


## Article

# Effects of Nitrogen Rate and Fertilizer Type on Gaseous Nitrogen Losses and Soil Nitrogen Storage in Alkaline Maize Fields of the Hetao Irrigation District

Yu Gao <sup>1,2,3</sup>, Yunfei Di <sup>1,2</sup>, Haibo Yang <sup>1,2</sup>, Yuzhe Tang <sup>1,2</sup>, Weijian Zhang <sup>1,2</sup>, Yuncai Hu <sup>4</sup>  and Fei Li <sup>1,2,\*</sup>

<sup>1</sup> College of Resources and Environment, Inner Mongolia Agricultural University, Hohhot 010011, China; gaoyu6670@163.com (Y.G.); diyunfei@emails.imau.edu.cn (Y.D.); haiboyang@imau.edu.cn (H.Y.); tangyuzhe9702@163.com (Y.T.); zhangwj1110@126.com (W.Z.)

<sup>2</sup> Inner Mongolia Key Laboratory of Soil Quality and Nutrient Resources, Hohhot 010018, China

<sup>3</sup> Institute of Resources, Environment and Sustainable Development, Inner Mongolia Academy of Agriculture and Animal Husbandry Sciences, Hohhot 010031, China

<sup>4</sup> Department of Life Science Engineering, School of Life Sciences, Technical University of Munich, 85354 Freising, Germany; yc.hu@tum.de

\* Correspondence: lifei@imau.edu.cn

## Abstract

Gaseous nitrogen losses and residual soil nitrogen accumulation are primary drivers of low nitrogen use efficiency in alkaline irrigated cropping systems. A two-year field experiment (2019–2020) in the Hetao Irrigation District under alkaline flood-irrigated maize evaluated the effects of nitrogen rate, fertilizer formulation, and enhanced-efficiency fertilizers—urea with urease inhibitor NBPT and ammonium sulfate with nitrification inhibitor DMPP—on  $\text{NH}_3$  volatilization,  $\text{N}_2\text{O}$  emissions, post-harvest soil mineral nitrogen, and grain yield. A soil pH manipulation sub-experiment ( $\pm 0.5$  units, ambient pH  $\sim 8.8$ ) was conducted to quantify the direct effect of alkalinity on volatilization.  $\text{NH}_3$  volatilization was insensitive to fertilizer formulation and inhibitor inclusion but strongly responsive to soil pH; a 0.5-unit increase in soil pH elevated volatilization efficiency by up to 25% relative to ambient conditions.  $\text{N}_2\text{O}$  emissions were around 18% higher under ammonium sulfate than under urea and were reduced by 21–32% with inhibitor treatments, without increasing  $\text{NH}_3$  volatilization. Inhibitor-assisted optimized management (urea + NBPT and ammonium sulfate + DMPP) achieved higher yields and lower emission intensity than urea alone. These results confirm that  $\text{NH}_3$  and  $\text{N}_2\text{O}$  losses are governed by distinct controls, and that concurrent mitigation of both pathways requires interventions that independently target each loss driver, beyond rate optimization and inhibitor application alone.

**Keywords:** ammonia volatilization; nitrous oxide emissions; soil nitrogen storage; enhanced-efficiency fertilizer; soil pH; maize; Hetao Irrigation district



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## 1. Introduction

Nitrogen fertilizer has been central to the dramatic increases in global crop production over the past half-century, yet a large fraction of applied nitrogen is not recovered by crops and instead escapes to the environment through gaseous losses—principally as ammonia ( $\text{NH}_3$ ) and nitrous oxide ( $\text{N}_2\text{O}$ ) [1,2]. Global agricultural  $\text{NH}_3$  emissions have grown rapidly since 1980, and synthetic fertilizers account for an average of 12–18% of the nitrogen applied that is lost as  $\text{NH}_3$  [1,2]. Excessive  $\text{NH}_3$  deposition drives soil acidification, eutrophication of water bodies, and—through secondary reactions with sulfur oxides and

nitrogen oxides—the formation of fine particulate matter, contributing substantially to regional haze events [3,4].  $\text{N}_2\text{O}$  is among the most potent anthropogenic greenhouse gases, with a 100-year global warming potential 273 times that of  $\text{CO}_2$  [5]; soils contribute more than 80% of total biospheric  $\text{N}_2\text{O}$  flux to the atmosphere [6], and agricultural production accounted for approximately 74% of human-driven  $\text{N}_2\text{O}$  emissions in the 2010s, with direct agricultural emissions reaching  $3.9 \text{ Tg N yr}^{-1}$  in 2020 [7]. Managing both loss pathways simultaneously is, therefore, a central challenge for sustainable intensification of crop production, yet the contrasting sensitivities of  $\text{NH}_3$  volatilization and  $\text{N}_2\text{O}$  emission to the same management inputs mean that interventions designed to reduce one gas frequently have limited or even counterproductive effects on the other [8,9].

Nitrogen application rate is the primary driver of both  $\text{NH}_3$  volatilization and  $\text{N}_2\text{O}$  emissions, operating by elevating soil ammonium ( $\text{NH}_4^+$ ) concentrations that simultaneously provide a substrate for both hydrolysis-driven pH increases favorable to  $\text{NH}_3$  formation and for nitrification–denitrification processes that produce  $\text{N}_2\text{O}$  [10]. Soil pH exerts a particularly strong additional control on  $\text{NH}_3$  volatilization: losses increase sharply above pH 7.5, as the thermodynamic equilibrium between  $\text{NH}_4^+$  and  $\text{NH}_3$  shifts progressively toward the gaseous form with increasing alkalinity [11]. Enhanced-efficiency fertilizers (EEFs), principally urease inhibitors (UI) such as N-(n-butyl) thiophosphoric triamide (NBPT) and nitrification inhibitors (NI) such as 3,4-dimethylpyrazole phosphate (DMPP), have been widely proposed as tools to reduce gaseous nitrogen losses by slowing urea hydrolysis and suppressing nitrification, respectively. Meta-analyses indicate that NBPT reduces cumulative  $\text{NH}_3$  emissions by approximately 52–61% on average [12,13], while DMPP and related NIs reduce  $\text{N}_2\text{O}$  by approximately 49% [14]. However, EEF effectiveness is strongly context-dependent, and the two inhibitor classes address fundamentally different loss pathways: field experiments in intensive maize cropping systems of northern China have demonstrated that the trade-off between  $\text{NH}_3$  volatilization and  $\text{N}_2\text{O}$  reduction persists when NIs are applied at the soil surface [9]. Critically, the NI application has been shown to increase  $\text{NH}_3$  emissions by an average of 35.7% globally—with this  $\text{NH}_3$  penalty positively correlated with soil pH—so that in alkaline soils the trade-off between  $\text{N}_2\text{O}$  reduction and  $\text{NH}_3$  amplification becomes particularly pronounced [8]. When indirect  $\text{NH}_3$ -derived  $\text{N}_2\text{O}$  is accounted for, the net climate benefit of NI application is substantially reduced [8]. Furthermore, a 30-year global synthesis found that DMPP was statistically ineffective for  $\text{N}_2\text{O}$  reduction in alkaline soils, where soil pH and microbial community composition reduced its efficacy [15]. These findings collectively indicate that the partial decoupling of  $\text{NH}_3$  and  $\text{N}_2\text{O}$  pathways is not merely a theoretical concern but a quantifiable management reality, particularly under the alkaline soil conditions that characterize large irrigated agricultural regions of northern China.

The Hetao Irrigation District of Inner Mongolia, one of the three largest irrigation districts in China, exemplifies the intersection of these challenges [16]. The district is characterized by alkaline soils (pH typically exceeding 8.0), conventional nitrogen application rates substantially above crop requirements, and seasonal flood irrigation delivering  $3600\text{--}4500 \text{ m}^3 \text{ ha}^{-1}$  through the Yellow River canal network—conditions under which crop nitrogen use efficiency averages well below 50% [17], leaving a large nitrogen surplus that accumulates in the soil profile or is lost to the environment [18,19]. Residual mineral nitrogen in the soil profile following harvest represents a legacy pool susceptible to subsequent leaching and denitrification losses, and has been shown to reach  $453\text{--}749 \text{ kg N ha}^{-1}$  in the 0–4 m soil profile of wheat and maize croplands across northern China under conventional nitrogen management, representing a substantial reservoir at risk of leaching under future rainfall intensification [20]. Beyond soil nitrogen accumulation, China's agricultural soils have been estimated to contribute approximately 23% of global  $\text{NH}_3$  and 20% of global

N<sub>2</sub>O emissions from synthetic fertilizer use, with maize among the three leading reactive nitrogen emitters among crop types [21]. Under flood irrigation, the rapid transition from aerobic to anaerobic soil conditions following inundation creates transient reducing environments that intensify denitrification and N<sub>2</sub>O production within narrow post-irrigation time windows [22], while the alkaline soil matrix simultaneously sustains high background rates of NH<sub>3</sub> volatilization [11]. The combination of high pH, high nitrogen inputs, and repeated flood irrigation events, therefore, creates a system in which the performance of EEFs and the relative magnitudes of competing nitrogen loss pathways may diverge substantially from predictions based on studies conducted in more neutral, rainfed, or drip-irrigated systems. Despite the agronomic and environmental importance of this region, the trade-offs among gaseous nitrogen losses, residual soil nitrogen accumulation, and crop productivity under different fertilizer management strategies are governed by interactions among nitrogen rate, soil properties, and irrigation-induced anaerobic dynamics that vary substantially across systems and are difficult to predict from models or studies conducted under contrasting conditions [23,24]. Nevertheless, studies that simultaneously quantify cumulative NH<sub>3</sub> volatilization, N<sub>2</sub>O emissions, and residual soil mineral nitrogen storage across a range of fertilizer management strategies—including direct assessment of soil pH effects on volatilization and evaluation of inhibitor performance—remain scarce for alkaline flood-irrigated maize production.

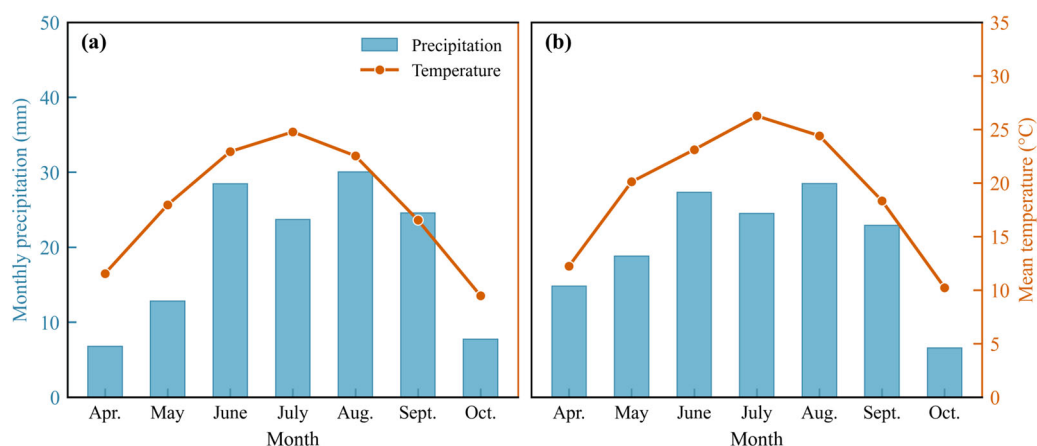
To address these gaps, we conducted a two-year field experiment (2019–2020) in the Hetao Irrigation District, comparing the effects of nitrogen application rate, fertilizer form, and enhanced-efficiency fertilizers on NH<sub>3</sub> volatilization, N<sub>2</sub>O emissions, post-harvest soil mineral nitrogen storage, and maize grain yield. A soil pH manipulation sub-experiment was embedded within the main trial to directly quantify the contribution of soil alkalinity to NH<sub>3</sub> volatilization efficiency under field conditions. The study addressed three questions: (1) Do nitrogen application rate, fertilizer formulation, and inhibitor inclusion exert differential effects on cumulative NH<sub>3</sub> volatilization and N<sub>2</sub>O emissions, and which factors are the primary determinants of each loss pathway under alkaline flood-irrigated conditions? (2) How sensitive is NH<sub>3</sub> volatilization efficiency to soil pH across a controlled gradient under field conditions? And (3) Which fertilizer management strategy most effectively balances the reduction of gaseous nitrogen losses, improvement of greenhouse gas emission intensity, and maintenance of grain yield in this system?

## 2. Materials and Methods

### 2.1. Study Site and Experimental Design

The field experiment was conducted at the Yonglian Village Research Base, Wuyuan County, Bayannur City, Inner Mongolia Autonomous Region, China (41°4′ N, 108°2′ E; altitude 1027 m a.s.l.). The site experiences a temperate continental climate with a mean annual temperature of 6.1 °C, a cumulative active temperature ( $\geq 10$  °C) of 3362.5 °C, and a frost-free period of 117–136 days. Mean annual precipitation is 170 mm, concentrated predominantly in the summer and autumn months. Monthly precipitation and mean air temperature during the two growing seasons are presented in Figure 1. The soil is classified as an Irragic Anthrosol (World Reference Base for Soil Resources, WRB), with a silt loam texture and a bulk density of 1.40 g cm<sup>-3</sup>. Prior to experimental establishment, the topsoil (0–20 cm) had an organic matter content of 21.64 g kg<sup>-1</sup>, alkaline-hydrolyzable N of 53.4 mg kg<sup>-1</sup>, available P of 20.08 mg kg<sup>-1</sup>, available K of 132.95 mg kg<sup>-1</sup>, CaCO<sub>3</sub> equivalent content of 105.27 g kg<sup>-1</sup>, CEC of 11.78 cmol(+) kg<sup>-1</sup>, exchangeable sodium percentage (ESP) of 8.9%, and a pH of 8.8. Soil organic matter was determined by the potassium dichromate oxidation method, alkaline-hydrolyzable N by the alkaline hydrolysis diffusion method, available P by the sodium bicarbonate extraction method (Olsen

method), available K by ammonium acetate extraction,  $\text{CaCO}_3$  equivalent by the volumetric calcimeter method, CEC by the ammonium acetate method, ESP by calculation from exchangeable sodium and CEC, and pH in a 1:2.5 soil-to-water suspension [25].



**Figure 1.** Monthly precipitation and mean air temperature during the growing season in (a) 2019 and (b) 2020.

A two-year field experiment was conducted during the 2019 and 2020 maize (*Zea mays* L.) growing seasons. The maize variety was Xinyu 12, sowing occurred at the end of April, and harvest occurred in early October in both years. Each experimental plot measured 6.5 m × 10 m. Six treatments were arranged in a randomized complete block design with four replications (Table 1): an unfertilized control (CK); an optimized nitrogen rate applied as urea (OPT-U; 180 kg N ha<sup>-1</sup>, determined by soil testing and fertilizer recommendation based on pre-experiment soil nutrient analysis and local maize nitrogen demand); a conventional high nitrogen rate applied as urea (CON-U; 400 kg N ha<sup>-1</sup>, reflecting the mean application rate established through pre-experiment farmer surveys in the local area); urea combined with the urease inhibitor NBPT at the optimized rate (OPT-IU); ammonium sulfate alone at the optimized rate (OPT-AS); and ammonium sulfate combined with the nitrification inhibitor DMPP at the optimized rate (OPT-IAS).

**Table 1.** Experimental treatment design showing nitrogen application rates and split-application schedule.

Treatment	Fertilizer Form	Basal N (kg ha <sup>-1</sup> )	Topdressing N (kg ha <sup>-1</sup> )	Total N (kg ha <sup>-1</sup> )
CK	—	0	0	0
OPT-U	Urea	54	126	180
CON-U	Urea	120	280	400
OPT-IU	Urea + NBPT	54	126	180
OPT-AS	Ammonium sulfate	54	126	180
OPT-IAS	Ammonium sulfate + DMPP	54	126	180

Nitrogen was applied with 30% as a basal fertilizer and 70% as a topdressing. Basal fertilizer was incorporated into the soil prior to the first irrigation event. Topdressing was applied at the V6 growth stage, followed immediately by flood irrigation. Both inhibitor-treated fertilizers (OPT-IU and OPT-IAS) were commercially available pre-incorporated products: NBPT (N-(n-butyl) thiophosphoric triamide; purity ≥98%, water-soluble powder formulation) was pre-incorporated into urea at a rate of 1% of the fertilizer-N applied (*w/w*), and DMPP (3,4-dimethylpyrazole phosphate; purity ≥98%, water-soluble formulation) was pre-incorporated into ammonium sulfate at a rate of 1% of the ammonium-N content (*w/w*).

Inhibitor-treated fertilizers were applied following the same split-application schedule as the corresponding non-inhibitor treatments, with no additional handling steps. Phosphorus ( $90 \text{ kg P ha}^{-1}$ ) and potassium ( $120 \text{ kg K ha}^{-1}$ ) fertilizers were applied uniformly as seed fertilizers in a single basal application across all treatments. Flood irrigation was applied three times per growing season, using water from the Yellow River supplied through the Hetao Irrigation District canal network. Irrigation volumes at crop growth stages V6, R1, and R4 were approximately  $1200$ ,  $1000$ , and  $1400 \text{ m}^3 \text{ ha}^{-1}$ , respectively, determined by flow meter readings at the branch canal, for a total seasonal irrigation volume of approximately  $3600 \text{ m}^3 \text{ ha}^{-1}$ . All other agronomic practices followed local conventional management.

## 2.2. Soil pH Microplot Experiment

To isolate and directly quantify the effect of soil pH on  $\text{NH}_3$  volatilization under field conditions, soil pH manipulation microplots ( $1.5 \text{ m} \times 0.8 \text{ m}$ ) were established within each plot of the five fertilized treatments (OPT-U, CON-U, OPT-IU, OPT-AS, and OPT-IAS) across all four replicate blocks. Within each plot, one microplot was adjusted to elevated pH ( $\text{pH}^+$ ) and one to reduced pH ( $\text{pH}^-$ ), yielding four microplots per pH condition per treatment. Soil pH was adjusted in the upper 0–30 cm layer relative to the ambient plot pH ( $\sim 8.8$ ) by approximately  $\pm 0.5$  units, yielding elevated ( $\text{pH}^+$ ,  $\sim 9.3$ – $9.4$ ) and reduced ( $\text{pH}^-$ ,  $\sim 8.3$ ) conditions. Soil pH elevation was achieved by thoroughly mixing  $110.16 \text{ g}$  of calcium oxide (CaO) dissolved in water with the excavated topsoil; soil pH reduction was achieved by mixing  $3.1 \text{ L}$  of  $0.368 \text{ mol L}^{-1}$  sulfuric acid ( $\text{H}_2\text{SO}_4$ ) with the excavated soil. Following the amendment, soil pH was analytically verified before backfilling.

The physical disturbance associated with soil excavation and backfilling required approximately 20 days for bulk density to return to values representative of the surrounding plot. Soil pH in each microplot was monitored continuously after adjustment; values converged to ambient plot levels by approximately day 22 post-adjustment and remained stable thereafter. Accordingly, differences in  $\text{NH}_3$  volatilization attributable to pH manipulation were detectable only within this  $\sim 22$ -day window. Because the settlement period precluded safe installation of static  $\text{N}_2\text{O}$  flux chambers, pH effects on  $\text{N}_2\text{O}$  emissions were not measured within the microplots and are evaluated only through the main-plot treatment comparisons.

## 2.3. $\text{NH}_3$ Volatilization Measurement

Ammonia volatilization was quantified using the ventilation (venting) method following the protocol of Wang [26]. Each capture device consisted of a rigid polyvinyl chloride (PVC) tube (inner diameter  $15 \text{ cm}$ , height  $15 \text{ cm}$ ) fitted with two polyurethane sponges (diameter  $16 \text{ cm}$ , thickness  $2 \text{ cm}$  each) pre-soaked in  $15 \text{ mL}$  of glycerol-phosphate absorption solution ( $50 \text{ mL}$  phosphoric acid and  $40 \text{ mL}$  glycerol, adjusted to  $1000 \text{ mL}$  with deionized water). The lower sponge was positioned  $5 \text{ cm}$  from the tube base; the upper sponge was placed flush with the tube rim (Figure 2). Each device was placed with its base resting on the soil surface, enclosing the measurement area in a cap-like configuration without soil insertion, thereby allowing free gas exchange at the soil–device interface. Three devices were deployed randomly per plot, positioned away from plot borders to avoid inter-plot contamination; the same deployment of three devices was applied within each microplot.

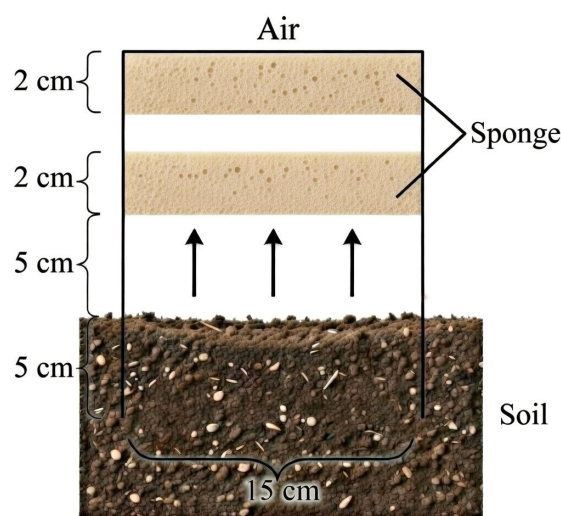
Monitoring commenced on the day following each fertilization event. During the 6-day intensive monitoring window after each fertilization and irrigation event, sampling was conducted every 2 days; during non-intensive periods, frequency was reduced to every 15–25 days. At each interval, sponges were retrieved and extracted with  $300 \text{ mL}$  of  $2 \text{ mol L}^{-1}$  KCl solution by oscillation for 1 h, and the  $\text{NH}_3\text{-N}$  concentration in the extract was determined using a continuous-flow analyzer (AA3, Bran + Luebbe GmbH,

Norderstedt, Germany). The  $\text{NH}_3$  volatilization rate ( $N$ ,  $\text{kg ha}^{-1} \text{d}^{-1}$ ) and cumulative seasonal losses were calculated by linear interpolation between consecutive sampling dates.  $\text{NH}_3$  volatilization efficiency (%) was expressed as the ratio of cumulative  $\text{NH}_3$  volatilization to total nitrogen applied.

$$N = \frac{M}{A \times D} \times 10^{-2} \quad (1)$$

$$\text{Cumulative } \text{NH}_3 = \sum [N \times D] \quad (2)$$

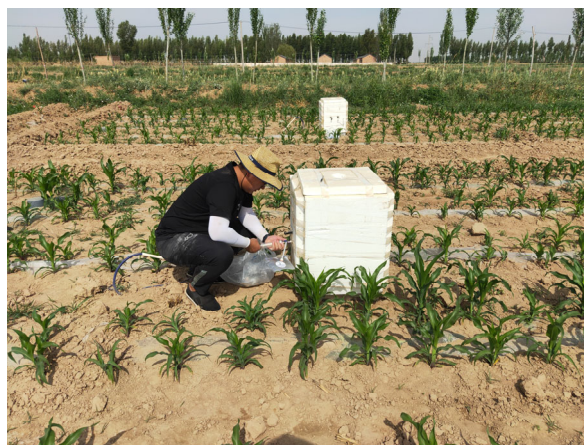
where  $M$  is the mean  $\text{NH}_3$ -N mass (mg) captured by a single device,  $A$  is the cross-sectional area of the capture device ( $\text{m}^2$ ), and  $D$  is the interval between successive measurements ( $d$ ).



**Figure 2.** Structure drawing of venting method for  $\text{NH}_3$  volatilization.

#### 2.4. $\text{N}_2\text{O}$ Flux Measurement

Nitrous oxide emission flux was measured using the static closed-chamber method. One chamber system was permanently installed per experimental plot, giving 24 chambers in total (6 treatments  $\times$  4 replicates). Each chamber system consisted of a stainless steel box (50 cm  $\times$  50 cm  $\times$  70 cm, wall thickness 1.2 cm) externally insulated with a foam layer, mounted on a permanently installed stainless steel base frame (50 cm  $\times$  50 cm  $\times$  15 cm, inserted 12 cm into the soil) fitted with a water-filled sealing channel (Figure 3). A three-way sampling valve and temperature port were positioned  $\sim$ 25 cm above the base frame.



**Figure 3.** Static dark box used to determine  $\text{N}_2\text{O}$ .

Gas samples were collected at 10-day intervals during the background monitoring period (i.e., all periods outside the post-event intensive windows), increasing to every 2 days during the 7-day period following each fertilization and flood irrigation event, as flood inundation independently generates transient anaerobic conditions that drive significant N<sub>2</sub>O pulses. In both years, the growing season extended from late April (sowing) to early October (harvest), approximately 155 days. Intensive sampling was therefore triggered on three occasions: (i) the first flood irrigation combined with topdressing in mid-May; (ii) the second flood irrigation in mid-June; and (iii) the third flood irrigation in early July. All sampling was conducted between 08:30 and 11:30 h local time, a window selected to approximate the daily mean N<sub>2</sub>O flux consistent with standard protocols widely adopted in static closed-chamber studies in northern Chinese cropping systems. Sampling was carried out simultaneously by multiple field teams, with chamber closure times staggered across plots to ensure all measurements fell within this window; each plot required approximately 35 min to complete. In total, approximately 24 sampling occasions were conducted per plot per growing season (12 intensive and approximately 12 background occasions), yielding approximately 96 gas samples per plot per season (4 time-point samples × 24 occasions) and approximately 4600 gas samples across all plots and both growing seasons.

Gas samples were withdrawn by syringe at 0, 10, 20, and 30 min after chamber closure and transferred to pre-evacuated vials; N<sub>2</sub>O concentrations were determined using a Picarro G2308 cavity ring-down spectroscopy analyzer (Picarro Inc., Santa Clara, CA, USA). Cumulative N<sub>2</sub>O emissions were calculated by trapezoidal interpolation between consecutive flux measurements.

The N<sub>2</sub>O emission flux ( $F$ ,  $\mu\text{g m}^{-2} \text{h}^{-1}$ ) was calculated as:

$$F = \rho \times h \times \frac{\Delta c}{\Delta t} \times \frac{273}{273 + T} \quad (3)$$

where  $\rho$  is the gas density of N<sub>2</sub>O under standard conditions ( $\text{kg m}^{-3}$ ),  $h$  is the effective headspace height of the sampling chamber (m), calculated as the total internal chamber height (0.70 m) minus the depth of the base frame inserted into the soil (0.12 m), giving  $h = 0.58$  m;  $\Delta c/\Delta t$  is the rate of N<sub>2</sub>O concentration change within the chamber ( $\mu\text{L L}^{-1} \text{h}^{-1}$ ), 273 is the absolute temperature at 0 °C (K), and  $T$  is the mean air temperature (°C) inside the chamber during sampling. All flux values were converted to daily emission flux units ( $\text{g ha}^{-1} \text{d}^{-1}$ ) to facilitate inter-period comparisons.

Cumulative N<sub>2</sub>O emissions ( $\text{g ha}^{-1}$ ) were calculated by trapezoidal interpolation between consecutive sampling dates:

$$CE = \sum \left( \frac{F_i + F_{i+1}}{2} \times 10^{-3} \times d \times 24 \times 10 \right) \quad (4)$$

where  $CE$  is the cumulative N<sub>2</sub>O emission ( $\text{g ha}^{-1}$ ),  $F_i$  and  $F_{i+1}$  are the emission fluxes at two consecutive sampling times ( $\text{mg m}^{-2} \text{h}^{-1}$ ), and  $d$  is the number of days between adjacent sampling intervals.

### 2.5. Post-Harvest Soil Mineral Nitrogen and Profile Nitrogen Storage

To assess the residual soil nitrogen remaining in the soil profile at the end of each growing season, soil samples were collected by depth increment (0–30, 30–60, and 60–90 cm) from each plot immediately following maize harvest in both 2019 and 2020. At each plot, three soil cores were extracted using a soil auger and composited by depth layer to yield one representative sample per layer per plot. Fresh soil subsamples were extracted with 2 mol L<sup>-1</sup> KCl solution (soil:solution ratio 1:5,  $w/v$ ) by shaking for 1 h at room temperature, followed by filtration through Whatman No. 42 filter paper. Ammonium

nitrogen ( $\text{NH}_4^+\text{-N}$ ) and nitrate nitrogen ( $\text{NO}_3^-\text{-N}$ ) concentrations in the filtrate were determined colorimetrically using a continuous flow analyzer (AA3, Bran + Luebbe GmbH, Norderstedt, Germany). Extract concentrations ( $\text{mg L}^{-1}$ ) were converted to a soil dry-mass basis ( $\text{mg kg}^{-1}$ ) using the 1:5 extraction ratio prior to calculation of nitrogen storage.

Soil nitrogen storage ( $S$ ,  $\text{kg ha}^{-1}$ ) within each depth increment was calculated as:

$$S = C \times \rho_b \times d \times 10 \quad (5)$$

where  $C$  is the  $\text{NH}_4^+\text{-N}$  or  $\text{NO}_3^-\text{-N}$  concentration expressed on a soil dry-mass basis ( $\text{mg kg}^{-1}$ ),  $\rho_b$  is the soil bulk density ( $\text{g cm}^{-3}$ ) of the respective layer, and  $d$  is the thickness of the soil layer (cm). The factor 10 converts units to  $\text{kg ha}^{-1}$ . Total profile nitrogen storage (0–90 cm) was obtained by summing the storage values across all three depth increments. Soil bulk density for each layer was determined from undisturbed core samples collected at the beginning of the experiment using the cutting-ring method.

### 2.6. Maize Grain Yield Measurement

Grain yield was determined at physiological maturity in a representative harvest area of  $6.6 \text{ m}^2$  per plot, avoiding border rows. Harvested grain was oven-dried to a constant weight, and yield was expressed as dry matter yield ( $\text{kg ha}^{-1}$ ) adjusted to a standard moisture content of 14%.

### 2.7. Greenhouse Gas Warming Potential and Emission Intensity

The global warming potential (GWP,  $\text{kg CO}_2\text{-eq ha}^{-1}$ ) attributable to  $\text{N}_2\text{O}$  emissions during the crop growing season was calculated as:

$$\text{GWP} = 273 \times R_{\text{N}_2\text{O}} + \text{GWP}_{\text{background}} \quad (6)$$

where  $R_{\text{N}_2\text{O}}$  is the cumulative  $\text{N}_2\text{O}$  emission during the growing season ( $\text{kg ha}^{-1}$ ), and the factor 273 reflects the 100-year global warming potential of  $\text{N}_2\text{O}$  relative to  $\text{CO}_2$  [5]. Background GWP from  $\text{CH}_4$  and  $\text{CO}_2$  soil respiration was determined from concurrent static chamber flux measurements at the experimental site during the 2019–2020 growing seasons, yielding a site-specific value of  $3023 \text{ kg CO}_2\text{-eq ha}^{-1}$ , which was incorporated into the total GWP calculation for each treatment.

The greenhouse gas emission intensity (GHGI,  $\text{kg CO}_2\text{-eq kg}^{-1}$  grain) was calculated as:

$$\text{GHGI} = \frac{\text{GWP}}{\text{Yield}} \quad (7)$$

where Yield is the grain yield ( $\text{kg ha}^{-1}$ ) of maize at harvest.

### 2.8. Statistical Analysis

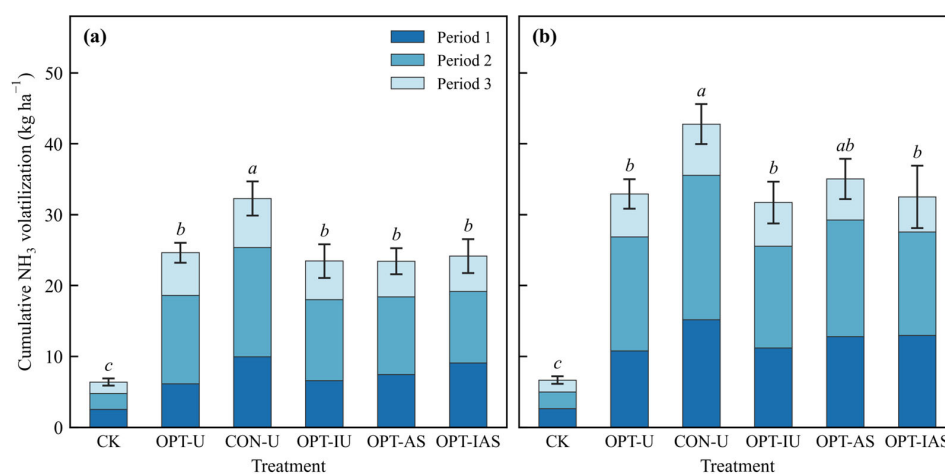
All data were compiled in Microsoft Excel 2024. Statistical analyses were performed in SPSS Statistics v30.0 (IBM Corporation, Armonk, NY, USA). Figures were produced using Python 3.13 with the Matplotlib library (v3.10.6). Each treatment was replicated in four independent blocks ( $n = 4$ ); sub-plot measurements within each plot—three  $\text{NH}_3$  capture devices and three composited soil cores per depth increment—were averaged to yield a single plot-level value prior to analysis, so that the plot mean constituted the unit of replication throughout. Because year represented a confounded climatic and pedological covariate rather than a controlled experimental factor, one-way ANOVA was conducted independently for each growing season, with treatment as the sole factor; cross-year robustness of treatment effects was evaluated by the consistency of treatment rank orders and significance patterns across both years. Differences among treatment

means for cumulative  $\text{NH}_3$  volatilization, cumulative  $\text{N}_2\text{O}$  emissions, post-harvest soil mineral nitrogen storage, GWP, GHGI, and grain yield were evaluated using one-way ANOVA for each year separately, followed by Duncan's multiple-range test for post-hoc pairwise comparisons ( $p < 0.05$ ). To evaluate the effects of fertilizer treatment and soil pH manipulation on  $\text{NH}_3$  volatilization efficiency, a two-way mixed ANOVA was performed for each year separately, with Treatment as the between-subject factor and pH condition ( $\text{pH}^-$ , ambient,  $\text{pH}^+$ ) as the within-subject factor, using block as the subject unit. To evaluate the effect of growth stage on cumulative  $\text{NH}_3$  volatilization and  $\text{N}_2\text{O}$  emissions, one-way ANOVA was performed with growth stage (Period 1, Period 2, and Period 3) as the independent variable across all treatments for each year separately. The significance threshold was set at  $p < 0.05$ . Pearson correlation coefficients were computed among nitrogen application rate, cumulative  $\text{NH}_3$  volatilization, cumulative  $\text{N}_2\text{O}$  emissions, post-harvest soil nitrogen storage, and grain yield using plot-level observations pooled across both years ( $n = 48$ ). Ninety-five percent confidence intervals for each correlation coefficient were calculated using the Fisher z-transformation. To assess interannual stability, correlations were additionally computed separately for each growing season ( $n = 24$  per year). Significance was assessed at  $p < 0.05$  and  $p < 0.01$ .

### 3. Results

#### 3.1. $\text{NH}_3$ Volatilization and Soil pH Effects

Nitrogen application rate was the primary factor associated with cumulative  $\text{NH}_3$  volatilization, with CON-U producing significantly greater seasonal losses than all optimized-rate treatments in both years ( $p < 0.05$ ), while the unfertilized CK lost less than  $7 \text{ kg ha}^{-1}$  in either year (Figure 4). Among optimized-rate treatments, neither fertilizer formulation nor inhibitor inclusion resulted in significant differences in cumulative emissions in either year ( $p > 0.05$ ), although OPT-AS exceeded OPT-U by 6.4% in 2020. Period 2 consistently accounted for the largest share of seasonal  $\text{NH}_3$  loss across fertilized treatments (41.8–50.8% in 2019; 44.9–48.8% in 2020). One-way ANOVA confirmed a significant effect of growth stage on cumulative  $\text{NH}_3$  volatilization in both years (2019:  $F = 247.85$ ,  $p < 0.001$ ; 2020:  $F = 159.54$ ,  $p < 0.001$ ). All fertilized treatments showed substantial year-on-year increases in cumulative  $\text{NH}_3$  (ranging from 32.6% to 49.6%), whereas the CK increased by only 4.4%.



**Figure 4.** Cumulative  $\text{NH}_3$  volatilization ( $\text{kg ha}^{-1}$ ; mean  $\pm$  SE) by treatment under ambient soil pH in (a) 2019 and (b) 2020. Stacked bars show contributions from Period 1 (sowing to first irrigation), Period 2 (first to second irrigation, encompassing topdressing), and Period 3 (second irrigation to harvest). Different lowercase letters above bars indicate significant differences among treatments (Duncan's test,  $p < 0.05$ ).

Two-way mixed ANOVA confirmed a significant main effect of soil pH condition on  $\text{NH}_3$  volatilization efficiency in both years (Table 2), with no significant Treatment  $\times$  pH interaction, indicating that the response to pH manipulation was consistent across all fertilizer treatments. An increase of approximately 0.5 pH units above ambient was associated with a 25% increase in volatilization efficiency relative to ambient conditions, whereas a comparable decrease reduced efficiency by up to 18% (Table 3). These responses were directionally consistent across all fertilizer treatments. Under ambient conditions, CON-U exhibited lower volatilization efficiency than the four optimized-rate treatments in both years; this pattern reflects the larger absolute nitrogen denominator in CON-U rather than a reduction in absolute  $\text{NH}_3$  losses, as absolute cumulative losses under CON-U were the highest among all treatments. The consistent sensitivity of  $\text{NH}_3$  losses to pH variation of less than one unit—at a site where ambient pH already exceeds 8.8—underscores the dominant control exerted by soil alkalinity on  $\text{NH}_3$  volatilization in this system.

**Table 2.** Two-way mixed ANOVA of  $\text{NH}_3$  volatilization efficiency in response to fertilizer treatment and soil pH manipulation.

Source	2019	2020
Treatment	F = 137.17, $p < 0.001$	F = 420.47, $p < 0.001$
pH condition	F = 117.78, $p < 0.001$	F = 82.49, $p < 0.001$
Treatment $\times$ pH	F = 1.43, $p = 0.226$	F = 2.17, $p = 0.080$

**Table 3.**  $\text{NH}_3$  volatilization efficiency (%) under ambient and pH-adjusted conditions for each treatment in 2019 and 2020 (mean  $\pm$  SE).

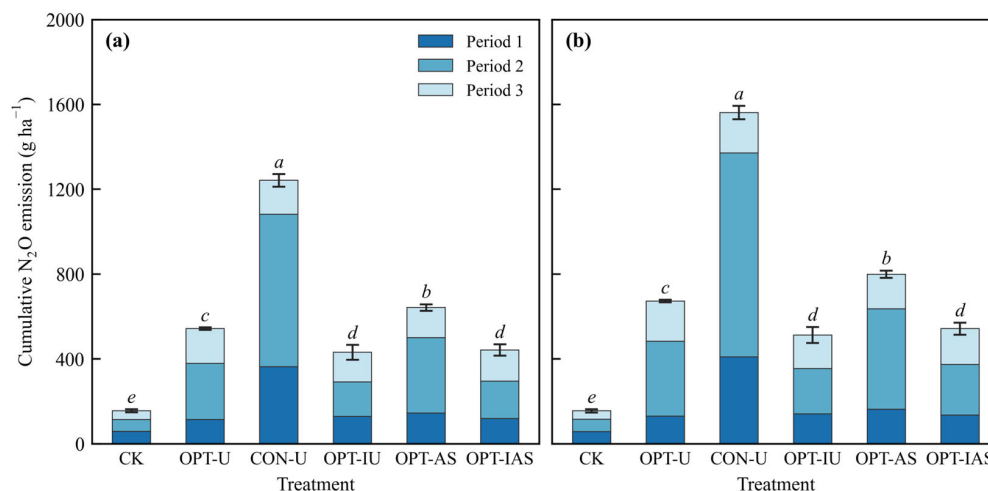
Year	Treatment	Effic. Ambient (%)	Effic. pH <sup>+</sup> (%)	Effic. pH <sup>-</sup> (%)
2019	OPT-U	13.67 $\pm$ 0.40 a	15.66 $\pm$ 0.88	12.51 $\pm$ 0.54
	OPT-IU	13.02 $\pm$ 1.04 a	16.09 $\pm$ 0.15	12.56 $\pm$ 1.57
	OPT-AS	13.00 $\pm$ 0.51 a	14.64 $\pm$ 1.13	11.28 $\pm$ 0.47
	OPT-IAS	13.41 $\pm$ 0.62 a	16.74 $\pm$ 0.09	12.84 $\pm$ 0.51
	CON-U	8.06 $\pm$ 0.54 b	9.85 $\pm$ 0.26	7.16 $\pm$ 0.31
2020	OPT-U	18.29 $\pm$ 0.91 a	20.09 $\pm$ 0.43	17.50 $\pm$ 0.28
	OPT-IU	17.61 $\pm$ 0.56 a	18.75 $\pm$ 0.30	16.10 $\pm$ 0.13
	OPT-AS	19.45 $\pm$ 0.93 a	22.05 $\pm$ 0.51	18.57 $\pm$ 0.57
	OPT-IAS	18.05 $\pm$ 1.93 ab	19.99 $\pm$ 0.50	16.33 $\pm$ 0.08
	CON-U	10.69 $\pm$ 0.08 b	11.05 $\pm$ 0.29	8.81 $\pm$ 0.30

Different lowercase letters within each year's ambient column indicate significant differences among treatments (Duncan's test,  $p < 0.05$ ). pH<sup>+</sup>: soil pH elevated ~0.5 units relative to ambient; pH<sup>-</sup>: soil pH reduced ~0.5 units. Values for pH<sup>+</sup> and pH<sup>-</sup> columns represent means  $\pm$  SE; no between-treatment statistical comparisons were performed for pH-adjusted conditions.

### 3.2. $\text{N}_2\text{O}$ Emissions in Response to Nitrogen Rate and Fertilizer Formulation

Cumulative  $\text{N}_2\text{O}$  emissions were associated with both nitrogen application rate and fertilizer formulation (Figure 5). CON-U produced the highest seasonal emissions in both years (1241.4  $\pm$  29.6 and 1560.9  $\pm$  32.0 g ha<sup>-1</sup> in 2019 and 2020), exceeding OPT-U by more than 129% and CK by eight- to tenfold ( $p < 0.05$ ). Among optimized-rate treatments, OPT-AS consistently generated higher  $\text{N}_2\text{O}$  emissions than OPT-U by approximately 18% in both years ( $p < 0.05$ ). OPT-IU reduced cumulative  $\text{N}_2\text{O}$  emissions relative to OPT-U by approximately 21–24%, and OPT-IAS reduced emissions relative to OPT-AS by approximately 31–32% ( $p < 0.05$  for both). The period between the first and second irrigations (Period 2) contributed the largest proportion of seasonal  $\text{N}_2\text{O}$  emissions across fertilized treatments (38–58% in 2019; 42–62% in 2020). One-way ANOVA confirmed a significant effect of growth stage on cumulative  $\text{N}_2\text{O}$  emissions in both years (2019: F = 1599.03,

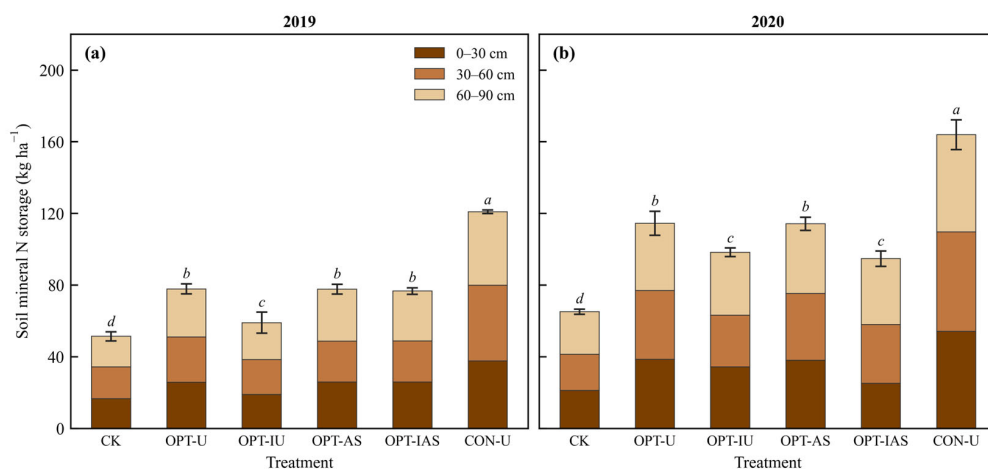
$p < 0.001$ ; 2020:  $F = 1727.93$ ,  $p < 0.001$ ). All fertilized treatments exhibited increases in  $N_2O$  emissions from 2019 to 2020 (approximately 19–26%), whereas CK showed minimal change.



**Figure 5.** Cumulative  $N_2O$  emissions ( $g\ ha^{-1}$ ; mean  $\pm$  SE) by treatment and period in (a) 2019 and (b) 2020. Different lowercase letters above bars indicate significant differences among treatments (Duncan’s test,  $p < 0.05$ ).

### 3.3. Post-Harvest Soil Mineral Nitrogen Storage

Post-harvest mineral nitrogen storage in the 0–90 cm soil profile differed significantly among treatments in both years (Figure 6), with all fertilized treatments retaining more mineral nitrogen than CK ( $p < 0.05$ ). Total profile storage increased from 2019 to 2020 across all treatments, with increases ranging from 26.7% (CK) to 66.6% (OPT-IU). CON-U retained the highest mineral nitrogen in both years ( $120.9 \pm 0.5\ kg\ ha^{-1}$  in 2019;  $163.9 \pm 4.8\ kg\ ha^{-1}$  in 2020), exceeding OPT-U by 55% and 43%, respectively. Among optimized-rate treatments, total storage in 2019 ranged from  $59.0 \pm 2.9\ kg\ ha^{-1}$  (OPT-IU) to  $77.8 \pm 1.4\ kg\ ha^{-1}$  (OPT-U), with OPT-AS and OPT-IAS not significantly different from OPT-U in either year ( $p > 0.05$ ). OPT-IU consistently retained less mineral nitrogen than OPT-U in both years ( $p < 0.05$ ), representing the lowest post-harvest soil nitrogen storage among all fertilized treatments and suggesting improved nitrogen capture by the crop under urease inhibition.



**Figure 6.** Post-harvest soil mineral nitrogen storage ( $kg\ ha^{-1}$ ; mean  $\pm$  SE) by treatment and depth layer in 2019 (a) and 2020 (b). Different lowercase letters above bars indicate significant differences among treatments (Duncan’s test,  $p < 0.05$ ).

### 3.4. Grain Yield, Global Warming Potential, and Emission Intensity

Grain yield increased with nitrogen application rate in both years, with CON-U achieving the highest values ( $11.22 \pm 0.45 \text{ t ha}^{-1}$  in 2019;  $15.03 \pm 1.05 \text{ t ha}^{-1}$  in 2020) and CK the lowest ( $p < 0.05$ ; Table 4). Among optimized-rate treatments, OPT-IU and OPT-IAS produced significantly higher yields than OPT-U in both years, while OPT-AS showed intermediate values not significantly different from OPT-U in 2019. Differences in GWP among treatments were small in absolute terms, varying by less than 15%, whereas grain yield varied by more than 50%. CON-U was the only treatment with GWP significantly higher than all others in both years ( $p < 0.05$ ), reflecting its substantially greater N<sub>2</sub>O emissions. GHGI showed an inverse relationship with yield, with CON-U exhibiting the lowest values in both years ( $0.300 \text{ kg CO}_2\text{-eq kg}^{-1}$  in 2019;  $0.229 \text{ kg CO}_2\text{-eq kg}^{-1}$  in 2020) and CK the highest. This outcome reflects the fact that the yield increment achieved under CON-U was proportionally larger than its GWP increment relative to other treatments—a yield-dilution effect on emission intensity—rather than a reduction in absolute N<sub>2</sub>O emissions, which were the highest of all treatments. Among optimized-rate treatments, OPT-IU and OPT-IAS achieved lower GHGI than OPT-U ( $p < 0.05$ ), driven primarily by their yield advantage rather than by large differences in GWP.

**Table 4.** GWP, Grain yield, and GHGI for all treatments in 2019 and 2020 (mean  $\pm$  SD).

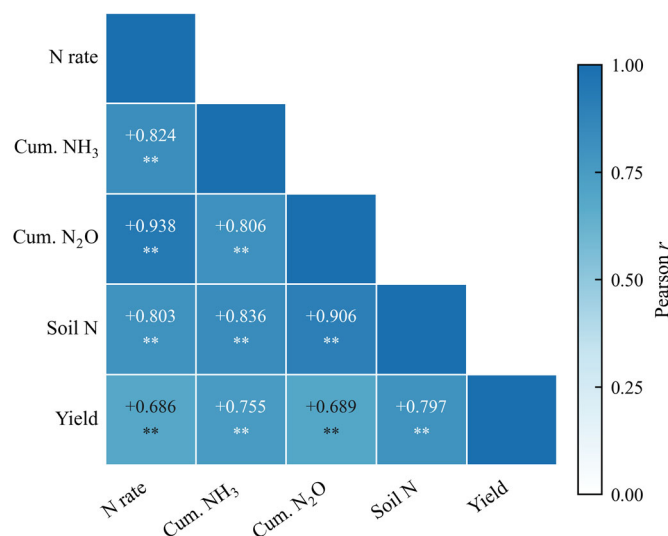
Year	Treatment	GWP (kg CO <sub>2</sub> -eq ha <sup>-1</sup> )	Yield (t ha <sup>-1</sup> )	GHGI (kg CO <sub>2</sub> -eq kg <sup>-1</sup> )
2019	CK	3065.32 $\pm$ 2.04 b	6.51 $\pm$ 0.42 d	0.471 $\pm$ 0.030 a
	OPT-U	3170.94 $\pm$ 1.56 b	7.36 $\pm$ 1.02 d	0.431 $\pm$ 0.060 b
	OPT-IU	3140.42 $\pm$ 9.55 b	9.37 $\pm$ 0.13 b	0.335 $\pm$ 0.005 d
	OPT-AS	3198.08 $\pm$ 4.23 b	8.11 $\pm$ 0.91 c	0.394 $\pm$ 0.045 c
	OPT-IAS	3143.56 $\pm$ 7.26 b	9.53 $\pm$ 0.49 b	0.330 $\pm$ 0.017 d
	CON-U	3361.90 $\pm$ 8.08 a	11.22 $\pm$ 0.45 a	0.300 $\pm$ 0.012 e
2020	CK	3065.18 $\pm$ 2.21 b	8.14 $\pm$ 0.57 e	0.377 $\pm$ 0.026 a
	OPT-U	3206.38 $\pm$ 1.67 b	9.94 $\pm$ 0.70 d	0.323 $\pm$ 0.023 b
	OPT-IU	3162.75 $\pm$ 10.32 b	12.18 $\pm$ 0.85 b	0.260 $\pm$ 0.018 d
	OPT-AS	3240.91 $\pm$ 4.56 b	10.54 $\pm$ 0.74 c	0.307 $\pm$ 0.022 c
	OPT-IAS	3170.94 $\pm$ 7.83 b	12.87 $\pm$ 0.90 b	0.246 $\pm$ 0.017 d
	CON-U	3449.13 $\pm$ 8.74 a	15.03 $\pm$ 1.05 a	0.229 $\pm$ 0.016 e

Different lowercase letters within the same column indicate significant differences among treatments (Duncan's test,  $p < 0.05$ ).

### 3.5. Relationships Among Nitrogen Loss Pathways and Yield

Pearson correlation coefficients were computed among nitrogen application rate, cumulative NH<sub>3</sub> volatilization, cumulative N<sub>2</sub>O emissions, post-harvest soil mineral nitrogen storage (0–90 cm), and grain yield based on plot-level observations across both years ( $n = 48$ ). Because all response variables share a common dependence on nitrogen application rate, these correlations should be interpreted as descriptive of co-variation across the management gradient rather than as evidence of independent mechanistic associations. Nitrogen application rate showed significant positive correlations with all response variables ( $r = 0.686$ – $0.938$ , all  $p < 0.01$ ; Figure 7), consistent with its role as the primary driver of co-variation across the management gradient. NH<sub>3</sub> volatilization and N<sub>2</sub>O emissions were positively correlated ( $r = 0.806$ , 95% CI [0.677, 0.887],  $p < 0.01$ ), as were NH<sub>3</sub> volatilization and post-harvest soil nitrogen storage ( $r = 0.836$ , 95% CI [0.723, 0.905],  $p < 0.01$ ), and N<sub>2</sub>O emissions and soil nitrogen storage ( $r = 0.906$ , 95% CI [0.838, 0.947],  $p < 0.01$ ). Grain yield was positively correlated with NH<sub>3</sub> volatilization ( $r = 0.755$ ,  $p < 0.01$ ), N<sub>2</sub>O emissions ( $r = 0.689$ ,  $p < 0.01$ ), and soil nitrogen storage ( $r = 0.797$ ,  $p < 0.01$ ). All correlation patterns

were directionally consistent between 2019 ( $n = 24$ ) and 2020 ( $n = 24$ ), indicating that the observed relationships were stable across years and not specific to either growing season.



**Figure 7.** Pearson correlation heatmap among nitrogen application rate (N rate), cumulative NH<sub>3</sub> volatilization (Cum. NH<sub>3</sub>), cumulative N<sub>2</sub>O emissions (Cum. N<sub>2</sub>O), post-harvest soil mineral nitrogen storage (Soil N), and grain yield. Correlations are based on plot-level observations ( $n = 48$ ). \*\*  $p < 0.01$ .

#### 4. Discussion

The results demonstrate that nitrogen application rate was the primary driver of both NH<sub>3</sub> volatilization and N<sub>2</sub>O emissions, consistent with its role in elevating soil ammonium concentrations, which simultaneously supply substrate for volatilization and for microbial nitrification–denitrification processes [27,28]. However, the two loss pathways responded differently to fertilizer formulation and inhibitor application, indicating that nitrogen rate alone does not determine their relative magnitudes. NH<sub>3</sub> volatilization was insensitive to fertilizer formulation and inhibitor inclusion, whereas N<sub>2</sub>O emissions differed significantly between ammonium sulfate-based and urea-based formulations and were substantially reduced when inhibitors were included. This divergence reflects the fundamental mechanistic distinction between NH<sub>3</sub> volatilization—a physicochemical equilibrium process governed by soil pH and ammonium availability—and N<sub>2</sub>O production—a biologically mediated process dependent on nitrifier and denitrifier activity [29]. The insensitivity of NH<sub>3</sub> volatilization to fertilizer formulation is consistent with evidence that the loss difference between urea and ammonium sulfate virtually disappears once soil pH exceeds 7.0, when thermodynamic control of the NH<sub>4</sub><sup>+</sup>–NH<sub>3</sub> equilibrium overrides substrate-level effects of fertilizer formulation [30]. Similar decoupling between the two pathways has been documented in northern Chinese cropping systems, where urease and nitrification inhibitors target biochemically distinct steps in the nitrogen transformation sequence, and NH<sub>3</sub> and N<sub>2</sub>O responses to the same management inputs consistently diverge [9].

Soil pH exerted a dominant influence on NH<sub>3</sub> volatilization efficiency, overriding the effects of fertilizer formulation and inhibitor application. The microplot pH manipulation experiment demonstrated that a change of only 0.5 pH units at the ambient site pH of ~8.8 altered volatilization efficiency by up to 25% upward and up to 18% downward, relative to ambient conditions. This asymmetric sensitivity is consistent with the nonlinear relationship between pH and the NH<sub>4</sub><sup>+</sup>–NH<sub>3</sub> equilibrium, in which the proportion of dissolved ammonium present as volatile NH<sub>3</sub> increases sharply above pH 7.5 [11]. Large-scale data syntheses across Chinese upland crops and Mediterranean calcareous systems

consistently identify soil pH as the most important edaphic variable controlling  $\text{NH}_3$  volatilization, surpassing fertilizer type, application method, and crop type in relative importance [31,32], and the within-site, within-season pH manipulation conducted here provides more direct evidence of this relationship under field conditions than cross-site comparisons allow.

The higher  $\text{N}_2\text{O}$  emissions observed under OPT-AS relative to OPT-U at equivalent nitrogen rates are consistent with the immediate availability of  $\text{NH}_4^+$  in ammonium sulfate, which enters the nitrification pathway without the hydrolysis step required for urea and may therefore sustain higher rates of ammonia oxidation during the critical post-irrigation period [33]. In alkaline soils, ammonia-oxidizing bacteria (AOB) dominate the nitrification process and are the primary contributors to nitrification-derived  $\text{N}_2\text{O}$  production, with  $\text{NH}_4^+$  amendment shown to stimulate AOB activity and  $\text{N}_2\text{O}$  emissions particularly at high pH [34]. In calcareous fluvo-aquic soils of northern China—geochemically comparable to the alkaline irrigated soils of the Hetao District—isotopomeric evidence has demonstrated that  $\text{NH}_4^+$ -based fertilizer application actively drives  $\text{O}_2$  consumption and induces suboxic conditions, under which ammonia oxidation and linked nitrifier denitrification together account for the large majority of total  $\text{N}_2\text{O}$  production [35]. Under transiently anaerobic conditions generated by flood inundation, both nitrifier denitrification and heterotrophic denitrification respond strongly to  $\text{NH}_4^+$  substrate availability at low  $\text{O}_2$  concentrations [36], providing a plausible mechanistic basis for the elevated  $\text{N}_2\text{O}$  observed under ammonium sulfate. A two-year field experiment in irrigated maize independently found that ammonium sulfate produced the highest cumulative  $\text{N}_2\text{O}$  losses among tested fertilizer types regardless of irrigation system, with DMPP being the most effective mitigation strategy [37], consistent with the pattern observed here. However, as direct measurements of soil  $\text{O}_2$  dynamics, nitrification rates, and denitrification rates were not conducted in the present study, the relative contributions of nitrifier denitrification and heterotrophic denitrification to the observed  $\text{N}_2\text{O}$  difference between OPT-AS and OPT-U cannot be quantified, and the pathway attribution remains inferential. Nonetheless, it should be noted that globally, ammonium sulfate does not consistently produce higher  $\text{N}_2\text{O}$  than urea, and the reverse pattern has been reported in several field studies under different irrigation and pH regimes [38]; the elevated emissions under OPT-AS are therefore likely context-specific, reflecting the rapid post-irrigation oxygen depletion and high ambient pH characteristic of flood-irrigated alkaline soils in the Hetao District.

Irrespective of fertilizer type, the pronounced concentration of  $\text{N}_2\text{O}$  emissions in Period 2—encompassing topdressing and the first flood irrigation event—across all fertilized treatments is consistent with transient anaerobic conditions generated by flood inundation, which intensify denitrification during a narrow post-irrigation window [39]. Mechanistic evidence confirms that transient anoxia stimulates denitrification capacity and reduces the  $\text{N}_2\text{O}$ -to- $\text{N}_2$  reduction ratio, producing short-lived but intense  $\text{N}_2\text{O}$  pulses that account for a disproportionate share of seasonal emissions [40]. OPT-IAS reduced  $\text{N}_2\text{O}$  emissions by 31–32% without increasing  $\text{NH}_3$  volatilization, contrasting with the global meta-analytic finding that nitrification inhibitors increase  $\text{NH}_3$  emissions by an average of 35.7%, particularly in alkaline soils. This discrepancy likely reflects the dominant thermodynamic control of soil pH on the  $\text{NH}_4^+$ – $\text{NH}_3$  equilibrium established above: at  $\text{pH} > 8.8$ , any secondary effect of DMPP on ammonium substrate availability is masked, consistent with the observation that the  $\text{NH}_3$  penalty of nitrification inhibitor application is amplified at lower pH where substrate-level effects become relatively more important [8]. OPT-IU similarly showed no increase in  $\text{NH}_3$  volatilization, consistent with the mechanism of NBPT action: delayed urea hydrolysis reduces the transient peak in  $\text{NH}_4^+$  concentration at the soil surface, which, at strongly alkaline pH, is the primary determinant of volatilization

flux [41]. The reduction in post-harvest soil mineral nitrogen under OPT-IU relative to OPT-U in both years further suggests that NBPT-mediated slowing of urea hydrolysis improved the synchrony between nitrogen supply and crop demand, reducing residual nitrogen accumulation susceptible to subsequent loss [42].

The inverse relationship between GHGI and grain yield across treatments—most pronounced for CON-U, which achieved the lowest GHGI despite the highest absolute N<sub>2</sub>O emissions—illustrates a well-recognized limitation of emission intensity metrics as standalone environmental indicators [43]. This yield-dilution effect is pervasive across global maize, wheat, and rice systems: yield-scaled N<sub>2</sub>O emissions decline mechanically as yields increase beyond system-specific thresholds, such that high-yielding treatments appear environmentally favorable even when their absolute emission loads are highest [44,45]. The yield-dilution effect under CON-U obscures its substantially greater absolute gaseous losses and post-harvest soil nitrogen accumulation (up to 163.9 kg N ha<sup>-1</sup> in 2020), which represents a legacy nitrogen pool at risk of leaching or denitrification in subsequent seasons. Comprehensive evaluation of nitrogen management strategies, therefore, requires simultaneous consideration of absolute emission loads, residual soil nitrogen, and emission intensity rather than reliance on any single metric. Among optimized-rate treatments, OPT-IU and OPT-IAS achieved the most favorable combination of outcomes—higher grain yield, lower N<sub>2</sub>O emissions, lower GHGI, and in the case of OPT-IU, reduced residual soil nitrogen—relative to OPT-U, without increasing NH<sub>3</sub> volatilization. However, as the preceding analysis shows, no single modification was sufficient to address all nitrogen loss pathways simultaneously.

The present study offers several methodological strengths: the simultaneous quantification of NH<sub>3</sub> volatilization, N<sub>2</sub>O emissions, and post-harvest soil mineral nitrogen storage within a single replicated field experiment enables direct evaluation of trade-offs among competing nitrogen loss pathways, rather than inference from separate studies conducted under different conditions; the embedded soil pH manipulation sub-experiment provides within-site, within-season evidence of pH control on NH<sub>3</sub> volatilization efficiency, free from the confounding by soil type, climate, and management history inherent in cross-site comparisons; and the two-year design, with consistent treatment rankings across contrasting climatic seasons, establishes temporal robustness beyond what single-year studies can support. Several limitations should nonetheless be acknowledged. The experimental design was not fully factorial with respect to fertilizer type and inhibitor type—NBPT was applied exclusively with urea (OPT-IU) and DMPP exclusively with ammonium sulfate (OPT-IAS), reflecting commercially available inhibitor–fertilizer pairings in the study region—and the independent contributions of these two factors to the observed differences in N<sub>2</sub>O emissions therefore cannot be fully disentangled; the comparisons are best interpreted as evaluations of fertilizer formulation effects under conditions representative of regional agricultural practice. The study was additionally conducted at a single experimental site within the Hetao Irrigation District; while the site conditions—alkaline soil pH, high nitrogen inputs, and seasonal flood irrigation—are broadly representative of the district, direct extrapolation of quantitative emission magnitudes to systems differing in soil texture, organic matter content, or irrigation regime should be made with caution. The two-year observation period, while sufficient to establish cross-year consistency in treatment rankings, does not capture the longer-term dynamics of residual soil nitrogen accumulation that may develop under sustained application of the evaluated management strategies. At the system level, a fundamental constraint is the pollution swapping characteristic of alkaline irrigated conditions [46]: NH<sub>3</sub> volatilization remained governed by soil pH and nitrogen input irrespective of fertilizer formulation or inhibitor application, whereas N<sub>2</sub>O reductions were achievable through inhibitor use, and trade-off analyses indicate

that inhibitor-induced  $\text{NH}_3$  increases can under some conditions offset the climate benefit of  $\text{N}_2\text{O}$  reduction [47]. Achieving concurrent mitigation of both pathways will therefore require complementary interventions—such as deep placement or drip fertigation, which have been shown to simultaneously reduce  $\text{NH}_3$ ,  $\text{N}_2\text{O}$ , and  $\text{NO}_3^-$  leaching in alkaline Chinese cropping systems [9,48]—beyond rate optimization and inhibitor application alone.

## 5. Conclusions

This two-year field study demonstrated that  $\text{NH}_3$  volatilization and  $\text{N}_2\text{O}$  emissions are governed by distinct controlling factors in alkaline, flood-irrigated maize systems.  $\text{NH}_3$  volatilization was primarily determined by nitrogen application rate and soil pH; a 0.5-unit increase in soil pH above the ambient value of  $\sim 8.8$  elevated volatilization efficiency by up to 25% relative to ambient, while fertilizer formulation and inhibitor inclusion produced no significant effect on cumulative losses.  $\text{N}_2\text{O}$  emissions were more sensitive to fertilizer formulation and were reduced by approximately 21–32% under inhibitor inclusion without a compensatory increase in  $\text{NH}_3$  volatilization. These contrasting responses indicate that the two gaseous loss pathways are subject to distinct regulatory controls beyond their shared dependence on nitrogen input.

Among the evaluated strategies, optimized nitrogen management combined with inhibitor application (OPT-IU and OPT-IAS) achieved the most favorable balance of outcomes: higher grain yields, lower  $\text{N}_2\text{O}$  emissions, lower GHGI, and reduced residual soil nitrogen accumulation relative to urea alone at equivalent nitrogen rates. However,  $\text{NH}_3$  losses remained governed by soil alkalinity irrespective of fertilizer formulation, indicating that concurrent mitigation of both gaseous loss pathways in alkaline flood-irrigated systems will require additional interventions targeting nitrogen delivery, beyond inhibitor application and rate optimization alone.

**Author Contributions:** Conceptualization, Y.G.; methodology, Y.G. and H.Y.; software, Y.G. and Y.T.; validation, H.Y. and W.Z.; formal analysis, Y.G., Y.D. and W.Z.; investigation, Y.D., H.Y., Y.T. and W.Z.; resources, F.L.; data curation, Y.T. and W.Z.; writing—original draft preparation, Y.G.; writing—review and editing, Y.G., Y.H. and F.L.; supervision, Y.H. and F.L.; project administration, F.L.; funding acquisition, F.L. All authors have read and agreed to the published version of the manuscript.

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## Abbreviations

The following abbreviations are used in this manuscript:

EEF	Enhanced-efficiency fertilizer
NBPT	N-(n-butyl) thiophosphoric triamide
DMPP	3,4-dimethylpyrazole phosphate
GWP	Global warming potential
GHGI	Greenhouse gas emission intensity
OPT-U	Optimized nitrogen rate urea
CON-U	Conventional nitrogen rate urea
OPT-IU	Optimized nitrogen rate urea + NBPT
OPT-AS	Optimized nitrogen rate ammonium sulfate

OPT-IAS	Optimized nitrogen rate ammonium sulfate + DMPP
CK	Control check
UI	Urease inhibitor
NI	Nitrification inhibitor

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