

Balancing nitrogen use efficiency, losses and soil nitrogen depletion to evaluate national scale agri-environmental performance ~~across spatial scales~~ over 40 years

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Abstract. Nitrogen (N) is essential for agricultural productivity, but excessive N inputs result in substantial losses to the environment. Conducting N assessments at national scales is challenging because observational data are limited, especially over long time periods. Here we compiled detailed datasets and performed high-resolution biogeochemical modelling to quantify N budgets for Switzerland's diverse agricultural ecosystems over four decades, with a focus on croplands and
15 grasslands (i.e., permanent managed meadows used for livestock feed). Between the 1980s and the 2010s, N use efficiency improved from 47% to 57% in croplands and from 63% to 71% in grasslands, while losses through leaching and gas emissions decreased by 24% in croplands and 4% in grasslands. These improvements are closely linked to the implementation of national-scale agri-environmental policies that reduced fertilizer use in the 1990s. However, despite increased efficiency, cropland soils experienced substantial N depletion between 1995 and 2011 ($-23 \text{ kg N ha}^{-1} \text{ yr}^{-1}$) in
20 croplands. Our results demonstrate that policy reforms have improved agricultural system functioning and reduced losses, but also reveal risks associated with unbalanced soil N, underscoring the need for integrated N management for sustainable agriculture.

1 Introduction

Nitrogen (N) is a vital element supporting life. Before synthetic fertilizers were developed, N was often a limiting factor for
25 agricultural productivity (Vitousek and Howarth, 1991). The discovery of the Haber–Bosch process in the early 20th century enabled inert di-nitrogen (N_2) gas to be converted to biologically available N (in the form of ammonia), so-called “reactive nitrogen” (N_r) (Erisman et al., 2008), and boosted fertilizer production. Since the 1970s, the rapid increase of synthetic fertilizer use has greatly facilitated crop production (Fowler et al., 2013; Galloway et al., 2013). Nearly half of the global population are nourished by the N fertilizer produced using the Haber-Bosch process (Erisman et al., 2008). However, large
30 amounts of N are unintentionally lost to the environment, causing a wide range of environmental damages (Galloway et al.,

2003; Sutton et al., 2011). These N losses comprise gaseous emissions to the atmosphere, such as ammonia (NH₃), nitric oxide (NO), nitrous oxide (N₂O) and N₂, and leaching of nitrate (NO₃⁻) into groundwater; these losses negatively affect air, water and soil quality (Anderson et al., 2003; Dodds and Smith, 2016; Moldanová et al., 2011; Sutton et al., 2013); damage ecosystems and biodiversity (Krupa, 2003; Sutton et al., 2020); and contribute to climate warming (Stocker et al., 2013; Zhu et al., 2025).

To evaluate the N performance of a system and the associated environmental impacts, N use efficiency (NUE) is a widely used indicator. NUE is defined as the ratio of N outputs to N inputs, with higher NUE indicating more of the inputs going toward their intended use. For cropping systems, NUE is calculated as the N in harvested products divided by total N input (i.e. synthetic and organic fertilizers, biological N fixation and atmospheric N deposition). It is usually reported with other indicators describing the magnitude of N use, such as N yield or N surplus (difference between the N inputs and outputs). The NUE approach can be extended to the national and global level to track efficiency changes over time. It has been reported that global average NUE declined from 68% to 45% between 1961 and 1980 and subsequently stabilized over the next three decades (Lassaletta et al., 2014). Global average N use efficiency (NUE; N in harvested products divided by total N input) declined from 68% to 45% between 1961 and 1980 and subsequently stabilized over the next three decades (Lassaletta et al., 2014). In many countries, marked reductions in NUE resulted from intensified fertilization, which also led to elevated N losses (Lassaletta et al., 2014). These issues underscore the need for evaluation of agri-environmental performance by means of robust N indicators and advanced methodologies. A common approach is N budgets (Oberson et al., 2024; Oenema et al., 2003; Zhang et al., 2015, 2021), which offer an insightful understanding of N sources and fates by quantifying N inputs (i.e. synthetic and organic fertilizers, biological N fixation and atmospheric N deposition) and outputs (i.e. N removed in harvest and various loss pathways). This approach is increasingly recognized by researchers, farmers, policy makers and other stakeholders as a critical tool for understanding the N cycle, informing decision-making and promoting better management practices for pollution mitigation (Quemada et al., 2020; Zhang et al., 2021). However, a major limitation of existing N budget studies is the lack of spatial and temporal data to estimate N budgets at regional scale over time. Hence, understanding how N budgets, and soil N changes, especially at large spatial and temporal scales, remains a critical research need. One way to overcome this lack of spatial and temporal data is to use well-calibrated and validated state-of-the-art biogeochemical ecosystems models, such as DayCent, DNDC, EPIC, etc., to provide reliable data on ecosystem-wide N budgets across space and time.

Swiss agriculture is fundamentally shaped by pronounced topographic heterogeneity, broad climatic gradients and long traditions of agro-pastoral management. The country's agricultural landscape covers approximately 1.5 million hectares (BFS, 2024) and can be structurally classified into three primary land use categories: **cropland**, **managed grassland** (meadow and pasture), and **summer pasture** (seasonal alpine pasture). Each of these systems fulfils distinct functional roles and exhibits unique spatial distributions. Croplands provide the basis for intensive arable production and grasslands constitute the primary resource base for Switzerland's ruminant livestock sector. From 1950 onwards, agricultural intensification generated substantial productivity gains but also exacerbated environmental problems (Spiess, 2011). Nitrate

65 leaching from farmland, for example, has increased N loads not only to local lakes and rivers (Gächter et al., 2004; Müller et al., 2022), but also to the river Rhine, contributing to eutrophication in the North Sea (Prasuhn and Sieber, 2005). These adverse impacts prompted revisions of Swiss agricultural and agri-environmental policies to mitigate N pollution (Decrem et al., 2007; Herzog et al., 2008).

70 In this study, we used the biogeochemical model DayCent to simulate N cycling for agricultural land across space and time, with Swiss agriculture as an exemplary case. We assembled spatially explicit datasets of N inputs, meteorological variables, soil properties, land use, crop rotations and local management practices at the national scale. Applying DayCent at a high spatial resolution of 1km×1km, we constructed N budgets and calculate soil N changes for two major agricultural ecosystems (croplands and grasslands; hereafter “grasslands” means “managed meadow”) over the period 1981-2020. We developed an informative analytical diagram that holistically evaluates NUE, N losses and soil N stocks. We found improved
75 NUE and decreased N losses, as well as a heightened risk of soil N depletion in Swiss agriculture. This result points to the need for a more integrated N assessment to balance agroecosystem performance, losses to the environment, and soil resource maintenance at regional and national scales.

2 Materials and Methods

2.1 DayCent model

80 DayCent is a process-based biogeochemical model that simulates the dynamics of both nitrogen and carbon cycling across various terrestrial ecosystems (Del Grosso et al., 2001). DayCent integrates environmental drivers to predict plant growth, soil organic matter (SOM) decomposition, trace gases and changes in other ecosystem parameters within the soil-plant-atmosphere continuum on a daily timestep. With intermediate complexity and the feasibility to be calibrated to local conditions, DayCent is widely used for evaluating ecosystem responses to land use change, management practices and
85 climate variability (Del Grosso et al., 2005; Gurung et al., 2020, 2021; Laub et al., 2024; McClelland et al., 2025).

In this study, we used the DayCent17centEVI model version to quantify N budgets for two Swiss land ecosystems: a) croplands and b) grasslands. The SOM in DayCent is split into three compartmental pools, namely active, slow and passive, with different potential decomposition rates. These SOM pools receive plant materials from above and belowground litter. Simulated N flows follow C flows depending on the C:N ratio which varies from 15 to 3 for the
90 active pool, from 20 to 12 for the slow pool, and from 10 to 7 for the passive pool. For newly formed surface biomass, the C:N ratio is a function of the N content of the decomposed material, with higher C:N ratio for lower N content. The model accounts for both organic and mineral N and the relevant processes, including mineralization-immobilization turnover, nitrification and denitrification. DayCent has a mechanistic representation of the N cycle. The model accounts for multiple N pools (mineral and organic) and processes, including mineralization immobilization turnover,
95 ~~nitrification and denitrification.~~ These transformations are regulated by environmental factors such as temperature, soil moisture, soil texture, SOM content and oxygen availability. Nitrogen inputs are modelled through simulated atmospheric

deposition, ~~and~~ biological N fixation, and through input data to the model that record application of both organic and synthetic fertilizers, ~~while-Simulated~~ losses are outgassing of N₂O, N₂, NO and NH₃ to the atmosphere and leaching (e.g., nitrate and organic forms). It is crucial to acknowledge that NH₃ volatilization in DayCent is not sophisticated, which may lead to underestimation of NH₃ emissions and overestimation of nitrate leaching.

DayCent enables explicit parameterization of agricultural management practices. It can dynamically accommodate crop systems (e.g., crop types and rotations), cultivation, irrigation, nutrient inputs (e.g., fertilization) and harvest, with the timing of each management event specified on a daily basis in simulations.

2.2 Model input data

DayCent is driven by weather data and soil data. Three basic meteorological variables include: daily maximum and minimum temperature, and precipitation. We used the 1km×1km resolution weather data from the Federal Office of Meteorology and Climatology (MeteoSwiss) (<https://hyd.ifu.ethz.ch/research-data-models/meteoswiss.html>; last access 12.01.2024). Inputs of site-specific soil properties such as soil texture (sand, silt, clay), soil pH and soil organic matter (SOM) were obtained from a recently developed national soil database by the National Competence Center for Soil (“Kompetenzzentrum Boden”) (Stumpf et al., 2024). Soil properties originally available at 30 m × 30 m resolution were aggregated to the 1 km model grid using conservative remapping implemented in the Climate Data Operator (CDO) (Schulzweida, 2023). This approach preserves area-weighted means and is useful in upscaling environmental datasets to coarser modelling resolutions (Hashimoto et al., 2025). Because the model operates at 1 km spatial resolution, harmonization of all input datasets was required to ensure consistency across input variables. While this aggregation inevitably smooths fine-scale heterogeneity in soil properties, the most prominent features of the geospatial distribution of soil variables at the national scale are preserved. ~~The original spatial resolution of these soil inputs were 30m×30m, which were resampled to 1km×1km using the conservative remapping method in Climate Data Operator (CDO) (Schulzweida, 2023).~~ Other required soil properties including bulk density, field capacity, wilting point and saturated hydraulic conductivity were determined by the pedotransfer functions embedded in the DayCent utility programme (Saxton et al., 1986; Saxton and Rawls, 2006), and root fractions were used the default values in DayCent.

In addition to meteorological and soil inputs, DayCent also needs data of land use and management practices. Our modelling simulations focused on two major Swiss land use categories: croplands and grasslands. Historical land use and areas of croplands and grasslands were from Federal Statistical Office (“Bundesamt für Statistik”, BFS) (BFS, 2024). Grasslands in Switzerland are categorised into three major types: 1) meadows, 2) pastures and 3) summer pastures. In this study, we focused on permanent meadows, which are managed for grass production for livestock feed. According to the definitions used for agricultural subsidies and the national fertilizer guidelines (Sinaj et al., 2017), meadows are further divided into intensively-managed, less intensively-managed and extensively-managed meadows, depending on the intensity of management practices (fertilization level and mowing events). These categories correspond to categories defined for agricultural subsidies and the national fertilizer guidelines (Sinaj et al., 2017), meaning category-specific data are available.

130 Annual areas of different cropping systems and grasslands are provided for 24 agri-climatic zones. These are defined in Wüst-Galley et al (2020) and have similar broad climatic conditions and agricultural management; they incorporate geographical (i.e. regions) as well as topographical differences in Switzerland (i.e. valley, hill, mountain and summer pastures).

A crucial input for DayCent simulations is the amount of C and N applied to soils from both organic and synthetic
135 fertilizers. Organic fertilizers comprise animal manure as well as compost, sewage sludge and digestates as assessed by the Swiss National Greenhouse Gas Inventory (<https://www.bafu.admin.ch/bafu/en/home/topics/climate/state/data/climate-reporting/ghg-inventories/latest.html>; last access: 28.10.2024). Animal manure, which represents by far the largest amount of organic fertilizer, is assessed considering livestock numbers, livestock species specific excretion rates for C and N and respective losses from stables and manure storage systems. The allocation of organic fertilizers to different crops or
140 grasslands is carried out in according with the following factors, as described in Wüst-Galley et al (2020): the tendency of farms to apply manure, slurry or poultry manure to different broad crop groups, from Kupper et al. (2022)–Sinaj et al (2017); the relative fertilizer requirements of different grassland types, across different elevation zones, as indicated in Sinaj et al (2017). For synthetic fertilizer, we used the fertilizer import data from the National Greenhouse Gas Inventory. We assumed that all imported fertilizer was applied to the fields in the corresponding year as ammonium nitrate which is the major type of
145 synthetic fertilizer. Other management data such as cultivation, planting, fertilization and harvest were taken from the national “Principles for the fertilisation of agricultural crops in Switzerland” book (Sinaj et al., 2017) and modelling setups (Lee et al., 2020b, a).

During the 1990s in Switzerland, important nationwide agri-environmental policy reforms introduced
150 subsidy schemes in order to mitigate nutrient losses. Meanwhile, there was a significant increase in livestock productivity (i.e., milk yield) during this period and in line with decreased livestock population without lowering production, resulting in decreased N inputs from livestock manure. These led to substantial reductions in synthetic and organic fertilizer application to agricultural soils. The fertilizer input data we used reflect these changes. In addition, the use of cover crops also forms part of the agri-environmental policy that was implemented in the late 1990s, with the aim of reducing soil erosion and nutrient leaching, particularly in winter months. However, this
155 potentially relevant measure was not covered by our study due to insufficient statistical data. Other agricultural practices were also assumed to remain unchanged.

2.3 Model simulations

The DayCent model that we used has been calibrated and tested by previous modelling studies (Dos Reis Martins et al., 2022, 2024; Necpalova et al., 2018). We used reported values of parameters controlling plant growth (both C and N yields)
160 and N processes (e.g., nitrification and denitrification) by these studies which evaluated against measurement data from several Swiss long-term experiments for croplands (Emmel et al., 2018; Hüppi et al., 2015; Krauss et al., 2017; Mayer et al., 2015) and Swiss Fluxnet sites for grasslands (Feigenwinter et al., 2023b). In this study, the regional simulations for

Switzerland were performed at 1km×1km grids for the time period from 1981 to 2020. In total, there are 14926 and 36140 simulated grids for croplands and grasslands, respectively. A complete round of DayCent simulations had two stages: historical spin-up and present baseline. We followed the Lee et al (Lee et al., 2020b, a) and assumed five spin-up phases characterised in Swiss agriculture: (1) native forest (between 0 and 1399; until equilibrium), (2) emergence of agriculture (between 1400 and 1750), (3) agricultural revolution (between 1751 and 1850), (4) agricultural intensification (between 1851 and 1950), and (5) modern agriculture (from 1950 to 1980). After the historical runs, DayCent was kept running for the studied time period (i.e., 1981 to 2020).

2.3.1 Simulations for croplands and crop rotation scheme

We included 15 crops in the simulations for croplands, which together account for over 95 % of Swiss croplands. These crops are grass-clover ley (temperate grass in crop rotation), winter wheat, silage maize, barley, rapeseed, sugar beet, grain maize, potato, triticale, spelt, sunflower, pea, rye, soybean and oat. Among these simulated crops, grass-clover ley and cereals (winter wheat, maize, barley) account for a dominant share of over 60% of cropland areas. In Switzerland, crop rotation is a common practice that is based on pedoclimatic conditions and production need. For example, a forage-crop rotation with two years of temporary grass followed maize and winter wheat in the next two years is a common sequence. However, national crop rotations data at fine spatial scale (e.g., 1km×1km) are not currently available. Therefore, we derived the crop rotations using the probability scheme in Lee et al (2020) for the entire country, which is developed from survey, existing literature, long-term experiments in the country and expert judgement (see Table S1 in Supplementary Materials). This crop rotation scheme incorporates nationwide guidance and recommendations for “best practice” aiming to avoid harmful development such as pests, diseases or pathogens. On the other hand, it also reflects what happened in real farming practices that were sourced from surveys of farmers. The purpose of developing such a scheme is to ensure that our simplified rotation in the model can replicate the reality as much as possible, given the limited information and resources. We then used this probability rotational scheme to predict sequences and determine the most likely crop rotation through an iterative process, while ensuring our modelling results for the areas of crops consistent with statistical data at both the national level and regional level (i.e. 24 agro-climatic zones) (Wüst-Galley et al., 2020). This approach distributed the crop types following a ranked order based on crop areas, i.e., the crop type with the largest area is selected first, then the crop type with the second largest area, until all crop types are selected.

2.3.2 Simulations for grasslands

For grassland (meadow) simulations, the number of mowing events and timing were sourced from grassland-use intensity maps for Switzerland (Weber et al., 2024). These grassland-use intensity maps were generated from Sentinel-2 and Landsat 8 satellite data for Switzerland using a rule-based algorithm that identifies drops in vegetation index time series (Weber et al., 2024). Since these maps were mainly produced for 2018-2021, we chose the data of year 2020 as the baseline because year 2020 has been assessed by independent publicly available reference data. In principle, fertilization levels such

195 as number of fertilization and application rates are influenced by the management intensity (more mowing events, higher
fertilizer inputs) and negatively related to altitude, with less fertilizer inputs in more elevated places (Sinaj et al., 2017).
Fertilizer application is assumed to take place within a two-week window after a mowing event. We developed a fertilization
timing scheme based on management practices between 2005 and 2020 of an intensively-managed grassland reported by
Feigenwinter et al. (2023a, b) and Hörtnagl et al. (2025)~~Feigenwinter et al. (2023)~~ and applied to the whole country (see
200 Fig. S1 in Supplementary Materials). Grazing is not simulated in this study, of which deposited livestock excretion has been
accounted in the C and N data we used.

2.3.3 Sensitivity tests and scenario simulations

To better understand the sensitivity of simulated N dynamics to environmental conditions, we conducted a series of
one-factor-at-a-time sensitivity tests using the baseline simulation as a reference. To assess the influence of
205 meteorological drivers, we applied uniform changes of (i) $\pm 2^{\circ}\text{C}$ in air temperature and (ii) $\pm 20\%$ in precipitation,
while keeping all other inputs unchanged. Air temperature and precipitation were selected because they are the
primary meteorological drivers controlling plant growth and soil C and N cycling in DayCent.

In addition, we performed three fertilizer management scenarios: (i) replacement of all organic fertilizer
with synthetic fertilizer while maintaining the same total N application rate, (ii) replacement of all synthetic fertilizer
210 with organic fertilizer while maintaining the same total N application rate, and (iii) a 20% reduction in both synthetic
and organic fertilizer application rates. Each scenario was simulated independently, with all other model inputs and
management practices identical to those of the baseline simulation. Changes in total N input, N harvest, NUE, total N
losses and soil N stock change were evaluated relative to the baseline simulation.

2.4 Nitrogen budgets construction

215 We analysed total N inputs, N outputs, N surplus, NUE, N losses and soil N balance from the DayCent simulations, and
quantified N budgets. The total N inputs (T_{in}) to agricultural systems (croplands and grasslands) include organic (N_{org})
and synthetic fertilizers (N_{syn}), biological N fixation (N_{BNF}) and atmospheric N deposition (N_{dep}). As mentioned in Section
2.1 and 2.2, BNF (including both non-symbiotic soil N fixation and symbiotic plant N fixation) and atmospheric
deposition are simulated by the DayCent model, whereas organic and synthetic fertilizers are model inputs.

$$220 \quad N_{in} = N_{org} + N_{syn} + N_{BNF} + N_{dep} \quad (1)$$

Total N outputs (T_{out}) include N yields of harvested products and all forms of N losses.

$$N_{out} = N_{yield} + N_{loss} \quad (2)$$

N surplus is total N inputs minus N yields.

$$N_{surplus} = N_{in} - N_{yield} \quad (3)$$

225 The NUE is defined as the harvested crop or grass N (N_{harv}) divided by total N inputs (Lassaletta et al., 2014; Zhang et al., 2021)

$$\text{NUE} = \frac{N_{\text{yield}}}{N_{\text{in}}} \times 100 \quad (4)$$

For N losses (N_{loss}), we included gaseous losses (N_{gas} ; NH_3 , NO , N_2O and N_2) and leaching (N_{leaching} ; e.g. nitrate and organic N compounds)

230
$$N_{\text{loss}} = N_{\text{gas}} + N_{\text{leaching}} \quad (5)$$

The soil N balance ($\Delta\text{soil N}$) is calculated by subtracting all N outputs including harvested crop N and N losses from total N input

$$\Delta\text{soil N} = N_{\text{in}} - N_{\text{out}} = N_{\text{org}} + N_{\text{syn}} + N_{\text{BNF}} + N_{\text{dep}} - N_{\text{yield}} - N_{\text{loss}} \quad (6)$$

Specifically, N flows expressed as percentages of total N inputs were calculated as:

235
$$\%N_{\text{flow}} = \frac{N_{\text{flow}}}{N_{\text{in}}} \times 100 \quad (7)$$

Similarly, the relative changes in soil N stock were calculated as:

$$\%\Delta\text{soil N} = \frac{\Delta\text{soil N}}{N_{\text{in}}} \times 100 \quad (8)$$

These results are shown in Figs. 5 and 6.

2.5 Uncertainty evaluation

240 We used a Monte Carlo ensemble approach (200 iterations) to quantify the uncertainty in simulated N losses, NUE and soil N balance arising from key processes represented in DayCent. Nine model parameters regulating N leaching, nitrification and denitrification were selected for the analysis (details are presented in Table S2). Parameter ranges were defined using probability distributions based on the default values and calibrated ranges reported in previous studies (Dos Reis Martins et al., 2022, 2024). Because these parameters also influence plant N uptake through their
245 effects on N loss pathways, plant-specific parameters were not evaluated separately.

To reduce the computational burden associated with the large number of simulations, we created a representative subset of grid cells using Conditioned Latin Hypercube Sampling (cLHS). This subset (n=400) preserved the spatial variability of climate, soil properties, and fertilization intensity across Switzerland. For each Monte Carlo iteration, DayCent was run for all selected grid cells using a unique set of parameter values. Simulated
250 N losses, NUE, and soil N balance were then calculated and compared with the baseline simulation. Uncertainty in national-scale estimates was quantified from the distribution of ensemble results and reported as 95% confidence intervals (CI).

3 Results

3.1 Nitrogen inputs through fertilization decreased in the 1990s

Our compiled N datasets show that livestock manure and synthetic fertilizers, the dominant N sources for crop and grass production in Switzerland, decreased substantially from 204 kt N yr⁻¹ to 156 kt N yr⁻¹ between the 1980s and the 2010s. This result reflects policy interventions introduced in the 1990s to reduce agricultural N losses. In Switzerland, this period marked a broader societal and political shift in Switzerland, moving priorities from a narrow focus on maximizing food production to more sustainable and environmentally responsible farming approaches. In 1993, agricultural policy was reframed (Decrem et al., 2007; Herzog et al., 2008; Spiess, 2011), with direct payments (subsidies) introduced within an agri-environmental scheme, replacing the earlier model of guaranteed government purchases (Herzog et al., 2008). Moreover, integrated and organic production systems were promoted by additional incentives, organic farming and other ecological programmes. The constitutional amendment in 1996 further reinforced this direction by formally recognizing the multiple roles of agriculture, including ecological stewardship. Cross-compliance were confirmed in 1998 through the Proof of Ecological Performance (PEP) (Decrem et al., 2007; Herzog et al., 2008; Spiess, 2011), which made direct payments conditional to farms maintaining balanced nutrient budgets. These reforms collectively led to a ~25% reduction in average synthetic N fertilizer consumption during the 1990s (Herzog et al., 2008), accompanied by a comparable decline in manure application due to decreased livestock numbers. By 2005, 97 % of agricultural land in Switzerland was reported being managed in accordance with PEP standards (Herzog et al., 2008).

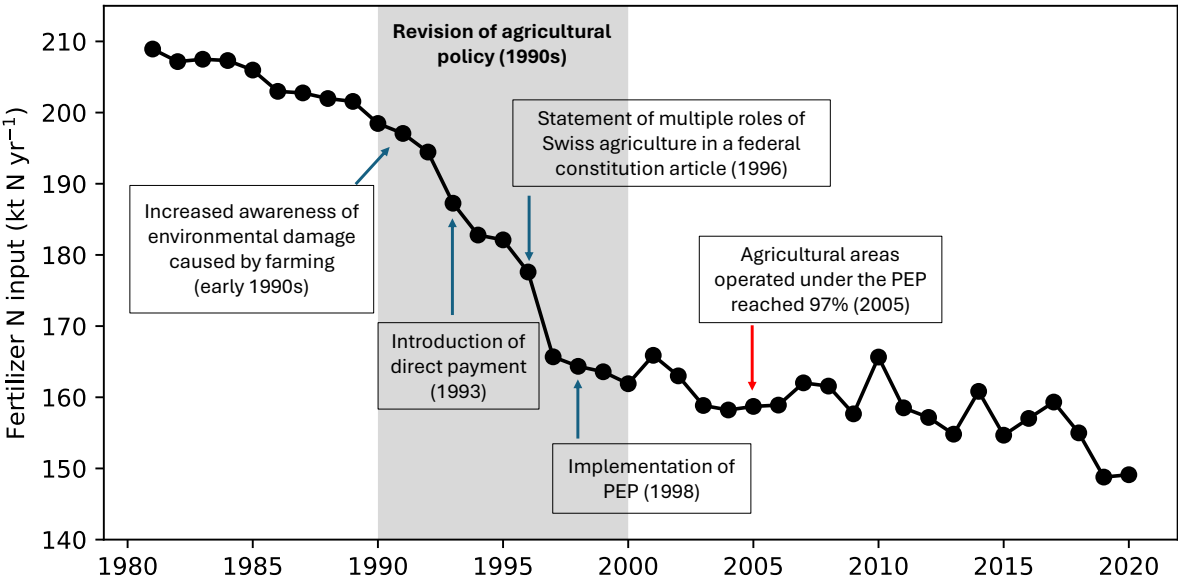


Figure. 1. Nitrogen inputs from fertilizers to agricultural land in Switzerland. Fertilizers include livestock manure and synthetic fertilizers. Values from compiled datasets (see Methods). Shaded grey area represents the period in the 1990s when national policies and measures were implemented in the agricultural sector. Important events are shown along the timeseries of fertilizer

275 nitrogen inputs. Values in the parenthesis are corresponding years. Note that the y-axis starts from 140 kt N yr⁻¹. PEP is Proof of Ecological Performance.

3.2 Nitrogen use efficiency of croplands and grasslands

DayCent simulations suggest that NUE increased in both croplands and grasslands over the simulated 40 years. Nitrogen yields in both agricultural ecosystems remained stable from 1981 to 2020, despite a decline in N inputs, resulting in notable increases in NUE (Figs. 2–4). The model results are consistent with data reported by the Swiss Farmers’ Union, which showed stable yields for major crops between 1991 and 2013 (Figs. S2–3 in Supplementary Materials). The most pronounced improvements occurred in the 1990s, coinciding with the implementation of policy measures aimed at controlling agricultural N surplus. Grasslands show higher N yields and NUE compared with croplands, with a steady increasing trend with relatively low inter-annual variability (Fig. 2C).

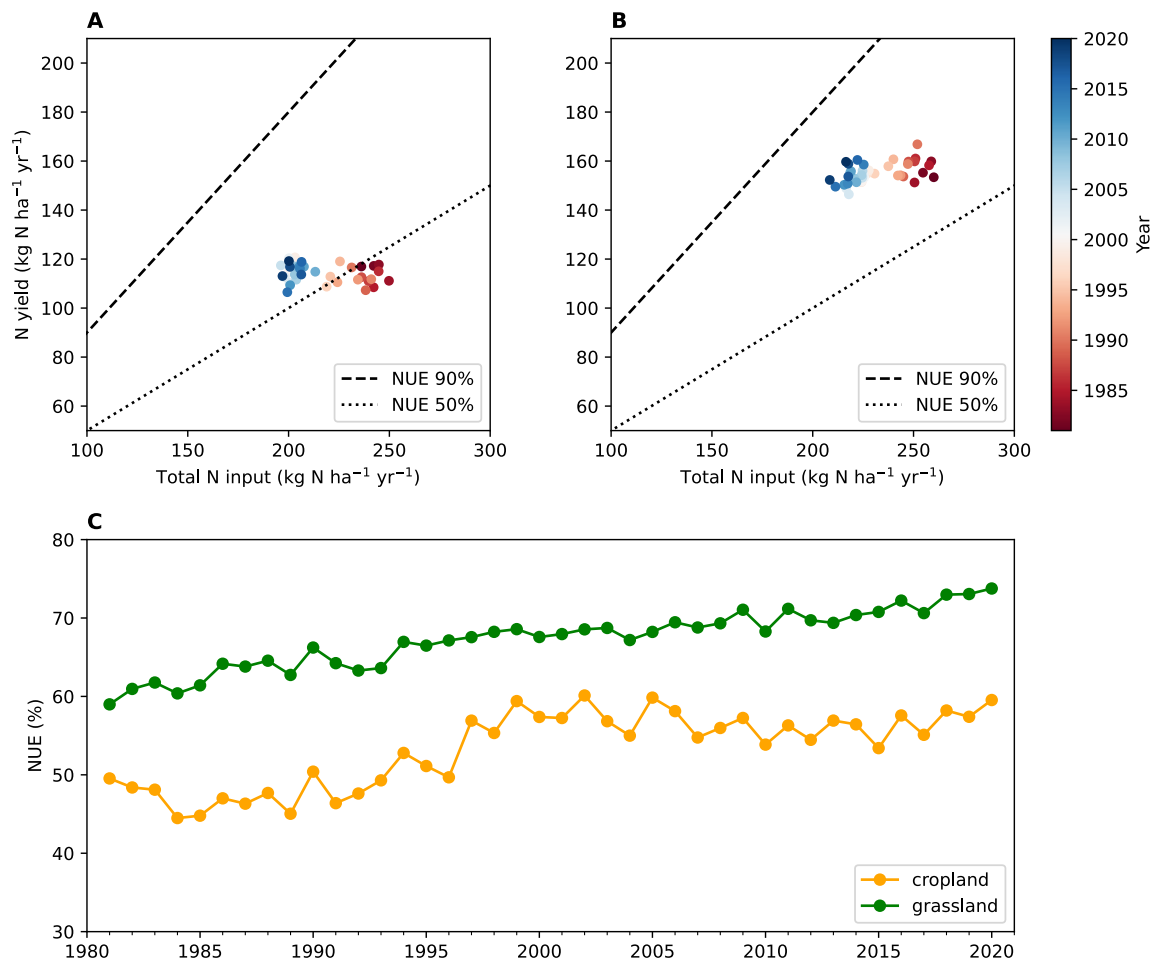
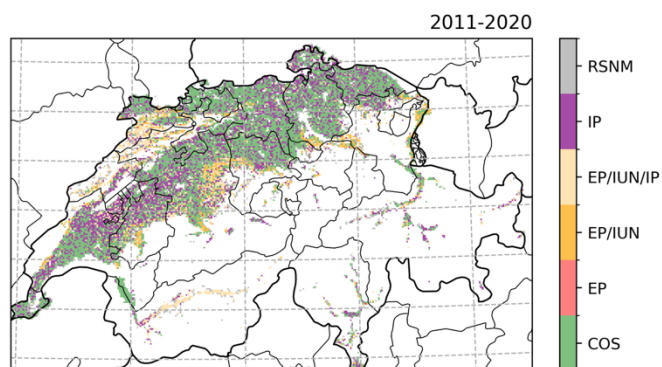
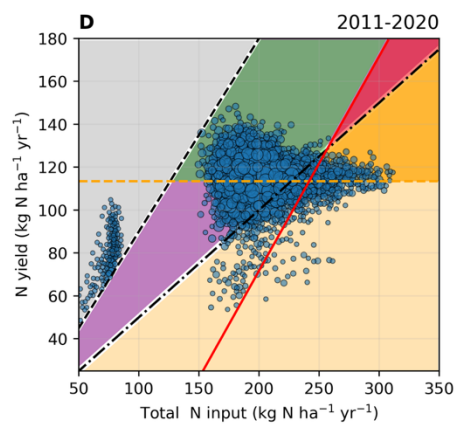
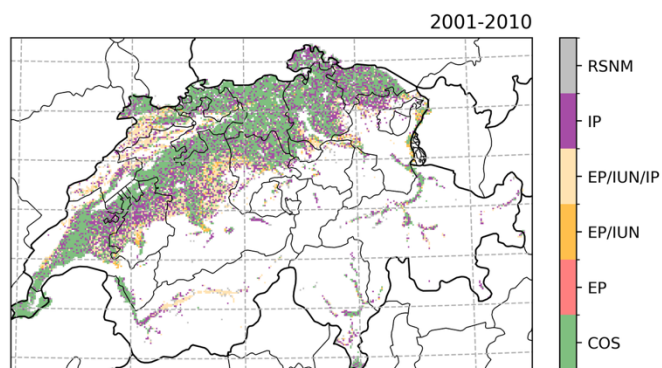
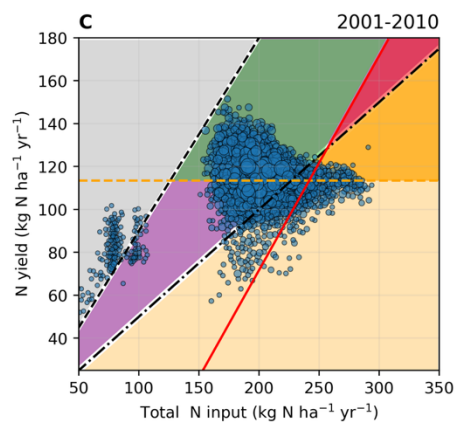
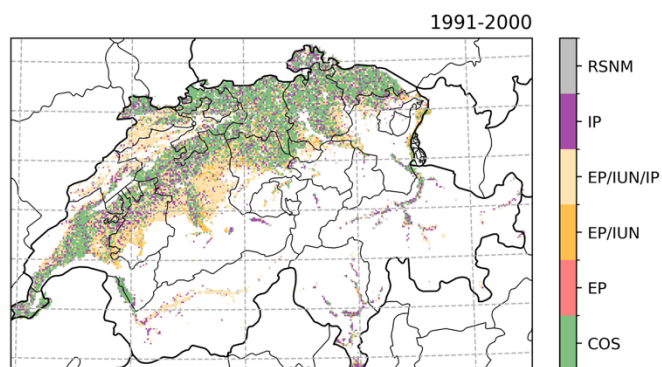
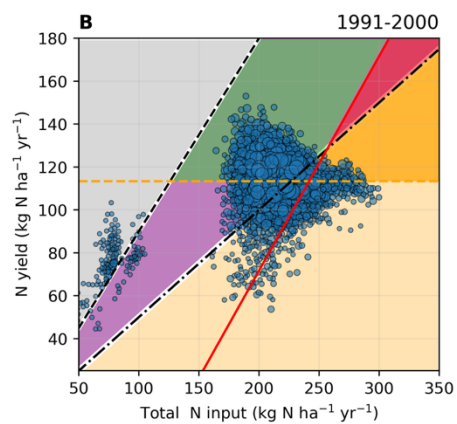
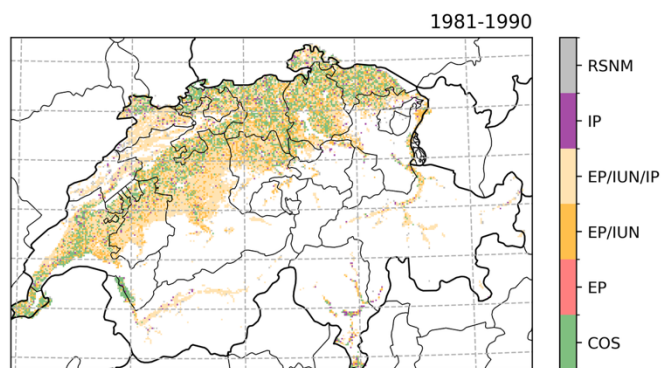
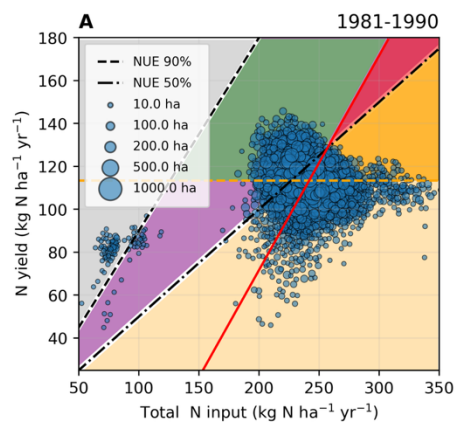


Figure. 2. Modelled annual mean total N input, N yield and NUE of Swiss croplands and grasslands from 1981 to 2020. Total nitrogen input and N yield of croplands (A) and grasslands (B). In panels a and b, the dashed black lines and dotted black lines represent 90% and 50% NUE, respectively and the red-blue scale shows the year. Note the axes do not start from zero. (C) NUE of croplands (yellow) and grasslands (green). Note the y-axis starts from 30%.

To further evaluate NUE, we applied the European Nitrogen Experts Panel (EUNEP) framework (*Nitrogen Use Efficiency (NUE) an Indicator for the Utilization of Nitrogen in Food Systems*) to assess improvements across four decades (from 1981 to 2020). The EUNEP framework classifies agricultural land into six regimes (Figs. 3–4 and Fig. A1 in Appendix), with the **characteristic operating space (COS)** representing the optimal agri-environmental performance. **COS** is defined by: 1) efficient use of nitrogen (NUE between 50% to 90%), 2) satisfactory N yield, and 3) controllable N surplus (total N input minus N yield). Other regimes outside the COS correspond to distinct agri-environmental issues (see Fig. 3 caption for details). Our results indicate remarkable progress in both croplands and grasslands: in the 1980s, only 26% of croplands and 15% of grasslands fell within COS. By the 2010s, these percentages increased to 56% and 77%, respectively (Table 1). Extremely high N inputs to grasslands (>350 kg N ha⁻¹ yr⁻¹) were largely abolished after 2011 (Fig. 4D and Fig. A1B in Appendix), and COS areas expanded geographically (into the central plateau where intensive crop production takes place) between 1981 and 2020, gradually becoming the dominant regime in both ecosystems (Figs. 3–4).

Table 1. Categorised agricultural land in Switzerland based on the EUNEP framework. Total areas of croplands and grasslands* and percentage of areas that belong to six EUNEP categories for in the 1980s and 2010s. *Grasslands refer to managed meadows only, while managed pastures and summer pastures for grazing are not included.

	Period	Areas (kha)	COS	EP	EP/IUN	EP/IUN/IP	IP	RSNM
Cropland	1981-1990	435	26%	0%	25%	46%	3%	<1%
	2011-2020	384	56%	0%	4%	8%	32%	<1%
Grassland*	1981-1990	394	15%	65%	2%	<1%	17%	<1%
	2011-2020	393	77%	9%	<1%	<1%	12%	1%



310 **Figure. 3. The EUNEP Framework of the NUE indicator diagram of Swiss croplands and decadal geographical distributions of six**
categorised agricultural land. (A)1981-1990 (B)1991-2000 (C) 2001-2010 (D) 2011-2020. For the NUE diagram, the dashed black
lines are 90% NUE, and dotted black lines are 50% NUE. The solid red lines and dashed orange lines represent desired maximum
N surplus and desired minimum N yield, which are the mean values of simulated N surplus and N yield for each agroecosystem in
the 1980s, which set goals for improvement in the following decades. Six regimes are (note the colour scheme is different from the
315 **originally proposed diagram(Anon, n.d.)): characteristic operating space (COS – shaded green area), excessive pollution (EP –**
shaded red area), insufficient productivity (IP – shaded purple area), EP and inefficient use of nitrogen (EP/IUN – shaded orange
area), EP/IUN/IP (shaded light yellow area), risk of soil nitrogen mining (RSNM – shaded grey area). The dark blue and light blue
circles represent data from 1981-1990 and 2011-2020, respectively. The size of the circle is proportional to the summed area of
320 **croplands or grasslands that is aggregated by N yield and total N input (precision at 0.1 kg N ha⁻¹), with legends shown in the**
figure.

Croplands and grasslands show distinct NUE patterns (Figs. 3–4 and Fig. A1 in Appendix). In grasslands, N yields generally increase with N inputs, whereas in croplands, higher N inputs often lead to larger N surplus and excessive pollution (Fig. A1 in Appendix). Croplands exhibit mixed improvements: although COS areas expanded in the 2010s, areas
325 with **insufficient productivity (IP)** also became more prevalent (Table 1), primarily due to a shift from the **excessive pollution/inefficient use of nitrogen/ insufficient productivity (EP/IUN/IP)** regime to the **IP** regime. By contrast, grasslands exhibit more consistent improvement, with higher percentage of **COS** areas and reduced prevalence of **EP**, **EP/IUN** and **IP** regimes (Table 1).

Risk of soil nitrogen mining (RSNM) is considered negligible (<1%) across Switzerland according to the
330 EUNEP framework (Fig. 3–4, Fig. A1 in Appendix and Table 1). A distinct cluster of cropland points shows low total N inputs (50–100 kgN ha⁻¹ yr⁻¹) but very high NUE (>90%) (Fig. 3 and Fig. A1A in Appendix). In the simulations this pattern reflects land-use changes. Throughout the whole simulation period, if land-use change took place, we assumed in the model that these areas were covered by grass-clover mixtures during the non-cropland years. This assumption keeps the model running in a consistent way. Fertilization was assumed to be absent for these non-cropland vegetated periods, with N inputs
335 only from BNF and atmospheric deposition. Consequently, N yields are low because of no additional anthropogenic N sources, while NUE is high, reflecting efficient N use under near-natural conditions.

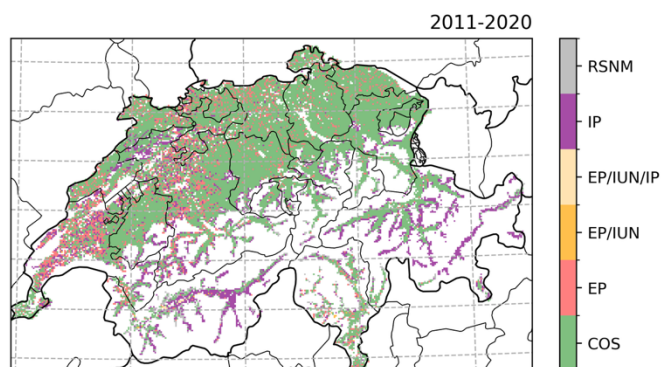
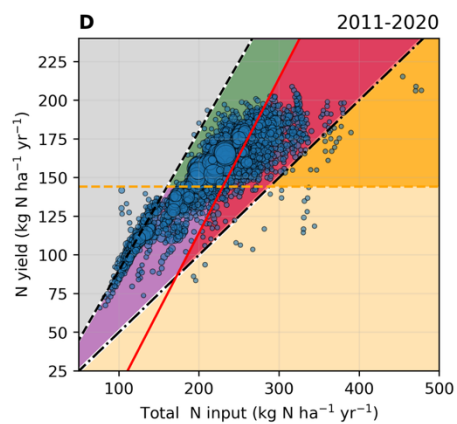
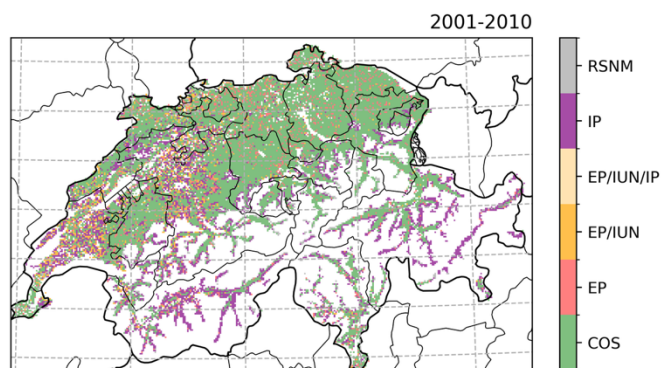
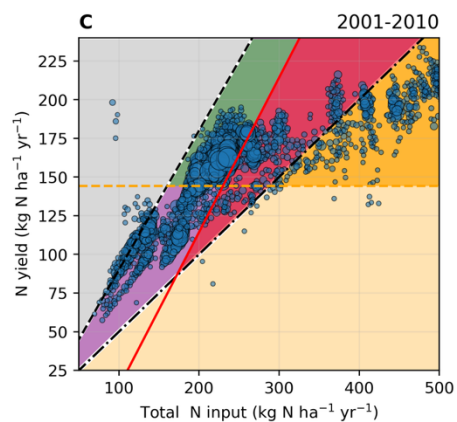
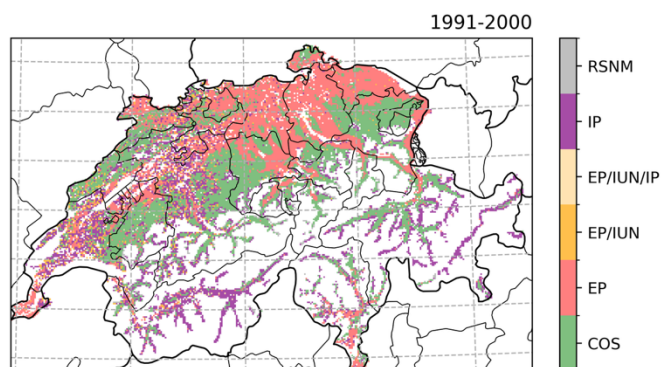
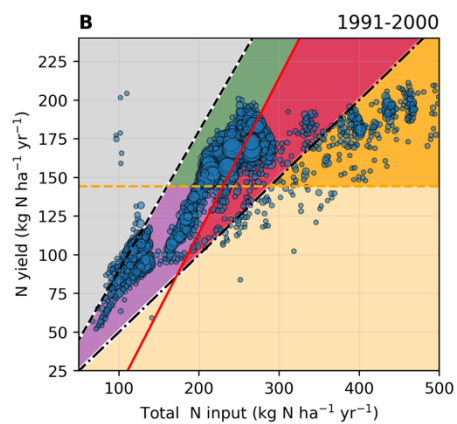
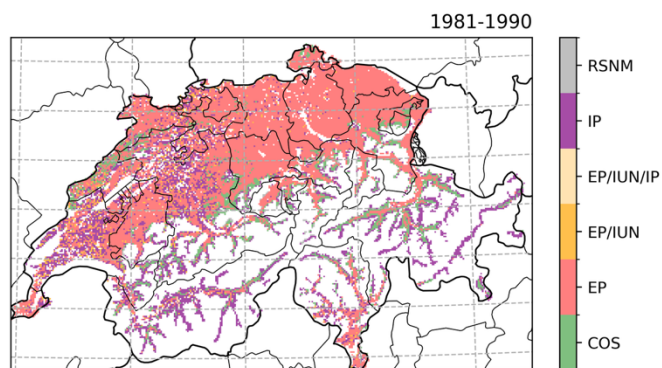
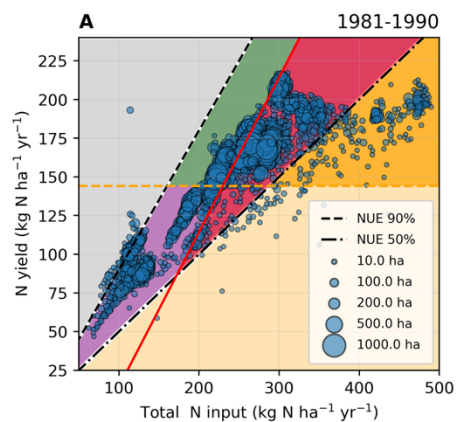


Figure. 4. The EUNEP Framework of the NUE indicator diagram of Swiss grasslands and decadal geographical distributions of six categorised agricultural land. (A)1981-1990 (B)1991-2000 (C) 2001-2010 (D) 2011-2020. Grasslands refer to managed meadows only, while managed pastures and summer pastures for grazing are not included.

3.3 Nitrogen losses and soil nitrogen stock changes

Aggregated N losses (gaseous emissions and leaching) from Switzerland’s croplands decreased markedly over the past four decades, from ~~134~~131 kg N ha⁻¹ yr⁻¹ in the 1980s to 100 kg N ha⁻¹ yr⁻¹ in the 2010s (Table 2 and Fig. S4A in Supplementary Materials), with leaching accounting for ~80% of total losses and for more than 40% of total N inputs (Fig. 5F). By comparison, N losses from grasslands are roughly half as much as croplands (Table 2 and Fig. S4A in Supplementary Materials), but only decreased slightly, from ~~59~~58 to ~~56~~55 kg N ha⁻¹ yr⁻¹, despite a substantial reduction in fertilizer inputs.

Table 2. Nitrogen budgets and soil N balance of Swiss agroecosystems. Nitrogen inputs include livestock manure, synthetic fertilizers, BNF and atmospheric N deposition. Nitrogen outputs include N removal through harvest, gaseous emissions and leaching. Soil N balance is N inputs minus N outputs. All variables are decadal mean for the 1980s and the 2010s, and have the unit kg N ha⁻¹ yr⁻¹. Values in parenthesis are uncertainty (95% confidence interval) due to model parameters estimated from the Monte Carlo approach (200 iterations). *Grasslands refer to managed meadows only, while managed pastures and summer pastures for grazing are not included.

	Period	N input				N output			Δsoil N
		N _{org}	N _{syn}	N _{BNF}	N _{dep}	N _{yield}	N _{gas}	N _{leaching}	
Cropland	1981-1990	131	80	19	11	113 (112-115) 113	24 (21-27) <u>24</u>	107 (93-116) <u>110</u>	-4 (-19-7) <u>7</u>
	2011-2020	129	36	27	11	115 (114-121) 115	22 (20-25) <u>22</u>	78 (70-85) <u>78</u>	-12 (-15 - -4) <u>-12</u>
Grassland*	1981-1990	124	57	60	12	158 (154-168) 158	34 (30-40) <u>35</u>	23 (14-30) <u>24</u>	36 (29-46) <u>25</u>
	2011-2020	89	53	64	12	155 (153-162) 155	34 (30-39) <u>35</u>	21 (12-27) <u>21</u>	6 (2-14) <u>6</u>

Soil N stock dynamics of croplands and grasslands show contrasting characteristics (Fig. S4B in Supplementary Materials). Widespread soil N depletion is found in croplands (Fig. 5H), resulting from larger N outputs (through harvest and losses) than inputs (as described Methods). It is estimated that croplands have lost a cumulative 537 kg N ha⁻¹ between 1981 and 2020. The most rapid depletion at -23 kg N ha⁻¹ yr⁻¹ occurred between 1995 and 2011. Long-term field monitoring also shows soil N depletion at several arable sites (Fig. S5 in Supplementary Materials). At the same time, nationwide long-term monitoring of soil C stocks reported that topsoil (0-20 cm) total organic carbon (TOC) in Swiss croplands have declined from 62 to 55 t TOC ha⁻¹ between 1985-1989 and 2015-2019 (Wollmann et al., 2025). These substantial decreases in soil TOC over time may indirectly provide some evidence for accompanied soil N depletion as pointed out by our modelling

365 results. By contrast, grassland soils showed positive N balance (Fig. 6H) and accumulated 728 kg N ha⁻¹ over the same period. The accumulation was the fastest in the 1980s and then gradually slowed down. In the final five years of the simulations, the soil N stocks in both ecosystems stabilized, suggesting that national mean soil N pools are approaching an equilibrium.

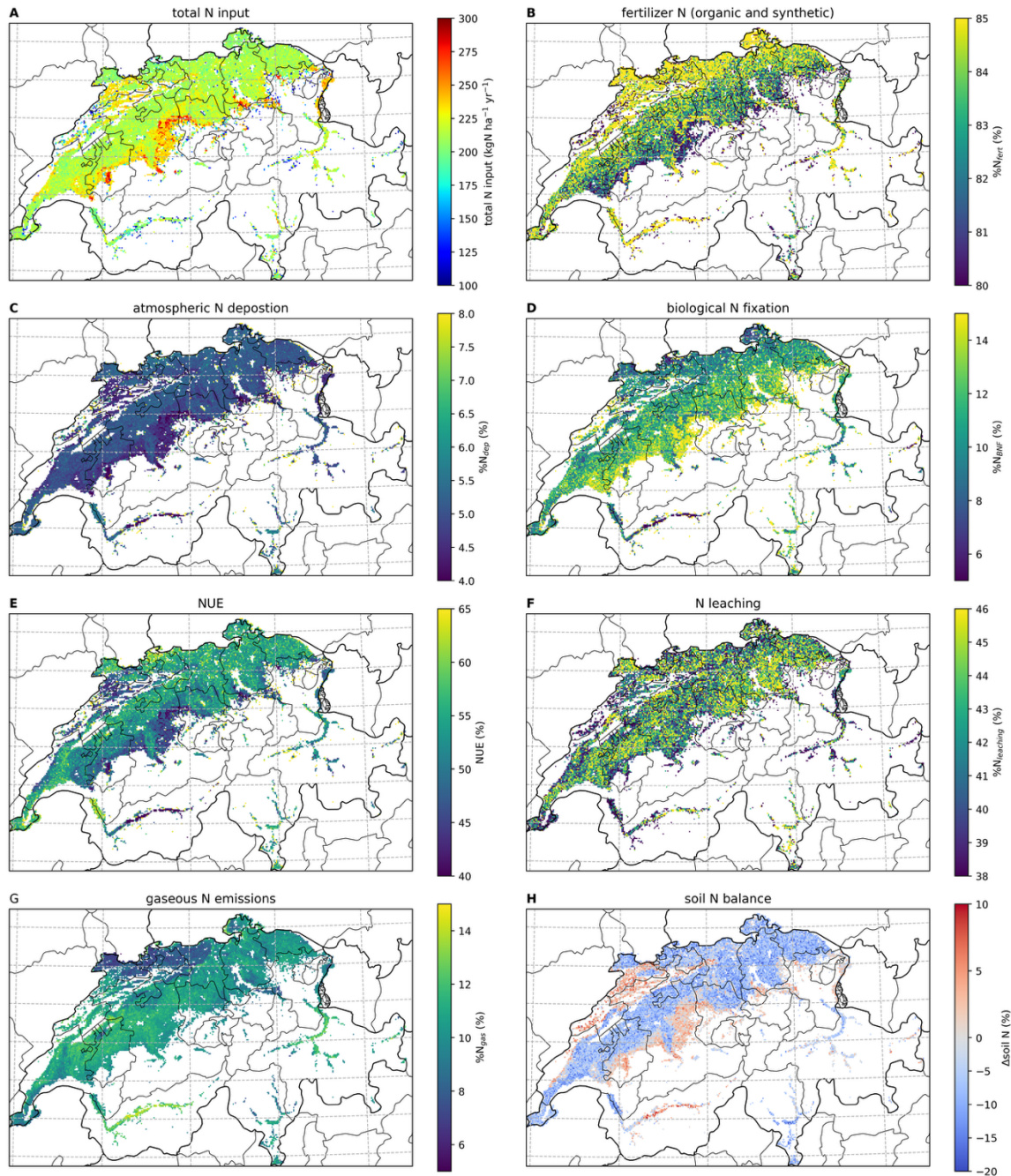
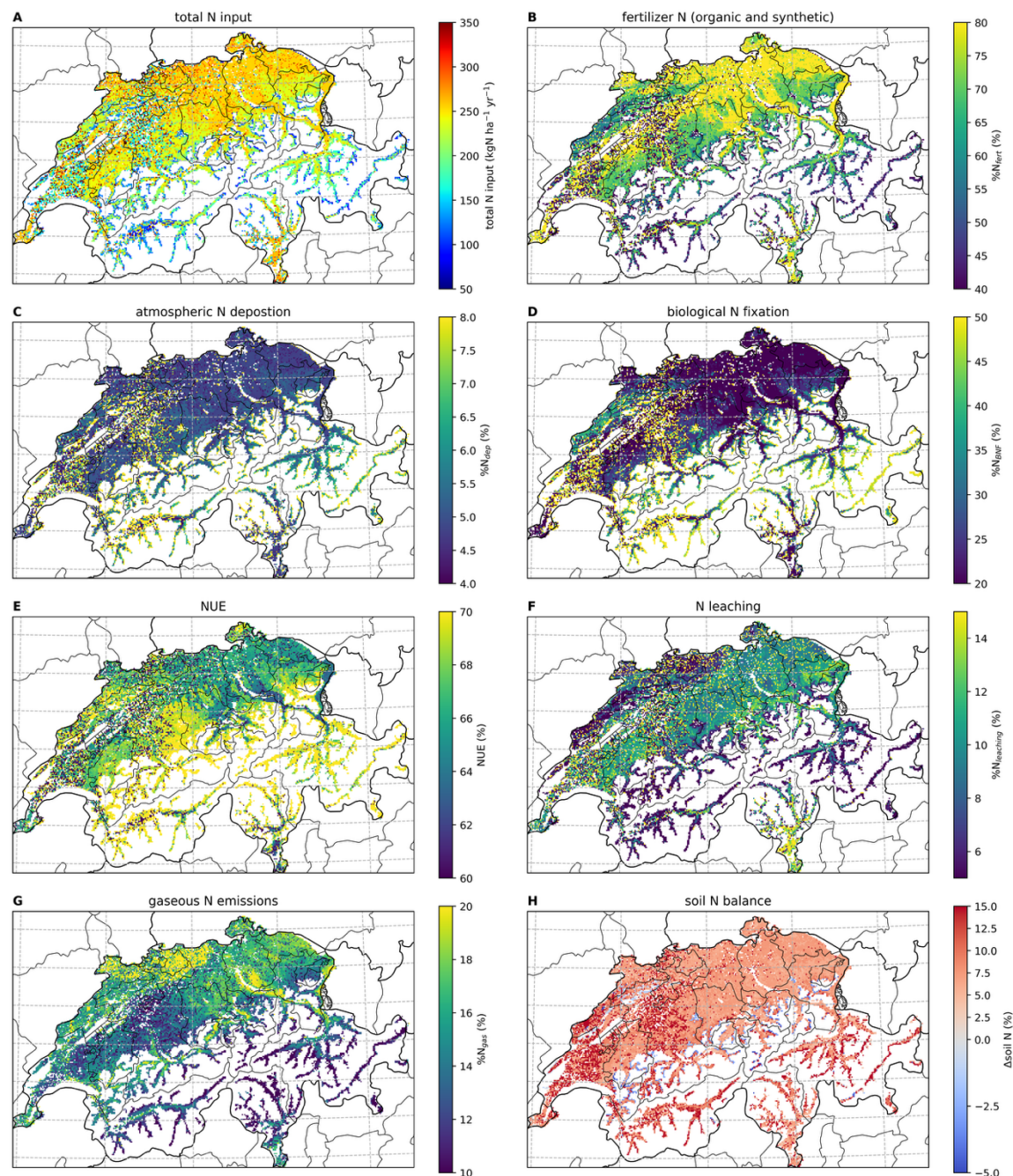


Figure. 5. Spatial maps of N budgets and soil N balance of Swiss croplands over 1981–2020. (A) total N input (B) fertilizer N including manure and synthetic fertilizers (C) atmospheric N deposition (D) BNF (E) NUE (F) N leaching (G) gaseous emissions (H) soil N balance. Total N inputs have the unit $\text{kg N ha}^{-1} \text{ yr}^{-1}$, and other variables are expressed as percentage relative to total N inputs (note the difference in scales). See Supplementary Fig. S6 in Supplementary Materials for maps showing absolute values.



375 **Figure. 6. Spatial maps of N budgets and soil N balance of Swiss grasslands over 1981-2020. (A) total N input (B) fertilizer N**
including manure and synthetic fertilizers (C) atmospheric N deposition (D) BNF (E) NUE (F) N leaching (G) gaseous emissions
(H) soil N balance. Total N inputs have the unit kgN ha⁻¹ yr⁻¹, and other variables are expressed as percentage relative to total N
inputs (note the difference in scales). See Supplementary Fig. S7 in Supplementary Materials for maps shown absolute values.

380 To synthesize N dynamics and assess agri-environmental performance of croplands and grasslands in Switzerland,
we use a novel analytical framework that jointly evaluates N inputs, NUE, N losses and soil N balance, with the latter two
expressed relative to total inputs. Compared with the EUNEP framework, this framework explicitly shows the magnitude of
N losses and soil N changes. In both agricultural ecosystems, NUE declines with increasing N inputs, while higher N inputs
are associated with larger N losses and shifts in the soil N balance (Fig. 7). At N inputs <100 kg N ha⁻¹ yr⁻¹, the two systems
385 operate efficiently: NUE largely exceeds 80%, N losses remain around 10%, and soil N pools are minimally disturbed. As N
inputs increase (150-200 kg N ha⁻¹ yr⁻¹), croplands frequently lose more than half of N inputs and experience substantial soil
N depletion, with decreases in soil N stocks reaching up to 30% of total inputs (45–60 kg N ha⁻¹ yr⁻¹). These N depletions in
croplands were not explicitly reflected in the EUNEP framework. At high inputs (>200 kg N ha⁻¹ yr⁻¹), outcomes diverge:
some sites accumulate N in soils when losses are <50%, while other places deplete soil N stocks. Under such high N inputs,
390 the magnitude of N losses affects the soil N balance, with positive soil N balance associated with lower N losses. Croplands
receiving >250 kg N ha⁻¹ yr⁻¹ N inputs were common in the 1980s but rare in the 2010s, reflecting the decline of input-
intensive practices. Grasslands consistently outperform croplands under comparable input levels, with higher NUE, lower
losses and predominantly positive soil N balances (Fig. 7B). Grassland soils retained more N in the 1980s than in the 2010s,
suggesting diminishing accumulation rates over time due to less fertilizer N inputs (see also Fig. 2, Fig. A1 in Appendix and
395 Fig. S4 in Supplementary Materials).

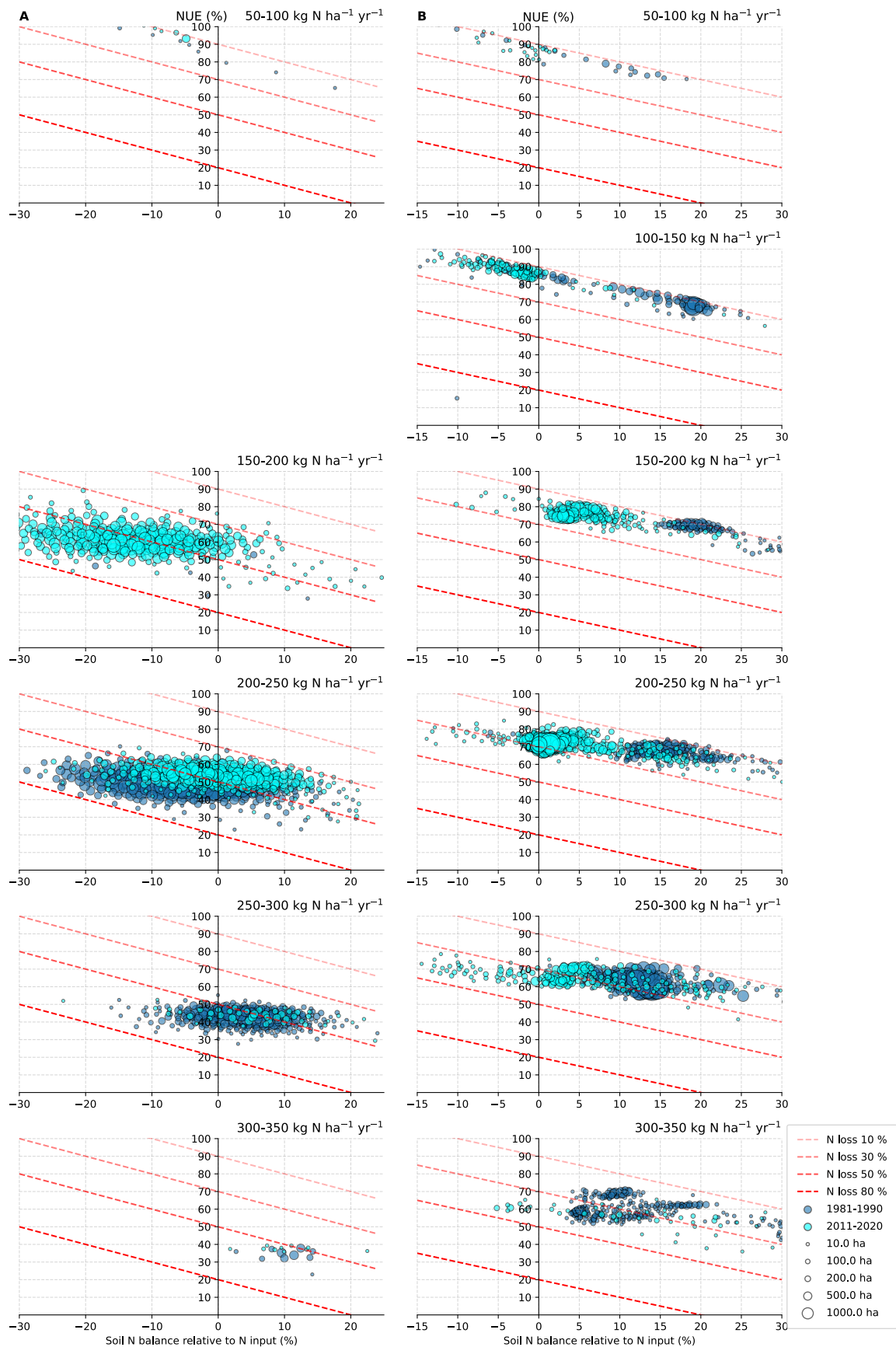


Figure. 7. An integrated N assessment framework of Swiss agroecosystems. NUE, N loss and soil N balance of croplands (A) and grasslands (B) under different levels of N inputs. N loss and soil N balance are expressed as percentage of total N inputs. In each individual panel, from top to bottom, the dashed red lines represent N loss at 10%, 30%, 50% and 90%. The dark blue and light blue circles represent data from 1981-1990 and 2011-2020, respectively. The size of the circle is proportional to the areas, with legends shown in the figure.

~~Our results also indicate that NUE increases with BNF derived N and decreases with fertilizer N (Fig. 8A, B). Under present conditions, the type of N input (i.e. BNF or fertilizer) has no significant influence on relative N losses ($\%N_{loss}$; Fig. 8C, D). The impact of reduction in fertilizer use on N losses is evident only in absolute losses (Fig. S4A in Supplementary Materials and Table 2). Importantly, $\%N_{loss}$ correlated negatively with soil N change, indicating that reducing N losses is crucial for preventing soil N depletion in agricultural land (Fig. 8E). Compared with croplands where soil N depletion may occur across a broad NUE range, surplus N in grasslands tends to stay in soils instead of getting lost to the environment, as lower NUE is linked to greater soil N accumulation (Fig. 8F).~~

Within the current model framework, our simulation results demonstrate at the national scale that NUE increases with a greater contribution of BNF-derived N and decreases with increasing fertilizer N (Fig. 8A, B). In DayCent, symbiotic N fixation occurs only when mineral N is insufficient to satisfy plant demand, leading to an inverse relationship between BNF-derived N and fertilizer N inputs. Although the model does not distinguish the source of mineral N during plant uptake or loss processes, the scenario simulations indicate that replacing fertilizer sources by only applying a single type of fertilizer results in modest changes in total N losses (ranging from -7 to 3%), whereas reducing the total fertilizer input substantially decreases N losses by ~20% (as presented in Table S3). These findings suggest that the amount of N input is a stronger determinant of N losses than the source of N input.

Although lower simulated relative N losses were associated with more positive soil N balances (Fig. 8E), the scenario simulations demonstrates that soil N stock changes are determined by the combined effects of N inputs, harvest removal, and N losses (Table S3). A 20% reduction in fertilizer application decreased total N inputs by 12–14% because increased BNF partially compensated for lower fertilizer inputs (Table S3). Nevertheless, soil N depletion in croplands increased and soil N accumulation in grasslands declined, indicating that reductions in N losses alone are insufficient to maintain soil N stocks if accompanied by reduced N inputs. These findings highlight the need to optimize both N inputs and N retention when improving N management.

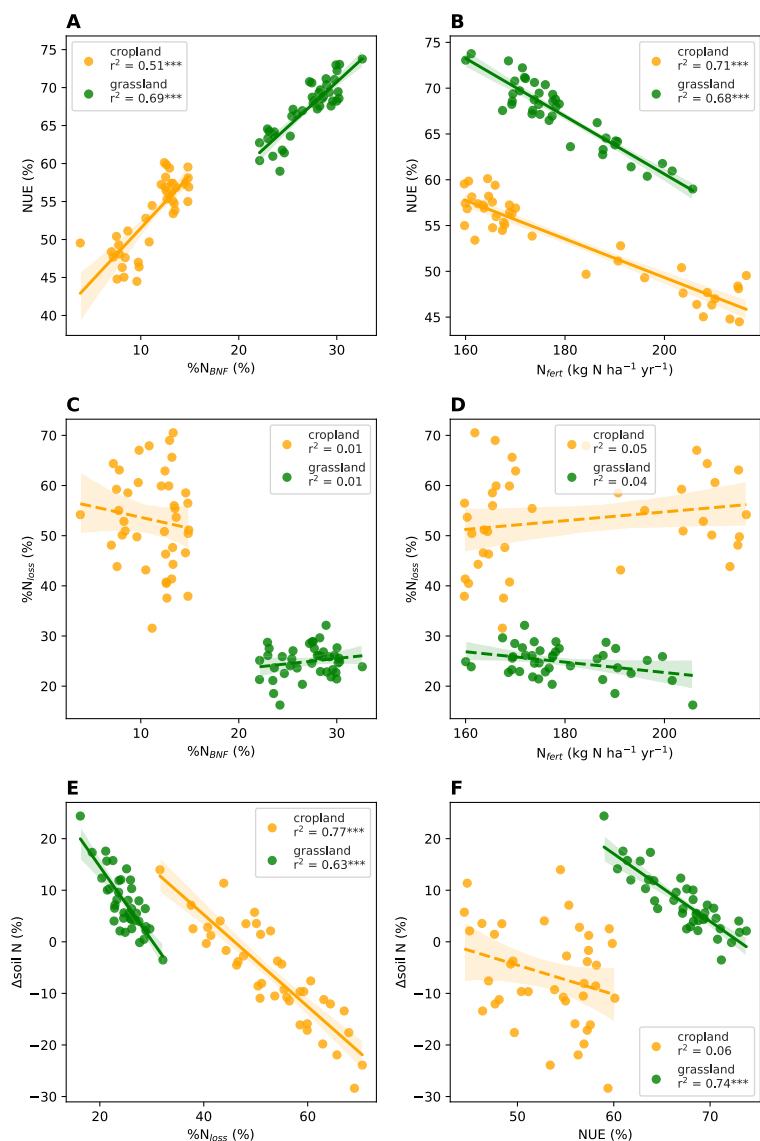


Figure 8. Analysis of N inputs, N loss and soil N balance of Swiss agroecosystems. Response of NUE to (A) relative BNF (%N_{BNF}) and (B) fertilizer N inputs (N_{fert}). No significant relationships between (C) relative BNF (D) fertilizer N inputs and relative N loss (%N_{loss}). Relationships between (E) relative N loss (F) NUE and soil N balance (Δsoil N). BNF, N loss and soil N balance are expressed as percentage relative to total N input. The points represent annual mean values. Croplands are shown in yellow colour, and grasslands are shown in green colour. Solid lines show significant relationships with significance level $***P < 0.001$, and dashed lines indicate insignificant relationships.

4 Discussion

This study presents an in-depth N assessment of two land ecosystems in Switzerland's agriculture. Using national-scale simulations evaluated against observational data (Figs. S2, S3 and S5), we construct spatially explicit N budgets over the past four decades and show that Switzerland's cropland and grassland systems have undergone profound transformations. ~~In the period 1981–2020, our results reveal stable yields despite declining inputs. Furthermore, NUE increased, and N losses reduced (especially in croplands). These findings can largely be explained by policy reforms in the 1990s. During 1981–2020, the simulations reproduced stable agricultural production despite declining fertilizer N inputs, accompanied by~~ increasing NUE and reduced N losses, particularly in croplands. The timing of these changes is consistent with the implementation of agricultural policies promoting less fertilizer use during the 1990s. However, the responses differed between ecosystems. In croplands, fertilizer reductions primarily involved synthetic fertilizers (Table 2), which likely contributed to the marked increase in NUE and decline in N losses. In contrast, grasslands experienced little change in N losses. This can be partly explained by fertilizer reductions dominated by decreases in organic fertilizer, while synthetic fertilizer inputs remained relatively stable (Table 2).

Determined through the EUNEP framework, agri-environmental performance of the two ecosystems shifted towards more efficient and sustainable regimes. These findings highlight the importance of policy intervention for agricultural N management. However, we identify prevalent negative soil N balance in Swiss croplands. Such soil N depletion problems have been studied but not linked to the N budgets approach, and are usually reported only at the site scale (Joris et al., 2020; Mulvaney et al., 2009; Schlingmann et al., 2020). Therefore, long-term monitoring of soil N stock with larger spatial coverage (e.g., regional/national scale), and more comprehensive N assessments and integrated N management are needed to address the risks of further N depletion in agricultural soils. Excessive soil N resulting from past overfertilization can be depleted through careful management, but having more agricultural land with negative a soil N balance should be avoided to ensure the long-term sustainability of agriculture.

4.1 Impacts of N inputs on yields

Swiss agriculture relies heavily on livestock manure (Table 2), which can supply much of crop and forage N demand. However, the heterogeneous distribution of manure across agricultural landscapes causes mismatches between supply and demand in space and time, and logistical constraints such as storage capacity and weather conditions complicate timely application. Hence, synthetic fertilizers remain a crucial supplementary N source to bridge these gaps, offering readily available N for plant uptake (Fig. 5B and Fig. 6B).

In addition to fertilizers, BNF is an important N input. In croplands, our simulations suggest that legumes and grass-clover mixtures in the crop rotation contribute 4–15% of total N inputs (Fig. 5D), while in grasslands BNF accounts for 22–33% (Fig. 6D). In both agroecosystems, we find that relying more on BNF and less on fertilizer tend to achieve higher NUE. Compared to fertilizers and BNF, atmospheric N deposition constitutes smaller share of N inputs in our

465 simulations (Fig. 5D, Fig. 6D and Table 2), which have less significant impacts on agricultural production at the national scale.

470 ~~In this study, we noticed a decoupling of yields from fertilizer inputs with increased NUE in both ecosystems between the 1980s and the 2010s, i.e. yields do not exhibit a linear dependence on N input. At the national scale, the simulations reproduced a progressive decoupling between crop yields and fertilizer N inputs from the 1980s to the 2010s. The simulated increasing NUE over recent decades reflects not only improvements in N management but also the diminishing marginal yield response to additional fertilizer inputs that exceed the agronomic optimum. This pattern is consistent with the well-established nonlinear response of crop yield to N fertilization, which indicates that further increases in fertilizer inputs are unlikely to translate into significant yield gains, suggesting that Switzerland has moved beyond the stage of input-driven intensification. The scenario simulations further support this interpretation: reducing fertilizer application rates by 20% results in only modest reductions in simulated N harvest (3–6%) while substantially decreasing N losses (~20%), suggesting considerable opportunity to further improve environmental performance without compromising agricultural production.~~ Our findings place Switzerland among the “type III” countries described in global analyses (Lassaletta et al., 2014)—those capable of maintaining (or increasing) productivity while reducing N inputs. ~~Our simulations indicate that further increases in fertilizer inputs are unlikely to translate into significant yield gains, suggesting that Switzerland has moved beyond the stage of input-driven intensification.~~

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4.2 N losses and soil N depletion remain challenges

While NUE improvements in Swiss agriculture are encouraging, the caveats are the persistent N losses (especially in grasslands) and negative soil N balance (in cropland). ~~Our modelling results show that N losses decreased only in absolute terms due to lower N inputs, while relative N losses show less significant reductions in croplands and even increased in grasslands (Table 2 and Fig. S4A in Supplementary Materials). Increases of relative N losses in grasslands might be attributed to the unstable N accumulation in soils that causes higher gaseous N emissions resulting from legacy effects (Qian et al., 2025).~~ Although absolute N losses declined with decreasing fertilizer inputs, relative N losses showed only modest reductions in croplands and even increased in grasslands. Our sensitivity analysis suggests that the weaker reduction in relative N losses in grasslands may partly reflect the greater sensitivity of grassland N cycling to increasing temperatures, which offset part of the reduction expected from lower fertilizer inputs. A uniform 2°C increase in air temperature increased simulated N losses by approximately 6% in croplands and 16% in grasslands, indicating that warming enhances soil N turnover and gaseous N emissions in DayCent. Consequently, the reduction in N losses resulting from lower fertilizer inputs may have been partially offset by temperature-driven increases in N cycling, particularly in grasslands. In addition, reducing fertilizer application substantially decreased absolute N losses but also altered BNF and soil N dynamics in DayCent, demonstrating that relative N losses are governed by the interaction among N inputs, plant response and internal soil N cycling rather than by fertilizer inputs alone.

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Most national- and regional-scale studies focus on N inputs, outputs or surplus, often neglecting soil N stock dynamics. Current knowledge suggests that N surplus is usually larger than changes in soil N stocks (Zhang et al., 2015) and that only countries with insufficient N inputs have been found to undergo soil mining and soil fertility loss (Lassaletta et al., 2014; Zhang et al., 2021), so regional-scale soil N balance is understudied, especially in places with high N inputs. Our modelling results reveal that soil N depletion can occur in cropland soils despite high N inputs and thus highlights that also in places with high N inputs the soil N balance should be evaluated, because negative soil N balances in croplands represents a waste of valuable nutrient resources and a threat to future soil fertility and productivity.

The spatially explicit simulations showed that higher N losses and negative soil N balances were mainly associated with intensively managed agricultural regions dominated by croplands, where fertilizer inputs and harvest removal were higher (Fig. 5). In contrast, grassland generally exhibited lower N losses and mostly positive soil N balance (Fig. 6), consistent with higher N retention associated with continuous vegetation cover. Although the simulations were conducted at a fine spatial resolution that enables analysis of regional differences across Switzerland, the primary objective of this study was to assess national-scale temporal trends and compare two agricultural systems. A more detailed investigation of regional hotspots of NUE, N losses and soil N changes represents an important direction for future research.

~~We found that reducing N losses can be critical for preventing negative soil N balance. In our case, the occurrence of soil N depletion in croplands could be explained by high N losses through leaching (Fig. 5F, Table 2 and Fig. S6 in Supplementary Materials). Croplands can be prone to leaching for several reasons. Cultivation involved in cropland management disrupts soil structure (Porwollik et al., 2022; Rupp et al., 2024), and annual crops leave long periods of bare soil post harvest, increasing vulnerability to leaching (Porwollik et al., 2022). Excessive fertilizer application, especially in the form of nitrate, significantly increases leaching risk and thereby degrades groundwater quality (Misselbrook et al., 1996; Shepherd et al., 2001; Vinten et al., 1994). This problem is reflected in the widespread (~35%) exceedance of groundwater nitrate guidelines (>25mg/L) across the Swiss Plateau (Covatti et al., 2025). In addition, wheat, which is effective at extracting nutrients from the soils (Kraaijevanger & Veldkamp, 2020), is the second most dominant crop in Swiss croplands. These factors together can contribute to a negative N balance in croplands.~~
Our simulations, supported by sensitivity analyses, indicate that soil N dynamics are governed by the combined effects of N inputs, harvest removal, climate and internal N cycling, rather than by any single environmental or management factor. Maintaining soil N stocks therefore requires balancing N inputs with harvest removal against N losses. Sensitivity analyses further showed that although reducing fertilizer application substantially decreased N losses, the accompanying reduction in total N inputs slightly accelerated soil N depletion in croplands. Moreover, increasing temperature enhanced simulated N losses, suggesting that climate warming may have partially offset improvements associated with reduced fertilizer use. Our simulations identified nitrate leaching as the dominant N loss pathway (Fig. 5F, Table 2; Fig. S6 in Supplementary Materials), explaining the persistence of N losses despite declining fertilizer inputs. This simulated pattern is consistent with previous studies showing that cropland systems are particularly susceptible to N losses

because annual cultivation disturbs soil structure and periods of bare soil increase vulnerability to nitrate leaching (Porwollik et al., 2022; Rupp et al., 2024). Excessive fertilizer application, especially in the form of nitrate, significantly increases leaching risk and thereby degrades groundwater quality (Misselbrook et al., 1996; Shepherd et al., 2001; Vinten et al., 1994), reflected by the widespread (~35%) exceedance of groundwater nitrate guidelines (>25mg/L) across the Swiss Plateau (Covatti et al., 2025). In addition, wheat as the most widely cultivated crop in Switzerland (and the second most common use of cropland), is known to extract large amounts of soil N (Kraaijvanger & Veldkamp, 2020), suggesting that harvested crops can represent a substantial pathway of N export. Together, these factors can contribute to a negative N balance in croplands. Compared with croplands, grasslands with a more continuous plant cover sustain year-round N uptake. Diverse plant communities in grasslands also improve nutrient retention (De Vries and Bardgett, 2016; Leimer et al., 2016), resulting in higher NUE and lower N losses (Fig. 6, Table 2 and Fig. S7 in Supplementary Materials). Extensive and deep rooting systems can also reduce leaching in grassland (Misselbrook et al., 1996).

4.3 Future N management for sustainable agriculture

For decades, agronomy has centred on enhancing crop productivity, but the adverse environmental consequences resulting from elevated N losses have shifted the paradigm towards agronomic sustainability, emphasizing ecological performance alongside yield optimization. The challenge, however, lies in simultaneously improving NUE and reducing N losses without inducing significant disturbances to soil N stocks. We propose that a robust framework for future N management must rest on the three interlinked pillars: NUE, N loss and soil N stock dynamics. NUE serves as an indicator of input efficiency, N losses capture environmental externalities, and soil N stocks indicate long-term system stability and resilience. By considering these dimensions together, policy makers and practitioners can design strategies that secure productivity without undermining ecological integrity.

This study demonstrates the effectiveness of coordinated policy intervention that resulted in improvements in NUE and reductions in losses. A successful case was also found in China, where targeted N management programmes between 2007 and 2017 led to simultaneous gains in agricultural and environmental outcomes (Duan et al., 2024). These examples underscore that system-level change is possible within decades when science, practice and policy are aligned.

Many high-income countries face similar challenges as Switzerland of balancing productivity with environmental goals, while low- and middle-income countries risk soil N depletion if inputs remain insufficient or face environmental penalties due to unsustainable intensification (Falconnier et al., 2023). In this study, we provide a transferable analytical approach that integrates NUE, N losses and soil N changes for evaluating agri-environmental performance at the national scale. Context-specific application of this framework could help identify “win-win” strategies that support food security while benefiting ecosystems functioning and resilience. Looking forward, future policy frameworks should encourage integrated N management that explicitly addresses productivity-pollution-soils nexus, possibly coupled with financial incentives to stimulate adoption of advanced nutrient management technologies.

5.4.4 Uncertainty, limitations and outlook

The Monte Carlo analysis showed that the uncertainty associated with simulated NUE, N losses and soil N stock changes was generally smaller than the observed long-term trends, increasing confidence that the simulated temporal changes are robust. Nonetheless, the quantified N budgets in this study were estimated primarily using a modelling approach so that the results should be interpreted with caution. In addition to the evaluated model parameters, other sources of uncertainty were not quantified, such model inputs (e.g., livestock C and N data, and soil data). It is also worth noting that DayCent uses a simple “tipping bucket” module to represent water movement in soil layers, which can overestimate percolation fluxes (i.e., predict quicker drainage). This influences soil water content and consequently affects water-dependent processes (such as crop N uptake, leaching and N₂O production). Moreover, DayCent lacks a sophisticated scheme for NH₃ volatilization, leading to substantially underestimated NH₃ fluxes. This can result in a larger soil nitrate pool because more ammonium is available for nitrification, subsequently enhancing nitrate leaching. Another uncertainty is related to absence of representing cover crops in the model, which have been found to improve multiple ecosystem functions, including total soil N (Liu et al., 2025) at a global level. For Switzerland, Herzog et al. (2005) estimated that the increased use of cover crops since the late 1990s reduced farm-level nutrient balances by 10% until 2005. The impact of not including them in this study is that we potentially underestimated the reduction in nitrate leaching and therefore also in NUE, but based on the above-mentioned study, the impact is not high.

Overall, our results and findings are robust and provide valuable insights for agricultural policy implementation and the assessment of agri-environmental performance. This work also demonstrates the feasibility of spatially explicit N flow quantification. Future work on monitoring national N use (e.g., constructing N budgets and calculate N balance) can incorporate biogeochemical modelling as part of the methodology.

The quantified N budgets in this study were estimated primarily using a modelling approach so that readers should interpret the results (and the associated uncertainty) and findings within that modelling context. In general, uncertainty arises from three major sources: 1) input data, 2) model parameters and processes and 3) spatial application of the model.

As described in Section 2.2, we used a wide range of data for this national-scale modelling. Livestock manure C and N data are by far the most reliable estimates available, yet uncertainty stems from various assumptions about management stages in livestock farming (e.g., the tendency of farms to apply manure; C and N losses during manure storage). A major input uncertainty is associated with the soil properties. We used the soil data developed by Stumpf et al (2024), which integrates field sampling, remote sensing and machine learning interpolation. While its full spatial coverage for the entire country is an advantage, areas with sparse sampling may have poorly represented soil properties. Additional soil parameters such as bulk density, field capacity and hydraulic conductivity of the soils derived using pedotransfer equations also

introduce uncertainty into the simulations. By contrast, the uncertainty concerning weather input data is expected to be comparatively small.

Parameter and process uncertainties are inherent in DayCent's representation of crop growth (including N uptake), nitrification and denitrification. Although the model has been calibrated and validated against measurements from several Swiss sites, variability in these parameters still contributes to overall uncertainty. Regarding N loss pathways, two major limitations exist. First, DayCent uses a simple "tipping bucket" module to represent water movement in soil layers, which can overestimate percolation fluxes (i.e., predict quicker drainage). This influences soil water content and consequently affects water-dependent processes (such as crop N uptake, leaching and N₂O production). Second, DayCent lacks a sophisticated scheme for NH₃ volatilization, leading to substantially underestimated NH₃ fluxes. This can result in a larger soil nitrate pool because more ammonium is available for nitrification, subsequently enhancing nitrate leaching.

The third source of uncertainty originates from the spatial application of the DayCent model at the national scale. Uncertainty here includes our crop rotation scheme, generalised timing for fertilizer application and mowing intensity of grasslands, which is difficult to evaluate systematically. In this study, we applied the most realistic scenarios identified to date that best represent real practices for national scale simulations. An improved crop rotation scheme will become available in the future as a long-term national crop rotation dataset is being assembled.

Overall, our results and findings remain robust and provide valuable insights for agricultural policy implementation and the assessment of agri-environmental performance. This work also demonstrates the feasibility of spatially explicit N flow quantification. Future work on monitoring national N use (e.g., constructing N budgets and calculate N balance) can incorporate biogeochemical modelling as part of the methodology. Importantly, long-term monitoring of soil N stock should also be integrated into future soil samplings and surveys.

6.5 Conclusions

In this study, we applied high-resolution, spatio-temporal process-based biogeochemical modelling to reconstruct four decades of N budgets across two major agricultural systems in Switzerland. Using this integrated modelling approach, we showed that agri-environmental policy interventions successfully reduced agricultural N losses while maintaining crop yields and substantially improving NUE. These improvements were largely driven by less fertilizer use throughout the 1990s. However, our modelling work also revealed a possibly **nationwide depletion of soil N stocks** occurring despite continued high fertilizer inputs, which is not reported in previous studies. This central finding may expose a critical vulnerability within intensively managed food-production systems: efficiency gains and pollution control can mask an overlooked issue of unbalanced soil N stock that underpins long-term productivity and resilience.

Using Switzerland as an exemplary case, the analytical framework used in this study and our policy-relevant findings are transferable to many agro-food systems with high nutrient inputs worldwide that are pursuing N loss mitigation targets while seeking to maintain food production. By jointly assessing productivity, N pollution and soil N stocks, our case

provides a systems-level diagnostic tool for evaluating the performance and sustainability of contemporary food-production systems.

630 **Appendix**

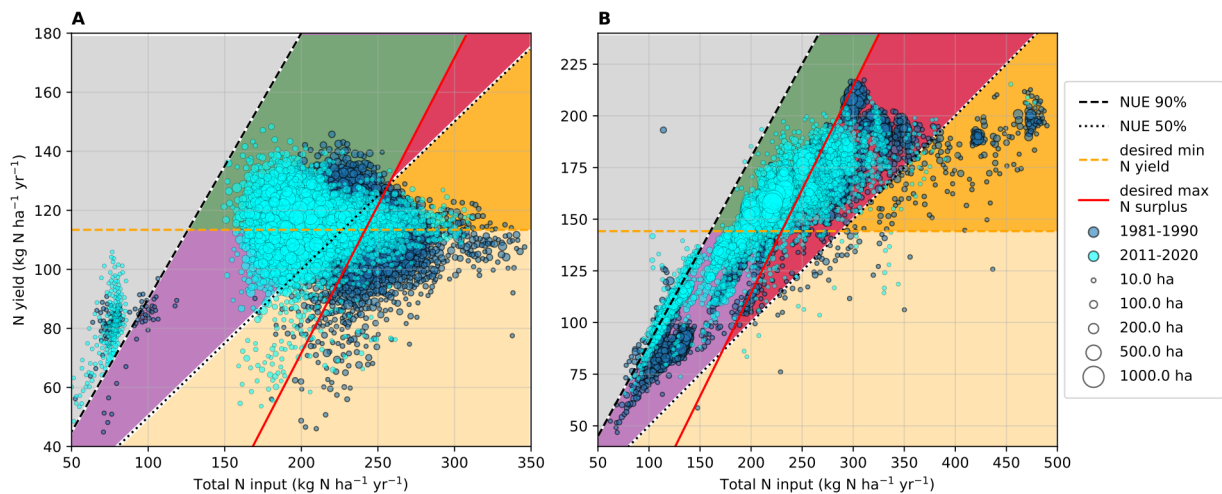


Figure. A1. The EUNEP Framework of the NUE indicator diagram for Swiss agroecosystems. Modelled total nitrogen input and N yield of croplands (A) and grasslands (B).

635 **Code, data, or code and data availability**

Modelling results of the nitrogen budgets presented in this study are in netCDF format and are deposited at <http://hdl.handle.net/20.500.11850/788419>. Code of the DayCent model is publicly available at <https://www.soilcarbonsolutionscenter.com/daycent>. For specific versions, please contact the developers directly.

Author contributions

640 J.J. conceived of the study. J.J. performed the simulations and analysed the data. C.W. and D.B. developed carbon and nitrogen data and provided model input. J.J., C.W. and D.B. compiled datasets. M.N. and A.S. assisted with software and modelling. J.J. wrote the original draft of paper. L.H.E.W. and J.S. supervised the project and acquired funding. All authors contributed to interpretation of results and critical revision of the paper.

Competing interests

645 The authors declare no competing interests.

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