



A statistical methodology to downscale aggregated land use data

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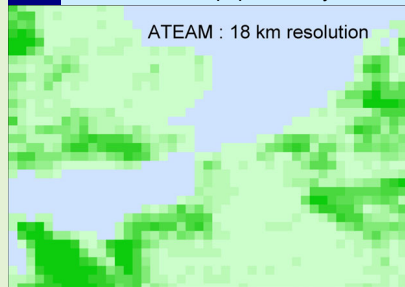
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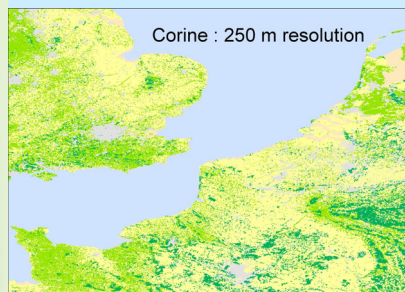
Downscaling – why?

Introduction

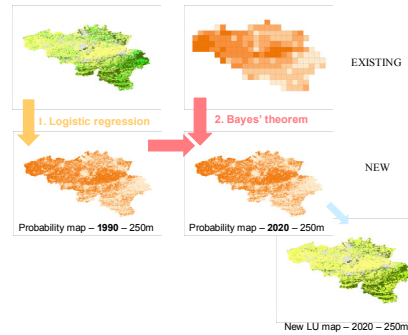
- Scenario and past Land Use (LU) data often at coarse resolution - reflecting global scale processes acting over large areas
- Fine resolution LU data is needed for ecological studies (e.g. population dynamics, biodiversity, carbon stocks...)
- In particular, there is a need for a fine resolution LU model over a large extent (the Svalbard migratory geese flyway) to model bird population dynamics



DOWNSCALING

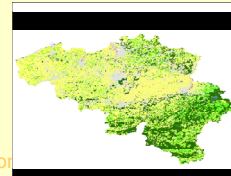


Downscaling – how?



1. Studying present LU patterns to derive probability maps of LU presence using statistical techniques

- Binomial and multinomial logistic regressions studying the presence/absence of each LU within each cell based on environmental variables and neighbourhood variables



Main conclusion from logistic regression

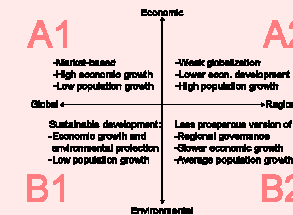
While the study of environmental variables provides insight on the factors that have led to the current LU patterns in Belgium, the **best statistical fit is achieved with a purely autoregressive model** (i.e. including only neighbourhood variables), for all LU studied, at the resolution of CORINE.

Consequently:

- Probability maps of land use presence can be derived using one LU dataset only (i.e. no ancillary data needed)
- A multinomial logit model can be applied
- This procedure is replicable over a wide area
- The resulting probability maps can serve as a baseline to downscale aggregated LU data

2. A method to update these probability maps based on Bayes' theorem and existing coarse scale LU scenario data

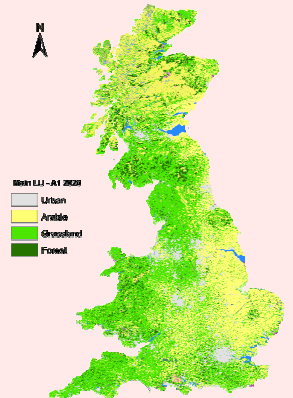
The ATEAM LUC scenario dataset: 4 scenarios based on the IPCC-SRES Framework – 18km resolution – 3 time steps (2020-2050-2080)



The multinomial logit model giving initial probability maps:
 $p(c|N_i) = \frac{\exp(\alpha_i N_i)}{\sum_{j=1}^k \exp(\alpha_j N_j)}$ $i = 1, \dots, k$
 Where:
 - $p(c|N_i)$ is the conditional probability that a cell i would take LU c among the set of k possible LU classes and given N_i
 - N_i is the neighbourhood variable
 From Bayes' theorem – the updating rule:
 $p(c|N_i) \propto p(c|N_i) p(c|c)$
 Updated conditional probabilities at time t' | Initial (time t) marginal probabilities (i.e. given by the CORINE dataset)
 Marginal probabilities at time t' (i.e. given by the ATEAM LU frequencies)

Results:

- For each cell LU with highest conditional probability is represented
- No border effects (between ATEAM cells) can be seen
- LU frequencies given by ATEAM are respected
- LU patterns are correctly represented despite increased clustering as a consequence of using neighbourhood variables



Conclusion

The statistical downscaling procedure is efficient and simple to implement (routine programmed in MATLAB):

- Only one baseline LU dataset is needed (i.e. Corine)
- Set to downscale the ATEAM scenarios over the whole of Europe15
- It could also downscale other aggregated LU data e.g. past statistics at the NUTS level

