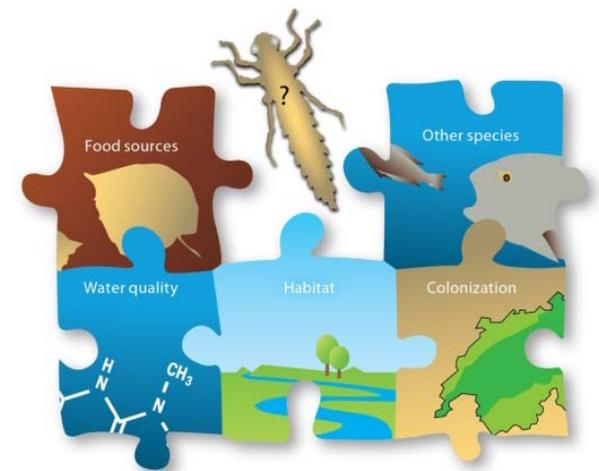
1. environmental conditions

- Temperature (mean water temperature, summer maximum)
- Flow conditions (no-slow-medium-high current class)
- Microhabitat/substrate
- Light conditions (shading, light intensity at the river bed)
- Input of organic material (leaf litter input, suspended OM)
- Water quality (nutrients, organic matter, organic contaminants)

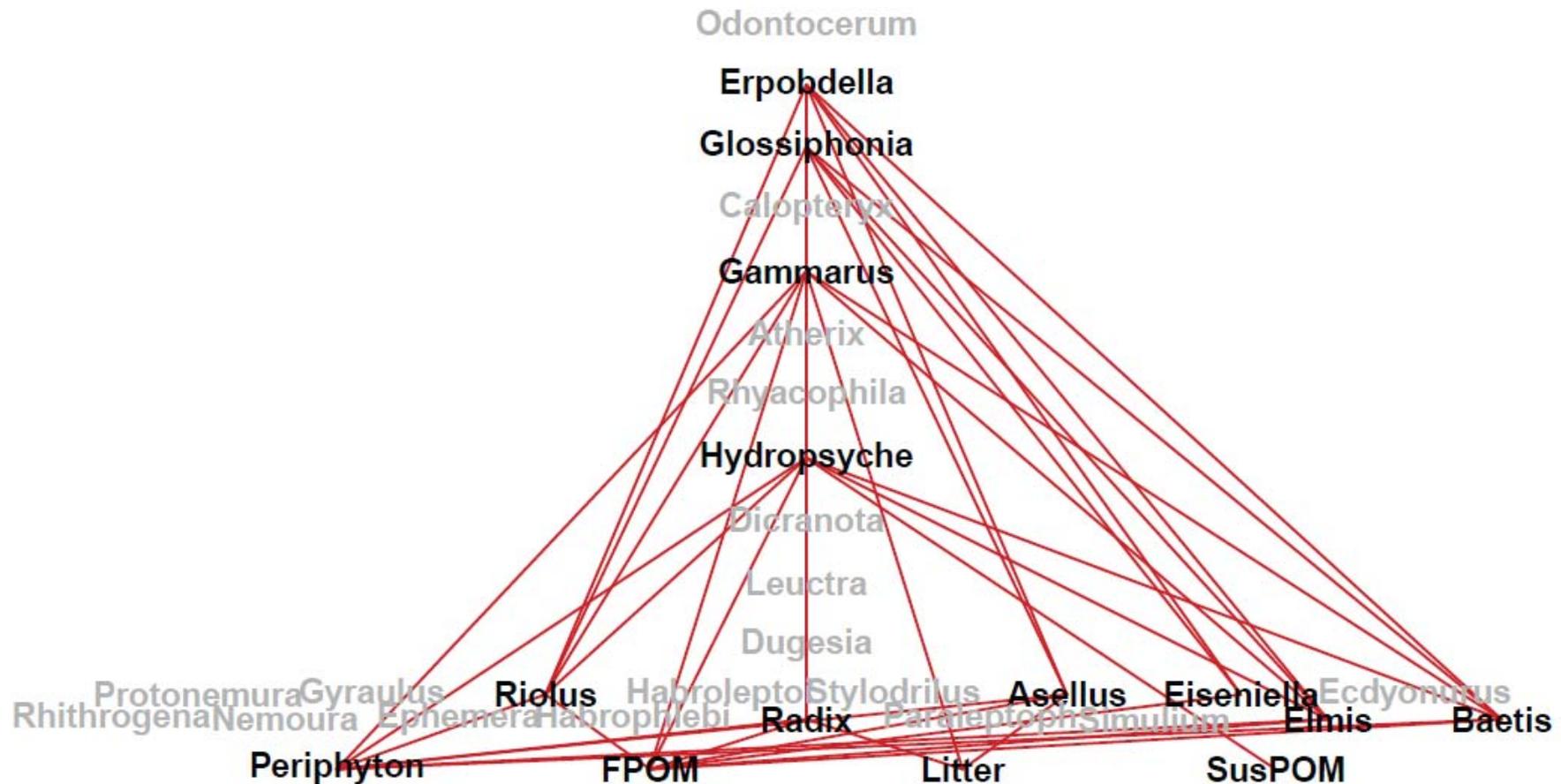
2. regional taxa pool (to exclude dispersal limitation)

Model output

- observation probability for each taxon at each site from equilibrium biomass density at constant environmental conditions

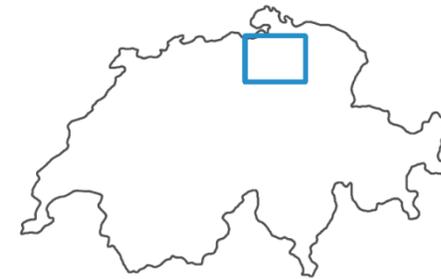
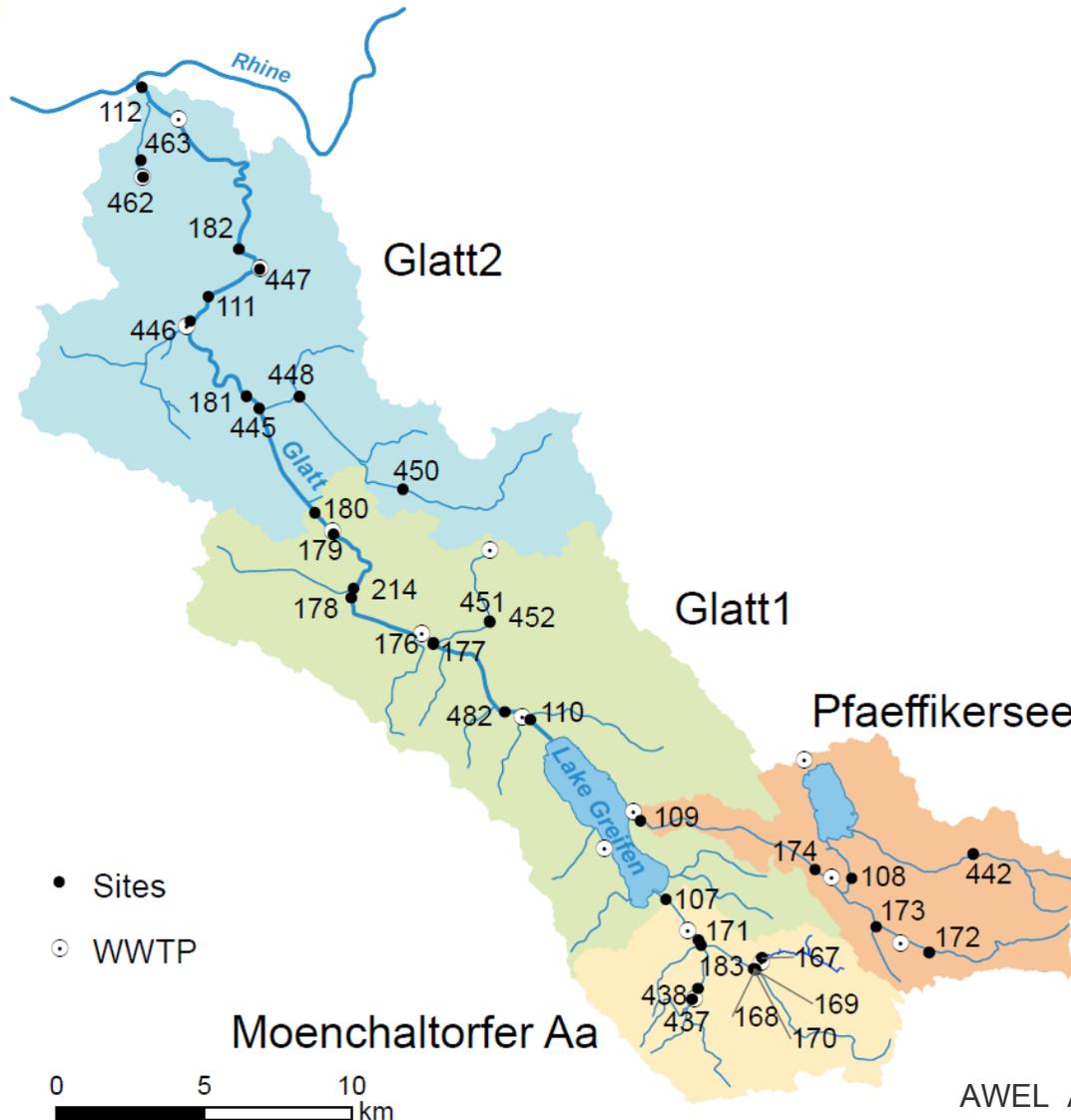


Model output - Food webs



taxa with $p(\text{obs}) > 0.5$ taxa with $p(\text{obs}) < 0.5$

Catchment Glatt - Greifensee, Canton Zurich



Glatt-catchment: 36 sites

- Sites
- WWTP

AWEL Amt für Abfall, Wasser, Energie und Luft:
Zürcher Gewässer 2012, Entwicklung – Zustand - Ausblicke

Results



Results with prior parameter distribution (without calibration)

	no of presence/ absence records*	max no. of results in agreement with observations	sum of taxa modelled correctly	% correct
sub-catchments	a	b	c	c/b
Pfäffikersee	1155	957	731	0.76
Mönchalt. Aa	1600	1296	966	0.75
Glatt 1	2275	1839	1325	0.72
Glatt 2	2030	1689	1253	0.74
total	7060	5781	4275	0.74

*6-11 sites per subcatchment, 4-9 observations per site, ca. 32 taxa

Model calibration with Bayesian inference at each sub-catchment increases compliance with data (to about 90%)

Importance of environmental influence factors - traits

% correct results with mean prior parameter values

sub-catchments	incl. all	excl. T-pref.	excl. sapro.	excl. current	excl. pest.	excl. feedt.	excl. TSCP	excl. TSCPF
Pfäffikersee	0.76	0.77	0.74	0.71	0.63	0.73	0.54	0.44
Mönchalt. Aa	0.74	0.74	0.72	0.69	0.60	0.73	0.50	0.48
Glatt 1	0.71	0.70	0.71	0.64	0.63	0.61	0.53	0.40
Glatt 2	0.72	0.71	0.71	0.63	0.66	0.60	0.55	0.39
total	0.73	0.73	0.72	0.66	0.63	0.65	0.53	0.42

- excl. **temperature** preference and **saprobic** valence has minor influence
- **current** preference, sensitivity to **pesticides** and feedingtypes are important factors to predict community composition of macroinvertebrates

Conclusions

- mechanistic models are valuable tools to integrate and communicate knowledge about cause-effect relationships and to test hypotheses
- most critical aspect is **treating uncertainty** in an adequate way
- multi-criteria decision support provides a framework to combine **objective** scientific **predictions** about future development and consequences of management alternatives with **subjective valuation** of stakeholders
- facilitates synthesis of large interdisciplinary projects

A scenic mountain landscape featuring a calm lake in the foreground that perfectly reflects the clear blue sky and the surrounding green mountains. The mountains are covered in lush green grass, and some peaks in the distance are still covered in snow. The overall atmosphere is peaceful and natural.

Thanks for your attention!

nele.schuwirth@eawag.ch

Acknowledgment

Thanks for your
attention



Funding

Project iWaQa: "Integrated River Water Quality Management"

SNF Swiss National Research Program 61 "Sustainable Water Management"

iWaQa Team:

Christian Stamm, Mark Honti, Chris Robinson, Peter Reichert, Anne Dietzel,
Rosi Siber, Nico Ghielmetti, Jörg Rieckerman, Mark Gessner

All people who gave access to databases and measured data!

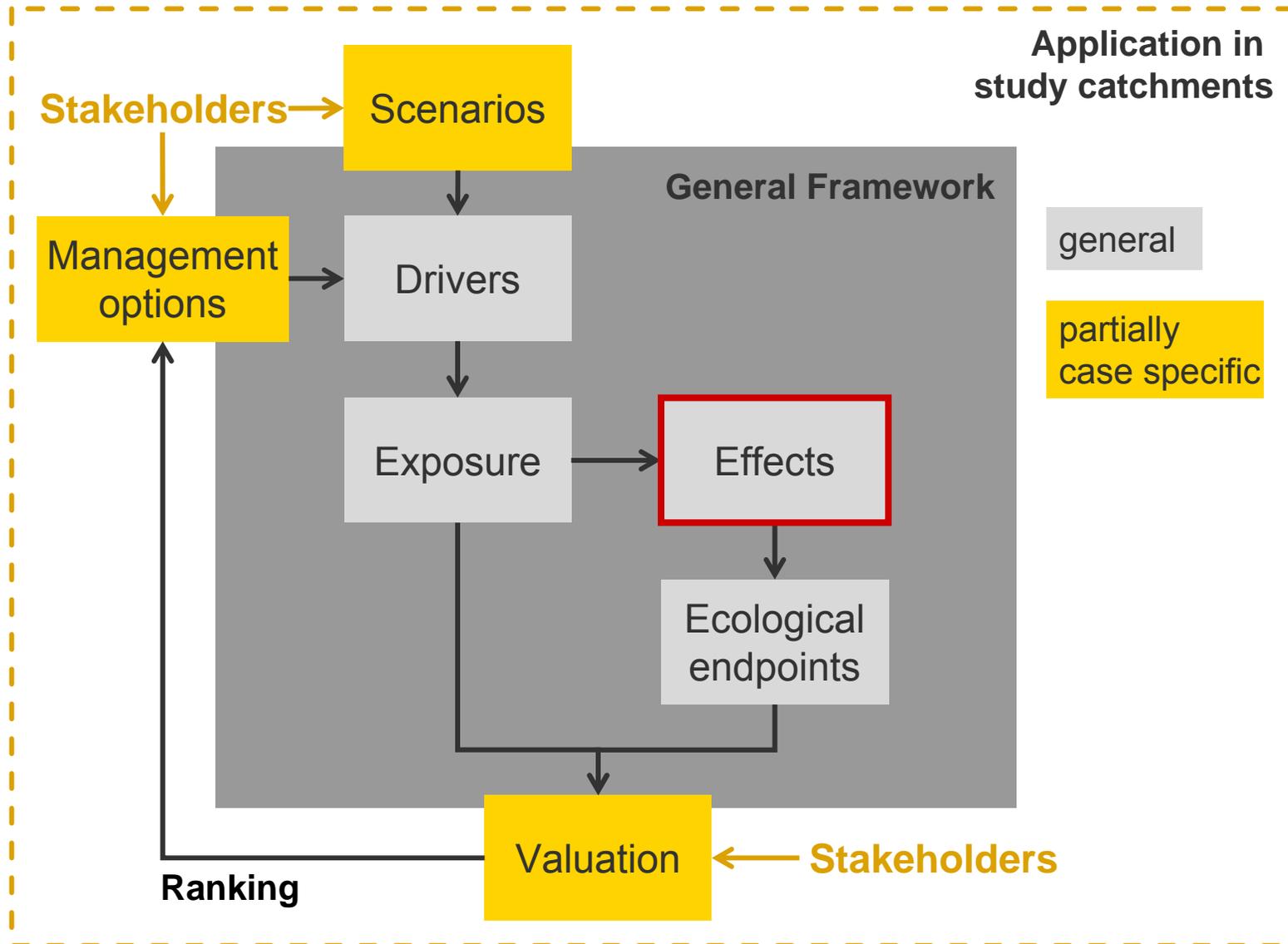


References MCDA in environmental management

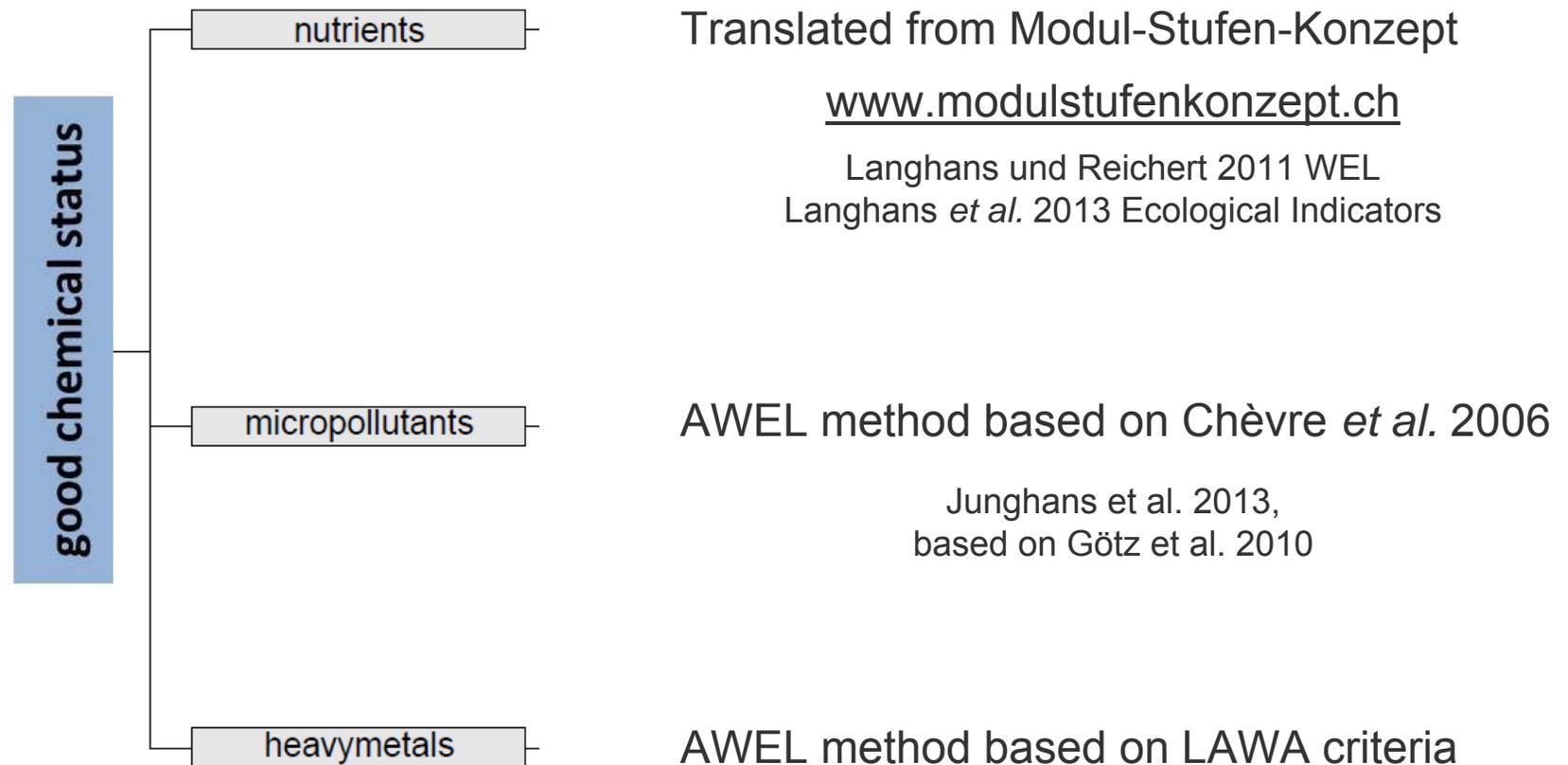
- Langhans, Lienert, Schuwirth, Reichert (2013): How to make river assessments comparable: A demonstration for hydromorphology, *Ecological Indicators* 32, 264-275.
- Reichert, Borsuk, Hostmann, Schweizer, Spörri, Tockner, Truffer (2007): Concepts of decision support for river rehabilitation, *Environmental Modelling & Software* 22, 188-201.
- Reichert, Schuwirth, Langhans (2013): Constructing, evaluating and visualizing value and utility functions for decision support, *Environmental Modelling & Software* 46, 283-291. [see also R-package “utility”.]
- Reichert, Schuwirth, Langhans (2011): MCWM - Ein Konzept für multikriterielle Entscheidungsunterstützung im Wassermanagement. *Wasser Energie Luft*, 103 (2), 139-148.
- Scholten, Reichert, Schuwirth, Lienert (in prep): Tackling uncertainty in multi-criteria decision analysis - Applied to water supply infrastructure planning.
- Schuwirth, Reichert, Lienert (2012): Methodological aspects of multi-criteria decision analysis for policy support: A case study on pharmaceutical removal from hospital wastewater. *European Journal of Operational Research* 220, 472-483.
- Schuwirth, Stamm, Reichert (2012): Incorporation of uncertainty in decision support to improve water quality. In: Seppelt, Voinov, Lange, Bankamp (Eds.): International Environmental Modelling and Software Society (iEMSs), 6th Biennial Meeting, Leipzig, Germany, 1005-1012.

Conceptual framework - project iWaQa

"Integrated management of river water quality"



F Value functions based on assessment programs



AWEL Amt für Abfall, Wasser, Energie und Luft, Kanton Zürich, Statusbericht 2006: Wasserqualität der Seen, Fließgewässer und des Grundwasser im Kanton Zürich.

LAWA: Übersicht über Qualitätsanforderungen der EG, der internationalen Flussgebietsgemeinschaften und der LAWA; www.umweltbundesamt.de/wasser/themen/ow_s2_2.htm

F Value function for lowest level sub-objectives

Translation of existing assessment procedure into measurable value functions

www.modulstufenkonzept.ch

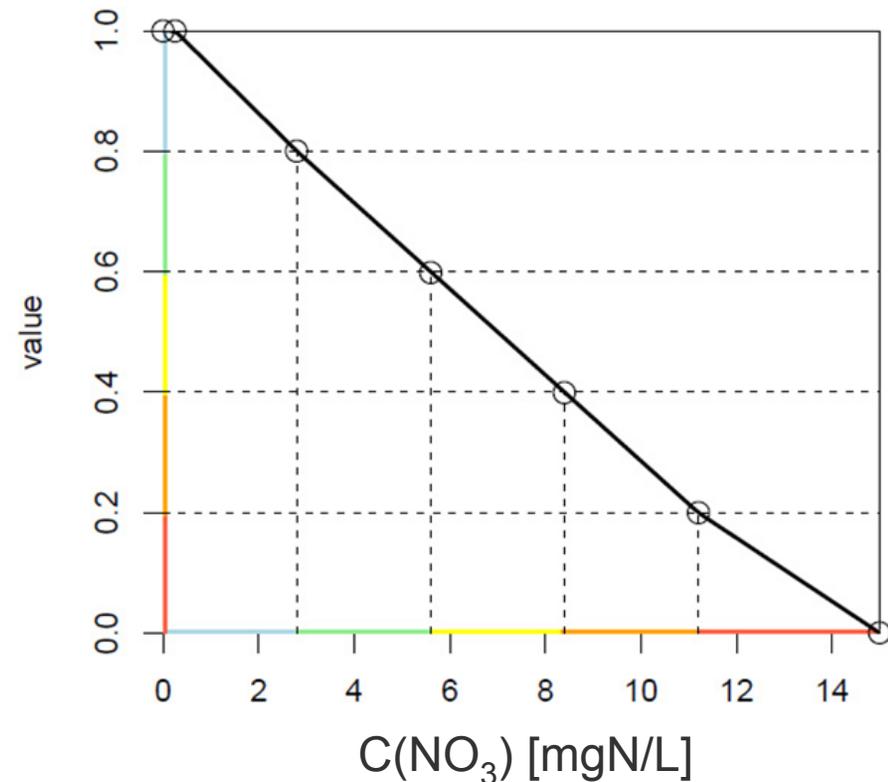
Quality class

very good	$C < \frac{1}{2} Z$
good	$\frac{1}{2} Z \leq C < Z$
moderate	$Z \leq C < 1.5 Z$
poor	$1.5 Z \leq C < 2 Z$
bad	$C \geq 2 Z$

Z: legal quality standard, C: concentration

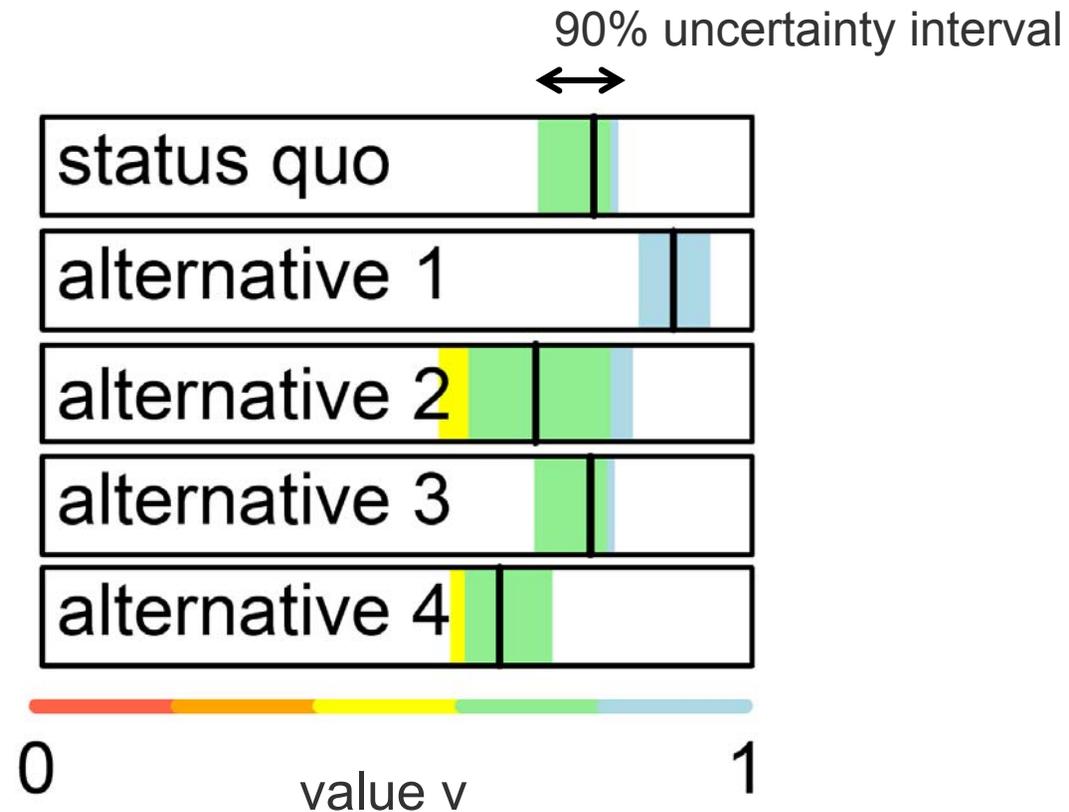
advantage: improvement within a class is visible, avoids rounding error

assumptions: all class transitions are equally valuable, piece-wise linear interpolation



G Visualization of results

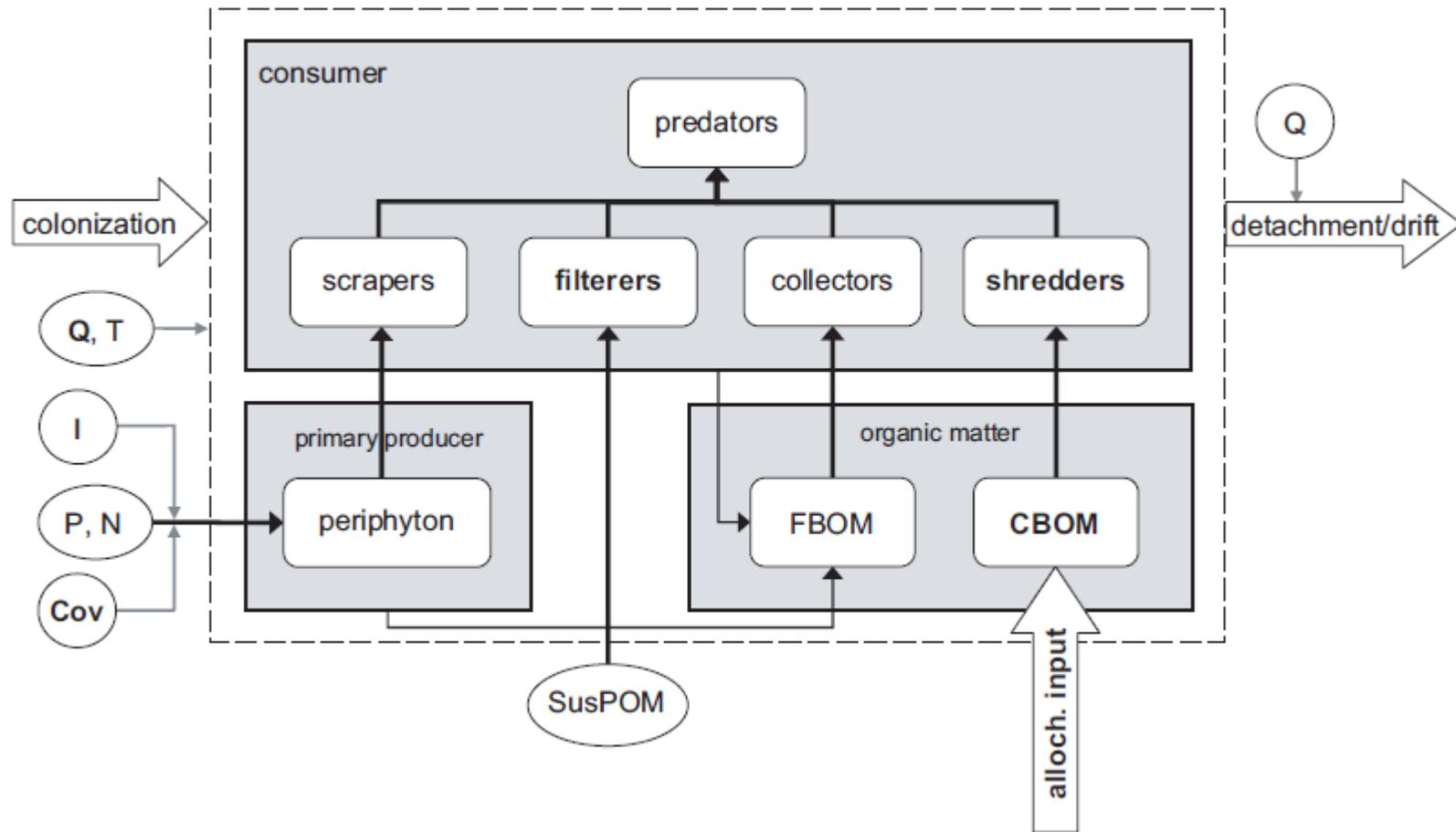
Propagating uncertainty to overall value



Compare alternatives, stakeholders, scenarios - identification of potential consensus solutions

ERIMO: Ecological River Model based on functional groups

N. Schuwirth et al. / Ecological Modelling 222 (2011) 91–104



Schuwirth, Kühni, Schweizer, Uehlinger, Reichert 2008 Freshwater Biology

Schuwirth, Acuña, Reichert, 2011 Ecological Modelling

ERIMO: Ecological River Model based on functional groups

- good representation of functional aspects
 - short term dynamics: seasonal effects, disturbance, ecosystem resilience
 - functional groups not (that) sensitive to water quality
 - omnivory can not be implemented
 - structural aspects hidden
- taxonomic resolution needed



Model formulation

Differential equations for the biomasses of all taxa and of organic matter per river length $\mathbf{B} = (B_1, \dots, B_n)$ [gDM/m]

stoichiometric coefficients $\mathbf{v} = \{v_{ij}\}$

process rates $\mathbf{r} = (r_1, \dots, r_m)$ [gDM/m²/a], which depend on parameters θ

$$\frac{d\mathbf{B}}{dt} = \mathbf{v} \mathbf{r} \left(\frac{\mathbf{B}}{w}, \theta \right) w$$

R-package "stoichcalc"

Reichert & Schuwirth 2010, Environmental Modelling and Software

Metabolic Theory of Ecology

T-corrected maximal rates of whole organism biomass production

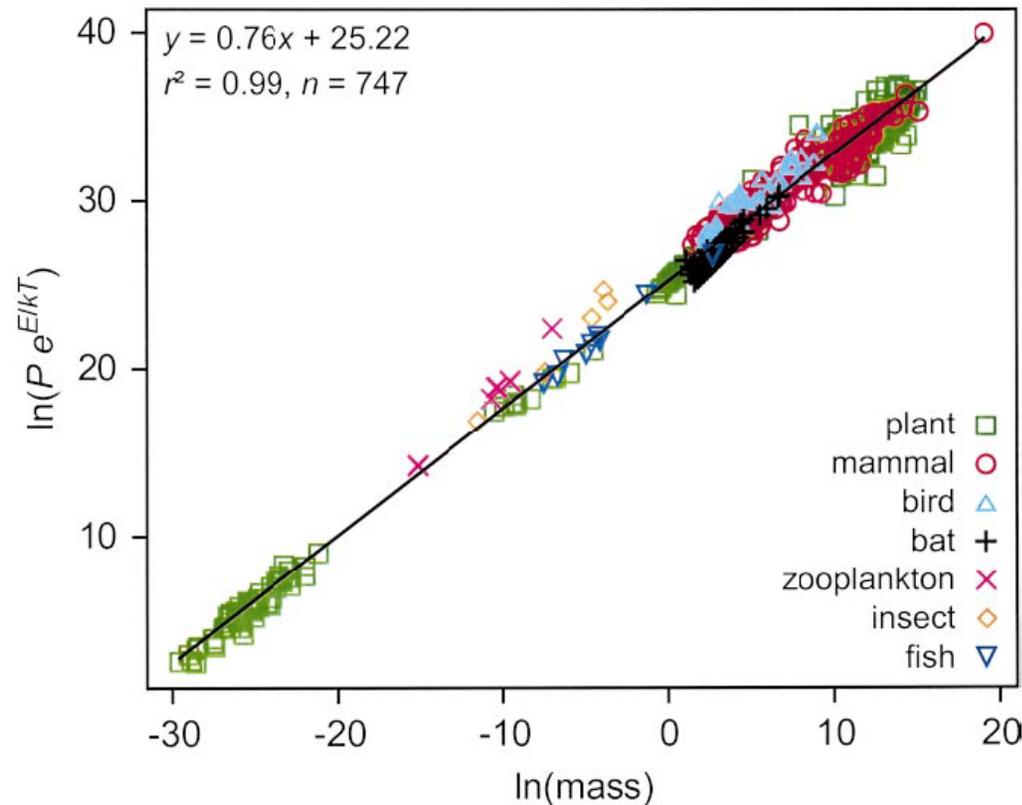


FIG. 2. Mass dependence (mass measured in grams) of temperature-corrected maximal rates of whole-organism biomass production ($Pe^{E/kT}$, measured in grams per individual per year) for a wide variety of organisms, from unicellular eukaryotes to plants and mammals (from Ernest et al. 2003). Data, which span >20 orders of magnitude in body size, have been temperature corrected using Eq. 6. The allometric exponent, indicated by the slope, is close to the predicted value of $\frac{3}{4}$ (95% CI, 0.75–0.76).

from Brown et al 2004: *Ecology*, 85(7), 2004, pp. 1771–1789



Growth, respiration and death rates

$$r_{\text{gro on } j} = f_{\text{gro}} \cdot f_{\text{gro } i} \cdot f_{\text{lim food}} \cdot f_{\text{pref } j} \cdot f_{\text{self inh}} \cdot f_{\text{basal } i} \cdot r_{\text{basal metab}}$$

$$r_{\text{resp } i} = f_{\text{resp}} \cdot f_{\text{basal } i} \cdot r_{\text{basal metab}}$$

$$r_{\text{mort } i} = f_{\text{mort}} \cdot f_{\text{basal } i} \cdot r_{\text{basal metab}} \cdot f_{\text{org contam}} \cdot f_{\text{saproby}}$$

$f_{\text{org contam}}$ factor dependent on pesticide pollution and sensitivity of the taxon

f_{saproby} factor dependent on organic matter pollution and classification of the taxon in the saprobic system

$f_{\text{self inh}}$ factor depending on temperature regime, current regime, substrate

Growth rate of consumers

$$r_{\text{gro on } j}^{\text{cons}} = \underbrace{f_{\text{gro}} \cdot f_{\text{gro } i} \cdot f_{\text{lim food}} \cdot f_{\text{pref } j} \cdot f_{\text{self inh}} \cdot f_{\text{basal } i}}_{\text{fit from large datasets}} \cdot \underbrace{r_{\text{basal metab}}}_{\text{basal metabolic rate}}$$

fit from large datasets

basal metabolic rate

Growth rate of consumers

$$r_{\text{gro on } j}^{\text{cons}} = f_{\text{gro}} \underbrace{f_{\text{gro } i}} \quad f_{\text{lim food}} \quad f_{\text{pref } j} \quad f_{\text{self inh}} \quad \underbrace{f_{\text{basal } i}} \quad r_{\text{basal metab}}$$

taxon specific parameters, prior mean = 1

account for variation of individual taxa around MTE predictions

Growth rate of consumers

$$r_{\text{gro on } j}^{\text{cons}} = f_{\text{gro}} \cdot f_{\text{gro } i} \cdot \underbrace{f_{\text{lim food}}}_{\text{food limitation}} \cdot f_{\text{pref } j} \cdot f_{\text{self inh}} \cdot f_{\text{basal } i} \cdot r_{\text{basal metab}}$$

$$f_{\text{lim food}} = \frac{D_{\text{food}}^q}{K_{\text{food}}^q + D_{\text{food}}^q}$$

One process for each food source j

D_{food} biomass density of the sum of all food sources

K_{food} halfsaturation density, at which growth rate is reduced to 50% of the max

Growth rate of consumers

$$r_{\text{gro on } j}^{\text{cons}} = f_{\text{gro}} \cdot f_{\text{gro } i} \cdot f_{\text{lim food}} \cdot \underbrace{f_{\text{pref } j} \cdot f_{\text{self inh}}}_{\text{food preference factor}} \cdot f_{\text{basal } i} \cdot r_{\text{basal metab}}$$

food preference factor

$$f_{\text{pref } j} = \frac{D_j \cdot p_j}{\sum_f D_f \cdot p_f}$$

One process for each food source j

D_j biomass density of food source j

p_j preference factor for food source j

Growth rate of consumers

$$r_{\text{gro on } j}^{\text{cons}} = f_{\text{gro}} \cdot f_{\text{gro } i} \cdot f_{\text{lim food}} \cdot f_{\text{pref } j} \cdot \underbrace{f_{\text{self inh}} \cdot f_{\text{basal } i}}_{\text{self inhibition}} \cdot r_{\text{basal metab}}$$

Two alternative formulations (Monod and Blackman)

$$f_{\text{self inh Monod}} = \frac{K_{\text{dens}}}{K_{\text{dens}} + D}$$

$$f_{\text{self inh Blackman}} = \begin{cases} 1 - \frac{D}{2 K_{\text{dens}}} & \text{for } D < 2 K_{\text{dens}} \\ 0 & \text{for } D \geq 2 K_{\text{dens}} \end{cases}$$

D biomass density of the taxon

K_{dens} half-inhibition density where growth rate is reduced to 50% of the max

$$K_{\text{dens}} = h_{\text{dens}} \cdot f_{\text{current}} \cdot f_{\text{temp}} \cdot f_{\text{substrate}}$$

depends on habitat conditions and specific preferences of the taxon

Catchment Glatt - Greifensee, Canton Zurich

