

# Using a Bayesian network model to predict effects of pesticides on aquatic community endpoints in a rice field – a southern European case study

➤ We propose a probabilistic approach to risk estimation incorporating temporal and spatial variability in risk estimation of pesticides using a Bayesian Network

➤ We use the Bayesian network as a meta-model that links the inputs and outputs for a process-based exposure model (RICEWQ) and a probabilistic case-based effect model (PERPEST).

## Background

- In the Mediterranean, climate change (CC) may affect the fate, transport and distribution of pesticides in aquatic environment (Noyes et al., 2009)
- Probabilistic approaches are recommended for risk assessment to account for uncertainty in pesticide exposure under future scenarios (Carriger & Newman, 2012; Mentzel et al., 2021)
- The goal of this study was to develop a model that offers a transparent way to estimate effects of pesticides on various aquatic taxonomic and functional (community) level.

## Approach

We developed a Bayesian network (BN) as a meta-model using outputs from two prediction models:

- RICE Water Quality (RICEWQ) model** - simulates chemical mass balance and aquatic exposure in rice paddies (Waterbourne Environmental Inc, 2011; Karpouzias & Capri, 2006), and
- Predicts the Ecological Risks of PESTicides (PERPEST) model** - simulates pesticide effect to various taxonomic groups using an existing database for micro and mesocosm studies (Van den Brink et al., 2002)

Concept model:

- RICEWQ was run for all scenarios (climate condition - 2008, 2050, & 2100, and dosage applied - baseline & baseline+50%), for each a exposure concentration distribution was derived (Fig. 1).
- The effect concentration and prior probabilities for each of the taxonomic groups was determined by the PERPEST model (Fig. 2 & Fig. 3).
- BNs are constructed for three different pesticide types commonly used in the study area “Albufera national park” near Valencia (Spain): azoxystrobin (fungicide), MCPA (herbicide), & acetamiprid (insecticide) (Fig. 3).

## Results

The BN model can derive outputs for the effect on each of the pesticide specific taxonomic groups. The displayed example shows the effect on insects by acetamiprid (Fig. 4) :

- It can be expected that insects are affected with a probability of 22 %, and slightly affected with a probability of 18 % (Fig. 4a).
- The likelihood of an effect on insects is true with approx. 30% and false with approx. 70% (Fig. 4b).

An assumption can also be made for the effect on functional groups (Fig. 5). We can compare the effect of acetamiprid on the different taxonomic groups (Fig. 5a):

- Macroinvertebrate are predicted not to be affected with a likelihood of almost 100 %.
- Insects, macro- and microcrustaceans, have higher probability to be affected with a likelihood of 25-30% .
- The probability for there being an effect on any of the taxonomic groups of the functional group of invertebrates is false with approx. 25 % and true with approx. 75%.

Furthermore, the functional groups of the different pesticide types can be compared (Fig. 6):

- It can be observed that the likelihood for there to be an effect on the plant community is lowest for acetamiprid with approx. 98%, and
- highest for MCPA with approx. 60%.
- Invertebrates have highest probability to be affected by azoxystrobin and lowest by acetamiprid (under climate conditions for 2008).

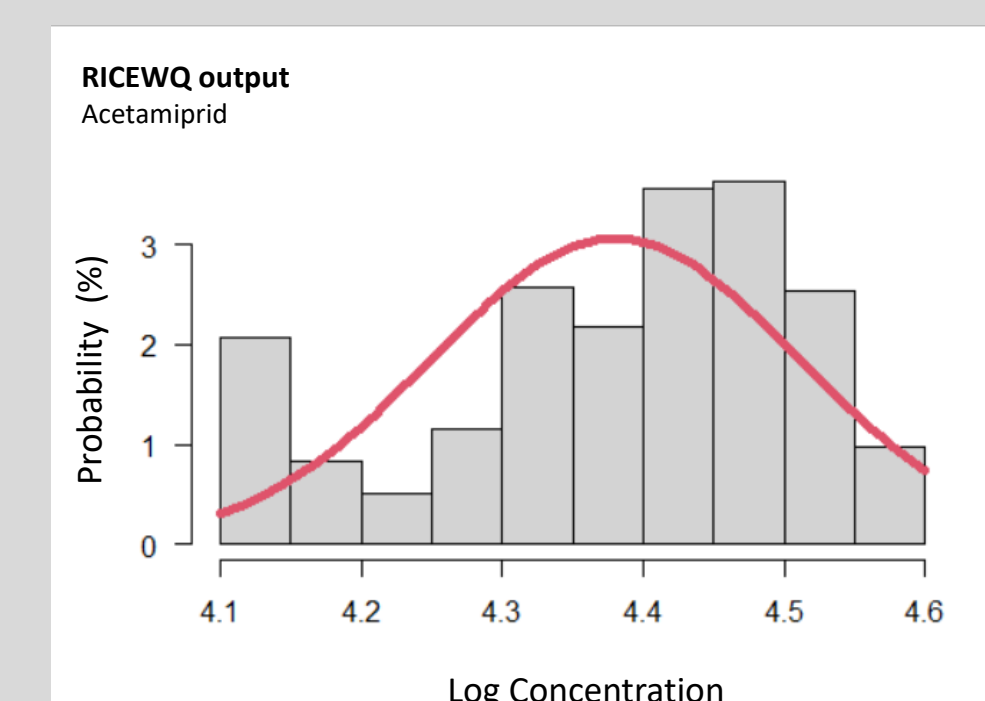
## Future perspectives

- Future efforts will aim to incorporate more scenarios such as other crop types, application patterns and climate projections to derive a more realistic pesticide risk assessment.
- We aim to carry out an effect assessment of the intentional mixtures to address cumulative risk of complex environmental mixtures.

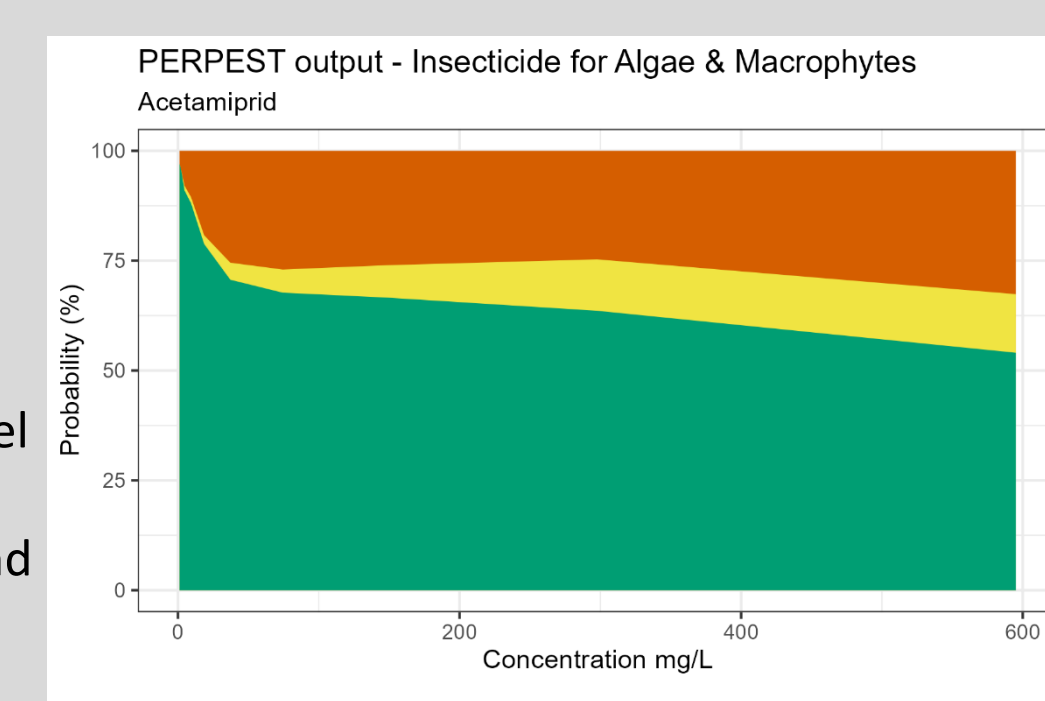
## References

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- Carriger, J.F. & Newman, M.C., 2012. <https://doi.org/10.1002/ieam.268>
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- Van den Brink, et al.(2002). <https://doi.org/10.1002/etc.5620211132>

## Bayesian network input information

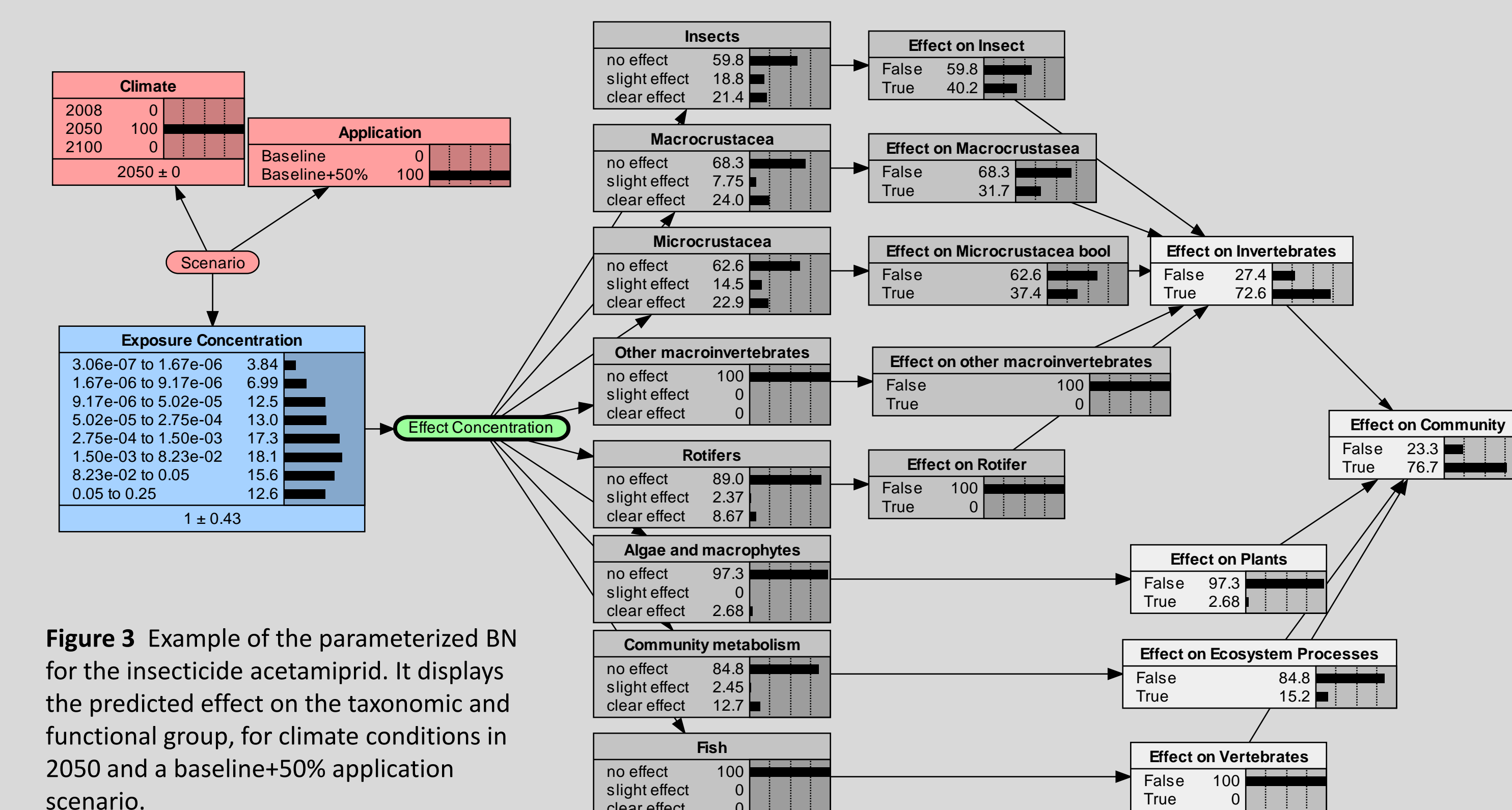


**Figure 1** Example RICEWQ model output, for each scenario a distribution is derived and used as input for the exposure module of the BN.



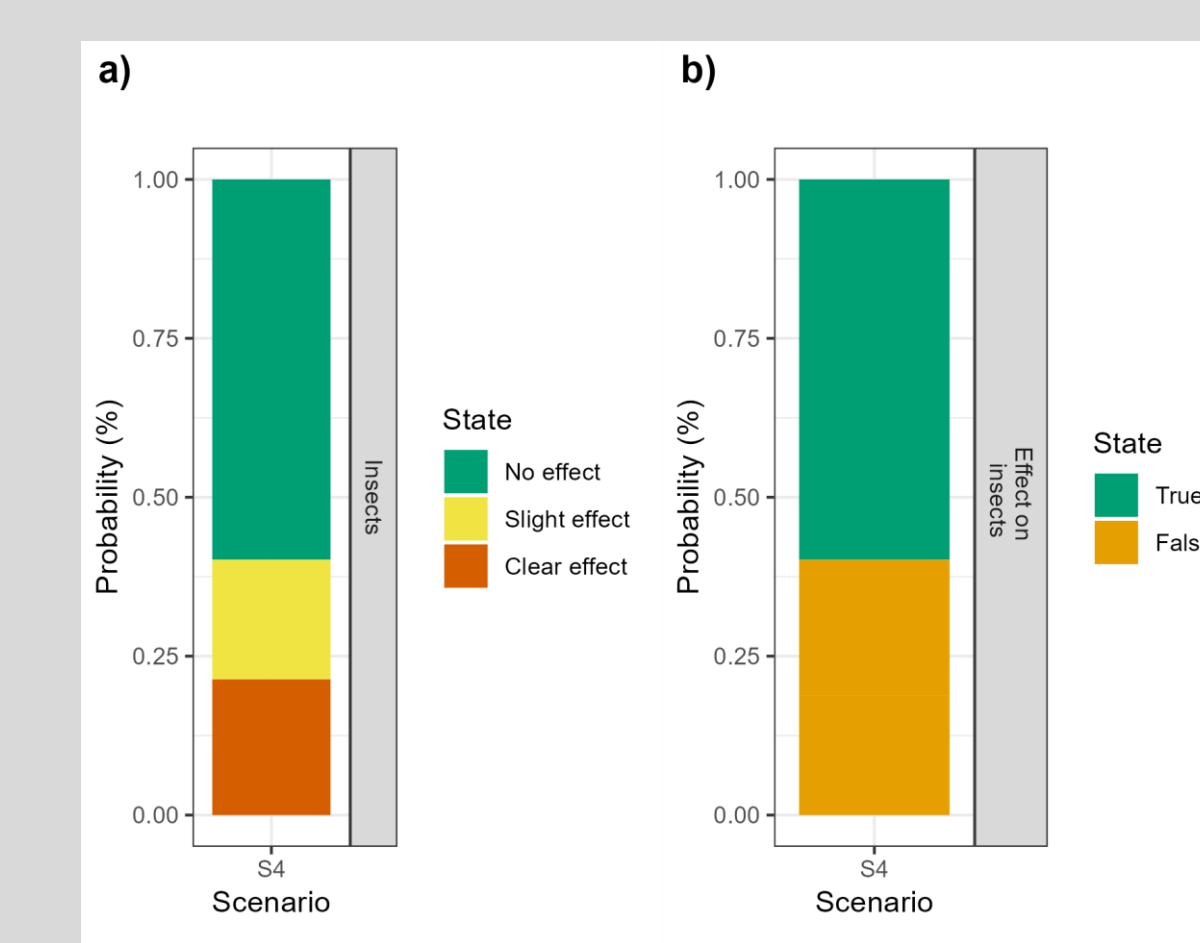
**Figure 2** Example PERPEST model output, for each taxonomic group a gradient was derived and used as input for the effect module of the BN.

## Parameterized model for an insecticide



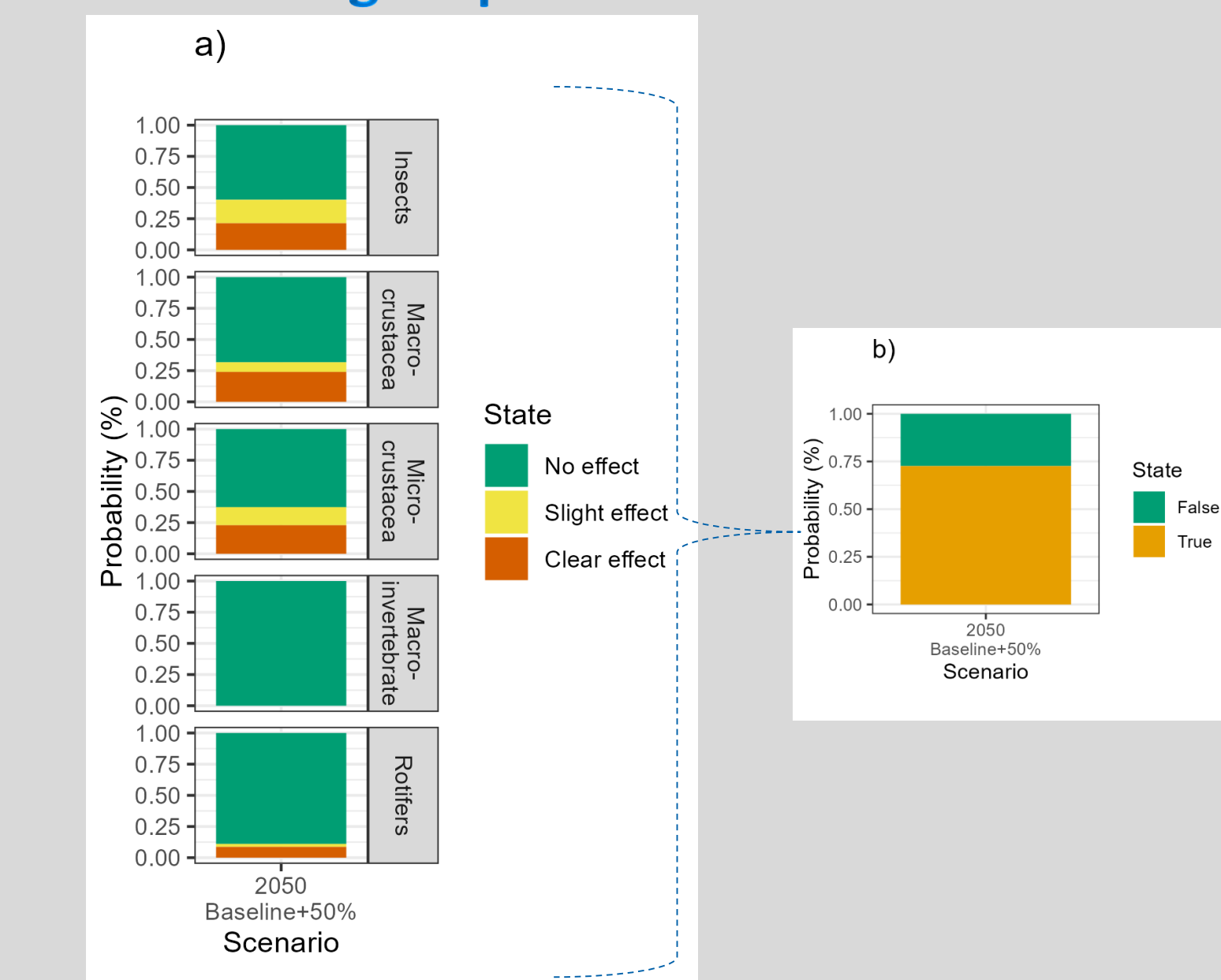
**Figure 3** Example of the parameterized BN for the insecticide acetamiprid. It displays the predicted effect on the taxonomic and functional group, for climate conditions in 2050 and a baseline+50% application scenario.

## Aggregation of the effect on taxonomic group



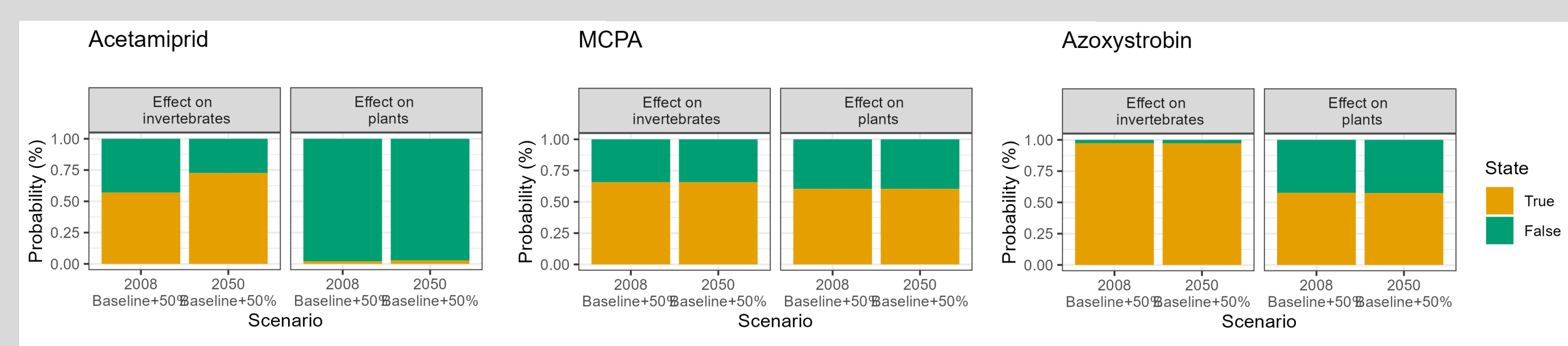
**Figure 4** Example of the predicted effect on insects by acetamiprid for a specific scenario (a) and summarizing boolean node output displaying whether or not an affect of the pesticide can be assumed (b), for climate condition in 2050 and a baseline+50% application.

## Aggregation of the effect on taxonomic to functional group



**Figure 5** Example of the predicted effect of the insecticide on the taxonomic groups (a), that are considered in the functional group “Invertebrates” (b), for climate conditions in 2050 and baseline+50% application.

## Comparison of the effect on functional groups /communities



**Figure 6** Examples of the predicted effect on the functional groups invertebrates and plants for the three selected pesticide (azoxystrobin, acetamiprid and MCPA), for climate conditions in 2008 and 2050 and a baseline+50 application

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