



Improving the parameter identifiability of a watershed scale onsite wastewater infiltration model

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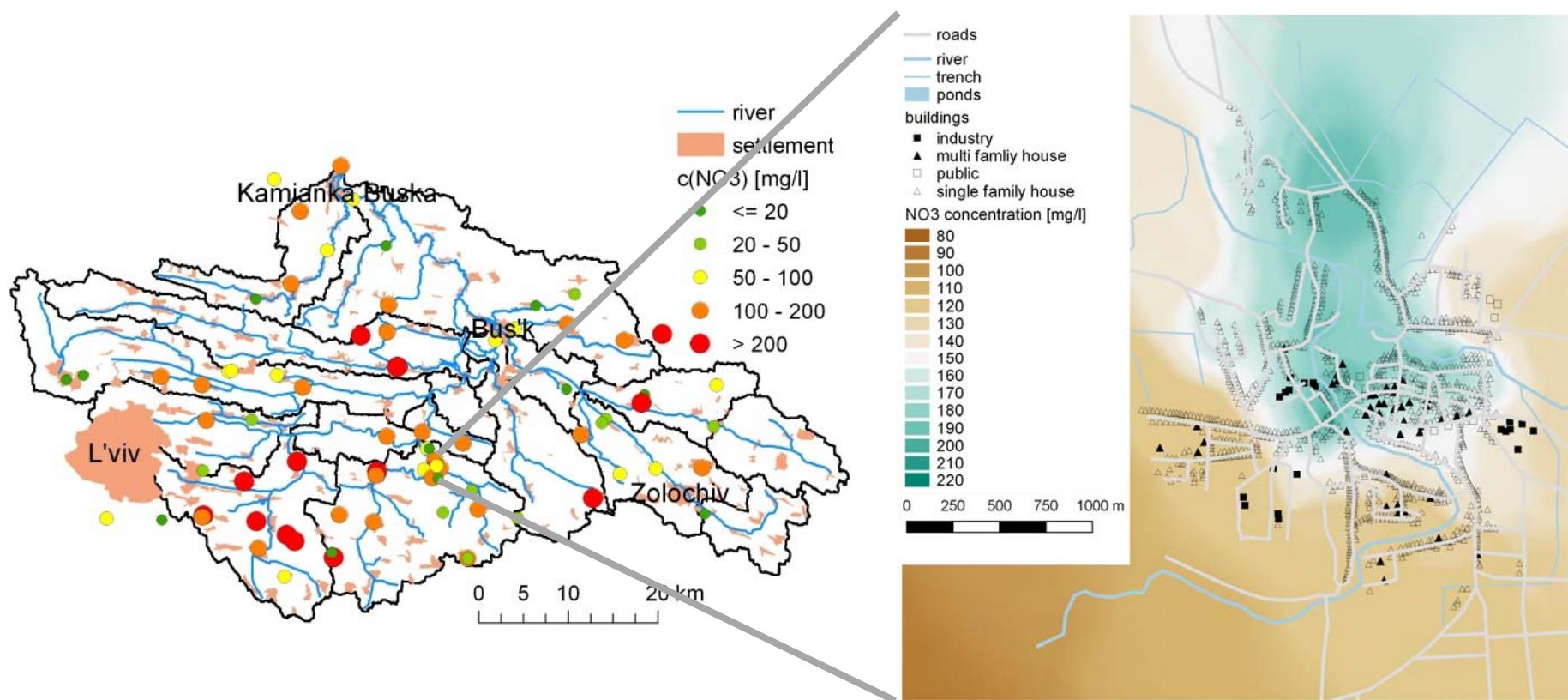
Athens, 16.09.2016

Motivation

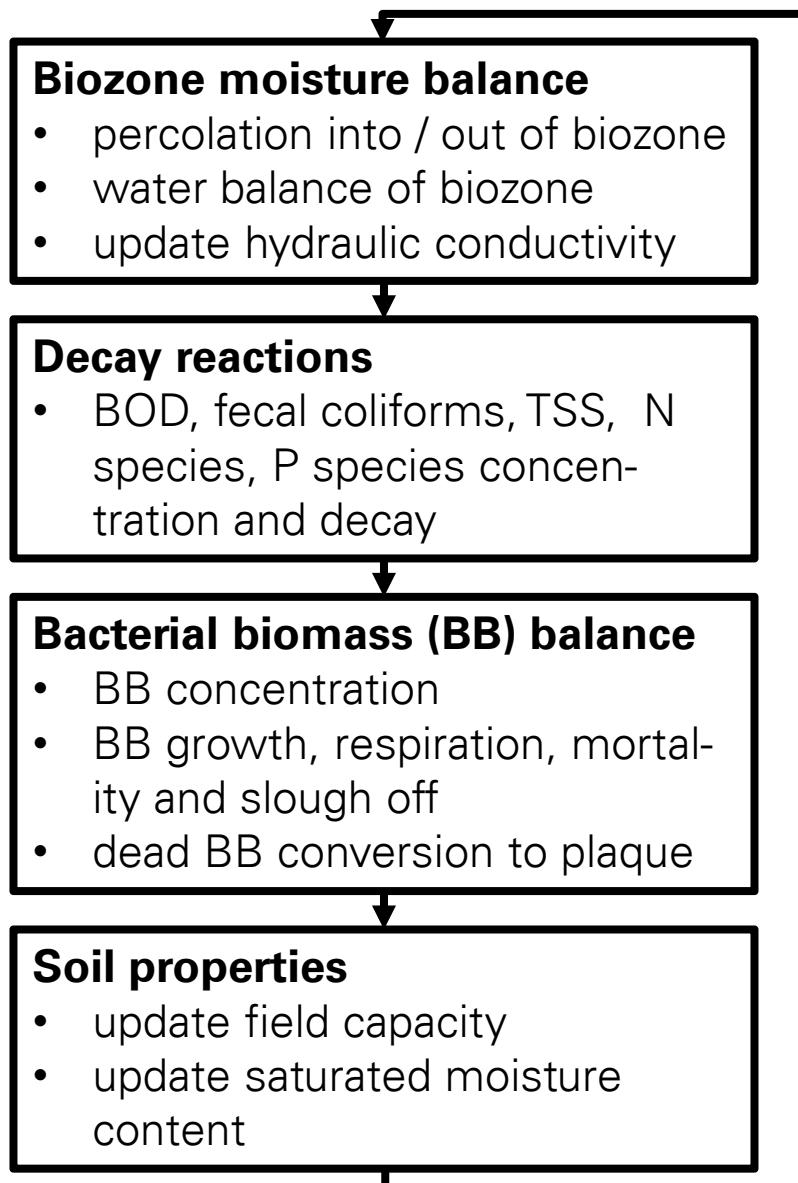
- Infiltration based wastewater disposal globally most frequent
- Pit latrines in low income countries, onsite wastewater systems (OWS) in high income countries
- high local impact on groundwater quality
- explanatory variables:
 - pit latrine density
 - groundwater level
 - hydraulic properties
- few systematic monitoring studies!

Motivation

Nitrate concentrations in house wells in Ukraine



- site scale models:
 - conceptual reactive transport (Wilhelm et al. 1994)
 - coupled vadose zone and kinetic reaction models (Heatwole et al. (2007), MacQuarrie et al. (2001))
- cross scale approaches:
 - simplified aquifer with reactive transport (Wang (2013))
- watershed models:
 - constant removal rate (Behrendt, 1998)
 - biozone mass balance (McCray et al. 2002).



- adapted from McCray (2002)
- model unit: HRU → aggregation of OWS in one unit
- daily time step
- biozone below OWS as additional soil layer
- mass balance of wastewater constituents, bacteria and interaction with soil properties
- OWS specific effluent values

Parameters

Parameter	Explanation	Unit	Min	Mean	Max
QSTE	Pit latrine effluent discharge	$\text{m}^3 \cdot \text{d}^{-1}$	0.05	0.1	0.2
cBODSTE	Pit latrine effluent BOD concentration	$\text{g} \cdot \text{m}^{-3}$	75	150	300
cTSSSTE	Pit latrine effluent TSS concentration	$\text{g} \cdot \text{m}^{-3}$	150	300	600
cFCSTE	Pit latrine effluent FC concentration	$\text{cfu} \cdot \text{ml}^{-1}$	50000	100000	200000
cNH4STE	Pit latrine effluent NH4 concentration	$\text{g} \cdot \text{m}^{-3}$	38	76	154
cNO3STE	Pit latrine effluent NO3 concentration	$\text{g} \cdot \text{m}^{-3}$	2	4	8
BzThk	Thickness of biozone	mm	25	50	100
BioD	Density of biomass	$\text{kg} \cdot \text{m}^{-3}$	900	1000	1100
CBODLBB	BOD to LBB conversion rate	-	0.21	0.42	0.84
CRespR	LBB respiration rate coefficient	d^{-1}	0.008	0.016	0.032
CMortR	LBB mortality rate coefficient	d^{-1}	0.0125	0.025	0.05
LCSIgR	LBB sloughing rate coefficient	d^{-1}	0.000002	0.000004	0.000008
ECSIgR	LBB sloughing rate exponent	-	1.2	1.5	1.875
SlgRPlqCF	slough off to plaque conversion coeff.	-	0.020	0.039	0.078
CTSSPlq	TSS to plaque conversion coefficient	-	0.05	0.1	0.2
LCFC	FC coefficient	-	345	690	1380
ECFC	FC exponent	-	0.64	0.8	1
CBODDR	BOD decay rate coefficient	d^{-1}	25	50	100
CFCBDR	fecal coliform decay rate coefficient	d^{-1}	2.5	5	10
CNitrR	nitrification rate coefficient	d^{-1}	0.193	3.2	53
CDenitrR	denitrification rate exponent	d^{-1}	0.0045	0.0416	0.385

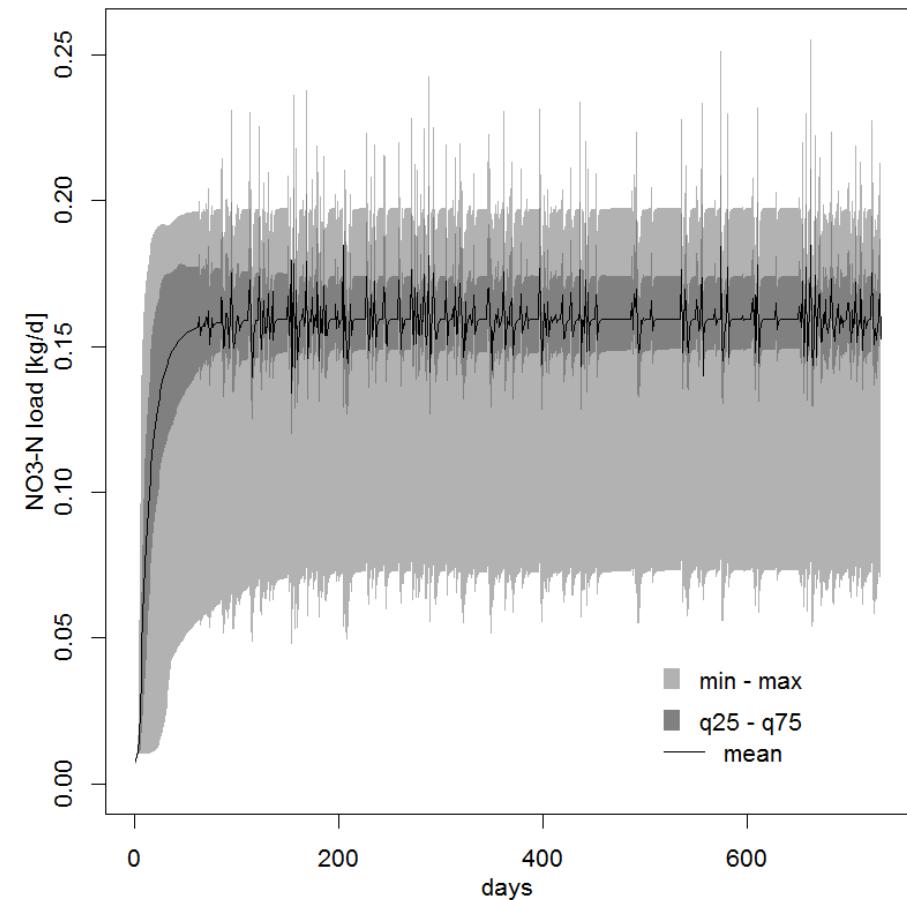
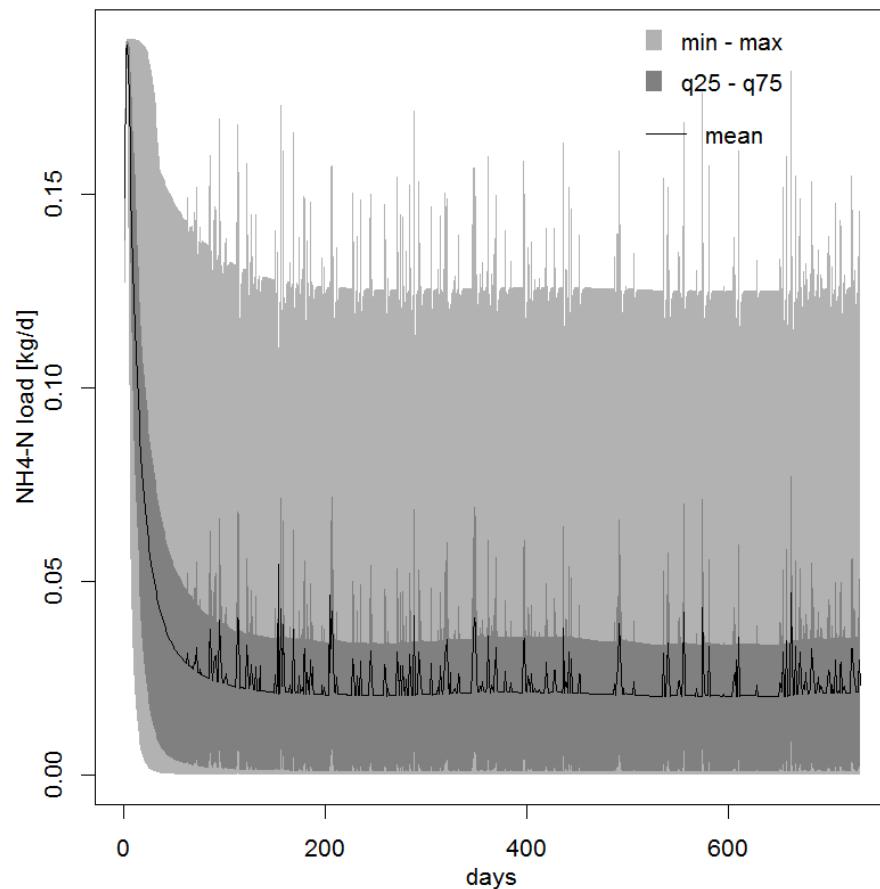
Parameters

Parameter interrelation in SoE

	BzThk	BODLB	RespR	MortR	LSlgR	ESlgR	TSSPlq	LFC	EFC	BioD	BODDR	FCDR	NitDR	DenDR
BzThk										1	1	1	1	1
BODLB			1	1	1	1								
RespR	1			1	1	1								
MortR	1	1			2	2	1							
LSlgR	1	1	2			2	1							
ESlgR	1	1	2	2				1						
TSSPlq					1	1	1							
LFC									1		1			
EFC								1			1			
BioD	1								1					
BODDR	1								1					
FCDR	1													
NitDR	1													
DenDR	1													
interaction	5	4	4	7	7	7	3	2	2	3	1	1	1	1

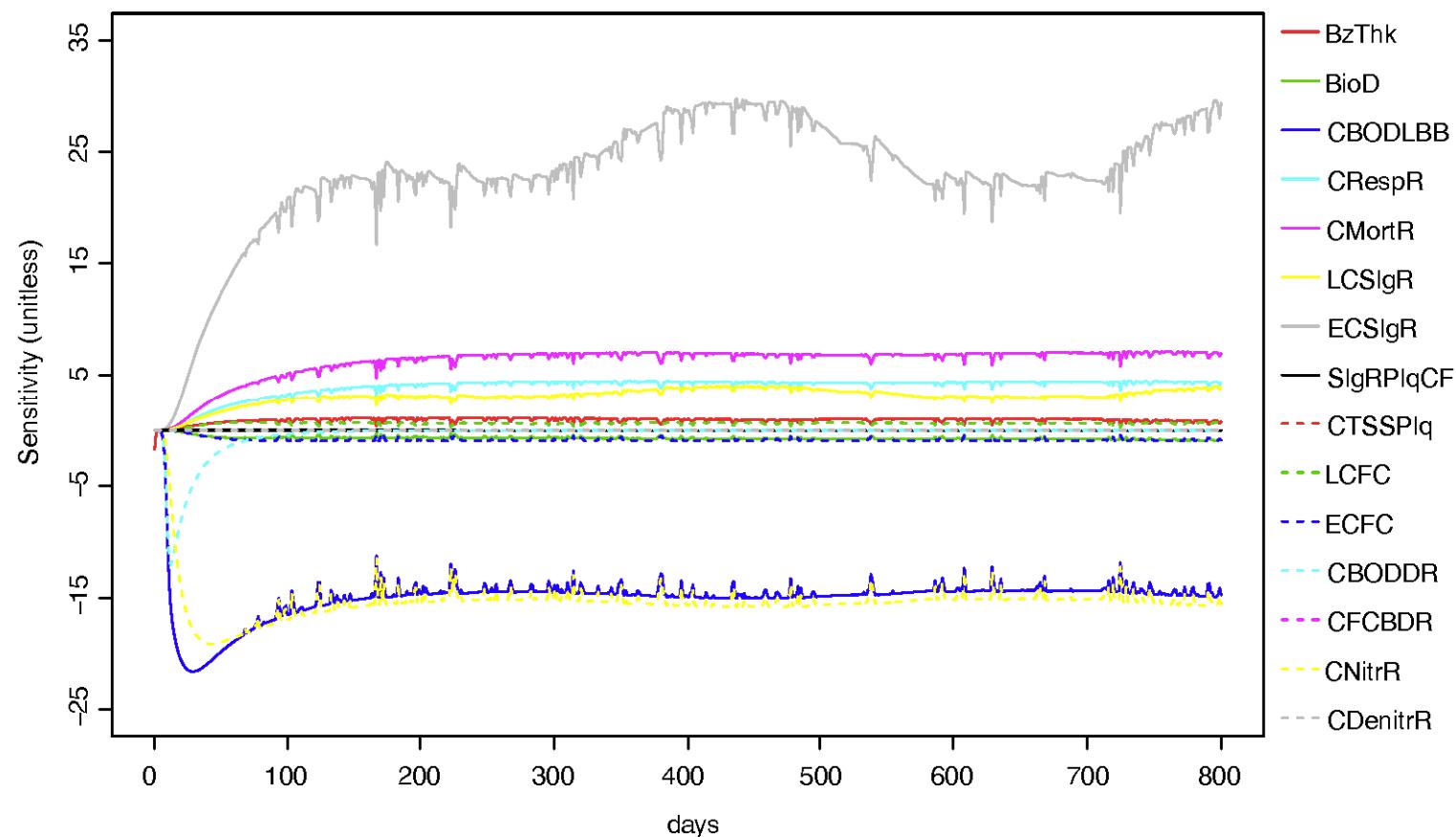
Sensitivity

Global sensitivity and model performance



Sensitivity

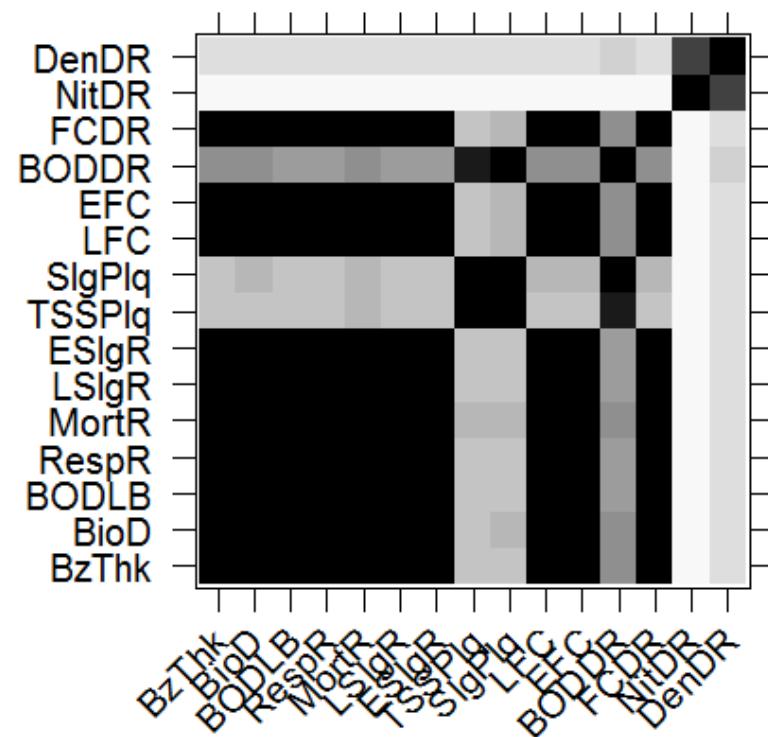
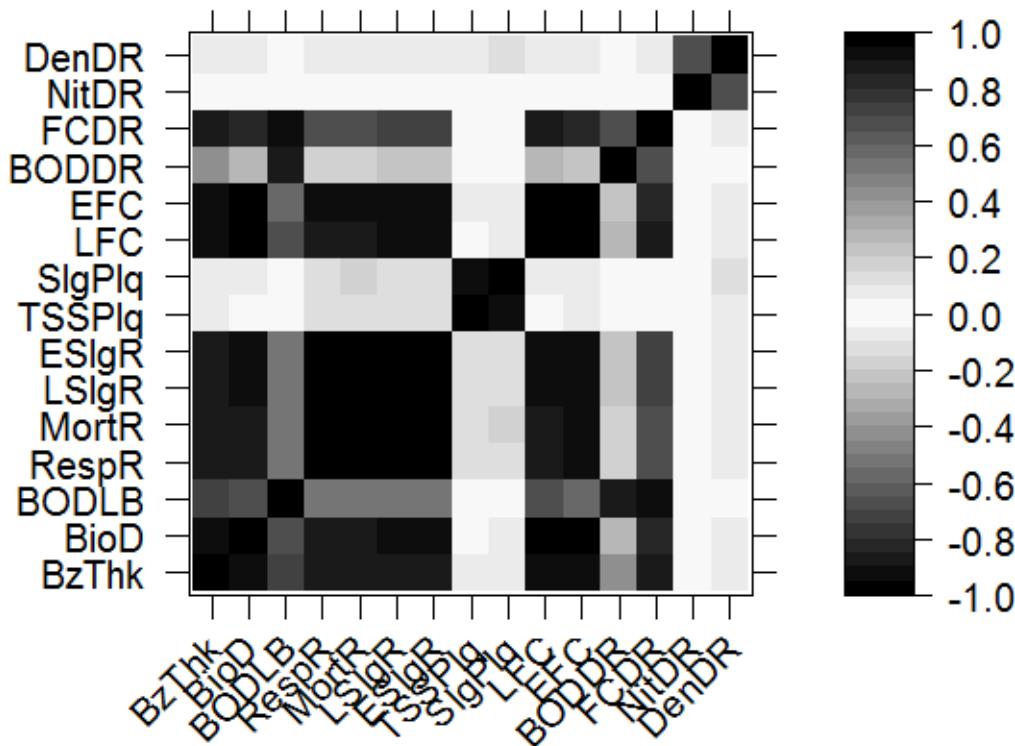
Local sensitivity of parameters to NH₄ concentration



Sensitivity

Correlation of local sensitivity indices for:
build-up phase

steady phase

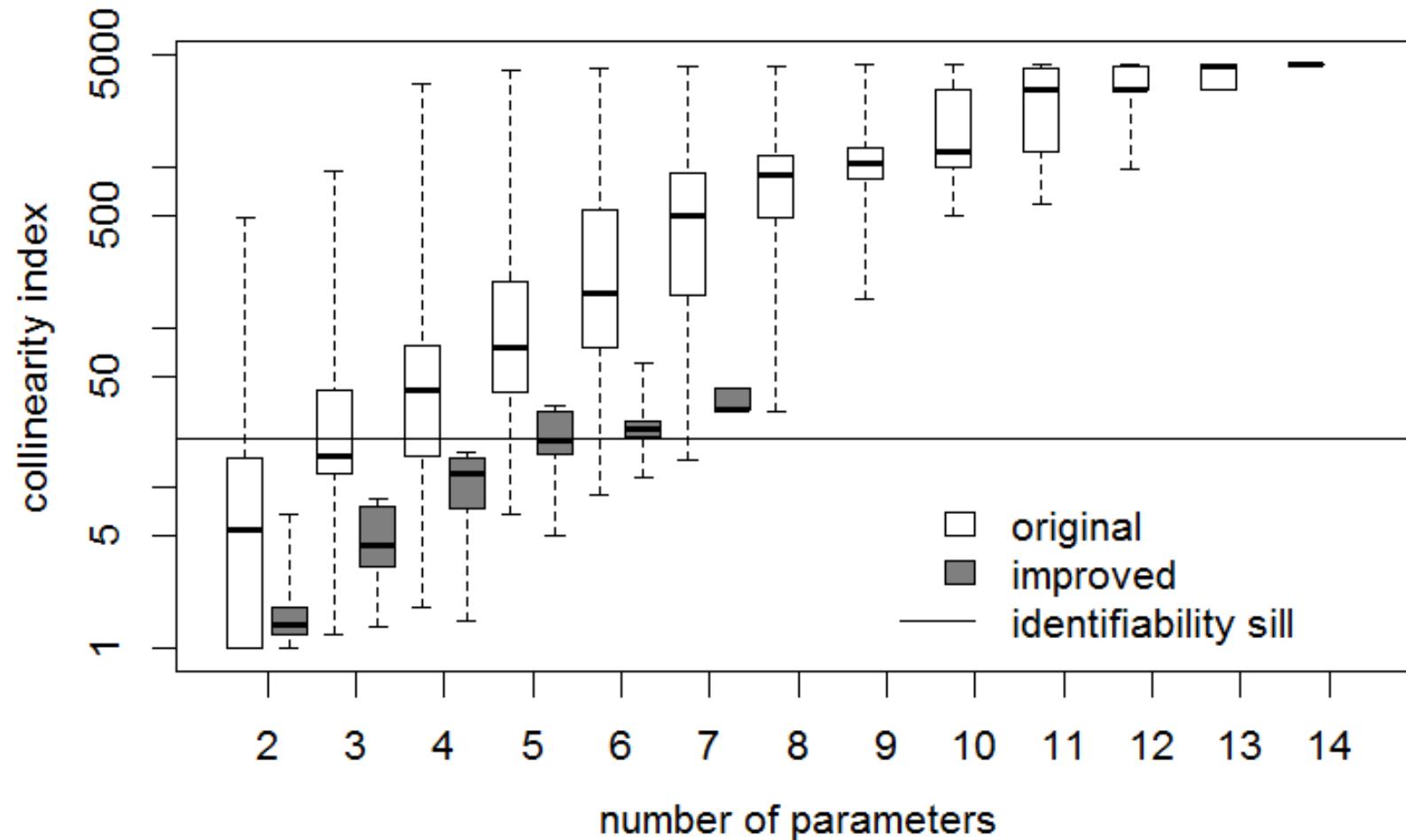


Model adaptations

- grouping of bacterial biomass parameters (ESIgR, LSIgR, MortR, RespR) to LBB decay
 - exponential sloughing rate constant (ESIgR)
 - biomass density (BioD) as a constant
 - exponential field capacity parameter (EFC) constant
 - BOD to biomass conversion constant
- seven out of 14 parameters preserved

Identifiability

Collinearity as measure of identifiability



Conclusions

- OWS algorithm in SWAT capable for regional impact modelling
- adaptation for other systems e.g. pit latrines possible
- algorithm highly collinear
- systematic procedure for model adaptation transferable to other models
- lag of benchmarking monitoring of OWS impact



Thank you for your attention