

Testing a regional scale pesticide fate model against monitoring data: necessity or endless quest?

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- Model purpose.
- MACRO-SE maps and databases.
- MACRO-SE workflow - Example for groundwater
- Simulated scenario (groundwater).
- Results
 - Maps of concentrations.
 - Confusion matrix.
 - Visual examination of the results
- General considerations on the validation of regional models.
- Conclusions.

Providing estimations of the **risks of pesticide losses to surface- and groundwater, from arable-land, in Sweden** (21 Swedish counties).

Help **understanding** (*some of*) the **factors affecting pesticide losses** at the regional scale, from **field to catchment**;

for:

1. **Swedish authorities.**
2. **Researchers.**

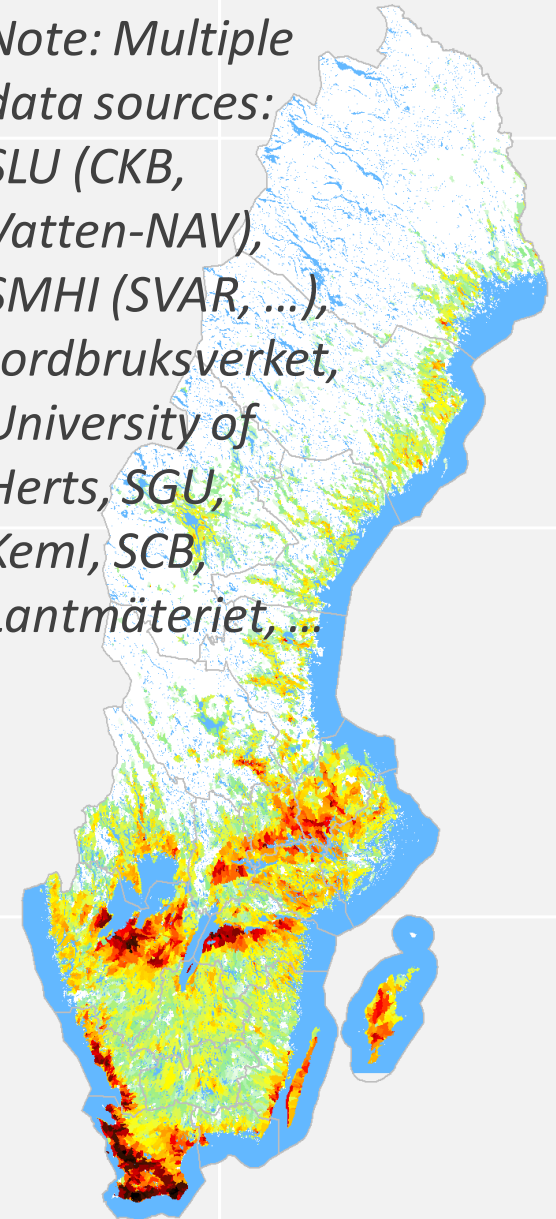


The model **MACRO** (*in short*)

- MACRO pesticide fate model.
- 1D.
- Water flow micropores (Richards) and macropores (kinematic wave) and convective-dispersive solute transport.
- Percolation (vertical) and losses to the drain (lateral; Houghoudt + seepage potential).
- Freundlich sorption; 1st order degradation.
- Penman-Monteith evapotranspiration.
- No runoff and erosion.
- Long simulation time so no uncertainty assessment (regional).

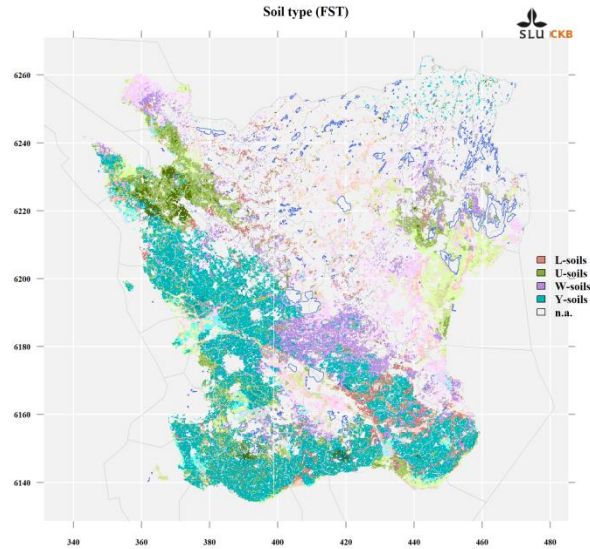
- (1) **Maps of arable soils.**
- (2) **Climate** maps & climate data series.
- (3) Statistics on **crop area (new)**.
- (4) Statistics on **pesticide usage**: What substance? On what crop? What time of the year? What dose? Where?
- (5) **Crop physiological stages** (emergence, harvest, ...).
- (6) Up-to-date **Pesticide Properties Database**.
- (7) *Simulated average water flow in catchment outlet (SMHI S-HYPE)*

Note: Multiple data sources: SLU (CKB, Vatten-NAV), SMHI (SVAR, ...), Jordbruksverket, University of Herts, SGU, KemI, SCB, Lantmäteriet, ...

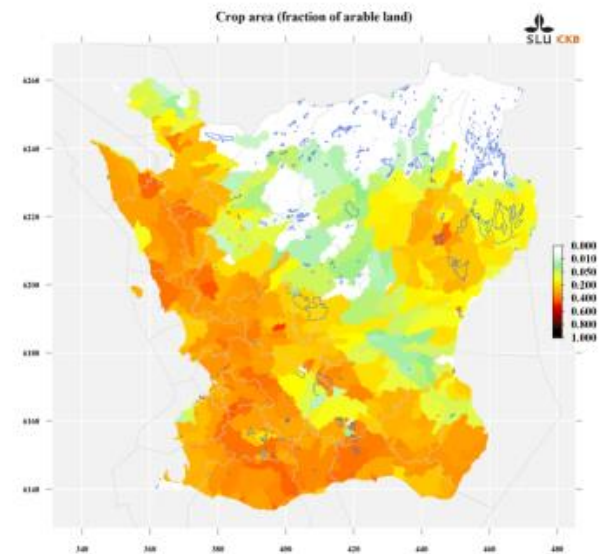


Groundwater risk:

- Scania county.
- Winter cereals.
- Isoproturon.
- ~500 g/ha.
- Spay: mid-October or mid-April



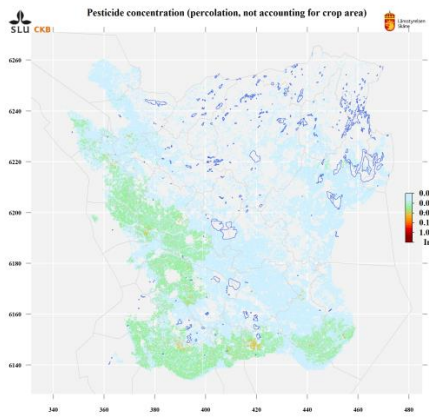
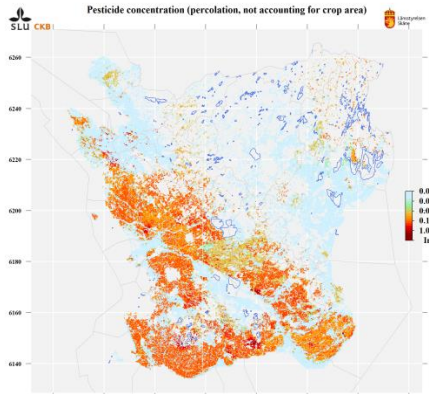
© SLU, CKB | Overlay and original data source SLU, SGU, Lantmäteriet



Autumn cereals; map: 14182
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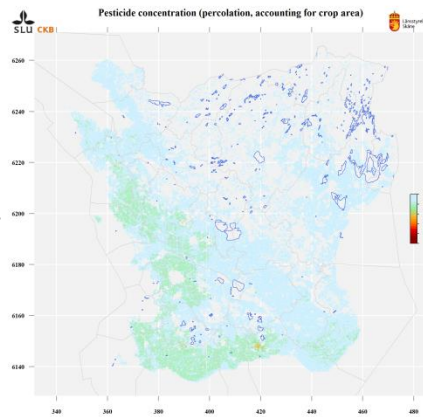
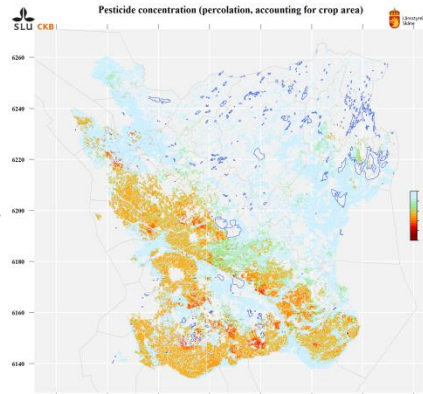
MACRO-SE workflow: example for groundwater

Edge of the field
Autumn

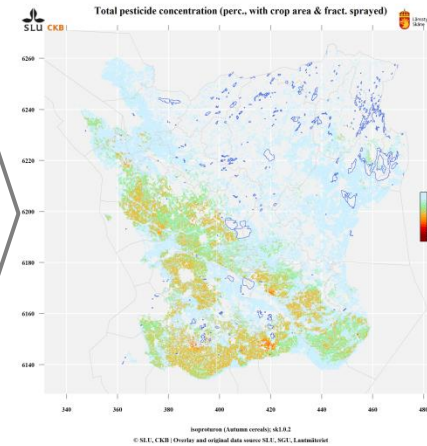


Spring

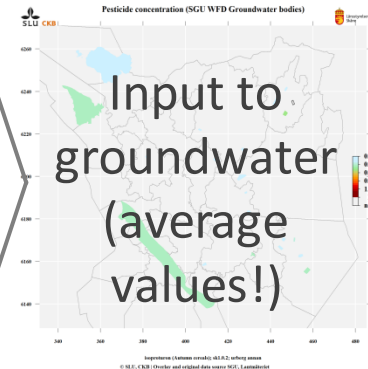
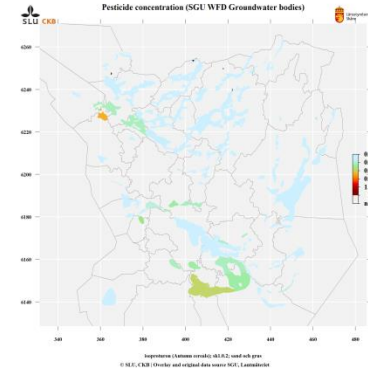
+ crop area



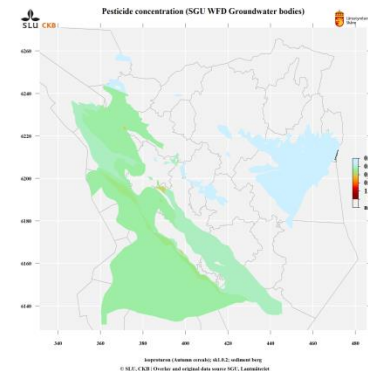
sum + fraction
sprayed



Note:
2 meters
depth!



Input to
groundwater
(average
values!)



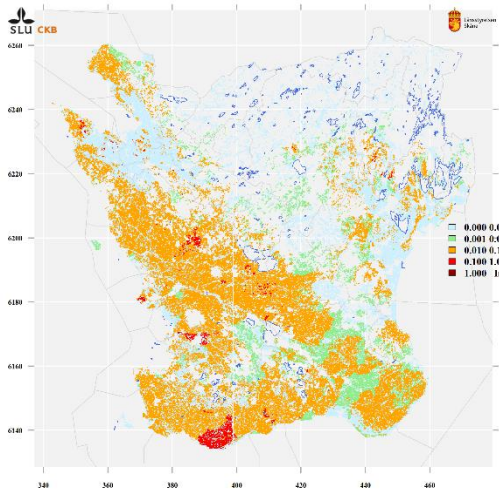
Monitoring: 128 water wells (georeferenced), in Scania, with one or several measurement, for one or several pesticides. Long term.

Substance	Type	Crop	Spray-season	Dose g/ha	% sprayed
bentazone	H	Peas and beans	Spr: 7 – 21 jun	435	95
Isoproturon	H	Win. cereals	Spr: 14 – 28 apr	555	4
Isoproturon	H	Win. cereals	Aut: 9 – 23 oct	494	23
MCPA	H	Spr. cereals	Spr: 19 may – 02 jun	500	63
MCPA	H	Win. cereals	Spr: 6 – 20 may	850	9
metazachlor	H	Spr. rapes	Spr: 15 – 29 may	950	20
metazachlor	H	Win. Rapes	Aut: 2 – 16 sep	888	62
quinmerac	H	Sugar beets	Spr: 11 – 25 may	150	1
quinmerac	H	Win. Rapes	Spr: 26 aug – 9 sep	250	56
metribuzin	H	Potatoes	Spr: 24 may - 7 jun	245	91

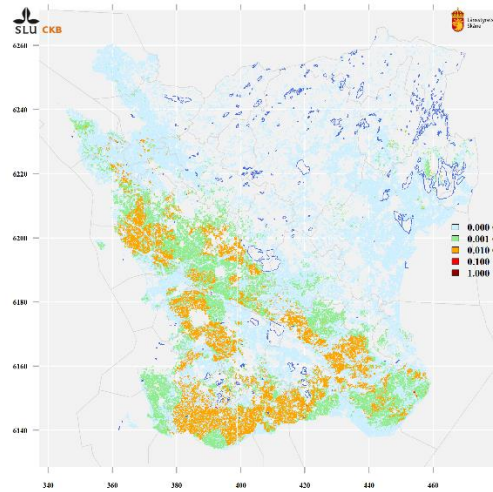
Statistics from 2 CKB monitoring catchments

Note: isoproturon & metazachlor: not re-registered (in 2014 and 2015, resp.)

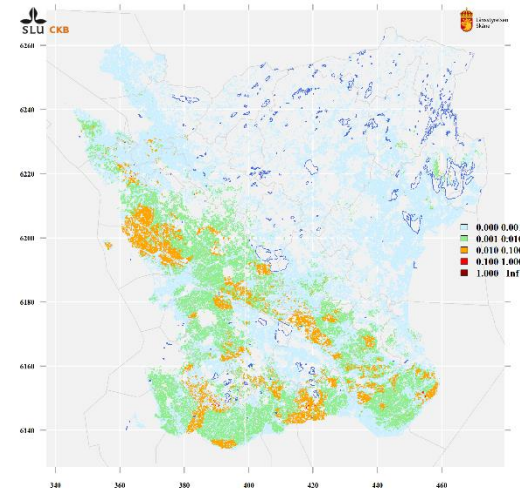
bentazone



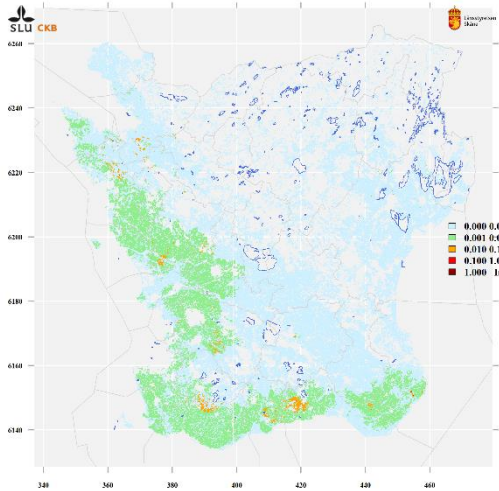
isoproturon (2)



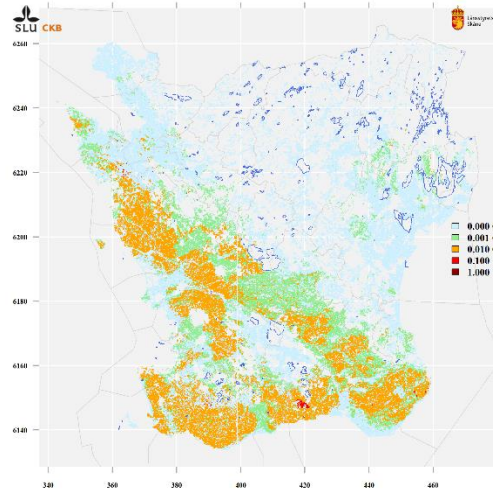
quinmerac (2)



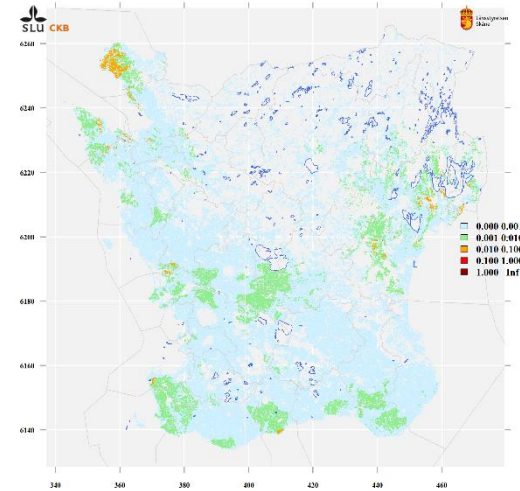
MCPA (2)



metazachlor (2)



metribuzin

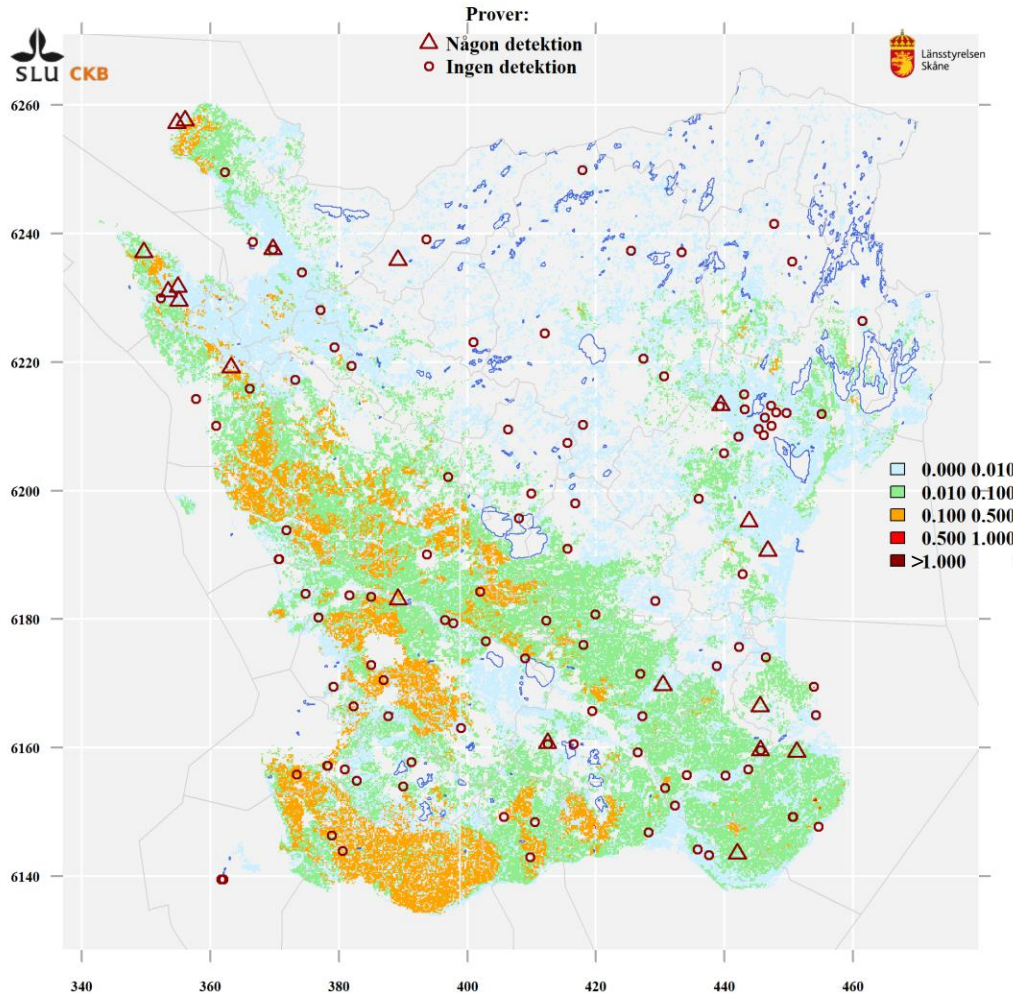


Results: confusion matrix (correct / incorrect classif.)

	Measured value	Simulated value		%tage correct clas.
		Detected	N. detected	
bentazon	Detected	6 (5%)	10 (8%)	
	N. detected	31 (24%)	81 (63%)	68%
isoproturon	Detected	0 (0%)	4 (3%)	
	N. detected	6 (5%)	118 (92%)	92%
kvinmerak	Detected	1 (1%)	1 (1%)	
	N. detected	4 (3%)	122 (95%)	96%
MCPA	Detected	0 (0%)	3 (3%)	
	N. detected	0 (0%)	95 (97%)	97%
metazaklor	Detected	0 (0%)	2 (2%)	
	N. detected	14 (11%)	112 (88%)	88%
metribuzin	Detected	0 (0%)	2 (2%)	
	N. detected	2 (2%)	93 (96%)	96%

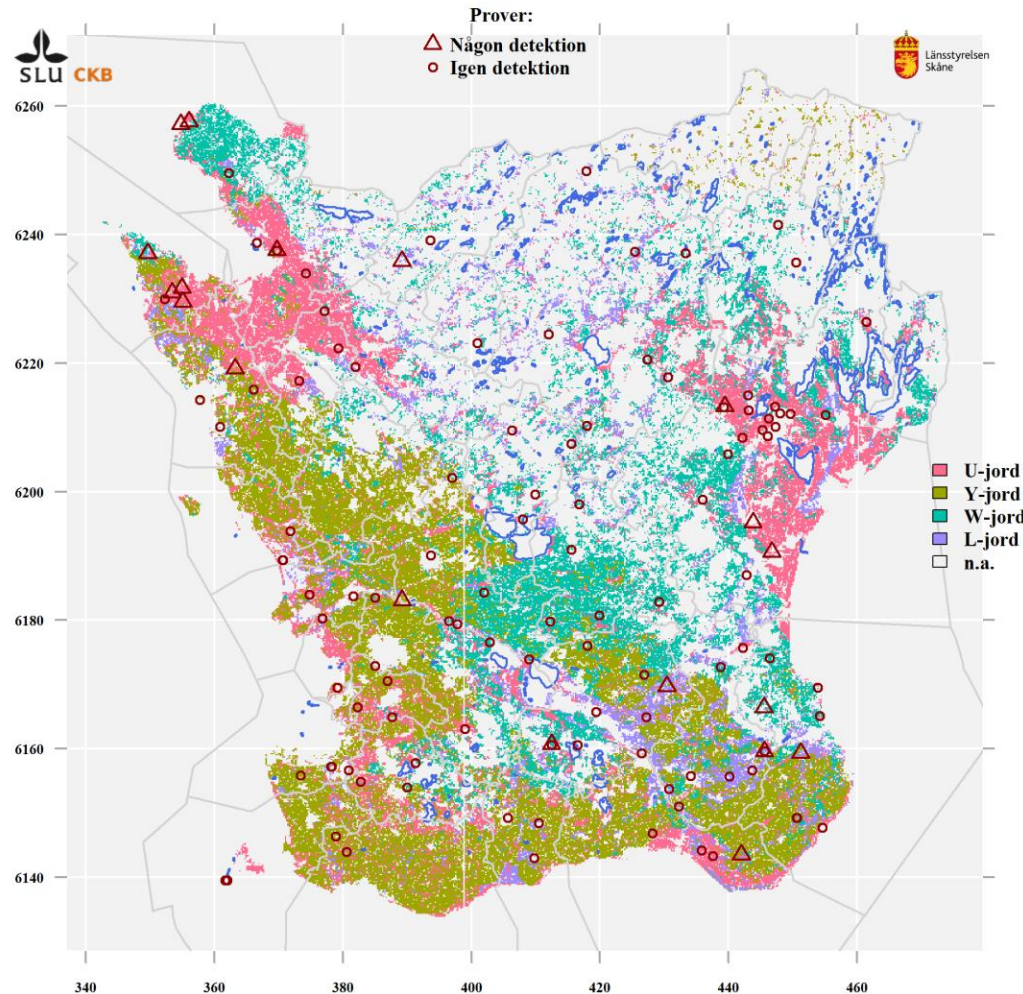
Green: correct classification; Red: incorrect classification.

→ High percentage of correct classification (also) because **low detection frequency** in observations: only 29/128 detections!



Cumulated concentration [$\mu\text{g}/\text{L}$] of all simulated herbicides at 2 m depth.

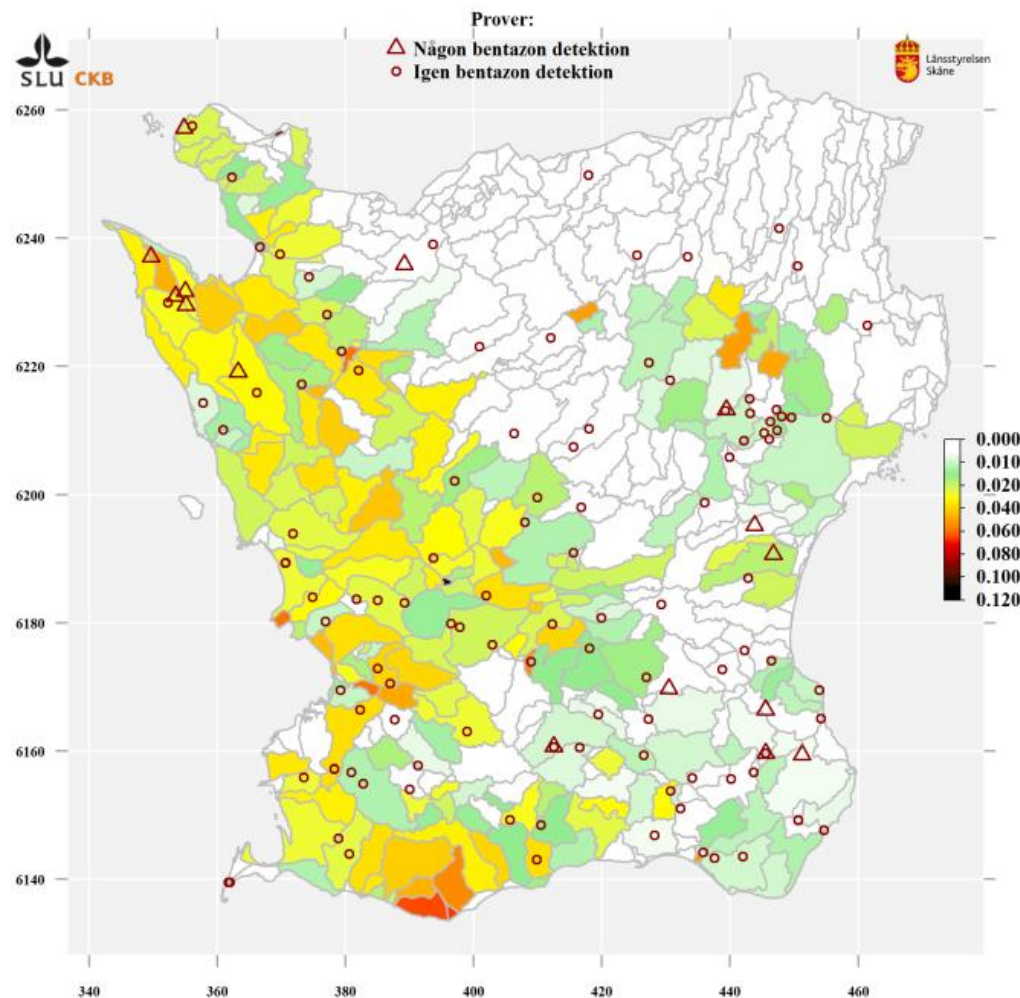
Red circles indicate wells where no pesticide was detected, triangles indicate wells with at least one detection.



Map of hydrological classes for arable soils. “U” (pinkish) indicates soils with losses to drainage only. “L” (purple) indicates losses to groundwater only. “Y” and “W” (yellowish; blue-green) indicate losses to both.

Red circles indicate wells where no pesticide was detected, triangles indicate wells with at least one detection.

→ Several detections in area where we don't predict flow to groundwater. Map error or **lateral flow towards the groundwater?**



Map of peas and beans area, as a fraction of the arable area in the catchment (average-statistics over 9 years).

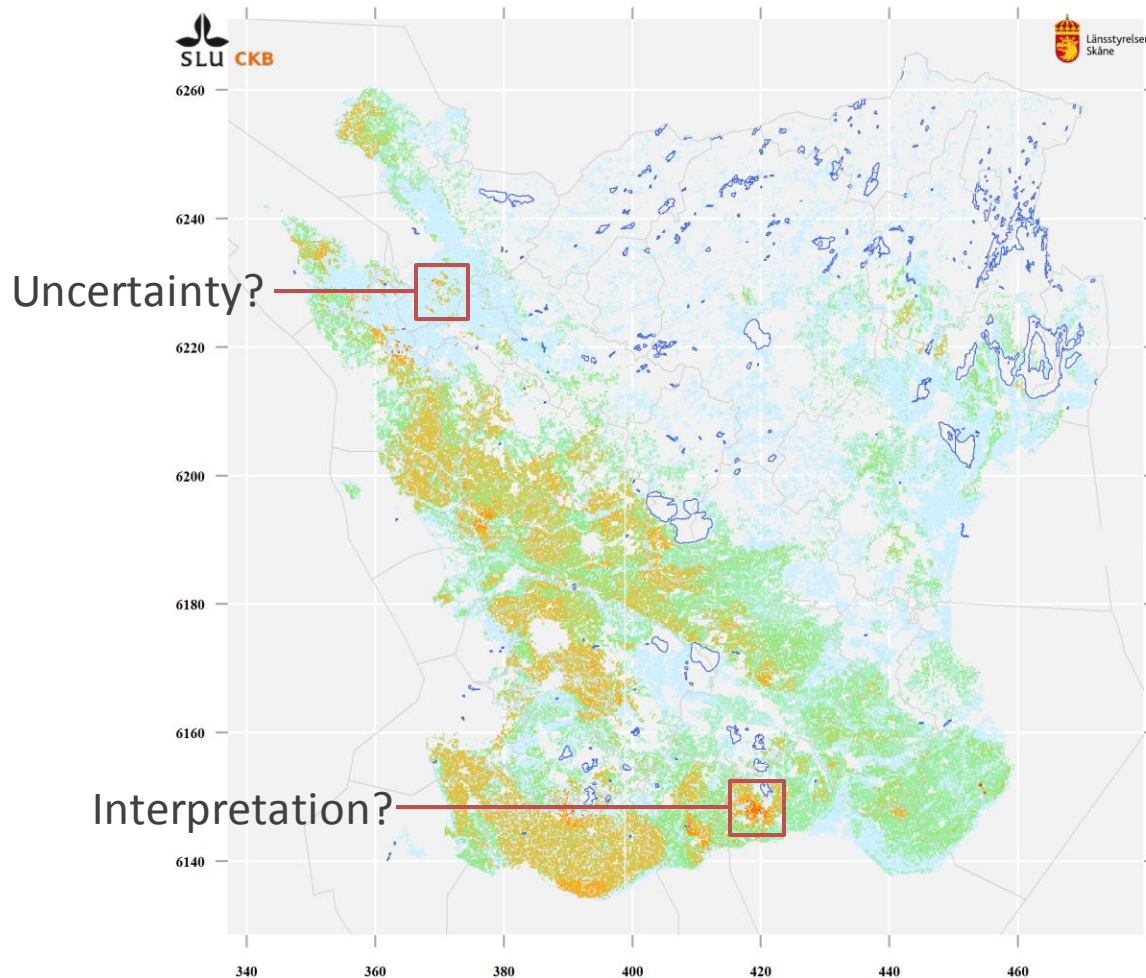
Red circles indicate wells where no bentazone was detected, triangles indicate wells with at least one detection.

→ Several detections in area where we don't observe a high fraction of peas or beans in the arable area. **Historical** or other usage?

Some consideration regarding the evaluation of regional models:

- Risk assessment models are **validated for a certain purpose**.
- Regional scale models are **not just numerical models**.
 1. Numerical model: collection of solutions.
 2. Model of the environment: collection of maps and databases.
 3. Parameter estimation routines.
- **Observations**, “ground-truth”, are **scarce and heterogeneous** (normalisation and aggregation needed?).
- Methodology for **matching simulation results and observations**:
 1. Aggregating and censoring simulated values.
 2. Evaluating (*or not*) ‘influence area’ around groundwater wells.
- Forecast vs history matching (historical maps & stats needed too).
- The regulatory framework, and practices, have changed.

→ **It's complicated!** We can discriminate leachable from non-leachable compounds (*Steffens et al. 2015*). But how accurate the results are is difficult to say (*blame the models, historical issues, ...?*)



- >> Need to define acceptable model usage.
- >> Model as a complex indicator?
- >> Combine with other sources of information?

Thank you for you attention!



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Water Management

SGU
Sveriges geologiska undersökning
Geological Survey of Sweden



Gustaf Boström, Julien Moeys, Nicholas Jarvis, Mikaela Gönczi och Jenny Kreuger, 2015. Riskkartering av bekämpningsmedel i Skånes Grundvatten - Simuleringar med MACRO-SE. CKB rapport 2015:1. (*in Swedish*).

<http://www.slu.se/sv/centrumbildningar-och-projekt/kompetenscentrum-for-kemiska-bekampningsmedel/publikationer/rapporter-fran-ckb/>

Gustaf Boström, 2014. Risk kartering av bekämpningsmedel i skånska ytvatten. Länsstyrelsen i Skåne län. Rapport 2014:5. (*in Swedish – surface water*).

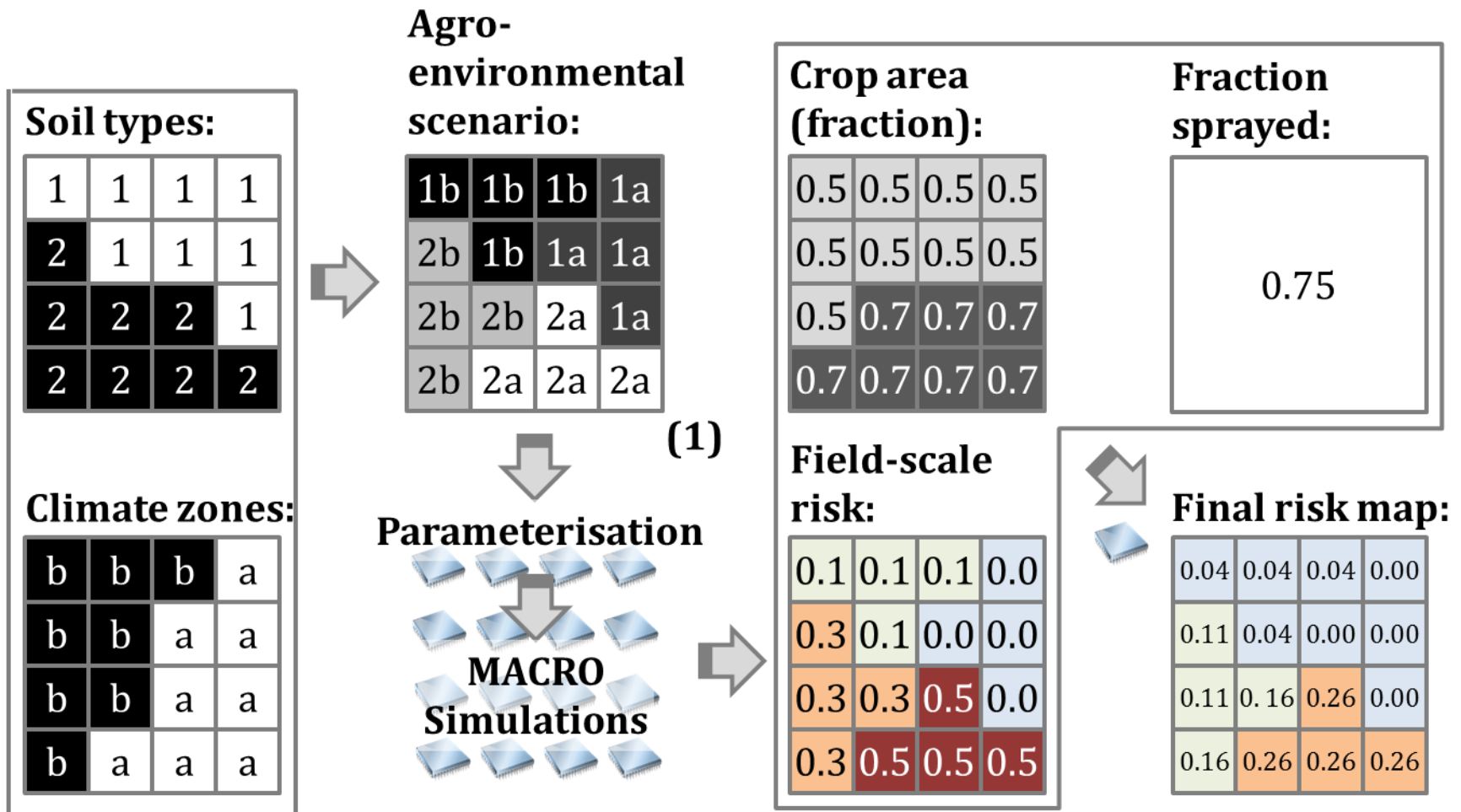
http://www.lansstyrelsen.se/skane/SiteCollectionDocuments/Sv/publikationer/2014/Riskkartering_bekampningsmede_skanska_ytvatten.pdf

Karin Steffens, Nicholas Jarvis, Elisabet Lewan, Bodil Lindström, Jenny Kreuger, Erik Kjellström, Julien Moeys, 2015. Direct and indirect effects of climate change on herbicide leaching—A regional scale assessment in Sweden. *Science of The Total Environment* 514, 239-249.

<http://dx.doi.org/10.1016/j.scitotenv.2014.12.049>

RPD. 2014. Regionala pesticiddatabasen. <http://www.slu.se/vaxtskyddsmedel>

Supplementary materials



+ additional aggregation & post-processing

Components of a regional scale pesticide fate model:

- **Numerical model of pesticide fate** (here MACRO, 1D).
 - Different processes → different sub-models.
 - More or less mechanistic or empirical.
 - More or less dynamic.
- **Model of the environment** (maps and databases).
 - Several variables (soil, climate, crop, pesticide).
 - **Discretised** time and space. Various resolutions, scales and levels of **aggregation**.
 - Note: Too few studies on its impact on model ‘validity’.
- **Parameter estimation routines**.
 - Continuous vs class-based? → Testing regional models = testing a local model-setup.
 - Fitted statistics vs expert judgment?
 - Sensitive parameter?

Observations (“ground truth”) against which models are tested are **also heterogeneous**:

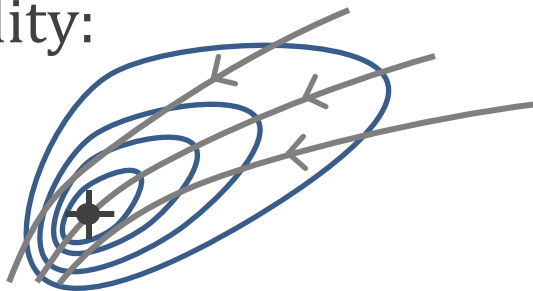
- Sampling design & frequency? Purpose?
- Different operators, laboratories, authorities.
- Observation replica: how to aggregate? min, max, average?
- Different detection limits (variable in time?). How to normalise?
- Different types of aquifer, different streams.
- Different depth (for groundwater).
- Poorly known historical pesticide usage crop area & groundwater ‘age’ (simplest: time stationarity is assumed)

Processing observations and simulated concentrations to make them comparable:

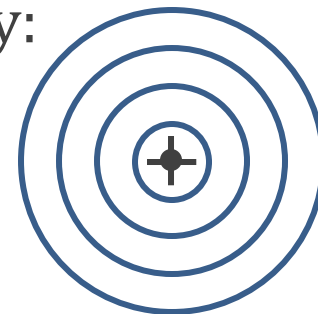
- Selecting a common detection limit (global, well-specific? ...).
- Censoring simulations (under detection \rightarrow non-detect).
- Re-censoring observations (if necessary).

Problem of the (unknown) groundwater influence area:

Reality:



Study:



Surface water: catchment outlet (estimation) vs observation point in the catchment; Only one sampling season.

