

## Article (refereed)

---

De Vries, W.; Wamelink, G.W.W.; van Dobben, H.; Kros, J.; Reinds, G.J.; Mol-Dijkstra, J.P.; **Smart, S.M.; Evans, C.D.; Rowe, E.C.**; Belyazid, S.; Sverdrup, H.U.; van Hinsberg, A.; Posch, M.; Hettelingh, J-P.; Spranger, T.; Bobbink, R.. 2010 Use of dynamic soil-vegetation models to assess impacts of nitrogen deposition on plant species composition: an overview. *Ecological Applications*, 20 (1). 69-79. [10.1890/08-1019.1](https://doi.org/10.1890/08-1019.1)

Copyright by the Ecological Society of America

This version available <http://nora.nerc.ac.uk/8737/>

NERC has developed NORA to enable users to access research outputs wholly or partially funded by NERC. Copyright and other rights for material on this site are retained by the authors and/or other rights owners. Users should read the terms and conditions of use of this material at <http://nora.nerc.ac.uk/policies.html#access>

**This document is the author's final manuscript version of the journal article, incorporating any revisions agreed during the peer review process. Some differences between this and the publisher's version remain. You are advised to consult the publisher's version if you wish to cite from this article.**

[www.esajournals.org](http://www.esajournals.org)

Contact CEH NORA team at  
[noraceh@ceh.ac.uk](mailto:noraceh@ceh.ac.uk)

1 Running Head. – modelling nitrogen deposition impacts

2  
3 Use of dynamic soil-vegetation models to assess impacts of nitrogen deposition on plant species  
4 composition and to estimate critical loads: an overview

5  
6 Wim de Vries<sup>1</sup>, Wieger Wamelink<sup>1</sup>, Han van Dobben<sup>1</sup>, Hans Kros<sup>1</sup>, Gert Jan Reinds<sup>1</sup>, Janet Mol-  
7 Dijkstra<sup>1</sup>, Simon Smart<sup>2</sup>, Chris Evans<sup>2</sup>, Ed Rowe<sup>2</sup>, Salim Belyazid<sup>3</sup>, Harald Sverdrup<sup>3</sup>, Arjen van  
8 Hinsberg<sup>4</sup>, Maximilian Posch<sup>4</sup>, Jean-Paul Hettelingh<sup>4</sup>, Till Spranger<sup>5</sup> and Roland Bobbink<sup>6</sup>

9  
10 <sup>1</sup> Alterra, Wageningen University and Research Centre, P.O. Box 47, 6700 AA Wageningen, The  
11 Netherlands

12 <sup>2</sup> Centre for Ecology and Hydrology, Orton Building, Deiniol Road, Bangor LL57 2UP, UK

13 <sup>3</sup> Lund University, Department of Chemical Engineering, Box 124, S-22100, Lund, Sweden

14 <sup>4</sup> Netherlands Environmental Assessment Agency (PBL), PO Box 303, 3720 AH Bilthoven, The  
15 Netherlands

16 <sup>5</sup> German Federal Environment Agency (UBA), Wörlitzer Platz 1, 06844 Dessau, Germany

17 <sup>6</sup> Research Centre B-WARE, Radboud University Nijmegen, PO Box 9010, 6500 GL Nijmegen

1 ***Abstract***

2 Field observations and experimental data of effects of nitrogen (N) deposition on plant species  
3 diversity have been used to derive empirical critical N loads for various ecosystems. The great  
4 advantage of such as approach is the inclusion of field evidence, but there are also restrictions,  
5 such as the absence of explicit criteria regarding significant effects on the vegetation, and the  
6 impossibility to predict future impacts when N deposition changes. Model approaches can account  
7 for this. In this paper, we review the possibilities of static and dynamic multi-species models in  
8 combination with dynamic soil - vegetation models to (i) predict plant species composition as a  
9 function of atmospheric N deposition and (ii) calculate critical N loads in relation to a prescribed  
10 protection level of the species composition. The similarities between the models are presented, but  
11 also several important differences, including the use of different indicators for N and acidity and  
12 the prediction of individual plant species versus plant communities. A summary of the strengths  
13 and weaknesses of the various models, including their validation status, is given. Furthermore,  
14 examples are given of critical load calculations with the model chains and their comparison with  
15 empirical critical N loads. We show that linked biogeochemistry-biodiversity models for N have  
16 potential for applications to support European policy to reduce N input, but the definition of  
17 damage thresholds for terrestrial biodiversity represents a major challenge. There is a also a clear  
18 need for further testing and validation of the models against long-term monitoring or long-term  
19 experimental datasets and against large-scale survey data. This requires a focused data collection in  
20 Europe, combing vegetation descriptions with variables affecting the species diversity, such as soil  
21 acidity, nutrient status and water availability. Finally, there is a need for adaptation and upscaling of  
22 the models beyond the regions for which dose-response relationships have been parameterised, to  
23 make them generally applicable.

24  
25 **Key words: soil-vegetation models, model validation, nitrogen deposition, critical loads,**  
26 **plant species composition, biodiversity, terrestrial ecosystems, plant communities**

1    **1    Introduction**

2

3    *Impacts of nitrogen deposition on plant species composition*

4    During the past two decades the reduction of sulfur (S) emissions and the persistence of a high N  
5    pressure on terrestrial and aquatic ecosystems shifted attention from effects of S-deposition and  
6    acidification towards effects of N-deposition and eutrophication. In Europe, N is the most  
7    important air pollutant affecting plant species diversity. Evidence suggests that increasing N  
8    availability often causes an overall decline in plant species diversity (Tilman, 1987; Bobbink et al.,  
9    1998) even at long-term low-N inputs (Clark and Tilman, 2008). In some cases, especially under  
10    very nutrient poor conditions, however, an increase in plant species diversity has been observed  
11    due to the expansion of nitrophilic species (Emmett, 2007). Effects of N deposition, either in the  
12    form of ammonia (NH<sub>3</sub>), ammonium (NH<sub>4</sub>), nitrogen oxide (NO<sub>x</sub>) or nitrate (NO<sub>3</sub>), are now  
13    recognised in nearly all oligotrophic and mesotrophic (semi-)natural ecosystems. An overview of  
14    effects on plant species diversity, including impacts on mosses, lichens and mycorrhizae, in forests,  
15    grasslands, heathlands, oligotrophic wetlands (mire, bog and fen), and coastal habitats, mainly in  
16    Europe, with related empirical critical N loads, is presented in Achermann and Bobbink (2003).  
17    More recently , an overview of effects of N deposition on a global scale, distinguishing Arctic and  
18    Alpine ecosystems, boreal forests, temperate forests and tropical forests, heathlands and  
19    grasslands, Mediterranean vegetation, tropical savannas and arid vegetation (desert and semi-  
20    desert) is presented in Bobbink et al. (2009).

21

22    *Critical loads and their use in policy making*

23    In order to set standards and targets for emission reduction policy, the concept of critical load has  
24    been developed. The general definition of a critical load is ‘a quantitative estimate of an exposure  
25    to one or more pollutants below which significant harmful effects on specified sensitive elements

1 of the environment do not occur according to present knowledge' (Nilsson and Grennfelt, 1988).  
2 Critical loads are defined for specific combinations of pollutants, effects and receptors. They  
3 reflect spatially variable sensitivities, thus leading to regionally defined emission-reduction needs.  
4 The concept is most commonly used in connection with the atmospheric deposition of S and N  
5 (acidification and eutrophication) and in these cases the critical load is the maximum flux (in kg N  
6  $\text{ha}^{-1}\cdot\text{yr}^{-1}$  or  $\text{keq H ha}^{-1}\cdot\text{yr}^{-1}$ ) that an ecosystem is able to sustain. Since 1994, critical loads for N and  
7 acidity have played an important role in European air pollution abatement (Hettelingh et al., 2001;  
8 Spranger et al., 2008). European critical load exceedances, calculated and mapped using the latest  
9 methods and datasets for critical loads, deposition, and emission scenarios, are presented in  
10 Hettelingh et al. (2007) and Slootweg et al. (2007). Results show that the area where critical loads  
11 of acidity are exceeded and will continue to decrease, even if no new legislation is implemented,  
12 while high ( $> 10 \text{ kg N ha}^{-1}\cdot\text{y}^{-1}$ ) exceedances for critical N loads remain widespread especially in  
13 north-western European areas dominated by ammonia emissions.

14  
15 Exceedances of the critical load of acidity and N have been used in European pollution abatement  
16 policy for defining emission-reduction targets, i.e. in the UNECE Convention on Long-range  
17 Transboundary Air Pollution (LRTAP Convention) and the European Union (National Emission  
18 Ceilings Directive, 2001; EC, 2005). Integrated assessment models (e.g. the RAINS model, Amann  
19 et al., 1999) use these data and methods in scenario analyses. The exceedance of critical loads of N  
20 is also used as an indicator for risk to biodiversity by the European Environment Agency (EEA,  
21 2007). The general organization of effects-based European air pollution policies is described in  
22 [www.unece.org/env/lrtap](http://www.unece.org/env/lrtap).

### 24 ***Critical load approaches for nitrogen used in policy making***

25 Critical loads for N, as used in European environmental policy, are estimated empirically or by  
26 simulation. The empirical approach uses experimental fields or 'mesocosms' where various levels

1 of N fertiliser have been added. In that case, the critical load is determined as the level of  
2 deposition where a decrease in biodiversity just starts to occur (Achermann and Bobbink, 2003).  
3 The great advantage of the use of empirical critical N loads, based on N-addition experiments, is  
4 that there is field and/or experimental evidence for a relationship between N deposition and  
5 effects. There are, however, several requirements for empirical critical N loads to be reliable,  
6 including: (i) long-term experiments (preferably > 4 year) to show long-term effects and (ii) studies  
7 in low background-N deposition areas to ensure that major effects have not already occurred. In  
8 high background N deposition, high-N additions may be needed before (additional) effects show  
9 up and N removal experiments should be used instead (Emmett, 2007). The empirical critical N  
10 loads are thus mainly based on long-term field-addition experiments and mesocosm studies in low-  
11 N-deposition areas with realistic N loads ( $<100 \text{ kg N}\cdot\text{ha}^{-1}\cdot\text{yr}^{-1}$ ). However, since such experiments  
12 are time- and labour-intensive, results are only available for a rather limited group of broadly  
13 defined ecosystems. The reliability (range) in empirical critical loads, being the level of the lowest  
14 N addition where effects occur, is mainly influenced by the chosen interval in N additions and by  
15 an uncertain background N deposition that has to be added to this level (Sutton et al., 2003). An  
16 aspect that also limits a strict comparability is the lack of fixed criteria regarding significant effects  
17 for which a critical load is derived (as e.g. in critical limits for toxic substances), and it is also  
18 impossible to predict future impacts when N deposition and other environmental conditions  
19 change simultaneously.

20  
21 The model-based critical load approach, used in European environmental policy making, is based  
22 on an ecosystem mass balance, which balances the deposition load to an ecosystem with its long-  
23 term capacity to buffer this input or to remove it from the system without harmful effects inside  
24 or beyond the system (Hettelingh et al., 2001; Spranger et al., 2008). The harmful effects are  
25 defined in terms of critical limits above which a negative effect is assumed to occur. An overview  
26 of those limits is given in De Vries et al. (2007). The model calculates a critical load as the

1 deposition level leading to soil conditions that are just tolerated by a given ecosystem. The model-  
2 based critical load approach as used up to now, however, is by definition based on the sustainable  
3 state of a given ecosystem that is invariable in time (steady-state) and excludes non-permanent  
4 buffering processes such as temporary N release and retention and cation exchange. This long-  
5 term critical load may therefore differ from the atmospheric deposition actually affecting the  
6 ecosystem, since ecosystems differ in sensitivity to perturbation depending on their current state  
7 and recent history. Nitrogen deposition thresholds may vary during forest stand development, as  
8 for example shown also with dynamic-model approaches (Tietema et al., 2002). These differences  
9 form the core of resilience and sustainability theories. This aspect has not been included in model  
10 based critical loads until now, and their steady state concept implies that the exceedance of such  
11 critical loads does not allow a prognosis of ecosystem status at any point in time.

12

13 ***Need for dynamic model approaches and aim of this paper***

14 Both empirical critical loads and steady-state models do not allow prediction of the temporal  
15 response of ecosystems to deposition scenarios, for example, in terms of impacts on plant species  
16 diversity. This requires the use of the dynamic integrated soil-vegetation models. Such models can  
17 also be used to assess critical loads, while accounting for differences in sensitivity to perturbation  
18 depending on their current state and recent history. In the context of ecological theory, N  
19 deposition is a form of disturbance; i.e. an external influence that moves the system away from its  
20 stable state (Gunderson, 2000). If such a disturbance is not too large, the ecosystem has the ability  
21 to return to its former state ('resilience', see e.g. Gunderson, 2000); in the case of N deposition,  
22 this might happen by incorporation of N in refractory soil organic material or N leaching.  
23 However, if the disturbance is larger or extends over a prolonged period, the system may move  
24 towards an alternative stable state, from which it will not be able to return to its former state  
25 without a new external influence (Ludwig et al., 1997). N deposition will ultimately stimulate the  
26 growth of more productive species, that usually produce more easily degradable litter and reach a

1 greater height, thus increasing both the deposition itself, and the amount of N cycling in the  
2 system. In these terms, the critical load is the highest deposition that will not cause an ecosystem  
3 to shift to an alternative (more productive, usually species-poorer) state. In this alternative state the  
4 quantity of N cycling through the system will be much larger than in its original state and, because  
5 of the tight cycling of N- a return to the former, N-poor state will only be possible by physically  
6 removing the excess N even if deposition decreases (Wamelink et al., 2008). Critical-load  
7 assessments including such aspects can be only be included in dynamic model approaches,  
8 simulating delays in damage due to buffering processes and delays in recovery due to time take to  
9 restore soils to its original state.

10  
11 In this overview, we describe the possibilities of multi-species models in combination with  
12 dynamic soil - vegetation models to (i) predict plant species composition or diversity as a function  
13 of atmospheric N deposition and (ii) calculate critical N loads in relation to an acceptable plant  
14 species diversity change. First, we present the two main model approaches that are presently  
15 employed in Europe: (i) a simple soil acidification and nutrient cycling model (SMART2 or  
16 MAGIC) combined with field-based empirical relationships with plant species responses (MOVE,  
17 GBMOVE or NTM) and (ii) a detailed, mechanistic, soil-acidification and nutrient-cycling model  
18 (ForSAFE) with a process-based description of plant species responses (VEG). An explanation of  
19 the model acronyms and main features appears in Table 4 and 5. The overview includes a  
20 description of each modelling approach, followed by application examples illustrating the model  
21 validation status and the use of the models in critical-load assessments. In a final section we  
22 discuss the potential of linked biogeochemistry-biodiversity models to support European pollution  
23 abatement policy, including (i) strengths and weaknesses of the two major model approaches, (ii)  
24 the use of different indicators for N availability, (iii) the validation status of each model, (iv) the  
25 potential of the models to assess critical loads, (v) the need for additional field surveys and (vi)  
26 relevant extensions to the modelling approaches.



1  
2  
3  
4  
5  
6  
7  
8  
9  
10  
11  
12  
13  
14  
15  
16  
17  
18  
19  
20  
21  
22  
23  
24  
25

**2     *Modelling approach***

Integrated soil-vegetation models are used at present in Europe to predict plant species composition as a function of atmospheric deposition of N and acidity, as illustrated in Figure 1. The principle of such model-based approaches is that a dynamic soil model (SMART2, MAGIC, (For)SAFE) predicts the changes in water and nutrient status (e.g. as N availability or C/N ratio) and soil acidity (e.g. as soil pH or base saturation) in response to atmospheric deposition, whereas a statistical model (NTM, MOVE) or a process-based model (SUMO, VEG) predicts vegetation succession or changes in plant species composition in response to the changes in water, nutrient and acidity status, using plant species-specific information on habitat preferences. Such coupled models can be used in an inverse way to determine critical loads. In that case, critical values for abiotic factors (e.g. N availability or soil pH ) have to be empirically determined per vegetation type, either directly or from information per species (step 1) and subsequently used in the coupled soil model (step 2) to back-calculate the critical N and acid loads.

Simulation of critical loads according to the above principle was carried out in the Netherlands (Van Dobben et al., 2006), where (i) the critical pH and N availability per vegetation type (association) were determined on the basis of a large set of vegetation relevés (vegetation description of a small plot), often , (cf. MOVE model, Latour and Reiling, 1993), and (ii) the dynamic soil model SMART2 (Kros et al., 1995) was used to calculate the critical loads at which the above critical limits were not exceeded in the long term. Other models use critical limits for other abiotic variables, such as the C/N ratio in the GBMOVE model (Smart et al., 2003) and the BERN model (Schlutow & Hübener, (2004). or the soil N, P, BC availability, soil moisture, pH, light and grazing pressure in the ForSafe-VEG model (Belyazid et al., 2006; Sverdrup et al., 2007).

1 Below, we discuss the two major model approaches used at present, i.e. (i) dynamic soil models  
2 linked with empirical static-vegetation models (the SMART2(-SUMO)-MOVE/NTM and  
3 MAGIC(-SUMO)-GBMOVE model chains) and (ii) a deterministic, dynamic ecosystem model  
4 integrating hydrology, growth, biogeochemical cycles and vegetation dynamics. (ForSafe-VEG)

5

## 6 **2.1 Linked dynamic soil models with empirical static vegetation models**

7 Two major comparable model chains of dynamic soil models linked with static vegetation models  
8 are SMART2(-SUMO)-MOVE/NTM and MAGIC(-SUMO)-GBMOVE, which are developed  
9 and used in the Netherlands (NL) and the United Kingdom (UK), respectively. The model chains  
10 consist of: (i) the soil models SMART2 (NL) or MAGIC (UK) that simulate the cycling of  
11 nutrients in the soil and predict soil acidity and N availability, (ii) the succession model SUMO that  
12 simulates the cycling of nutrients (N, P, K, Ca, Mg) in the plant-soil system, including biomass  
13 growth through photosynthesis and biomass removal through management and (iii) multiple  
14 regression equations between species presence and abiotic factors that define the realized niches of  
15 a substantial proportion of the vascular flora of each country (and in the UK also bryophytes)  
16 (MOVE in NL and GBMOVE in UK), or of plant communities (NTM in NL). In the Dutch  
17 MOVE and NTM models, abiotic factors are groundwater table, soil pH and N availability,  
18 derived via Ellenberg indicator values, whereas a version of the UK GBMOVE model also  
19 includes three climatic variables, i.e. the minimum January temperature, maximum July  
20 temperature and precipitation. Ellenberg's indicator values are classes of plants species with  
21 similar ecological niches, which are derived for about 2720 central-European vascular plants  
22 (Ellenberg et al., 1992). Ellenberg derived values for the following ecological factors: light ( $E_L$ ),  
23 temperature ( $E_T$ ), continentality ( $E_K$ ), moisture ( $E_F$ ), soil pH ( $E_R$ ), nutrients/nitrogen ( $E_N$ ), and  
24 others (salinity, heavy metal resistance) ( $E_{\text{sonst}}$ ). SMART2 can be used both in its original, dynamic  
25 form, allowing the calculation of target loads, and as a steady state version, allowing the calculation  
26 of steady state critical loads, as used in policy making. The use of SUMO is optional in both model

1 chains. Changes in species composition are modelled by first simulating the effects of N and S  
2 deposition on soil conditions followed by simulating the impacts of changed soil conditions on  
3 species composition.

4

#### 5 **Modelling the relation between atmospheric deposition and soil conditions**

6 *SMART2* and *MAGIC* are dynamic, process-oriented models that predict changes in soil  
7 chemistry at a given level of N and S deposition. Changes of N and S deposition on soil variables  
8 such as pH, C/N ratio or N availability by *SMART2* (Kros et al., 1995; Kros, 2002) or *MAGIC*  
9 (Cosby et al., 2001). Both *SMART2* and *MAGIC* include the major hydrological and  
10 biogeochemical processes in the soil compartment, to calculate the long-term effects of  
11 atmospheric deposition of  $\text{NO}_x$ ,  $\text{NH}_y$ ,  $\text{SO}_x$  and base cations ( $\text{BC}^{2+}$ ) on soil-solution chemistry, and  
12 in case of *MAGIC* also the surface-water chemistry. The models have a high degree of process  
13 aggregation to minimize their data requirements, which allows application on a regional scale.  
14 They consists of a set of mass-balance equations, describing the soil input-output relationships,  
15 and a set of equations describing the rate-limited and equilibrium soil processes. Apart from pH,  
16 the models predict changes in aluminium, base cation, ammonium, nitrate and sulfate  
17 concentrations in the soil solution and solid phase. Key parameters include the input and output  
18 fluxes of base cations and strong acid anions, the soil cation exchange capacity, and the fraction of  
19 this capacity that is occupied by Ca, Mg, Na and K ions. Nitrogen dynamics in *MAGIC* are based  
20 on empirical relationships between net N retention and the current C/N ratio in the soil, whereas  
21 in *SMART2* litterfall, mineralization, root uptake and immobilization are modelled explicitly. Both  
22 *SMART2* and *MAGIC* have an internal simplified growth module which enable the models to  
23 calculate nutrient cycling detached from SUMO. The detached version of *SMART2* has been used  
24 for the calculation of critical loads and target loads. For the computation of target loads, a  
25 procedure was developed to iteratively run *SMART2* until the N and S deposition used, lead to the  
26 critical pH or N availability for a given vegetation type. Furthermore, a steady-state version of

1 SMART2 has been developed that computes the critical N and acid load that in steady state leads  
2 to a given combination of N availability and pH. A complete overview of these models, and the  
3 differences between SMART2 model and MAGIC can be found in the various references  
4 mentioned and in De Vries et al. (2007).

5  
6 *SUMO* is a process-based model that simulates biomass growth under given soil, climate and  
7 management conditions (Wamelink, 2007). The basis of the model is a maximum growth that is  
8 being reduced by a series of linear and non-linear reduction factors to constrain growth. These  
9 reduction factors convey the effect of changes in the availability of light, N, phosphorous, water  
10 and temperature. SUMO distinguishes five functional plant types (climax trees, pioneer trees,  
11 shrubs, dwarf shrubs and herbs) that compete for light and nutrients. Their competitive balance is  
12 governed by vegetation structure, i.e. canopy height and biomass of roots and leaves per functional  
13 type. Management is simulated as biomass removal by mowing, grazing, cutting or turf stripping.  
14 The accumulation of biomass in the five functional types determines the succession stage (e.g.  
15 pioneer, grassland, heathland, forest). SUMO can be coupled to niche models. e.g. NTM or  
16 MOVE, through vegetation structure and soil chemical conditions (pH and nutrient availability)  
17 simulated by SMART2 or MAGIC. For these soil models litterfall is a crucial input term that is  
18 generated by SUMO. In each time step there is feedback between SMART2 or MAGIC and  
19 SUMO; the models exchange information about N and P, litterfall, and vegetation structure.

20

### 21 **Relationships between plant species occurrence and soil conditions**

22 *MOVE and NTM:* The models MOVE (Latour and Reiling, 1993; Latour et al., 1994) and NTM  
23 (Schouwenberg et al., 2000; Wamelink et al., 2003a) are based on response curves in which the  
24 probability of plant species (MOVE) or plant community (NTM) occurrence is determined by  
25 vegetation structure and the abiotic site conditions of groundwater table, soil pH and N  
26 availability. The probability of occurrence is a simple bell-shaped curve derived for 914 species by

1 2nd-order logistic regression based on presence/absence, representing species occurrence along an  
2 environmental gradient. These relationships are based on the realised niche, i.e. they account for  
3 competitive exclusion, rather than responses of the species in isolation. Since MOVE and NTM  
4 focus on more than one abiotic factor, the curves are multi-dimensional. The probabilities of  
5 occurrence are determined per vegetation type relative to soil pH and N availability, estimated on  
6 the basis of Ellenberg's (1992) indicator values for N ( $E_N$ ) and acidity ( $E_R$ ). In the critical-load  
7 approach of Van Dobben et al. (2006), the 20- and 80-percentiles of these frequency distributions  
8 were used as the critical limits, i.e. the range between these percentiles was considered as the  
9 optimal range for each vegetation type.

10  
11 The above frequency distributions were determined in a database of 160 000 vegetation relevés  
12 that were labelled in terms of vegetation type (Schaminée et al., 1989), originally developed for a  
13 revision of the Dutch classification of plant communities. In a separate procedure, the Ellenberg  
14 values (which are on an arbitrary scale) were translated into physical units that can be used as input  
15 to dynamic models. This translation requires a training set where vegetation and soil conditions (at  
16 least pH and N availability) have been recorded simultaneously. In the past few years, much effort  
17 has been put into the collection of such data (Wamelink et al., 2007, see [www.abiotic.wur.nl](http://www.abiotic.wur.nl)).

18 Various translation functions between Ellenberg values and physical units have been derived (e.g.  
19 (Alkemade et al., 1996; Ertsen et al., 1998; Wamelink et al., 2002; Van Dobben et al., 2006). Those  
20 of Van Dobben et al. (2006) for example run:

21

$$22 \text{ pH} = 3.1 + 0.53 \cdot E_R \quad R^2 = 0.43; n = 3630 \quad (1)$$

23

$$24 N_{av} = 6.19 + 0.64 \cdot E_N + c \cdot \text{vegtype} \quad R^2 = 0.24; n = 6911 \quad (2)$$

25

1 Where  $N_{av} = N$  availability ( $\text{kMol}\cdot\text{ha}^{-1}\cdot\text{yr}^{-1}$ ) and the constants  $c$  per vegetation type are -1.182 for  
 2 grass, -1.898 for heath, -0.274 for coniferous forests and 0 for deciduous forest

3  
 4 *GBMOVE*: As with *MOVE*, multiple logistic regression was used to construct empirical equations  
 5 that predict habitat suitability for higher and lower plants representative of British plant  
 6 communities, based on their abundance along key environmental gradients as recorded by  
 7 extensive relevé data (e.g. Roy et al., 2000). Each equation consists of regression coefficients that  
 8 apply to either four or seven explanatory variables, depending on whether climate variables  
 9 (minimum January temperature, maximum July temperature and precipitation) are included or not.  
 10 Important interaction terms are also included. These quantify the extent to which a species'  
 11 response on one gradient is conditioned by another gradient (e.g. Pakeman et al., 2008). The data  
 12 used to derive each equation were assembled from a variety of sources as described in De Vries et  
 13 al. (2007) and covered more than 40 000 vegetation relevés. The regression was based on  
 14 presence/absence data for each plant species in each plot paired with values of climatic variables  
 15 (derived from the plot's geographical position) and plot-averaged Ellenberg indicator values. The  
 16 final number of species having *GBMOVE* regression models is 327 for bryophytes and 803 for  
 17 vascular plants in non-coastal habitats (74 in coastal habitats).

18  
 19 As with *MOVE*, soil pH and soil C/N ratio are translated instantaneously into mean Ellenberg  $E_R$   
 20 and  $E_N$  values respectively, using paired soil measurements and mean Ellenberg values from the  
 21 Countryside Survey 1998 database (Smart et al., 2003). The limitations of the assumption that  
 22 species presence immediately changes in response to soil conditions are discussed in detail in  
 23 Section 5. Relationships thus obtained are:

24  
 25 
$$\ln(C/N) = 3.61 - 0.63 \cdot \ln E_N \quad R^2 = 0.62; n = 256 \quad (3)$$

26 
$$\text{pH} = 2.5 + 0.61 \cdot E_R \quad R^2 = 0.61; n = 256 \quad (4)$$

1

2 The mean Ellenberg  $E_R$  and  $E_N$  values per relevé are terms in the GBMOVE regression equations.  
3 At each time step the simulated values of soil C/N, soil pH, % soil moisture and cover-weighted  
4 canopy height are translated into Ellenberg units and put in the regression equation, resulting in  
5 predicted probability of species occurrence over time. Changes in soil pH and C/N ratio are  
6 predicted with the dynamic soil model MAGIC. Canopy height can be changed arbitrarily using  
7 pre-existing knowledge of the pace of succession in a particular location, or on a more process-  
8 linked basis, by the SUMO succession model. Climate variables can be changed to mimic  
9 expectations under different climate change scenarios. Likewise, soil moisture can also be changed  
10 to mimic drainage or drought.

11

## 12 **2.2 The integrated dynamic FORSAFE-VEG model**

13 The only fully integrated dynamic soil and plant species diversity (vegetation) model that is  
14 presently available in Europe is the ForSAFE-VEG model chain. This model chain, developed in  
15 Sweden, consists of: (i) the ForSAFE model, aimed at the dynamic simulation of changes in soil  
16 chemistry, soil organic matter, hydrology and tree biomass growth in relation to changes in  
17 environmental factors (Wallman et al., 2005); and (ii) the VEG submodel, which simulates changes  
18 in the composition of the ground vegetation in response to changes in biotic and abiotic factors  
19 such as light intensity at the forest floor, temperature, grazing pressure, soil moisture, soil pH and  
20 alkalinity in addition to competition between species based on height and root depth (Belyazid et  
21 al., 2006; Sverdrup et al., 2007). For each time step, defined by the resolution of the input data,  
22 ForSAFE simulates the changes in state variables in response to environmental changes  
23 (temperature and precipitation, atmospheric deposition, forest management). These state variables  
24 are read by the VEG module, where the occupancy strength is calculated for each plant group.  
25 The plant groups are defined by the user. The single occupancy strengths are then used to  
26 calculate the relative occupancy of each plant group.

1  
 2 If a stress factor would eliminate a certain species, the disappearance of this species will not be  
 3 instantaneous, but will happen with a delay, which depends on the lifespan of the species. Unlike  
 4 the model chains with MOVE and GBMOVR, this aspect is included in ForSAFE-VEG. The  
 5 change in occupancy of a specific plant group,  $dX/dt$ , depends on the actual occupancy of the  
 6 plant group ( $X$ ), the target occupancy (referred to as equilibrium occupancy  $X_{eq}$ ), and the specific  
 7 regeneration time of the plant group ( $\tau$ ) according to the following equation:

8

$$9 \quad \frac{dX}{dt} = \frac{1}{\tau} \cdot (X_{eq} - X) \quad (5)$$

10  
 11 The regeneration time  $\tau$  is related to the lifespan of a specific plant group. The life span depends  
 12 on site factors, such as drought. The equilibrium occupancy of a plant group  $i$ ,  $X_{eq,i}$  is the ratio  
 13 between the strength of the species under the specific environmental conditions and the sum of  
 14 the strengths of all present species according to the following equation:

15

$$16 \quad X_{eq,i} = \frac{S_i}{\sum_{j=1}^{j=\text{plantgroup}} S_j} \quad (6)$$

17  
 18 where  $S_i$  is the individual strength of the plant group  $i$ . The sum of plant group strengths is also  
 19 used as an indicator of the density of the ground cover, referred to as the mass index (MI). The  
 20 strength of each plant group is the product of the following drivers: (i) soil solution N  
 21 concentration ( $\text{mol.l}^{-1}$ ), (ii) soil solution phosphorus concentration ( $\text{mol.l}^{-1}$ ), (iii) soil acidity ( $[\text{H}^+]$ ,  
 22  $[\text{BC}^{2+}]$ ,  $[\text{Al}^{3+}]$  ( $\text{eq.l}^{-1}$ ), (iv) soil water content ( $\text{m}^3 \text{ water.m}^{-3} \text{ soil}$ ), (v) soil temperature ( $^{\circ}\text{C}$ ), (vi) light  
 23 reaching the ground ( $\mu\text{mol photon.m}^{-2}.\text{s}^{-1}$ ), (vii) grazing (moose units  $\text{km}^{-2}$ ), (viii) wind tatter and  
 24 wind chill damage, (ix) plant competition based on above-ground competition for light and below-



1 ground competition for water and nutrients and (x) air CO<sub>2</sub> concentration. ForSAFE-VEG thus  
2 simulates the ground vegetation occupancy based on the individual response of plant groups to  
3 these controlling factors. The effects are multiplicative and have the same weights in affecting the  
4 plant strength. . Each plant group represents various individual plant species, varying from less  
5 than ten up to several hundreds.

6  
7 For each plant group indicator that has been selected, response functions were parameterised for  
8 Sweden from published laboratory and field data, by approximations from empirical data, or by  
9 scaling the response with respect to other plant groups for which the response is known. Scaling  
10 of plant groups towards known responses was based on generic knowledge and expert opinions  
11 from Swedish plant ecologists. However, the basic shape of each response function does not vary  
12 between the plant groups. For example, all plant groups will respond positively to an increase in  
13 water availability in the soil up until a certain level where anaerobic conditions in the saturating soil  
14 may hinder the plant's growth. The distinction between the plant groups is the minimal water  
15 content required for survival, optimal water content for growth, and the point at which water  
16 becomes damaging. The individual response functions are described in detail in Belyazid (2006)  
17 and De Vries et al. (2007).

18

### 19 **3 *Model validation on changes in soil and vegetation data***

20

21 Results of both types of model chains were compared with measurements on changes in soil and  
22 vegetation data. The performance of the models was calculated by two measures that are often  
23 applied for this purpose; the Normalized Root Mean Square Error (NRMSE; eq. 7), and the  
24 Normalized Mean Error (NME; eq. 8) (Janssen and Heuberger, 1995).

25

$$\text{NRMSE} = \frac{\sqrt{\frac{1}{N} \sum_{i=1}^N (P_i - O_i)^2}}{\bar{O}} \quad (7)$$

$$\text{NME} = \frac{\bar{P} - \bar{O}}{N} \quad (8)$$

Where:

$P_i$  = predicted value: model output for N<sub>2</sub>O emissions

$O_i$  = observed value: field value for N<sub>2</sub>O emissions

$\bar{P}$  = average for the predicted values

$\bar{O}$  = average for the observed values

$N$  = number of observations

NRMSE describes the deviations between the measurements and the predictions in a quadratic way and is thus rather sensitive to extreme values. Optimally, it should be 0. The NME compares predictions and observations over the entire time-span, on an average basis. It expresses the bias in average values of model predictions and observations, and gives a rough indication of overestimation (NME>0) or underestimation (NME<0).

### 3.1 Linked dynamic soil models with static vegetation models

Below, examples are given of the validation of the SMART2/MAGIC-GBMOVE/NTM model chains with respect to soil chemical data (focus on MAGIC), aboveground biomass data (SUMO) and time series of observed species composition and species richness (focus on GBMOVE in combination with MAGIC).

1 *Validation of MAGIC on time series for soil chemical data:* Data from plot-scale N-manipulation studies  
2 in the UK have been used to test the ability of MAGIC to predict changes in soil C/N under  
3 different addition levels (Evans et al., 2006). For two sites with high-quality soil C and N data  
4 (Figure 2), the model successfully reproduced observed decreases in C/N under three treatments.  
5 These simulations incorporated an (observed) increase in C storage as a consequence of N  
6 deposition, which slowed down the rate of C/N change. The NME's derived were 0.0388 for the  
7 Ruabon site and 0.0813 for the Budworth site, indicating a slight overestimation of the predictions.  
8 The calculated NRMSE's were very low, i.e. 0.0065 and 0.014, respectively.

9  
10 MAGIC was also validated on data on C/N ratios for the Parkgrass experimental site at  
11 Rothamsted, which are available for a 100-year period. N removal was calculated by multiplying  
12 hay removal, for which accurate measurements are available, by the proportion of N in hay  
13 biomass. The uncertainty in N concentration in hay has a large effect on the net addition  
14 (deposition minus removal) and thus on the historic C/N trajectory (Table 1). Note that raising N  
15 deposition would have the same effect as reducing N removal.

16  
17 *Validation of SUMO on time series of above-ground biomass:* Biomass growth predicted by SUMO was  
18 validated using data collected at two unfertilized grassland sites, one in the Netherlands and one in  
19 the UK, using site specific historical deposition data. The first grassland site is situated near  
20 Wageningen in the Netherlands and is part of a long-term field experiment started in 1958 on  
21 former agricultural land (Elberse et al., 1983). The second grassland site is the Parkgrass  
22 experimental site at Rothamstead in the UK mentioned before. The Parkgrass site was mown  
23 twice a year and the harvested biomass was weighed and averaged over ten year periods. The  
24 experiment started in 1856 and still continues today. The trends in herbage yields are extensively  
25 described by Jenkinson et al. (1994). At the Dutch site, the measured biomass varies greatly  
26 between years due to yearly differences in rainfall and temperature. The simulated biomass does

1 not vary much and remains within the range of the measured biomass (Figure 3). The high  
2 measured biomass in the first year is probably caused by the former agricultural use of the field.  
3 Both the measured and the simulated biomass show a decrease over the years, due to the yearly  
4 biomass removal. The results for Rothamsted show that the harvested biomass is fairly well  
5 simulated by SUMO. The reduction in biomass harvest between 1850 and 1900, due to exhaustion  
6 of the soil, the stabilisation of the harvest when the effect of N deposition compensated for the  
7 exhaustion between 1900 and 1950, and the increase of the harvest later on due to the further  
8 increase in deposition since 1950 are simulated quite well. The effect of N deposition since  
9 approximately 1960 is underestimated, however (Figure 3 right). Overall, the NME of the  
10 Ossekampen was -0.012 and -0.0033 for Rothamsted, implying that on average, the predictions are  
11 almost equal to the observations. The values for the NMRSE are 0.290 and 0.195, respectively  
12 indicating a considerable deviation for defined years. SUMO was also validated on a heathland and  
13 a forest site in the Netherlands, as described in De Vries et al. (2007).

14  
15 *Validation of MAGIC-GBMOVE on time series of observed species composition and species richness: A*  
16 number of tests have been carried out to determine how successfully MAGIC-GBMOVE could  
17 reproduce the observed species composition in sampled plots. Observations were compared with  
18 predictions generated initially by populating a simulated set of plots with species conditioned on  
19 probability of occurrence values generated by GBMOVE and a Poisson distribution of mean  
20 species-richness values with proportional variance predicted by a separate General Linear Mixed  
21 Model using the same explanatory variables as GBMOVE. This statistical model is fully described  
22 in Smart et al. (2005). To assess the influence of uncertainty in the calibration equations relating  
23 soil properties to mean Ellenberg scores, predictions of species composition based on soil C/N  
24 and pH generated by MAGIC were compared with predictions based on observed mean Ellenberg  
25 scores. These comparisons were carried out for control plots at the long-term continuous Park  
26 Grass hay experiment at Rothamsted (unimproved neutral grassland) and the Hard Hills grazing

1 and burning experiment at Moorhouse National Nature Reserve (ombrogenous bog), described in  
2 Smart et al. (2005). The results for Rothamsted indicated that when mean Ellenberg scores based  
3 on observed species composition were used as input to GBMOVE, on average 67% of species  
4 observed were actually predicted. However, when predictions were based on MAGIC simulations  
5 of soil C/N and pH as input to GBMOVE, the percentage of all species correctly predicted  
6 decreased substantially (Figure 4 left). The main reason for the poor performance at Rothamsted  
7 appears to be that observed soil changes were inconsistent with observed vegetation changes. This  
8 is most likely due to sampling practices in the experimental plots, avoiding a thin mat of persistent  
9 litter that had developed in the O horizon over the course of the experiment. This resulted in C/N  
10 measurements indicating a higher fertility in the rooting zone than that encountered by at least  
11 some of the more shallow rooting species present. Also at Moorhouse, both the predictions from  
12 MAGIC linked to GBMOVE and the predictions based solely on observed mean Ellenberg values  
13 did not compare well with the species actually observed in the control plots, although the key  
14 dominants in the vegetation were predicted to be present by both models. (Figure 4 right). When  
15 predicted species lists for both GBMOVE and MAGIC+GBMOVE were examined, key absences  
16 included a range of bryophytes. Possibly, bryophytes are more responsive to direct deposition  
17 effects, and less so to changes in soil chemistry, making the approach less useful for lower plants.  
18

19 From this model comparison it was concluded that it is unlikely that both generally applicable yet  
20 highly accurate models can be developed, because of the dependence of current species  
21 composition on site-specific aspects of patch and wider landscape history. Because of this,  
22 probabilities of occurrence from GBMOVE are no longer used as expectations of species  
23 presence, but rather interpreted as indices of habitat suitability where target species ought to be  
24 able to persist and increase in population size in the absence of constraints to dispersal and  
25 establishment. The further validation work therefore focuses on comparing predicted trends

1 through time with observed changes, based on species present on-site and in the local species  
2 pool, rather than attempting to predict the entire species assemblage on a specific site.

3  
4 *Validation of MAGIC-GBMOVE on temporal changes among plant species:* At Moorhouse, observed  
5 versus predicted species changes over time were summarised as slope coefficients for each species  
6 in a linear regression line relating (i) observed abundance over the years and (ii) predicted habitat  
7 suitability from MAGIC and GBMOVE across the same time period. Despite considerable scatter  
8 there was a positive correlation between observed change in species frequency and predicted  
9 change in habitat suitability (Figure 5). A chi-square test of observed versus predicted directions of  
10 change was significant ( $p=0.016$ ). While the correlation between observed and predicted slopes  
11 was also significant, predicted rates of change covered a narrower range than observed species  
12 changes, which may be due to weather fluctuations or sampling errors.

13

### 14 3.2 The integrated dynamic FORSAFE-VEG model

15 The ForSAFE-VEG model was validated on changes in soil chemistry (Belyazid et al., 2006),  
16 standing wood biomass (Belyazid, 2006) and in ground vegetation cover (Sverdrup et al., 2007) at  
17 16 Swedish forest sites that are part of the ICP Forest level II monitoring network (International  
18 Co-operative Program on Assessment and Monitoring of Air Pollution Effects on Forests,  
19 <http://www.icp-forests.org>). At the sites, 42 plant groups and 9 tree seedling types have been  
20 identified. These plant groups were assumed to be potentially present throughout Sweden, but are  
21 only expected to manifest where environmental conditions are favourable. The sites cover a wide  
22 range of climatic conditions, soils, fire regimes, atmospheric deposition gradients and management  
23 histories. ForSAFE-VEG was used to simulate the changes in soil chemistry, hydrology and tree  
24 biomass according to these conditions, and the composition of the ground vegetation was  
25 subsequently derived. Atmospheric deposition data for  $\text{NO}_3^- + \text{NH}_4^+$  and  $\text{SO}_4^{2-}$  were derived on the  
26 basis of EMEP model estimates according to the 1999 LRTAP Gothenburg protocol (Schöpp et

1 al., 2003). The sites were subject to different histories of fire regimes, alterations between open  
2 fields and forest cover as well as different harvesting regimes depending on the location of each  
3 site.

4  
5 *Validation on soil chemical data.* Simulated soil organic matter contents showed a reasonable  
6 correlation between the measured and modelled values of soil organic carbon (C) and N at the 16  
7 study sites. The NME's derived were  $-79 \text{ g/m}^2$  for organic C and  $-3.04 \text{ g/m}^2$  for organic N,  
8 indicating a slight underestimation of the predictions. The calculated NRMSE's were quite high,  
9 i.e. 0.58 for both organic C and N.

10  
11 The model reconstructs the pH profiles at the 16 study sites quite well (Figure 6). The NME for  
12 the 16 sites varied from  $-0.056$  to  $+0.097$  while the NMRSE varied from 0.016 to 0.210, indicating  
13 an appropriate prediction of the average pH and a limited deviation with measurements at various  
14 depth. The model, however underestimates the acidity at the deeper soil layers (Figure 6). This  
15 inconsistency is probably due to the fact that the model considers only a limited amount of roots  
16 at the deep layers, thus underestimating uptake and the presence of organic matter and its  
17 decomposition. Also important for the ground-vegetation community is the soil base cation to  
18 aluminium ratio (BC/Al ratio). The variation in both the measured and modelled BC/Al ratios was  
19 large for most of the sites, but the correspondence between the model and the measurements was  
20 reasonably good. More information on the validation of soil organic C and N and of soil pH is  
21 given in Belyazid (2006) and Belyazid et al. (2006), respectively.

22  
23 *Validation of standing tree biomass and ground vegetation composition:* Predicted values for the ground  
24 occupancy of the 42 identified plant groups calculated with ForSAFE-VEG for the year 1995 were  
25 plotted against measurements from the same year to establish the validity of the model outputs at  
26 the 16 sites (Sverdrup et al., 2007). Results are presented for two representative sites, i.e. Brattfors

1 (Figure 7 left) and Svartberget (Figure 7, right). The model predicts fairly well the occupancy of the  
2 present vegetation groups. The NMRSE varied from 0.167 for Brattfors to 0.197 for Svartberget,  
3 while the NME is close to 0. Recently, the model has also been validated on two Swiss forest  
4 plots (Aeschau and Bachtel). A comparison of the model output for these test sites to ground  
5 vegetation assessment showed that only 40-55% of the species present at these two sites were also  
6 modelled.. The presence of major species, i.e. *Vaccinium myrtillus*, *Blechnum spicant*, *Dryopteris dilatata*,  
7 *Polytrichum formosum*, *Rubus fruticosus* and *Oxalis acetosella* were, however, forecasted correctly in both  
8 sites, for the latter two species even predicted with a correct estimate of the cover degree. The  
9 observed sensitive reaction of *Rubus fruticosus* cover to N deposition, was also predicted well  
10 ForSAFE-VEG. Finally, tree biomass was predicted well for the two Swiss test sites.

11

#### 12 **4 Model application: assessment of critical nitrogen loads**

13

##### 14 **4.1 Application of the SMART2-MOVE model for Dutch vegetation types**

15 To date, the MAGIC-GBMOVE model has not been applied in 'inverse mode' to estimate critical  
16 loads based on biodiversity targets. The SMART2-MOVE model, however, has been used in an  
17 inverse way to assess critical loads and target loads for major vegetation types in the Netherlands  
18 and to compare results with empirical critical N loads (Van Dobben et al., 2006). The empirical  
19 critical loads and the calculated critical loads correspond reasonably well (Table 2), i.e. their ranges  
20 usually overlap, although there is no significant correlation between the range midpoints of the  
21 two methods. The comparison of methods is slightly hampered by the fact that simulated values  
22 are determined for a very detailed typology (associations or sub-associations in the sense of  
23 Braun-Blanquet, 1964; Schaminée et al., 1995) while the empirical values were determined for so-  
24 called 'EUNIS' classes, where EUNIS stands for European Nature Information System (Davies  
25 and Moss, 2002). On average the midpoints of the empirical ranges are 3.4 kg N ha<sup>-1</sup>.y<sup>-1</sup> below the  
26 simulated range midpoints, and this difference is nearly significant (P≈0.07). There are various



1 reasons for the lower empirical values compared to the simulated ones, the most important  
2 probably being that the empirical critical loads tend to be based on the most sensitive components  
3 of an ecosystem, often under abiotic conditions that enhance sensitivity still further (cf.  
4 Achermann and Bobbink, 2003). In contrast, in the simulation approach all environmental  
5 conditions are usually set to 'mean' or 'most probable' values.

6  
7 In the Netherlands recent attempts to integrate the empirical and the simulation method have  
8 made use of the virtues of both: the broad scientific acceptance (at least in Europe) of the  
9 empirical values, and the ecological detail of the simulated ones (Van Dobben and van Hinsberg,  
10 2008). To this end, both the EUNIS typology and Schaminée et al.'s (1995) typology were  
11 translated into the European Habitat typology (Commission of the European Communities, 2003),  
12 and critical load ranges were determined according to both methods. For each Habitat type a  
13 unique critical load value was determined as the midpoint of the simulated range when this  
14 midpoint was within the empirical range; otherwise either the upper or the lower extreme of the  
15 empirical range was used. This method was developed in response to the policymaker's need for  
16 unique critical load values per Habitat type, to be used for the assessment of human activities in  
17 European 'Natura 2000' areas.

18

#### 19 **4.2 Application of the ForSAFE-VEG model at 16 Swedish forest sites**

20 ForSAFE-VEG does not run in an inverse mode to derive critical loads. Actually, this is  
21 impossible, as the model is too complex to be used in an inverse way. Instead the “critical load” is  
22 determined to be passed at the time one can observe unwanted significant shifts in vegetation  
23 composition or abundance. This time is used for estimating the critical load, which is defined in  
24 this case as the deposition of N at the point in time of significant unwanted vegetation change.  
25 Actually, this value is dependent on the site history. To estimate the critical loads of N, a  
26 preliminary definition was adopted by which a 95% of the natural ground vegetation composition

1 is preserved. This definition excludes the effect of other factors than N on the ground vegetation  
2 composition.

3  
4 Critical-load estimates for 16 forested sites in Sweden thus derived are given in Table 3. The table  
5 presents the year when the acceptable change in ground vegetation composition occurred, and the  
6 value of the deposition at that year. A reduction from today's deposition values can then be  
7 deduced to lower the deposition to the historic value that preceded the undesired change in the  
8 ground-vegetation composition (Table 3). The estimates set the critical load as the deposition at  
9 the time the change occurs, probably leading to a slight overestimate of the critical load. Results  
10 show that all sites have significant exceedance, and in order to protect 95% of the area, a 90%  
11 reduction of present deposition is required, implying an average atmospheric deposition in  
12 southern Sweden of 1.1 kg N.ha<sup>-1</sup>.yr<sup>-1</sup>. Using a protective level to 50%, still 55% reduction in  
13 present deposition will be required, implying an average deposition in southern Sweden of 2.8 kg  
14 N.ha<sup>-1</sup>.yr<sup>-1</sup>.

15

## 16 **5 Comparison and evaluation of the model chains**

17

### 18 *Comparison of model approaches and evaluation of the model chains*

19 The modelling approaches described in this article consist of a combination of a biogeochemical  
20 model of nutrient (including N) behaviour in the soil, connected with a vegetation model  
21 predicting water, N and acidity impacts on biodiversity. The biogeochemical models discussed are  
22 SMART2 (either or not in connection with SUMO), MAGIC and ForSAFE. These models differ  
23 with respect to the included processes and management options (Table 4). Models of vegetation  
24 succession are included in ForSAFE, and in the model chain SMART2-SUMO-MOVE/NTM,  
25 with SUMO being a specific model for vegetation succession. Vegetation-succession models are  
26 intermediates between biogeochemical models and species-composition models since they

1 simulate changes in elemental budgets and biomass distribution. Both SUMO and FORSAFE thus  
2 simulate the development of vegetation biomass and stocks of nutrient elements in relation to  
3 events such as fire, grazing, mowing or turf stripping. For example, grazing increases light  
4 availability and thus favours the growth of short-growing plants.

5  
6 A comparison of the characteristics of the vegetation models and succession models predicting N  
7 impacts on biodiversity (MOVE/NTM, VEG and SUMO) is given in Table 5. The major strength  
8 of the SMART2/ MAGIC-GBMOVE/NTM approach is the empirical determination of the  
9 relation between plant species composition and soil moisture, nutrient availability and soil acidity.  
10 Furthermore, the relationships are based on species-response curves of a large number of higher  
11 and lower plant species (e.g. in MOVE about 900 plant species are covered (Wiertz et al., 1992) .  
12 By using vegetation relevés identified on the level of vegetation types, it was possible to estimate  
13 critical limits for these vegetation types based on percentile values of the Ellenberg indicators for  
14 nutrients and pH. Thus the strength of the resulting empirical niche models is that the weight of  
15 data reduces noise relative to species-environment relationships, at least in the Ellenberg domain.  
16 GBMOVE also includes climate and management besides N and acidity, again based on survey  
17 data, and thus incorporates the impacts of climate and management on plant species diversity and  
18 its modifying effect on critical loads.

19  
20 The major weakness of the SMART2/MAGIC-GBMOVE/NTM approach is that a relationship  
21 is needed between Ellenberg indicators for N and moisture availability and acidity and measured  
22 values for these abiotic variables. Such calibration equations increase uncertainty because soil pH,  
23 soil C/N and soil moisture do not explain the total variation in mean Ellenberg scores. The greater  
24 the scatter about each regression line the more likely it is that predictions of mean Ellenberg values  
25 from soil measurements will differ from actual observations. The relationship with N indicators,  
26 such as N availability used in MOVE and soil C/N ratio used in GBMOVE, is rather weak,

1 especially in high-fertility ecosystems. Especially the uncertainty in the Ellenberg indicator for  
2 nutrient availability is large and can be the main source of uncertainty in the end result  
3 (Schouwenberg et al., 2000; Wamelink et al., 2002). Ideally the use of Ellenberg indicator values  
4 should thus be avoided and response curves should be estimated from actual measurements of soil  
5 pH and N availability (Wamelink et al., 2005). Furthermore, it is not likely that the relations  
6 between Ellenberg indicator values and actual conditions derived for the Netherlands or the  
7 United Kingdom are valid for other countries. Therefore, to use these models in other countries it  
8 is necessary to analyse local vegetation relevés in order to assign critical site factors to ecosystems.  
9 Finally, the output of the model chains is the potential vegetation on a site, whereas the observed  
10 vegetation may differ due to time lag effects. The MOVE and GBMOVE models are based on  
11 empirical observations recorded at different times in the past 30-70 years across Dutch and British  
12 ecosystems while the resulting regression models assume equilibrium between species and  
13 environment and the niche of each species is thus static. Another weakness is the lack of feedback  
14 between vegetation change and the soil model, at least when SUMO is not included.

15  
16 The major strength of ForSAFE-VEG is the mechanistic approach relating (many) abiotic  
17 parameters to plant species diversity including: (i) ground vegetation community competition,  
18 feedbacks from climate and from grazing animals and forest management and (ii) the mechanistic  
19 integration of the N cycle with process kinetics and feedbacks to the chemistry, organic matter  
20 decomposition and growth cycles. Furthermore, the model is field tested in Sweden. The major  
21 weakness of the ForSAFE-VEG approach is the high data demand. This holds specifically for the  
22 driving variables that consist mainly of descriptions of events, in particular the timing and intensity  
23 of grazing and other management events, which is also a limitation of SUMO. Furthermore, the  
24 complexity of the model makes interpretation of the results difficult, especially how different  
25 factors like acidity, nitrogen, management and climate change are all linked to biodiversity.

26

1 *Use of nitrogen indicators in view of impacts on plant species occurrence*

2 To connect theory on N dynamics in soils with models of plant species occurrence, a measure of  
3 N exposure, i.e. of plant-available N, is required. There are different measures to integrate N  
4 exposure into a single indicator. Some of these indicators give direct information on an N flux to  
5 the ecosystem, whereas other indicators only give indirect information based on correlations with  
6 fluxes. Most effects of N are due to an excess of N, either in the form of  $\text{NH}_4$  or  $\text{NO}_x$ . In the soil  
7 models, a differentiation is made between  $\text{NH}_4$  and  $\text{NO}_3$  but not in the vegetation response  
8 models, even though there are indications that plants are more sensitive for  $\text{NH}_4$  than for  $\text{NO}_3$   
9 (see e.g. Bobbink et al., 2003). The knowledge is, however, considered too limited to include in the  
10 vegetation effect models.

11  
12 Major direct indicators of N availability are: (i) gross mineralization and nitrification rates,  
13 reflecting the internal N cycling and thus potentially the maximum inorganic-N pool available to  
14 plants in competition with microbial uptake and (ii) N deposition. The N-deposition flux  
15 accurately reflects the exposure of species with limited root systems, particularly bryophytes and  
16 lichens. In other systems, however, the transformation of N by soil microorganisms modifies plant  
17 exposure and N-deposition flux is thus not a good measure of exposure for plants rooting in soil.  
18 A better indicator is the sum of N deposition and N mineralization, as used in the SMART2-  
19 SUMO-MOVE approach, although the link to biodiversity is only expert-based, namely through  
20 the Ellenberg N indicator.

21  
22 Indirect indicators, which are correlated with N availability, include (Rowe et al., 2005): (i) soil  
23 solution N concentration, (ii) plant tissue N concentration, (iii) soil C/N ratio and (iv) indicators  
24 based on the plant species assemblage. Apart from various other factors, use of a soil solution N  
25 concentration forms the basis of the ForSAFE-VEG model approach. The advantage of using this  
26 indicator is that the soluble N pool is immediately available to plants. However, soil solution only

1 reflects the N in excess of uptake and leaching losses and thus may underestimate total N  
2 availability to plants. Concentrations are also very dynamic, both spatially and temporally, and  
3 single measurements of soil-solution N concentrations are thus of limited use. Measures integrated  
4 over time are thus more reliable indicators of N status. Furthermore, species differ in their ability  
5 to use different forms of soluble N, i.e.  $\text{NO}_3$ ,  $\text{NH}_4$  and DON and the ratio of ammonium to  
6 nitrate in solution may provide information relevant to species occurrence and also the potential  
7 for microbial uptake of nitrate. Measures of plant chemistry are not yet used in any of the models.  
8 A disadvantage of the use of tissue concentrations is that they vary considerably in time  
9 (seasonally), among species, plant parts, tissue age/phenological stage and with nutrient supply,  
10 grazing or other management (Rowe et al., 2005). Nevertheless, if these factors can be controlled  
11 (e.g. by sampling a standard part, from a single species or group, at a standard time of year), tissue  
12 concentrations of N and amino acids may be good indicators of N exposure and in principle could  
13 be outputs from biogeochemical models. The soil C/N ratio, as used in the GBMOVE approach,  
14 is not directly influencing plant response but represents a readily measurable proxy for important  
15 processes (e.g. mineralization or nitrification). In general, the relationship is weak, since total soil  
16 N is largely inactive, and it is not a good indicator of N availability (Tamm, 1991). Finally,  
17 Ellenberg indicators (Ellenberg et al., 1992) are used in GBMOVE, MOVE and NTM to describe  
18 assemblages of European vascular plants and bryophytes. Mean Ellenberg fertility scores have  
19 been shown to be reasonable indicators of soil N availability (Van Dobben, 1993), although the  
20 relationship usually shows large variation (Wamelink et al., 2002) and appears to correlate best with  
21 annual above-ground biomass production rather than soil nutrient status (Hill and Carey, 1997).  
22 However, in systems which are limited by other nutrients, e.g. phosphorous or potassium, this may  
23 not be the case.  
24  
25 In summary, plants do not respond to a single abiotic variable, and there are problems with all  
26 variables that could potentially be used as input to the vegetation models. Those considered most

1 useful are direct indicators of N availability, such as N deposition plus N mineralization, followed  
2 by indirect indicators, such as soil solution N concentrations in the rooting zone, foliar N  
3 concentration or soil C/N ratio.

4

#### 5 ***Model validation status***

6 The validation status of the various models differs, specifically with respect to the comparisons  
7 between measured and modelled changes in plant species composition. In general, the  
8 biogeochemical models used (SMART2, MAGIC and ForSAFE) have a good validation status.  
9 For example, SMART2 has been validated on the German ‘Solling site’ and hundreds of  
10 intensively monitored forest plots (De Vries et al., 2003). Here we show that MAGIC is able to  
11 predict changes in observed N leaching and soil C/N in plot-scale N manipulation studies under  
12 different addition levels. The ForSAFE model also shows a good correlation between simulated  
13 and measured values of tree biomass, pools of soil organic C and N, soil pH and BC/Al ratios at  
14 16 Swedish forest sites.

15

16 The validation status of the vegetation models is, however, much less advanced. Biomass growth  
17 of SUMO has until now only been validated on data collected at a grassland site, a heathland site  
18 and a forest site in the Netherlands and a grassland site in the UK. Regarding (GB)MOVE, a  
19 preliminary test was made whether GBMOVE in combination with MAGIC could reproduce the  
20 observed species composition in test plots, including an unimproved neutral grassland and a  
21 blanket bog. The comparisons of predicted species occurrence using measured soil C/N and pH  
22 versus mean Ellenberg scores indicate that the greatest uncertainty in model predictions, is due to  
23 the weak calibration relationships, especially those between soil C/N and mean Ellenberg N at  
24 high fertility. Current model development is therefore focusing on the establishment of a direct  
25 relation between N indicators and species composition. Finally, the ForSAFE-VEG model has  
26 until now been validated at 16 Swedish forest sites, by comparing simulated and measured values

1 on the ground occupancy of 42 plant groups, and in Switzerland. Considering the sometimes poor  
2 correlations in all model applications, as described above, there is still a need for substantial model  
3 validation and model improvement. In this context, good model-testing data are urgently needed.

4

5 ***Reliability of biogeochemistry-biodiversity modelling approaches to assess critical loads***

6 Up to now, only the SMART-MOVE model chain has been used to assess critical loads by using  
7 the model in an inverse way. Due to its complexity, the ForSAFE-VEG model is more suited to  
8 predict response of plant species composition to environmental change than to predict critical  
9 loads. A major disadvantage is the fact that the critical N load is influenced by the N deposition  
10 history. The uncertainties in the assessed critical loads, using SMART-MOVE model chain, are  
11 specifically due to uncertainties in the calibration equations between abiotic conditions and  
12 Ellenberg scores, used to transfer information between soil models and plant-species models. A  
13 detailed uncertainty analysis, focusing on this aspect, was carried out by van Dobben et al. (2006).  
14 Results show that the uncertainty in critical loads per vegetation type is quite high (generally in a  
15 range of 15-40% of the average value), but the ranges of simulated and empirical values usually  
16 overlap, implying that the results are applicable for practical purposes. However, at the site level,  
17 uncertainty becomes very large and thus it is not yet possible to determine critical loads with any  
18 practical significance (Van Dobben et al., 2006). The uncertainties can only be reduced if more  
19 data become available on the abiotic response per species under field conditions, at least to N  
20 availability and soil pH, as described below.

21

22 Despite the uncertainties in the described model approaches, the approach provides a relevant  
23 addition to empirical critical-load estimates. First of all, even though empirical critical loads may  
24 be as good or even better than modelled critical loads at present, dynamic models allow us to  
25 explore impacts of future scenarios, where habitats may face completely novel configurations of  
26 multiple drivers, so-called “no analogue” states (Steffen et al., 2004). In these situations empirical



1 critical loads that are based on reviews of contemporary and historical data become may be  
2 increasingly inappropriate and process based, detailed models are necessary. Critical loads can be  
3 interpreted as a means of identifying thresholds of ecosystem resilience (Gunderson, 2000).  
4 Estimating resilience thresholds using linked soil-vegetation models is therefore particularly  
5 appropriate, because of several key aspects associated with ecosystem responses to perturbation.  
6 Threshold changes and non-linear responses to disturbance, species invasion and changes in  
7 nutrient availability can result from the dynamic interplay between above- and below-ground biota  
8 and differences in the extent to which soil and vegetation store or more rapidly process and release  
9 excess nutrients or buffer pH changes (Evans et al., 2001; Craine et al., 2002). Hence resilience is a  
10 property related to the ability of soil and vegetation to buffer or to amplify the response to  
11 changing conditions. Quantifying resilience and proximity to thresholds of change then requires  
12 quantification of the dynamic relationships between ecosystem compartments (e.g. Suding et al.,  
13 2008). Policy interest also focuses on timescales for recovery; hence dynamic modelling is required  
14 to model the persistence of alternative stable states that can result when perturbations drive  
15 ecosystems into new domains of stability (Gunderson, 2000; Suding et al., 2004). While it may be  
16 possible to identify critical ecosystem state variables that are closely correlated with damage and  
17 loss of adaptive capacity, these may be above or below ground (e.g. Strengbom et al., 2001) once  
18 more emphasising the importance of jointly modelling soil and vegetation. Lastly, the critical-load  
19 approach increasingly recognises that resilience is not realistically thought of as a threshold situated  
20 along single or multiple, yet independent, abiotic axes. Multiple drivers and their interactions are  
21 important. For example, land use around a nature reserve can change the composition of the local  
22 species pool, increasing availability of species that could capitalise on changes in soil conditions  
23 driven by atmospheric pollutant deposition or conversely, reducing the availability of desirable  
24 species for recolonisation following recovery and remediation (Lindborg and Eriksson, 2004).  
25 Dynamic model development offers a way of incorporating other drivers as modifiers of the

1 critical-load range and allowing multiple drivers to dynamically interact. The fully integrated,  
2 mechanistic ForSafe-VEG model is most appropriate to investigate such interactions.

3

#### 4 *Measurements of plant species response to environmental variables*

5 Expert-based estimates of plant species responses to environmental variables form the basis of all  
6 models. Although the Ellenberg indicator system (Ellenberg et al., 1992) or its derivatives  
7 (Diekmann, 2003) are the most frequently used systems for this, their uncertainty is quite large, it  
8 is sometimes unclear what the indicator values represent, and they may be biased (Ertsen et al.,  
9 1998; Schaffers and Sykora, 2000; Wamelink et al., 2002; Wamelink et al., 2003b; Witte and von  
10 Asmuth, 2003; Smart and Scott, 2004; Wamelink et al., 2004). Furthermore, it requires a translation  
11 from indicator values into soil chemical variables.

12

13 The most logical solution to avoid this highly uncertain step is to replace the Ellenberg indicator  
14 system by a system based on measurements (Wamelink et al., 2002). On a small scale this was  
15 carried out for France (Gégout et al., 2003) and the Netherlands (Wamelink et al., 2005). Results  
16 are promising, but Europe-wide and international data are needed, instead of only national data to  
17 ensure a wide application. Such data should consist of a vegetation description (relevés) and at  
18 least the following measured variables: geographical coordinates, soil acidity, nutrient status, and  
19 water availability. Based on this, plant species response per abiotic variable can be estimated,  
20 reviewed and tested on independent datasets. The hypothesis that plant species have different  
21 responses in different regions can be tested on the basis of such data.

22

#### 23 *Relevant extensions to the modelling approaches.*

24 As mentioned above, in the SMART2/ MAGIC- (GB)MOVE/ NTM model chain, model output  
25 is the potential vegetation on a site, not accounting for time-lag effects. Priorities for future work  
26 on modelling N impacts on biodiversity in this model chain thus include: (i) inclusion of species-

1 response curves based on combined field measurements of vegetation relevés and abiotic data and  
2 (ii) representation of lag times (e.g. due to species persistence, dispersal).

3  
4 The present linked-model approaches further centre upon the impact of N deposition on existing  
5 species assemblages. However, changes in resource availability are also predicted to increase  
6 susceptibility to invasion by immigrants, some of which could possess traits associated with  
7 suppression of resident species and changes in nutrient cycling (Evans et al., 2001; Davis et al.,  
8 2005). If information about the composition of the regional species pool is available, then model  
9 application could be extended to estimate changes in habitat suitability for potential immigrants.  
10 This would amount to a risk assessment of the establishment of potential invaders. Incorporating  
11 the dynamic impact of invasive species on above and below-ground ecosystem compartments is  
12 possible in the presented model approaches but would require modification of the biomass growth  
13 models in SUMO and For-SAFE/VEG.

14  
15 ***Conclusions on the potential of linked biogeochemistry-biodiversity modelling approaches***

16 The following conclusions can be drawn regarding the modelling approaches presented here:

- 17 - Vegetation models, based on either large-scale vegetation surveys (MOVE/ GBMOVE,  
18 NTM) or mainly experimental data (VEG), have been developed and integrated with  
19 biogeochemical models, such as SMART2 (either or not in connection with SUMO), MAGIC  
20 and ForSAFE.  
21 - There are large similarities between the models, particularly those based on survey data, but  
22 there are also several important differences, including: (i) use of different abiotic variables for  
23 N, such as N availability and Ellenberg N indicators in SMART2-MOVE/NTM, soil C/N  
24 ratio in GBMOVE and soil-solution N in ForSAFE, (ii) prediction of individual plant species  
25 (e.g. MOVE/GBMOVE) versus plant communities (NTM) and (iii) calibration based on  
26 different (national) soil and vegetation datasets.

- 1 - At their current level of development, most integrated models focus primarily on predicting  
2 the biodiversity impacts of different scenarios in terms of air-quality change and climate  
3 change.
- 4 - The model chains can also predict biodiversity-based critical loads or target loads. In doing  
5 this, the definition of reference conditions and damage thresholds for terrestrial biodiversity  
6 represents a major challenge. Although the definition of biodiversity targets is an issue for  
7 policy-makers, dynamic models can provide valuable information on realistic reference  
8 conditions and achievable recovery targets.
- 9 - The reliance on Ellenberg indicator values, used as a proxy for abiotic conditions in survey-  
10 based models such as SMART2-MOVE and MAGIC-GBMOVE, adds uncertainty to model  
11 predictions. However, Ellenberg values are likely to remain necessary in many areas due to the  
12 insufficient European coverage of combined vegetation and soil survey data.
- 13 - Models based on survey data have largest potential for country-wide mapping of critical loads  
14 in view of their limited data demands. This holds not only for Europe but also for other areas  
15 coping with elevated N deposition, such as North America and China. However, testing and  
16 adaptation of the linkage between vegetation and soil, using e.g. Ellenberg indicators is needed  
17 for other countries or ecosystems before large-scale applications can be made. Furthermore,  
18 results are quite reliable for vegetation types but at site level, the uncertainty in critical loads  
19 becomes too large to be of practical significance.
- 20 - In combination, both empirical and model-based critical loads, are powerful tools to assess a  
21 reliable value for defined ecosystems. An example of such a combination approach is  
22 presented by Van Dobben and van Hinsberg (2008).

23  
24 Despite the various limitations mentioned, this overview shows that linked biogeochemistry-  
25 biodiversity models for N have great potential for applications to support policies to reduce N  
26 inputs. Apart from further model development, there is a need for further testing and validation of

1 the models against long-term monitoring or long-term experimental datasets and against large-  
2 scale survey data. In this context, the continuation of existing programmes, where possible with  
3 improved integration of biotic and abiotic measurements, is essential to the future development of  
4 this work. Finally, there is a need for adaptation and upscaling of the models beyond the regions  
5 for which dose-response relationships have been parameterised, such as Mediterranean and Alpine  
6 regions, and Eastern Europe, based on a focused data collection combining vegetation descriptions  
7 with variables affecting the species diversity. A similar approach outside Europe would also allow  
8 the use of these models, both for predicting impacts of scenarios and assessing critical loads.

9

#### 10 *Acknowledgments*

11 The preparation of this paper has been partially supported by the European Commission LIFE  
12 programme within the framework of the European Consortium for Modelling Air Pollution and  
13 Climate Strategies (EC4MACS) and the trust fund for the partial funding of effect-oriented  
14 activities under the Convention on Long-range Transboundary Air Pollution. We also thank the  
15 two anonymous reviewers for their detailed and helpful comments which largely improved the  
16 paper.

17

18

1 **References**

- 2 *Achermann, B., Bobbink, R. (Eds.), 2003. Empirical critical loads for nitrogen: Expert workshop, Berne, 11-13*  
3 *November 2002. Environmental Documentation 164, Swiss Agency for the Environment, Forests and Landscape.*  
4  
5 *Alkemade, J.R.M., Wiertz, J., Latour, J.B., 1996. Kalibratie van Ellenbergs milieu-indicatiegetallen aan*  
6 *werkelijk gemeten bodemfactoren. Rijksinstituut voor volksgezondheid en milieu, Bilthoven, The Netherlands.*  
7 *Rapport nr. 711901 016.*  
8  
9 *Amann, M., Cofala, J., Heyes, C., Klimont, Z., Schöpp, W., 1999. The RAINS model: A tool for assessing*  
10 *regional emission control strategies in Europe. Pollution Atmosphérique 20, 41-64.*  
11  
12 *Belyazid, S., 2006. Dynamic Modelling of biogeochemical processes in forest ecosystems. Doctoral Thesis. Reports in*  
13 *Ecology and Environmental Engineering 2006:1, Department of chemical Engineering, Lund University, Sweden.*  
14  
15 *Belyazid, S., Westling, O., Sverdrup, H., 2006. Modelling changes in forest soil chemistry at 16 Swedish*  
16 *coniferous forest sites following deposition reduction. Environmental Pollution 144, 596-609.*  
17  
18 *Bobbink, R., Hornung, M., Roelofs, J.G.M., 1998. The effects of air-borne nitrogen pollutants on species diversity*  
19 *in natural and semi-natural European vegetation. Journal of Ecology 86, 717-738.*  
20  
21 *Bobbink, R., Ashmore, M., Braun, S., Flückiger, W., van den Wyngaert, I.J.J., 2003. Empirical nitrogen critical*  
22 *loads for natural and semi-natural ecosystems: 2002 update. In: Achermann, B., Bobbink, R. (Eds). Empirical*  
23 *critical loads for nitrogen. Swiss Agency for Environment, Forest and Landscape SAEFL, Berne, pp. 43-170.*

1

2 *Bobbink, R., Hicks, K., Spranger, T., Galloway, J., Alkemade, R., Ashmore, M., Bustamante, M., Davidson,*  
3 *E., Dentener, F., Emmett, B., Fenn, M., Gilliam, F., Nordin, A., de Vries, W., 2009. Atmospheric nitrogen*  
4 *deposition, biodiversity effects and critical loads - a synthesis. Ecological Applications (submitted).*

5

6 *Braun-Blanquet, W., 1964. Pflanzensoziologie. Grundzüge der Vegetationskunde. 3. Aufl. Springer, Wien/New*  
7 *York.*

8

9 *Clark, C.M., Tilman, D., 2008. Loss of plant species after chronic low-level nitrogen deposition to prairie*  
10 *grasslands. Nature 451, 712-715.*

11

12 *Commission of the European Communities, 2003. Interpretation manual of European Union habitats - EUR 25.*  
13 *DG Environment - Nature and Biodiversity.*

14

15 *Cosby, B.J., Ferrier, R.C., Jenkins, A., Wright, R.F., 2001. Modelling the effects of acid deposition: refinements,*  
16 *adjustments and inclusion of nitrogen dynamics in the MAGIC model. Hydrology and Earth System Sciences 5,*  
17 *499-517.*

18

19 *Craine, J.M., Tilman, D., Wedin, D., Reich, P., Tjoelker, M., Knops, J., 2002. Functional traits, productivity*  
20 *and effects on nitrogen cycling of 33 grassland species. Functional Ecology 16, 563-574.*

21

- 1 *Davies, C.E., Moss, D., 2002. EUNIS Habitat Classification. Centre for Ecology and Hydrology. 2001 Work*  
2 *Programme, Final Report to the European Environment Agency European Topic Centre on Nature Protection and*  
3 *Biodiversity.*
- 4
- 5 *Davis, M.A., Thompson, K., Grime, J.P., 2005. Invasibility: the local mechanism driving community assembly*  
6 *and species diversity. Ecography 28, 696-704.*
- 7
- 8 *De Vries, W., Reinds, G.J., Posch, M., Sanz, M.J., Krause, G.H.M., Calatayud, V., Renaud, J.P., Dupouey,*  
9 *J.L., Sterba, H., Gundersen, P., Voogd, J.C.H., Vel, E.M., 2003. Intensive Monitoring of Forest Ecosystems in*  
10 *Europe. Technical Report 2003. UN/ECE and EC, Forest Intensive Monitoring Coordinating Institute,*  
11 *Geneva and Brussels.*
- 12
- 13 *De Vries, W., Kros, J., Reinds, G.J., Wamelink, W., Mol, J., van Dobben, H., Bobbink, R., Smart, S., Evans,*  
14 *C., Schlutow, A., Kraft, P., Belyazid, S., Sverdrup, H.U., van Hinsberg, A., Posch, M., Hettelingh, J.P., 2007.*  
15 *Developments in modelling critical nitrogen loads for terrestrial ecosystems in Europe. Alterra Wageningen UR,*  
16 *Wageningen, the Netherlands. Report 1382.*
- 17
- 18 *Diekmann, M., 2003. Species indicator values as an important tool in applied plant ecology - a review. Basic and*  
19 *Applied Ecology 4, 493-506.*
- 20
- 21 *EC, 2005. Thematic Strategy on Air Pollution. Communication from the Commission to the Council and the*  
22 *European Parliament. COM(2005) 446 final.*



- 1 EEA, (European Environment Agency), 2007. *Halting the loss of biodiversity by 2010: proposal for a first set of*  
2 *indicators to monitor progress in Europe. EEA Technical report 11/2007.*
- 3
- 4 Elberse, W.T., van den Bergh, J.P., Dirven, J.P.G., 1983. *Effects of use and mineral supply on the botanical*  
5 *composition and yield of old grassland on heavy-clay soil. Netherlands Journal of Agricultural Science 31, 63-88.*
- 6
- 7 Ellenberg, H., Weber, H.E., Düll, R., Wirth, V., Werner, W., Paulißen, D., 1992. *Zeigerwerte von Pflanzen*  
8 *in Mitteleuropa. Scripta Geobotanica 18, 1-248.*
- 9
- 10 Emmett, B.A., 2007. *Nitrogen saturation of terrestrial ecosystems: some recent findings and their implications for*  
11 *our conceptual framework. Water, Air, and Soil Pollution: Focus 7, 99-109.*
- 12
- 13 Ertsen, A.C.D., Alkemade, J.R.M., Wassen, M.J., 1998. *Calibrating Ellenberg indicator values for moisture,*  
14 *acidity, nutrient availability and salinity in the Netherlands. Plant Ecology 135, 113-124.*
- 15
- 16 Evans, C.D., Caporn, S.J.M., Carroll, J.A., Pilkington, M.G., Wilson, D.B., Ray, N., Cresswell, N., 2006.  
17 *Modelling nitrogen saturation and carbon accumulation in heathland soils under elevated nitrogen deposition.*  
18 *Environmental Pollution 143, 468-478.*
- 19
- 20 Evans, R.D., Rimer, R., Sperry, L., Belnap, J., 2001. *Exotic plant invasion alters nitrogen dynamics in an arid*  
21 *grassland. Ecological Applications 11, 1301-1310.*

- 1 Gégout, J.C., Hervé, J.C., Houllier, F., Pierrat, J.C., 2003. Prediction of forest soil nutrient status using  
2 vegetation. *Journal of Vegetation Science* 14, 55-62.
- 3
- 4 Gunderson, L.H., 2000. Ecological resilience - in theory and application. *Annual Review of Ecology and*  
5 *Systematics* 31, 425-439.
- 6
- 7 Hettelingh, J.-P., Posch, M., de Smet, P.A.M., 2001. Multi-effect critical loads used in multi-pollutant reduction  
8 agreements in Europe. *Water, Air, and Soil Pollution* 130, 1133-1138.
- 9
- 10 Hettelingh, J.-P., Posch, M., Slootweg, J., Reinds, G.J., Spranger, T., Tarrasón, L., 2007. Critical loads and  
11 dynamic modelling to assess European areas at risk of acidification and eutrophication. *Water, Air, & Soil*  
12 *Pollution: Focus* 7, 379-384.
- 13
- 14 Hill, M.O., Carey, P.D., 1997. Prediction of yield in the Rothamsted Park Grass Experiment by Ellenberg  
15 indicator values. *Journal of Vegetation Science* 8, 579-586.
- 16
- 17 Janssen, P.H.M., Heuberger, P.S.C., 1995. Calibration of process-oriented models. *Ecological Modelling* 83, 55-  
18 66.
- 19
- 20 Jenkinson, D.S., Potts, J.M., Perry, J.N., Barnett, V., Coleman, K., Johnston, A.E., 1994. Trends in herbage  
21 yields over the last century on the Rothamsted long-term continuous hay experiment. *Journal of Agricultural Science*  
22 122, 365-374.
- 23

- 1 Kros, J., 2002. *Evaluation of biogeochemical models at local and regional scale*. Wageningen, Alterra, 2002,  
2 *Alterra scientific contributions, Wageningen, The Netherlands*.
- 3
- 4 Kros, J., Reinds, G.J., de Vries, W., Latour, J.B., Bollen, M.J.S., 1995. *Modelling abiotic site factors in response*  
5 *to atmospheric deposition and upward seepage*. In: Schoute, J.F.T., Finke, P.A., Veeneklaas, F.R., Wolfert, H.P.  
6 (Eds). *Scenario studies for the rural environment*. Kluwer, Dordrecht, pp. 445-448.
- 7
- 8 Latour, J.B., Reiling, R., 1993. *A multiple stress model for vegetation ('move'): a tool for scenario studies and*  
9 *standard-setting*. *Science of the Total Environment* 134, 1513-1526.
- 10
- 11 Latour, J.B., Reiling, R., Slooff, W., 1994. *Ecological standards for eutrophication and dessication: perspectives for*  
12 *a risk assessment*. *Water, Air, and Soil Pollution* 78, 265-277.
- 13
- 14 Lindborg, R., Eriksson, O., 2004. *Historical landscape connectivity affects present plant species diversity*. *Ecology*  
15 85, 1840-1845.
- 16
- 17 Ludwig, D., Walker, B., Holling, C.S., 1997. *Sustainability, stability, and resilience*. *Conservation Ecology*  
18 [online] 1, 7.
- 19
- 20 *National Emission Ceilings Directive, 2001. Directive 2001/81/EC of the European Parliament and of the*  
21 *Council of 23 October 2001 on national emission ceilings for certain atmospheric pollutants*.
- 22

- 1 Nilsson, J., Grennfelt, P., 1988. *Critical loads for sulphur and nitrogen. Report from a Workshop held at*  
2 *Skokloster Sweden March 19-24 1988. Miljø rapport 1988: 15. Copenhagen Denmark Nordic Council of*  
3 *Ministers.*
- 4
- 5 Pakeman, R.J., Reid, C.L., Lennon, J.J., Kent, M., 2008. *Possible interactions between environmental factors in*  
6 *determining species optima. Journal of Vegetation Science 19, 201-208.*
- 7
- 8 Rowe, E.C., Moldan, F., Emmett, B.A., Evans, C.D., Hellsten, S., 2005. *Model chains for assessing impacts of*  
9 *nitrogen on soils, waters and biodiversity: a review. Centre for Ecology and Hydrology. Contract report project no*  
10 *C02887 for DEFRA (UK) project no. Cpea 19.*
- 11
- 12 Roy, D.B., Hill, M.O., Rothery, P., Bunce, R.G.H., 2000. *Ecological indicator values of British species: an*  
13 *application of Gaussian logistic regression. Annales Botanici Fennici 37, 219-226.*
- 14
- 15 Schaffers, A.P., Sykora, K.V., 2000. *Reliability of Ellenberg indicator values for moisture, nitrogen and soil*  
16 *reaction: a comparison with field measurements. Journal of Vegetation Science 11, 225-244.*
- 17
- 18 Schaminée, J.H.J., Westhoff, V., van Wirdum, G., 1989. *Naar een nieuw overzicht van de*  
19 *plantengemeenschappen van Nederland. De Levende Natuur 90, 204-209.*
- 20
- 21 Schaminée, J.H.J., Stortelder, A.H.F., Westhoff, V., 1995. *De vegetatie van Nederland 1. inleiding tot de*  
22 *plantensociologie: grondslagen, methoden en toepassingen. Opulus Press, Uppsala.*
- 23

- 1 *Schlutow, A., Hübener, P., 2004. The BERN Model: Bioindication for Ecosystem Regeneration towards*  
2 *Natural conditions. Umwelt bundesamt. Research report 20085221.*
- 3
- 4 *Schöpp, W., Posch, M., Mylona, S., Jobansson, M., 2003. Long-term development of acid deposition (1880-*  
5 *2030) in sensitive freshwater regions in Europe. Hydrology and Earth System Sciences 7, 436-446.*
- 6
- 7 *Schouwenberg, E.P.A.G., Houweling, H., Jansen, M.J.W., Kros, J., Mol-Dijkstra, J.P., 2000. Uncertainty*  
8 *propagation in model chains: a case study in nature conservancy. Alterra, Green World Research, Wageningen.*  
9 *Alterra rapport 1.*
- 10
- 11 *Slootweg, J., Posch, M., Hettelingh, J.-P. (Eds.), 2007. Critical loads of nitrogen and dynamic modelling: CCE*  
12 *Progress Report 2007. Coordination Centre for Effects. MNP Report 500090001, Bilthoven, Netherlands.*
- 13
- 14 *Smart, S.M., Scott, W.A., 2004. Bias in Ellenberg indicator values - problems with detection of the effect of*  
15 *vegetation type. Journal of Vegetation Science 15, 843-846.*
- 16
- 17 *Smart, S.M., Scott, W.A., Whittaker, J., Evans, C., Emmett, B., 2003. Development and testing of linked soil-*  
18 *plant model systems. Pollution Umbrella Progress Report. May 2003.*
- 19
- 20 *Smart, S.M., Evans, C., Rowe, E., Wamelink, W., Roy, D., Preston, C., Hill, M., Rothery, P., Bullock, J.,*  
21 *Moy, I., Scott, A., Emmett, B., Wright, S., 2005. Atmospheric nitrogen pollution impacts on biodiversity: Phase*  
22 *1 - Model development and testing. Final report to Department of the Environment, Food and Rural Affairs, Joint*  
23 *Nature Conservation Committee and English Nature.*

1

2 *Spranger, T., Hettelingh, J.-P., Slootweg, J., Posch, M., 2008. Modelling and mapping long-term risks due to*  
3 *reactive nitrogen effects: An overview of LRTAP convention activities. Environmental Pollution 154, 482-487.*

4

5 *Steffen, W., Sanderson, A., Tyson, P.D., Jager, J., Matson, P.A., Moore III, B., Oldfield, F., Richardson, K.,*  
6 *Schellnhuber, H.J., Turner, B.L., Wasson, R.J., 2004. Global change and the earth system: a planet under*  
7 *pressure. Springer-Verlag, Berlin.*

8

9 *Strenghom, J., Nordin, A., Näsholm, T., Ericson, L., 2001. Slow recovery of boreal forest ecosystem following*  
10 *decreased nitrogen input. Functional Ecology 15, 451-457.*

11

12 *Suding, K.N., Gross, K.L., Houseman, G.R., 2004. Alternative states and positive feedbacks in restoration*  
13 *ecology. Trends in Ecology & Evolution 19, 46-53.*

14

15 *Suding, K.N., Ashton, I.W., Bechtold, H., Bowman, W.D., Mobley, M.L., Winkelman, R., 2008. Plant and*  
16 *microbe contribution to community resilience in a directionally changing environment. Ecological monographs 78,*  
17 *313-329.*

18

19 *Sutton, M.A., Cape, J.N., Rihm, B., Sheppard, L.J., Smith, R.I., Spranger, T., Fowler, D., 2003. The*  
20 *importance of accurate background atmospheric deposition estimates in setting critical loads for nitrogen. In:*  
21 *Achermann, B., Bobbink, R. (Eds). Empirical critical Loads for Nitrogen. Proceedings of an Expert Workshop*  
22 *in Berne, Switzerland, November 11-13. Swiss Agency for the Environment, Forests and Landscape (SAEFL)*  
23 *Environmental Documentation 164, pp. 231-257.*

1

2 *Sverdrup, H., Belyazid, S., Niblgård, B., Ericson, L., 2007. Modelling change in ground vegetation response to*  
3 *acid and nitrogen pollution, climate change and forest management at in Sweden 1500–2100 A.D. Water, Air,*  
4 *and Soil Pollution: Focus 7, 163-179.*

5

6 *Tamm, C.O., 1991. Nitrogen in terrestrial ecosystems. Questions of productivity, vegetational changes and*  
7 *ecosystem stability. Springer Verlag, Berlin.*

8

9 *Tietema, A., Mol Dijkstra, J.P., Kros, J., de Vries, W., 2002. Dynamic nitrogen deposition thresholds during*  
10 *forest stand developments in a Douglas fir forest analysed with two nitrogen models SMART2 and MERLIN.*  
11 *Hydrology and Earth System Sciences 6, 375-382.*

12

13 *Tilman, D., 1987. Secondary succession and the pattern of plant dominance along experimental nitrogen gradients.*  
14 *Ecological monographs 57, 189-214.*

15

16 *Van Dobben, H.F., 1993. Vegetation as a Monitor for Deposition of Nitrogen and Acidity. PhD Thesis,*  
17 *University of Utrecht.*

18

19 *Van Dobben, H.F., van Hinsberg, A., 2008. Overzicht van kritische depositiewaarden voor stikstof, toegepast op*  
20 *habitattypen en Natura 2000 gebieden. Alterra, Wageningen. Alterra rapport 1654.*

21

- 1 *Van Dobben, H.F., van Hinsberg, A., Schouwenberg, E.P.A.G., Jansen, M.J.W., Mol-Dijkstra, J.P.,*  
2 *Wieggers, H.J.J., Kros, J., de Vries, W., 2006. Simulation of critical loads for nitrogen for terrestrial plant*  
3 *communities in the Netherlands. Ecosystems 9, 32-45.*
- 4
- 5 *Wallman, P., Svensson, M., Sverdrup, H., Belyazid, S., 2005. ForSAFE - an integrated process-oriented forest*  
6 *model for long-term sustainability assessments. Forest Ecology and Management 207, 19-36.*
- 7
- 8 *Wamelink, G.W.W., 2007. Simulation of vegetation dynamics as affected by nitrogen deposition. Pons & Looijen,*  
9 *Wageningen.*
- 10
- 11 *Wamelink, G.W.W., ter Braak, C.F.J., van Dobben, H.F., 2003a. Changes in large-scale patterns of plant*  
12 *biodiversity predicted from environmental economic scenarios. Landscape Ecology 18, 513-527.*
- 13
- 14 *Wamelink, G.W.W., van Dobben, H.F., Berendse, F., 2003b. Apparently we do need phytosociological classes to*  
15 *calibrate Ellenberg indicator values! Journal of Vegetation Science 14, 619-620.*
- 16
- 17 *Wamelink, G.W.W., Goedhart, P.W., van Dobben, H.F., 2004. Measurement errors and regression to the mean*  
18 *cannot explain bias in average Ellenberg indicator values. Journal of Vegetation Science 15, 847-851.*
- 19
- 20 *Wamelink, G.W.W., van Dobben, H.F., Berendse, F., 2008. Vegetation succession as affected by decreasing*  
21 *nitrogen deposition, soil characteristics and site management: a modelling approach. Forest Ecology and*  
22 *Management (accepted).*
- 23



- 1 *Wamelink, G.W.W., Joosten, V., van Dobben, H.F., Berendse, F., 2002. Validity of Ellenberg indicator values*  
2 *judged from physico-chemical field measurements. Journal of Vegetation Science 13, 269-278.*
- 3
- 4 *Wamelink, G.W.W., Goedhart, P.W., van Dobben, H.F., Berendse, F., 2005. Plant species as predictors of soil*  
5 *pH: replacing expert judgement with measurements. Journal of Vegetation Science 16, 461-470.*
- 6
- 7 *Wamelink, G.W.W., Goedhart, P.W., Frissel, J.Y., Wegman, R.M.A., Slim, P.A., van Dobben, H.F., 2007.*  
8 *Response curves for plant species and phytosociological classes. Alterra, Wageningen. Alterra report 1489.*
- 9
- 10 *Wiertz, J., van Dijk, J., Latour, J.B., 1992. De MOVE-vegetatie module: De kans op voorkomen van 700*  
11 *plantesoorten als functie van vocht, pH, nutriënten en zout. Wageningen Bilthoven. RIVM report 711901006 /*  
12 *IBN-DLO report no. IBN 92/24.*
- 13
- 14 *Witte, J.P.M., von Asmuth, J.R., 2003. Do we really need phytosociological classes to calibrate Ellenberg indicator*  
15 *values? Journal of Vegetation Science 14, 615-618.*

1 *Table 1. Comparison of MAGIC-simulated and measured C/N ratios at Park Grass experimental site at Rothamsted, UK,*  
 2 *using 100% and 50% of the approximate estimate for nitrogen offtake in hay.*

Year	Measured C/N ratio (g C g <sup>-1</sup> N)	Simulated C/N ratio (g C g <sup>-1</sup> N)	
		100% N offtake	50% N offtake
1876	12.8	9.7	12.3
1959	12.3	11.2	12.4
2000	12.4	12.2	12.0

3  
 4  
 5 *Table 2 Empirical (Achermann and Bobbink, 2003) and average modelled (using SMART2) critical N loads (CL) and target*  
 6 *N loads for 2030 and 2100 (in kg.ba<sup>-1</sup>.yr<sup>-1</sup>) for European Nature Information System (EUNIS) classes.*

EUNIS Class	Empirical	Modelled	Modelled target load	Modelled target load
	CL	CL <sup>1)</sup>	(2030)	(2100)
Forest (G)	10-20	16.8 (12.9 - 18.2)	8.4 (7.4 – 16.8)	14.0 (13.0 – 16.8)
Raised bogs (D1)	5-10	6.1 (6.1 – 6.1)	4.5 (3.8 – 6.1)	5.7 (5.0 – 6.1)
Salt marsh (A2.64/65) <sup>2)</sup>	30-40	30.0 (30.0 – 34.1)	33.7 (29.9 – 33.9)	34.1 (34.0 – 34.1)
Dry and neutral grasslands (E1.7) <sup>2)</sup>	10-20	8.0 (8.0 – 8.0)	1.4 (0.2 – 3.1)	7.9 (4.4 – 10.9)
Semi-dry calcareous grasslands (E1.26) <sup>3)</sup>	15-25	12.4 (12.4 – 12.4)	-	-
Moist and wet oligotrophic grasslands (E3.5)	10-20	12.6 (12.6 – 12.6)	1.4 (0.5 – 6.7)	1.2 (0.4 – 12.6)
Coastal dune heaths (B1.5) <sup>4)</sup>	10-20	15.5 (14.4 – 15.5)	3.3 (3.1 – 5.0)	12.9 (12.6 – 12.9)
Dry heaths (F4.2)	10-20	11.2 (9.4 – 17.1)	19.8 (17.0 – 21.7)	19.8 (18.5 – 21.7)

<sup>1)</sup> Values in bracket refer to the 5 and 95 percentile

<sup>2)</sup> Consists of a few nature types only with similar requirements regarding N status, leading to very similar values for the various percentiles.

<sup>3)</sup> Consists of one nature type only, so all critical nutrient N load computations yield equal results

<sup>4)</sup> Consists of a few receptors only, leading to strongly skewed distribution

- 1 *Table 3 Preliminary critical loads for N based on preservation of the ground vegetation biodiversity according to the set conditions*  
 2 *for non-effect for 16 Swedish study sites*

Site	Time of vegetation response	Critical load deposition kg.ha <sup>-1</sup> .yr <sup>-1</sup>	Present deposition kg.ha <sup>-1</sup> .yr <sup>-1</sup>	Excess deposition kg.ha <sup>-1</sup> .yr <sup>-1</sup>	Required deposition reduction %
Högbränna	1910	1.1	1.5	0.4	27
Brattfors	1890	0.9	2.0	1.1	55
Storulvsjön	1925	2.0	3.5	1.5	43
Högskogen	1928	4.8	7.9	3.2	40
Örlingen	1910	3.6	8.5	3.9	52
Edeby	1918	3.9	7.8	3.9	50
Blåbärskullen	1880	1.6	8.5	6.9	81
Höka	1920	4.0	8.9	4.9	55
Hensbacka	1922	7.4	18.0	10.6	59
Söstared	1868	2.1	20.0	17.9	89
Gynge	1870	2.8	8.3	5.5	66
Fagerhult	1915	3.7	7.5	3.8	51
Bullsäng	1870	2.1	15.0	12.9	86
Timrilt	1889	3.6	23.0	19.4	84
Vång	1910	7.8	17.0	9.2	54
Västra Torup	1866	2.4	27.0	24.6	91

3

1 *Table 4 Key processes represented in the biogeochemical models used in model chains for assessing impacts of nitrogen on*  
 2 *biodiversity. SMART stands for simulation model for acidification's regional trends, SUMO for succession model, MAGIC for*  
 3 *Model for Acidification of ground water in catchments and SAFE for Soil acidification in forest ecosystems ● = modelled*  
 4 *dynamically; ○ = modelled indirectly or in a simplified way; k = included as constant or fitted term; - = not modelled.*

Process	SMART2 <sup>1</sup>	SMART2/SUMO <sup>1</sup>	MAGIC <sup>1</sup>	ForSAFE-VEG
Photosynthesis / tree growth	k	●	-	●
Competition / succession	-	●	-	●
Plant N uptake	●	●	○	●
Symbiotic nitrogen fixation	k	●	-	k
Litterfall	●	●	○	●
Decomposition	●	●	○	●
N mineralization	●	●	○	●
Nitrification	●	●	○	●
Denitrification	●	●	○	○
Inorganic N leaching	●	●	●	●
Organic N leaching	-	-	○	○
N immobilization	●	●	●	●
Soil carbon dynamics	●	●	○	●
SOM pools with different reactivity	●	●	-	●
Major ion chemistry/acidity	●	●	●	●
Base cation weathering	○	●	k	○
Grazing	-	●	○	●
Fire	-	●	○	●
Sod cutting	-	●	○	-
Tree felling	-	●	○	●

5 <sup>1)</sup> The combination of the vegetation model (GB)MOVE or NTM with either SMART2 or

6 MAGIC does not include any additional process compared to the use of the individual models

1 *Table 5 Comparison of the characteristics of MOVE/NTM, GBMOVE, VEG and SUMO where MOVE stands for*  
 2 *Model of Vegetation and GBMOVE for the Great Britain version of MOVE, NTM for Nature technical model, VEG for*  
 3 *vegetation model and SUMO for succession model .*

Characteristic	MOVE/NTM	GBMOVE	VEG	SUMO
<b>Methodology</b>				
Relation between abiotic conditions and species diversity	Statistical (Logistic & Splines)	Multiple Logistic Regression	Mechanistic competition model (growth functions)	Mechanistic competition model (growth functions)
Abiotic conditions as single stressors, combined single stressors or multistressors	Multistress (water content, pH, N-availability)	Multistress (% soil moisture, pH, C/N ratio, cover-weighted canopy height)	Combined single stressors (water content, pH, N, P, light, temperature, grazing)	Combined single stressors (water content, pH, N, P, light, grazing, management)
Crucial factor for critical load calculations	pH & N-availability	pH & N-availability	N-concentration in soil solution & pH & Al-concentration & plant competition (light & nutrients)	N-availability & pH & plant competition (light & nutrients)
Link between environment and biodiversity	<b>(In)direct</b> (correlations between mean Ellenberg-indicator values of plant releveés and abiotic measurements)	<b>(In)direct</b> (correlations between mean Ellenberg-indicator values of plant releveés and abiotic measurements)	<b>Indirect:</b> model-outcome as a result of differences in species-specific growth functions	<b>Indirect:</b> model-outcome as a result of differences in plant type-specific growth functions
<b>Applicability</b>				
Link with biodiversity targets	<b>Direct</b> (via protected species or protected habitat types of EU-	<b>Direct</b> (via indicator species designated by statutory agencies by	Indirect by calculating a relevant indicator.	Indirect: Only possible after link with a species model

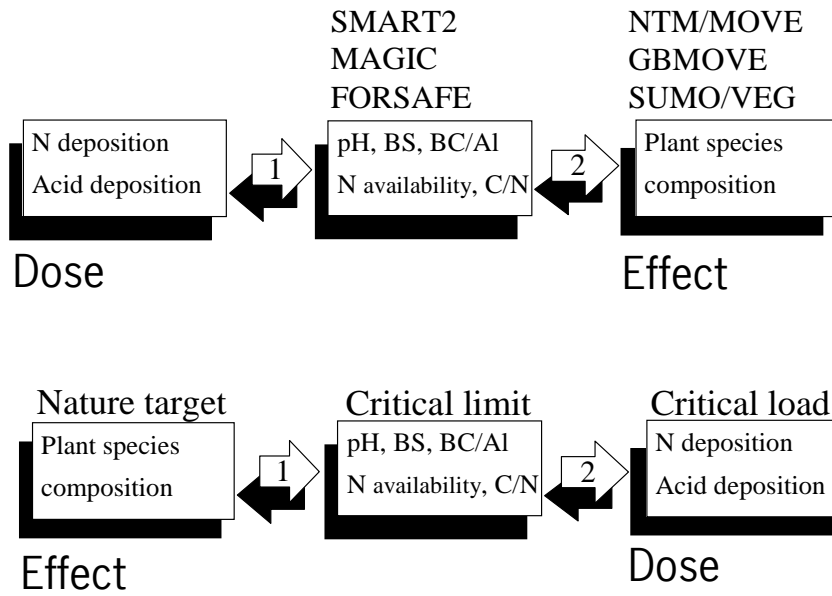
## Wim de Vries

Characteristic	MOVE/NTM	GBMOVE	VEG	SUMO
	habitat directive)	habitat)	-	-
Link with SMB for CL- calculations	<b>Possible</b> via acceptable N-leaching			
Link with dynamic soil models for critical load and target-load calculations	<b>Operational</b> (SMART2)	Operational (MAGIC)	Link with dynamic models is operational (SAFE) and used for target-load calculations	Link with dynamic models is operational (SMART2) but can't be used for target- load calculations
Countries for which the method is developed and tested/used	Methodology developed for the Netherlands, tested in UK, Denmark	Methodology developed for, and tested in, the UK	Methodology developed and tested in Nordic countries	Methodology developed for Netherlands, tested in UK.

1

2

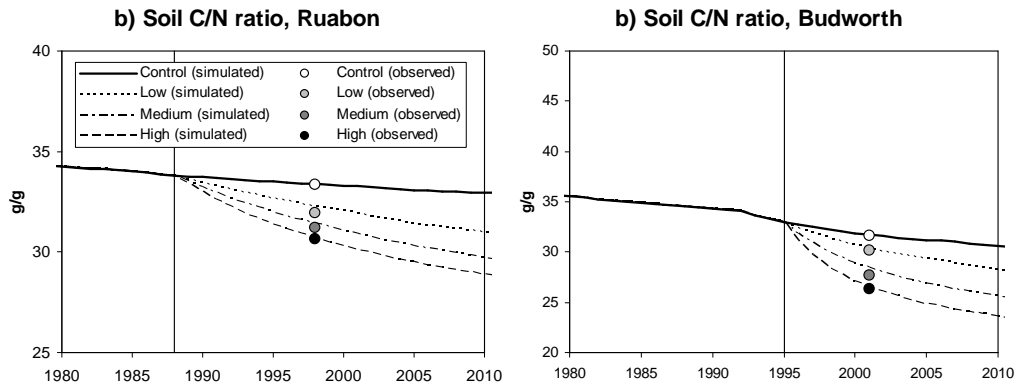
- 1 Figure legends
- 2 *Figure 1 Method to predict plant species composition as a function of atmospheric deposition (top) and to calculate critical loads for*
- 3 *nitrogen and acid deposition (bottom). The model abbreviations are explained in Table 4 and 5*
- 4 *Figure 2 Simulated and observed organic soil C/N ratio under ambient N deposition and three levels of long-term NH<sub>4</sub>NO<sub>3</sub>*
- 5 *addition at two heathland experimental sites. Vertical line indicates start of experiment.*
- 6 *Figure 3 Measured and simulated biomass harvest for a mown grassland site near Wageningen in the Netherlands (left) and for an*
- 7 *experimental grassland site at Rothamstead in the UK (right).*
- 8 *Figure 4 Percentage of species correctly predicted in the three Park Grass control plots (left) and Moorhouse (right) based predictions*
- 9 *by MAGIC+GBMOVE versus predictions based on observed mean Ellenberg scores only, as input to GBMOVE.*
- 10 *Figure 5 Predicted versus observed change in individual species in the Moorhouse Hard Hills control plots. Predicted change is the*
- 11 *slope coefficient of a linear regression on occurrence probabilities predicted by MAGIC+GBMOVE for each year between 1973*
- 12 *and 2001. Observed change is the slope coefficient of a linear regression on % frequency in sample plots in each survey year. Pearson*
- 13 *correlation coefficient = 0.568, p=0.002.*
- 14 *Figure 6 Modelled and measured pH values through the soil profile at 16 Swedish study sites*
- 15 *Figure 7 Modelled and measured ground vegetation occupancy of different plant groups at 2 Swedish study sites, i.e. Brattfors (left)*
- 16 *and Svartberget (right). The included line is the 1:1 relation..*



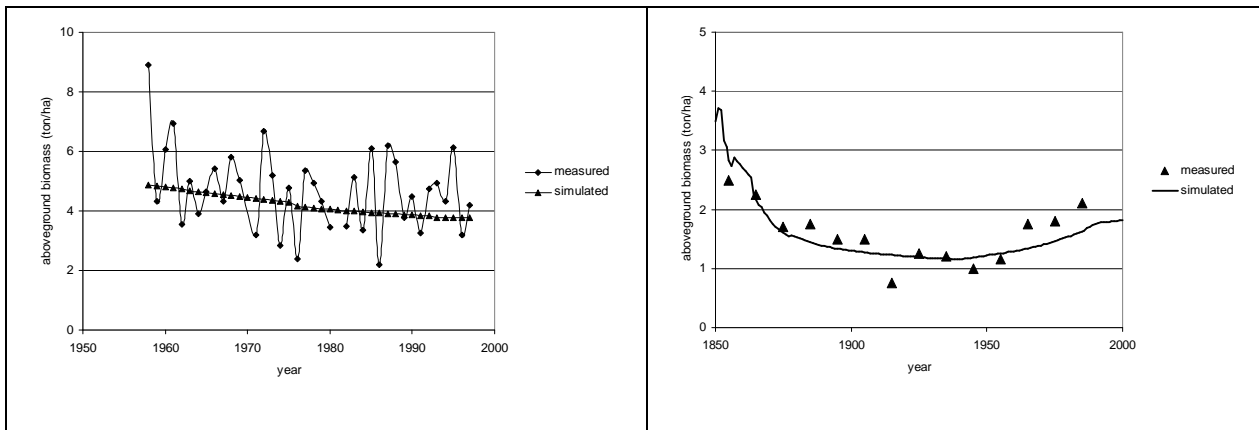
1  
 2 *Figure 1 Method to predict plant species composition as a function of atmospheric deposition (top) and to calculate critical loads for*  
 3 *nitrogen and acid deposition (bottom). The model abbreviations are explained in Table 4 and 5*

4



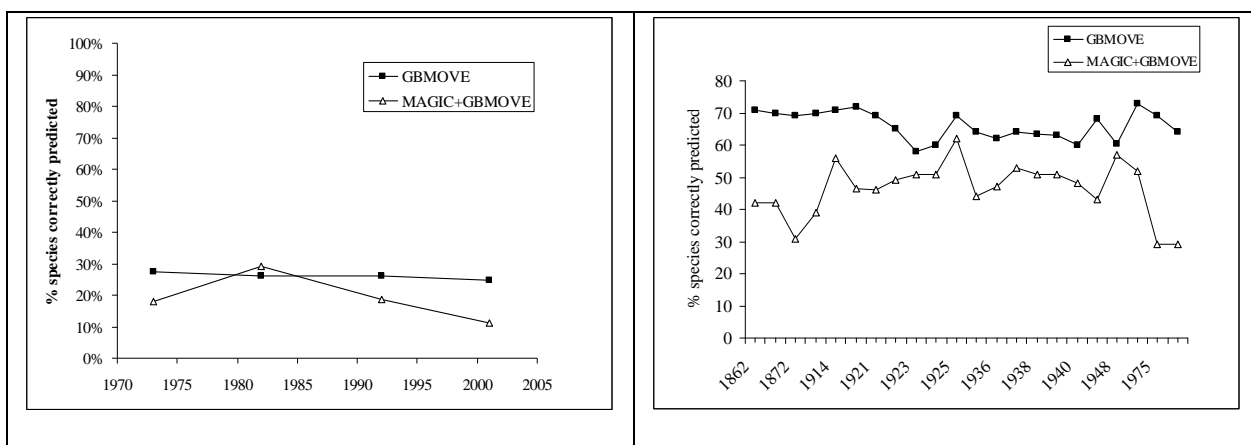


1  
 2 *Figure 2 Simulated and observed organic soil C/N ratio under ambient N deposition and three levels of long-term NH<sub>4</sub>NO<sub>3</sub>*  
 3 *addition at two beathland experimental sites. Vertical line indicates start of experiment.*  
 4



5 *Figure 3 Measured and simulated biomass harvest for a mown grassland site near Wageningen in the Netherlands (left) and for*  
 6 *an experimental grassland site at Rothamstead in the UK (right).*

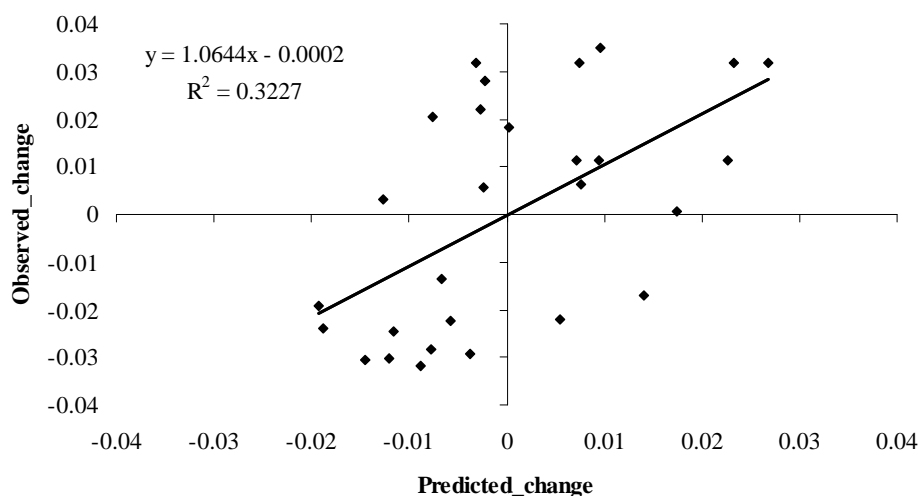
1



2 *Figure 4 Percentage of species correctly predicted in the three Park Grass control plots (left) and Moorhouse (right) based predictions*  
 3 *by MAGIC+GBMOVE versus predictions based on observed mean Ellenberg scores only, as input to GBMOVE.*

4

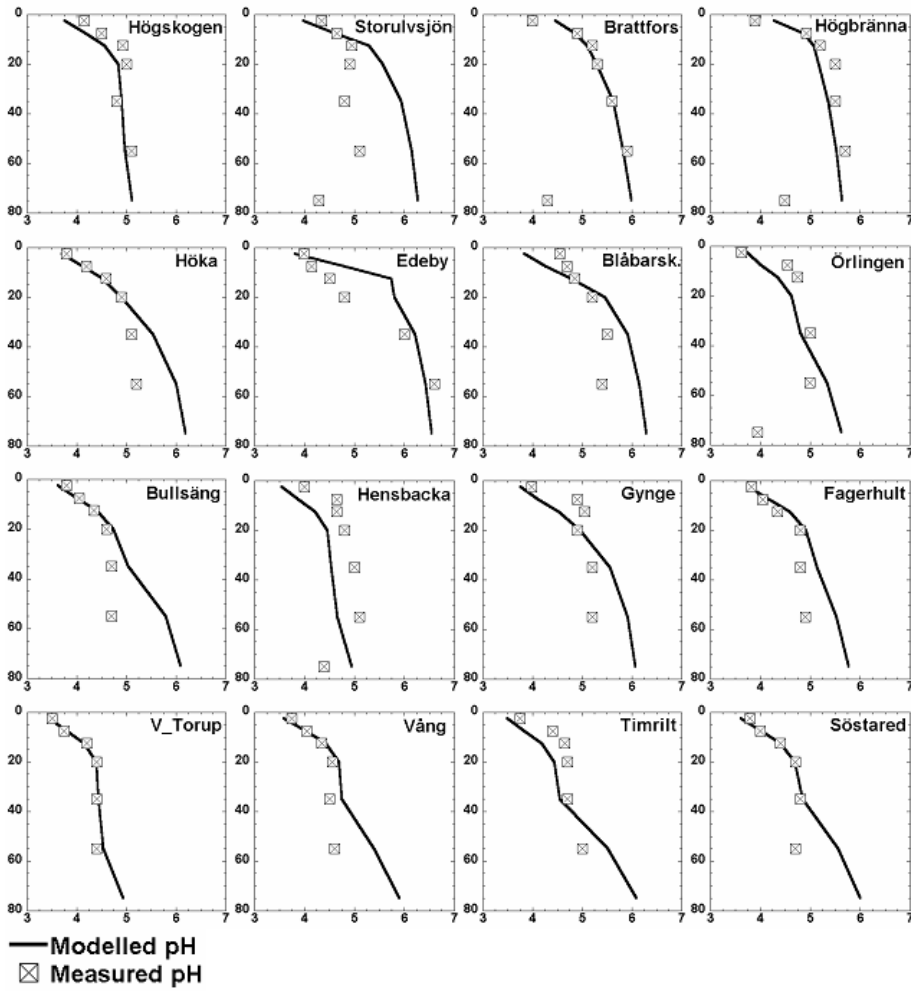
5



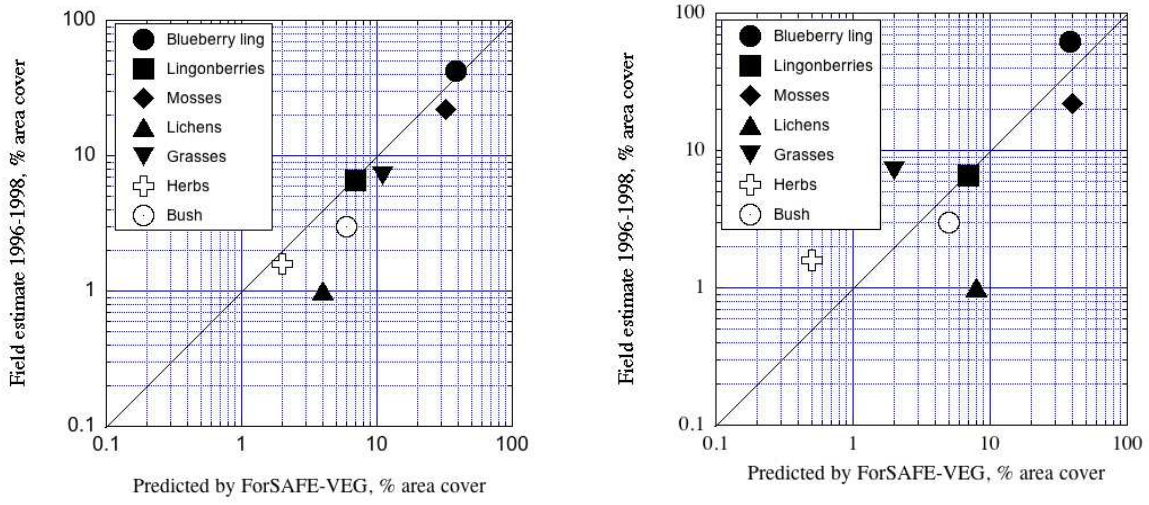
6

7 *Figure 5 Predicted versus observed change in individual species in the Moorhouse Hard Hills control plots. Predicted change is the*  
 8 *slope coefficient of a linear regression on occurrence probabilities predicted by MAGIC+GBMOVE for each year between 1973*  
 9 *and 2001. Observed change is the slope coefficient of a linear regression on % frequency in sample plots in each survey year. Pearson*  
 10 *correlation coefficient = 0.568, p=0.002.*

11



2 *Figure 6 Modelled and measured pH values through the soil profile at 16 Swedish study sites*



1

2

3 *Figure 7 Modelled and measured ground vegetation occupancy of different plant groups at 2 Swedish study sites, i.e. Brattfors (left)*

4 *and Svartberget (right). The included line is the 1:1 relation.*

5