

Modeling impacts of farming management practices on greenhouse gas emissions in the oasis region of China

Y. Wang¹, G. J. Sun¹, F. Zhang¹, J. $Qi^{2,1}$, and C. Y. Zhao¹

¹Key Laboratory of Arid and Grassland Agro-Ecology (MOE), Lanzhou University, Lanzhou, Gansu, 730000, China ²Center for Global Change and Earth Observations, Michigan State University, East Lansing, MI, 48823, USA

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Abstract. Agricultural ecosystems are major sources of greenhouse gas (GHG) emissions, specifically nitrous oxide (N_2O) and carbon dioxide (CO_2) . An important method of investigating GHG emissions in agricultural ecosystems is model simulation. Field measurements quantifying N2O and CO₂ fluxes were taken in a summer maize ecosystem in Zhangye City, Gansu Province, in northwestern China in 2010. Observed N₂O and CO₂ fluxes were used for validating flux predictions by a DeNitrification-DeComposition (DNDC) model. Then sensitivity tests on the validated DNDC model were carried out on three variables: climatic factors, soil properties and agricultural management. Results indicated that: (1) the factors that N2O emissions were sensitive to included nitrogen fertilizer application rate, manure amendment and residue return rate; (2) CO₂ emission increased with increasing manure amendment, residue return rate and initial soil organic carbon (SOC); and (3) net global warming potential (GWP) increased with increasing N fertilizer application rate and decreased with manure amendment, residue return rate and precipitation increase. Simulation of the long-term impact on SOC, N2O and net GWP emissions over 100 yr of management led to the conclusion that increasing residue return rate is a more efficient method of mitigating GHG emission than increasing fertilizer N application rate in the study area.

1 Introduction

The observed increases in global temperature and global sea level during the past 150 yr as well as the well-documented large-scale melting of snow and glaciers during recent decades are unequivocal evidence of global warming (Ver-



Correspondence to: G. J. Sun (sungj@lzu.edu.cn)

meer et al., 2009; Vinnikov et al., 2003; IPCC, 2007a). These changes are the result of human-induced increases in greenhouse gases (e.g., CO₂, CH₄, N₂O) (IPCC, 2001; IPCC, 2007a). Besides fossil fuel burning, agricultural activities are another major contributor to the rise of these greenhouse gases (Watson et al., 1996). It is estimated that global agricultural emissions of CO₂, CH₄ and N₂O account for 20 %, 15 % and 90 % of total human emission, respectively (Bouwman, 1990a; IPCC, 2007a). The contributions of CO₂, CH₄ and N₂O to global warming are 60 %, 20 % and 6 %, respectively (IPCC, 2001; IPCC, 2007a; Oenema et al., 2001). It is thus widely recognized that reducing the agricultural greenhouse gas (GHG) emissions is of great importance in the management of global climate change.

China is an agricultural country with centuries of history of agricultural development. To support the increasingly growing population since the middle of the 20th century, Chinese agricultural area has expanded dramatically, reaching approximately 140 M ha (China statistical yearbook, 2006). The areal expansion has been accompanied with increasingly intensive managements including fertilizer applications. However, these achievements have come with a great cost in exhausting natural resources and in degrading the ecosystems (Huang, 2008a). Although the areal extent of the agricultural lands has been recently decreasing due to acceleration of industrialization and urbanization (Huang, 2008b), fertilizer application has been recently intensified to increase agricultural productivity. An undesired consequence of the intensified application of fertilizer is the increase of the agricultural emission of GHG (FAOSTAT, 2003). Globally speaking, since the industrial revolution the contents of the atmospheric CO₂, CH₄ and N₂O increased 100 ppm, 1000 ppb, 50 ppb, respectively (IPCC, 2007b; Lal, 2004; Mosier et al., 1998). Since the study area of this research is situated in an arid region where soil CH₄ oxidation rate was negligible (Li et al., 2010a), our following discussion thus only focuses on soil N2O and CO2 emissions. Soil N2O comes from two processes in soil: nitrification and denitrification, and these two processes can be affected by climate changes and agricultural activities. For example, the optimal temperature for both nitrification and denitrification is 25-35 °C (Bouwman, 1990b) and nitrogen fertilizer application increases soil N2O emission (Chen, 1989; Li, 1993; Wang, 1994; Hou et al., 1998; Chen, 1995). Soil CO₂ emission is also controlled by climate changes and agricultural activities. For example, temperature rising can effectively enhance the soil CO₂ emission (Han, 2007) and nitrogen fertilizer application can stimulate soil CO₂ emission (Liu et al., 2008; Xing, 2006). The above studies were about the individual effects of nitrogen fertilizer or organic fertilizer on the N2O or CO₂ emissions; however, few studies have considered the associated impact of the two factors (i.e., nitrogen fertilizer and organic fertilizer) on the N₂O and/or CO₂ emissions.

In order to assess the potentials of reducing agricultural N₂O and CO₂ emissions through changing management practices and evaluate the possible responses of agricultural N2O and CO2 emissions to different management practices, we utilized the DeNitrification-DeComposition (DNDC) model to simulate soil carbon sequestration potentials and greenhouse gas emissions brought about by different farmland management practices in agroecosystems (Li, 2004a). In the past 20 yr, the DNDC model has been proven to be effective in many places around the world, such as North America, Europe, Asia and Oceania (Li et al., 1996; Plant, 1999; Stange et al., 2000; Li, 2000; Zhang et al., 2002; Xu et al., 2003; Cai et al., 2003; Frolking et al., 2004; Grant et al., 2004; Smith et al., 2004; Butterbach-Bahl et al., 2004; Pathak et al., 2005; Jagadeesh Babu et al., 2005; Beheydt et al., 2007; Smith et al., 2010). Application of this model in China began in the late 1990s. The study of Xu et al. (2000, 2001) in Guizhou Province suggested that soil N₂O fluxes can be well simulated by the DNDC model for corn-rape rotation, soybean-winter wheat rotation and fallow fields. A study of soil N₂O emission in soybean fields also lends a strong support to the acceptability of the DNDC model in simulating N₂O emission flux during the soybean growing period (Xie and Li, 2004). The strongest support to the acceptability of DNDC model came from the Quzhou experiment station of China Agricultural University where longterm observational data of soil organic carbon variations with different treatments of fertilization and tillage were consistent with the corresponding results of the DNDC modeling (Wang et al., 2004).

We choose an oasis (i.e., Heihe Oasis) as our study area (Fig. 1) because nearly all of the oases in arid and semi-arid areas of northwestern China are facing serious challenges in developing a sustainable economy and in maintaining health ecosystems. Oases account only for about 4% of the total arid and semi-arid area in China, but they support over 90% of the population of the entire area (Wang, 2010). The objectives of the study are: (1) to evaluate the acceptability of the DNDC model using field-observed data, (2) to test the sen-



Fig. 1. Location of the Heihe River Basin and study field.

sitivity of factors affecting the agricultural GHG, and (3) to assess the possible responses of agricultural N_2O and CO_2 emissions to different management practices. We hope that our study can provide scientific references for the optimization of the management practices and the development of sustainable strategies.

2 Materials and methods

A one-year experiment was conducted at a field with summer maize in Zhangye City, Gansu Province, in the Northwestern China. During the experiment, soil CO_2 and N_2O fluxes were measured, and information about local climate, soil and farm management was collected.

2.1 Field site and measurement

Field measurements were conducted during April to October 2010 at an agricultural experimental station run by the Zhangye City Agricultural Science Research Institute. The station is located in an irrigated area within Heihe oasis $(38.91^{\circ} \text{ N}, 100.36^{\circ} \text{ E})$ (Fig. 1). The elevation of the site is 1560 m. The climate is continental, with an approximate annual mean temperature of 7.0 °C, an annual mean rainfall of 127 mm, an annual evaporation of 2345 mm, and 153 frost-free days per year. From April to October, the active accumulated temperature above 5 °C is 3223 °C. The soil in the experimental field is irrigated desert soil with a bulk density of 1.38 g cm⁻³ and pH of 8.6. The initial soil organic carbon (SOC) content of 0.0138 kg C kg⁻¹ for the top 20 cm of the soil profile and the texture of the soil is sandy loam.

Summer maize was the main crop in the area. It was sown on 28 April 2010 and harvested on 3 October 2010 with a sowing rate of 50 kg ha^{-1} . The experiment included four treatments: (1) OM (manures at $2000 \text{ kg C ha}^{-1}$); (2) N

(nitrogen applied as urea at 300 kg N ha^{-1}); (3) MN (manures at $2000 \text{ kg C ha}^{-1}$ and nitrogen applied as urea at 300 kg N ha^{-1}); and (4) B (neither fertilizer nor manure). The experiment was a randomized block design with three replications. That is, 12 plots were used and each plot was $4 \text{ m} \times 8 \text{ m}$. All of the treatments had the same K fertilizer, P fertilizer, tillage and irrigation. The animal manure was applied as a basic fertilizer to the soil on 28 April 2010. Urea application was conducted three times: 28 April (planting), 18 June (jointing stage) and 26 July (silking stage), with a mass ratio of 2:2:1. During the growing season, the field was irrigated 4 times: 18 June, 26 July, 6 August, and 26 August. After the harvest, 15 % of the above-ground maize residue was left in the field and later incorporated into the soil during the subsequent tillage.

Gas samples were collected using the closed-chamber method. Each of the chambers consisted of two parts, one is the chamber cylinder $(30 \text{ cm} \times 50 \text{ cm} \times 70 \text{ cm})$ made of organic glass, and the other one is the base collar with 5 cm internal diameter. The base collars for gas collection chambers were installed in each plot 24 h before the sampling. One base collar was installed in each one of the 12 plots. Permanent boardwalks were set before the cropping season to minimize soil disturbance during gas sampling. The gas sampling started at 09:00 a.m. and ended at 11:00 a.m. (local time). Each sampling lasted for 20 min and 5 samples were taken at an interval of 5 min during each sampling. The field measurement was conducted once per week from April to May, twice or more per week from June to August, and again once per week from September to October. The N2O concentration of gas samples was measured using a GC Agilent 7890 equipped with a ⁶³Ni electron capture detector (ECD) in the laboratory within 2-3 days after sampling. The column for measuring N₂O was packed with Porapak Q (80-100 mesh), and the length of the column was 3 m. The temperature of ECD was 350° and the temperature of column was 55°. The flow rate of carrier gas was 30 ml min⁻¹. The N₂O concentration of each sample was quantified against the concentration of the calibration gas. The N_2O emission flux (F) was calculated with the equation as follows (Li et al., 2010a):

$$F = 60 \times 10^{-5} \times \left[\frac{273}{(273+T)}\right] \times \left(\frac{P}{760}\right) \times \rho H \times \left(\frac{dc}{dt}\right)$$
(1)

where *F* is the N₂O emissions flux $(mg N_2 O m^{-2} h^{-1})$, $\rho(g l^{-1})$ represents N₂O density at 0° and 760 mm Hg, *T* (°) is the mean value of air temperature inside the chamber measured during the closure, *H* (cm) is the height of chamber headspace, *t* (min) is the time for sampling, *dc/dt* $(10^{-9} min^{-1})$ is the increase of the N₂O concentration per minute in the closed chamber, *P* (mm Hg) is the air pressure of experimental site.

Soil CO₂ flux in the field was determined with open-type soil carbon flux monitoring instrumentation of LI-8100 (LI-COR, Lincoln, NE, USA). Three steel collars were installed for each treatment as duplicates. That is, one steel collar was

installed in each one of the 12 plots. To avoid short-term fluctuation in the respiratory rate of soil caused by human disturbance, we inserted all of the steel collars into the soil, with a 5 cm wall exposed above the soil surface for installing the monitoring chamber, and cleared the litter and the newlygerminated weeds in the steel collars 24 h before measurement (Zhang, 2008). Each measurement was commenced at 09:00 a.m. and ended at 11:00 a.m. (local time). The field measurement was conducted twice or three times per month during May, August and September, and twice or more per week from June to July.

2.2 The DNDC model

To assess the impacts of climate change and human activities on ecosystems, biogeochemicalcycles become foci of recent scientific research and a need for developing a mathematical descriptions or model has thus been pressing to quantitatively evaluate the combined influence of various factors (e.g. climate, soil, vegetation and human activity). The DNDC model, a biogeochemical model, was developed to meet this need (Li, 2001).

2.2.1 Model introduction

Denitrification and decomposition are the main chemical reactions that bring nitrogen and carbon in soil into the atmosphere. These two reactions change the soil fertility while releasing CO_2 , N_2O and CH_4 to the atmosphere. The model consists of six sub-models: soil climate, crop growth, decomposition, nitrification, denitrification and fermentation. The functional equations of the six sub-models are primarily derived from basic physical, chemical and biological theories or from empirical relationships based on observed data. The detailed model structure can be seen in Li et al. (1992).

2.2.2 Model application

To simulate tracer gas emissions for a specific site, the DNDC model requires a number of input parameters, including climate conditions, soil properties and agricultural management practices. The basic meteorological data (e.g., daily maximum temperature, daily minimum temperature, daily precipitation, air pressure) were acquired from the China Meteorological Data Sharing Service Network (http: //cdc.cma.gov.cn/). The initial physical and chemical properties of the soil (e.g., SOC, bulk density, pH, clay content) were acquired from conventional field sampling and laboratory analysis. The agricultural management practices (e.g., tillage, fertilization, and irrigation) were determined by the experimental design.

2.2.3 Model validation

The DNDC model can simulate the major GHG emission fluxes of N_2O , CO_2 and CH_4 , crop yields, carbon/nitrogen

content in soil and plants, and soil temperature and humidity in different levels of a soil profile. Observed values were compared with simulation values to evaluate the accuracy of modeling results using statistical metrics (Rykiel Jr, 1996; Smith et al., 1997; Tonitto et al, 2007). The statistics include modeling efficiency (EF) (Loague and Green, 1991), relative error (E) and mean difference (M) (Addiscott and Whitmore, 1987), and they were applied to determine whether there are significant differences between modeled and measured values. Other statistical tests were also conducted to assess the effects of different treatments on modeled and measured results.

2.2.4 Sensitivity tests

The purpose of sensitivity tests is to find those factors that have a major impact on the variables of interest and to quantify the degree of sensitivity. Baseline scenarios were dictated by local climate, soil properties and agricultural management practices. Alternative scenarios were constructed by changing the values of a single input factor while keeping all other input parameters constant. According to the research by Xing and Shen (1998), eight variables were included for sensitivity test and they were mean annual temperature, annual precipitation, soil texture, SOC, pH, residue incorporation, and N fertilizer and organic manure application. The annual mean temperature and annual precipitation in the baseline scenario were acquired from the China Meteorological Data Sharing Service Network (http://cdc.cma.gov.cn/). Soil physical and chemical properties were acquired from the Institute of Soil Science, Chinese Academy of Sciences, where they were compiled from the second national soil survey during 1979–1994 covering all counties. Farm management data come from household surveys. In the sensitivity test, we adopted the same strategies of nitrogen fertilizer and animal manure applications as in the experimental study.

2.2.5 Long-term impacts of management practices

A 100-yr simulation was conducted with the DNDC model to study the greenhouse gas emissions under different scenarios in the oasis summer maize system. The purpose was to identify practices that might reduce greenhouse gas emissions while soil potentials for reasonable productivity were elevated.

The global warming potential (GWP) is an estimate of the degree of contribution to global warming from a given amount of greenhouse gas. It is on a relative scale that compares the effect of the gas in question to that of the same mass of CO_2 . The GWP is calculated using the following equation (Li et al., 2004b):

 $GWP_{i} = CO_{2i} / 12 \times 44 + N_{2}O_{i} / 28 \times 44 \times 298 + CH_{4i} / (2)$ 12 × 16 × 25 where GWP_{*i*} (kg CO₂ ha⁻¹ yr⁻¹) is the global warming potential induced by scenario *i*; CO_{2*i*}, N₂O_{*i*} and CH_{4*i*} are the CO₂ flux (kg Cha⁻¹ yr⁻¹), the N₂O flux (kg Nha⁻¹ yr⁻¹) and the CH₄ flux (kg Cha⁻¹ yr⁻¹), respectively, induced by scenario *i*. For this research which focused on an arid region, the CH₄ oxidation rate was negligible in comparison with the CO₂ and N₂O fluxes from the agricultural soil (Li et al., 2010a).

3 Results and discussion

Field observations provided the primary data of greenhouse gas emissions in a specific environment and were utilized for model validation first. They were then extrapolated through sensitivity analysis and long-term prediction with the validated model.

3.1 Model validation

To simulate daily N₂O and CO₂ fluxes from summer maize field, the DNDC model was run with the following input data: (1) local meteorological data (e.g. maximum and minimum air temperatures and precipitation) in 2010; (2) soil physical and chemical properties (e.g. SOC, soil bulk density, pH, soil clay content); and (3) agricultural management information (e.g. crop type, planting and harvest dates, tillage, fertilization and irrigation), which has already been mentioned in detail in Sect. 2.1. Temperature and water availability of soils have significant impacts on all processes involved in C- and N-cycling in agroecosystems (Frolking et al., 1998; Wang et al., 2010). Therefore, exact simulation of soil temperature and hydrology in agroecosystems is a prerequisite to successful modeling of C- and N-cycling in agricultural soils. In order to validate the DNDC model, we compared modeled and observed values of soil temperature and water-filled pore space (WFPS) at 10 cm soil depth for a field with summer maize (Fig. 2). The results indicated that DNDC was able to capture the seasonal changes in soil temperature and WFPS. Between modeled and observed values, the r^2 values were 0.98 for soil temperature and 0.67 for WFPS, and the relative deviations were about 3.5 % for soil temperature and 26.2 % for WFPS.

The results in Figs. 3 and 4 show that there was a strong correlation between modeled and observed daily N_2O and CO_2 emission fluxes for different fertilizer rates. The relative deviations were about 45 % for N_2O and 25 % for CO_2 . The DNDC-modeled daily N_2O is in the form of peak emissions, mainly coming from both nitrification and denitrification at daily (dry period) and hourly (rainfall period) time steps. The model captured the main peak emissions of N_2O , which matched well the field observations in discharge time, and the peak emissions of N_2O mainly occurred after fertilization and irrigation or during soil freezing and thawing (Fig. 3). The modeled and observed CO_2 fluxes included



Fig. 2. Comparison of observed and modeled soil temperature between 0-10 cm (A), water-filled pore (WFPS) between 0-10 cm(B), and climatological measurements (C).

autotrophic respiration by plant roots and heterotrophic respiration by soil microorganisms. The observed and modeled daily CO₂ emission rates had similar seasonal patterns for four different treatments (Fig. 4). CO₂ fluxes increased in the spring, reached their highest values in the summer, and then decreased rapidly in the autumn. The model results showed that autotrophic respiration of plant roots is the main source of soil CO₂ emission during the maize growing season (Table 1) (Li et al., 2010b; Moyes et al., 2010). The relationships between observed and modeled nitrate (NO₃⁻) and ammonium nitrogen (NH₄⁺) for the top 10 cm of the soil profile in summer maize fields are shown in Figs. 5 and 6. The relative deviations were about 43 % for NO₃⁻ and 40 % for NH₄⁺.

The results of statistical metrics are shown in Table 2. The modeling efficiency, EF, provides a comparison of the efficiency of the chosen model to the efficiency of describing the data as the mean of the measurements. A positive modeling efficiency (EF>0) indicates that the trend is better described with a modeled value rather than a mean value of the measurements (Smith et al., 1997). Moreover, the EF values of DNDC modeling tend to indicate that the DNDC can describe the trend of N₂O, CO₂, NO₃⁻ and NH₄⁺ very

Table 1. Modeled soil CO_2 flux with autotrophic respiration by plant roots and heterotrophic respiration by soil microorganisms.

Treatment	Root-respiration $kg C ha^{-1} yr^{-1}$	Soil-heterotrophic-respiration $kg C ha^{-1} yr^{-1}$
ОМ	3416	1524
Ν	4654	939
MN	4635	1751
В	595	359

OM was a traditional agricultural fertilization mode when there was no chemical fertilizer provided; N was a fertilization mode with high input and intensive agriculture; MN was a fertilization mode recommended by local experts; B was a controlled trial.

well (Table 2). The bias in the total difference between simulation and measurements is determined by calculating the relative error, E, and comparing it with E95 %, calculated from the standard error and the critical value of Student's tdistribution with n-2 degree of freedom and two-tailed 0.05 probability. The biases of DNDC-modeled CO2 and N2O fluxes are not significant (at a 0.05 significance level), but for NO_3^- and NH_4^+ , there are significant differences between model and measured flux in treatments OM and B. The nature of the bias was further examined using the mean difference, M. Because the value of M can be related to Student's t-distribution, a t-test was conducted to determine if there are significant differences between modeled and measured values. In line with statistical convention, a t-value less than the critical two-tailed 2.5 % t-value was taken to indicate that the model did not show a significant bias when compared to measured values, and a value greater than the critical value was taken to indicate that the simulation showed a significant bias towards over- or under-estimation. Results showed that, for CO_2 , N_2O and NH_4^+ , modeled values for all four treatments did not have significant bias, but for NO_3^- , there was significant bias for treatments OM and B, which tend to be similar to the result of the t-test for E (Table 2). The possible reason for these results is that the local rural irrigation method is flooding, that is, direct flooding of the crop field with water. The irrigation water mainly comes from irrigation wells and an upstream irrigation channel. In Heihe Oasis agricultural areas, groundwater nitrate pollution is very serious, and the mean NO₃⁻-N concentration is $11.44 \pm 1.70 \text{ mg L}^{-1}$ in irrigation wells (Yang and Su, 2008). Therefore part of the nitrogen could be supplied to the maize field with irrigation water. The DNDC model version 9.2 did not consider this load of nitrogen; therefore, the modeled values of NO_3^- and NH_4^+ in treatments OM and B are underestimated. Other than the differences between modeled and measured values, the model performance in simulating the trend and patterns should be considered. The results further support the acceptability of the DNDC model.

Table 3 indicated the N_2O and CO_2 fluxes, the crop yield, the annual SOC and the net GWP values obtained with the



Fig. 3. Comparison of observed and modeled N₂O emissions in summer maize fields.



Fig. 4. Comparison of observed and modeled CO₂ emissions in summer maize fields.

DNDC model under four different management scenarios. The results demonstrated that when the crop yield was high, the net GWP under the MN treatment was much smaller than it was under the N treatment. Statistical tests suggested that there was a significant N₂O and CO₂ concentration difference between experiments with and without fertilizer application ($P \le 0.01$), which indicated that more N₂O and CO₂ were generated in experiments with fertilizer application than in the control. This was a result caused by fertilizer application, not by errors in sampling and measurement.

3.2 Sensitivity tests

Three sets of variables considered in sensitivity test are climatic factors, soil properties and agricultural management. The ranges of all variable factors were based on the field survey. The DNDC model was run with different scenarios regarding to climatic factors, soil properties and management practices (Table 4). Relative sensitivities for the tested factors were determined by the comparison of modeled results. A quantitative, relative sensitivity, referred to as the sensitivity index, was calculated to quantify the impacts of input factors on certain output results (Eq. 3) (Walker et al., 2000; Li et al., 2006):

$$S = \frac{\left(\frac{O_2 - O_1}{O_{avg}}\right)}{\left(\frac{I_2 - I_1}{I_{avg}}\right)} \tag{3}$$

where *S* is the relative sensitivity index, I_1 and I_2 are the minimum and maximum input values tested for a given parameter, I_{avg} is the average of I_1 and I_2 , O_1 and O_2 are the model output values corresponding to I_1 and I_2 , and O_{avg} is the average of O_1 and O_2 . A higher absolute value of *S* indicates greater impact of input data on the output, and a



Fig. 5. Comparison of observed and modeled the nitrate (NO_3^-) for the top 10 cm of the soil profile in summer maize fields.

Table 2. The results of statistical metrics related to the performance of models in simulating the summer maize fields.

	EF				E				E _{95 %}			М			t			t _{0.95}			
	OM	Ν	MN	В	OM	Ν	MN	В	OM	Ν	MN	В	OM	Ν	MN	В	OM	Ν	MN	В	
N ₂ O	0.27	0.63	0.79	0.60	38.23	40.49	32.00	29.91	46.23	42.54	45.23	43.06	0.00	0.04	0.01	0.00	0.02*	1.13*	0.56*	1.27*	2.06
CO_2	0.65	0.73	0.67	0.64	8.36	0.52	2.98	15.28	12.80	11.56	17.54	31.86	4.39	1.09	-0.66	1.34	0.72*	0.14^{*}	0.09^{*}	1.72*	2.08
NO_3^-	0.83	0.67	0.42	0.77	56.82	16.59	1.23	59.59	35.89	26.67	25.56	47.20	9.54	3.21	-4.36	11.75	3.42	0.50^{*}	0.46^{*}	2.21	2.14
NH [‡]	0.98	0.86	0.71	0.93	33.78	37.41	33.05	41.64	40.82	38.63	29.91	40.51	0.05	1.07	1.06	0.03	1.19*	1.35*	0.85*	1.03*	2.13

* means t-test values for mean difference (M) are less than critical two-tailed 2.5 % t-value.

Table 3. Modeled N_2O and CO_2 fluxes, crop yields, SOC values and net GWP under four different treatments of summer maize in the field.

	OM	Ν	MN	В
$N_2O (kg N ha^{-1} yr^{-1})$	0.31	3.6	4.57	0.1
$CO_2(kgCha^{-1}yr^{-1})$	4940.17	5592.85	6386.35	953.82
Crop yield $(\text{kg C ha}^{-1} \text{ yr}^{-1})$	1484	2303	2301	293
SOC $0-20 \text{cm} (\text{kg} \text{C} \text{kg}^{-1})$	0.0141	0.0138	0.0142	0.0135
Net GWP (kg CO ₂ -equivalent ha ⁻¹)	-6722	-2084	-6331	346

OM was a traditional agricultural fertilization mode when there was no chemical fertilizer provided; N was a fertilization mode with high input and intensive agriculture; MN was a fertilization mode recommended by local experts; B was a controlled trial.

negative value shows that there is an inverse relationship between the input and the output. In the DNDC model, CO₂ flux includes photosynthesis, plant autotrophic respiration, microorganism heterotrophic respiration, and dissolved organic carbon (DOC) leaching.

3.2.1 Climate effects

The results of the sensitivity tests showed that temperature and precipitation have significant effects on N_2O and CO_2 emissions (Fig. 7). For the two considered climate factors, annual mean temperature had a greater impact on N2O and CO₂ emissions than did annual precipitation (Table 4). Many studies have shown that N2O and CO2 fluxes have significant positive correlations with air temperature or soil temperature if soil moisture is not a limiting factor (Han et al., 2007; Yamulki et al., 1997). However, in this study the sensitivity tests indicated that the N2O flux decreased and the CO2 flux increased with an increase of temperature at the study site. There are two major reasons for decreased N₂O emission at higher temperatures: one is decreased soil nitrogen concentration due to inhibition of vegetation at high temperatures; the other is inhibition of soil mineralization due to decreased water content at high temperatures. The CO₂ flux increased due to the increased decomposition rate of organic matter caused by higher microbial activity, which increased with temperature.

The impacts of precipitation change on gas emissions were mainly due to changes in soil moisture. Model simulations showed that when the precipitation changed from 80% to 120% of the baseline value, the annual N₂O emissions decreased from 3.61 to $3.54 \text{ kg N ha}^{-1} \text{ yr}^{-1}$ and the CO₂ emissions increased from 380 to $412 \text{ kg C ha}^{-1} \text{ yr}^{-1}$. Increased precipitation stimulated denitrification, which was the main process of N₂O production, but with the extension of hypoxia, denitrification could produce more N₂ than N₂O.

Parameter	Baseline	Range tested	Sensitivity index (S)			
			Net GWP	N ₂ O flux	CO ₂	
	0.04	Decrease by 2°C and 4°C	0.67	0.12	0.42	
temperature (°C)	9.04	and increase by 2 °C and 4 °C	-0.07	-0.13	0.43	
Total annual precipitation (mm)	163.8	Decrease by 10 % and 20 % and increase by 10 % and 20 %	0.76	-0.05	0.20	
Soil texture	Sandy loam	Sand, loamy sand, loam, clay	0.68	0.21	0.44	
SOC content (%)	1.38	0.5, 0.75, 1.0, 1.25, 1.5, 1.75, 2.0	-1.86	0.58	1.78	
Soil pH	8.7	7.3, 8, 9.4	1.39	-2.52	-0.01	
Total fertilizer N input $(kg N ha^{-1} yr^{-1})$	300	200, 250, 350, 400, 450	-1.36	1.35	0.01	
Manure amendment $(\text{kg C ha}^{-1} \text{ yr}^{-1})$	0	2000, 4000	1.34	0.41	1.29	
Residue incorporation (%)	15	0, 50, 90	1.64	0.02	0.97	

Table 4. Calculated sensitivity indices quantifying impact of the tested input parameters on output Net GWP, N₂O and CO₂ fluxes.



Fig. 6. Comparison of observed and modeled ammonium nitrogen (NH_{4}^{+}) for the top 10 cm of the soil profile in summer maize fields.

There are three main driving factors that control the production process: soil redox potential (Eh), DOC concentration and available N (i.e. ammonium or nitrate). When the environmental conditions change, these driving factors will also change. If any one factor becomes a restrictive factor, N₂O flux will be reduced. The relationship between soil moisture and CO₂ emission is more complicated. When soil moisture is lower than the field capacity, CO₂ flux increases with precipitation; when soil moisture ranges between the field capacity and the wilting point, there is no significant correlation between precipitation and CO₂ flux; when soil moisture is higher than the field capacity, the diffusion of oxygen in the soil is restrained, as is autotrophic and heterotrophic respiration, and CO_2 flux decreases with increasing precipitation (Davidson et al., 1998; Lavigne et al., 2004; Chen et al., 2004; Reichstein et al., 2003). In the DNDC model, Li et al. (2006) used a recession curve to describe the drainage after rainfall.

3.2.2 Impacts of soil properties

Based on the analysis of sensitivity tests on the soil properties, N₂O emission is very important to SOC factor (Fig. 7). As initial SOC increased from 0.5 % to 2 %, N₂O emission increased from $3.11 \text{ kg N ha}^{-1} \text{ yr}^{-1}$ to $4.43 \text{ kg N ha}^{-1} \text{ yr}^{-1}$,



Fig. 7. Sensitivity of net GWP, N₂O flux and CO₂ flux to change of each of the following factors: climate, soil and management.



Fig. 8. Influence of management practices on long-term SOC content.

which is identical to the result of Jagadeesh Babu et al. (2006), and the range of change covered the range of N_2O flux change caused by the change of other soil factors. Within the DNDC model, dissolved organic carbon (DOC) is the only energy source for the entire denitrification process. Higher SOC generates more DOC, which in turns drives denitrification until the final product N_2 is produced.



Fig. 9. Influence of management practices on long-term N_2O fluxes.



Fig. 10. Influence of management practices on long-term net GWP.

Soil texture is determined mainly by the ratios of sand, silt and clay. The influence of soil clay content on N2O emission is mainly through its effect on soil hydraulic characteristics and soil aeration. Soil aeration gradually deceases as soil texture changes from sand to clay. When soil moisture is high, more N₂O will be produced in fine-textured soil. In this situation, an anaerobic environment is formed, and denitrification becomes the major factor determining N₂O emission (Bollmann and Conrad, 1998). Table 4 indicateds that N₂O emission is more sensitive to the change in soil pH than to other soil properties. Because soil in arid areas is usually alkaline, the pH range set in the sensitivity tests was 7.3–9.4, which was intended to represent normal soil conditions in the Zhangye Region. N2O flux decreased when pH was increasing (Fig. 7). It is commonly agreed that soil denitrification is not directly controlled by pH, but it is indirectly influenced by it because effective carbon control by pH is required for microbial denitrifiers, a decrease in which will lead to a decrease in denitrification (Koskinen and Keeney, 1982). It is generally agreed that the optimal pH for denitrifying bacteria is 6-8 and that their total pH range of activity is 3.5-11.2 (Chen, 1989).

From the sensitivity tests, it can be seen that there was a significant positive correlation between SOC and CO_2 emission rate (Fig. 7), and the SOC content was the most sensitive factors governing soil properties for CO_2 emission (Table 4).

When the initial SOC content increased from 0.5% to 2%, CO₂ emission was increased by 229%. Because the initial SOC is the determining material basis of CO₂ production by microbial decomposition, it is essential to soil respiration (Sikora and McCoy, 1990). Bazzaz and Willians (1991) predicted that the CO₂ emission rate increases with SOC content. In the present study, CO₂ flux is 215, 607 and 506 kg C ha⁻¹ yr⁻¹ when soil texture is sand, loam and clay, respectively. This might be because SOC absorption by clay minerals can inhibit SOC decomposition in the soil and further reduce CO₂ emission.

3.2.3 Impacts of management options

Results from the sensitivity tests indicated that the N₂O and CO₂ emissions were most sensitive to N fertilizer and manure amendment (Fig. 7 and Table 4). Increasing manure amendment from 2000 to $4000 \text{ kg C ha}^{-1} \text{ yr}^{-1}$ increased annual N2O and CO2 emission rates from 4.51 to $5.42 \text{ kg N ha}^{-1} \text{ yr}^{-1}$ and from 1113 to $1843 \text{ kg C ha}^{-1} \text{ yr}^{-1}$, respectively. These increases occurred because manure amendment or return of crop residue to the soil could increase the SOC content and because of the positive correlation between SOC and the emissions of N2O and CO2 (Smith et al., 1997). Application of N fertilizer to agricultural soil affected the N₂O emission rate, and the magnitude of this effect mainly depended on the nitrogen utilization efficiency of the plants. N fertilizer increased the substrate concentrations (NO_3^-, NH_4^+) of nitrification and denitrification, which in turn promoted the production of N₂O, yielding the observed linear increase in N2O emission with increasing N fertilizer (Gregorich et al., 2005). Excess of N fertilizer could significantly promote N₂O emission (Hou and Chen, 1998; Dobbie et al, 1999). There is a complex interaction between N fertilizer application and CO₂ emission rate (Kowalenko et al., 1978; Paustian et al., 1990; Jacinthe et al., 2002). In our analysis, the CO₂ flux increased when N fertilizer application rate increased; however, when the N fertilizer application rate reached a certain value $(250 \text{ kg N ha}^{-1} \text{ yr}^{-1})$, the increase rate of the CO2 flux decreased. One reason for this phenomenon could be a decrease in nitrogen utilization efficiency followed by a slowdown of biomass accumulation and a decrease in autotrophic respiration rate when N fertilizer application rate reaches a certain value. The other reason is that N fertilizer applied to the soil will reduce the C/N ratio, initially promoting microbial decomposition of SOC and increasing the CO_2 emission rate, but the CO_2 emission rate will decrease when the DOC consumption is excessive and there is no supplementation.

3.3 Long-term impacts of management practices

The one-year sensitivity tests identified the factors to which the N_2O and CO_2 fluxes were most sensitive on a short time scale in the Hexi oasis summer maize ecosystem.

For the same soil C and N accumulation and consumption, the results of the DNDC model simulation may show a difference between a long time sequence (100 yr) and a short time sequence (1 yr). To test the long-term impacts, four long-term (100 yr) alternative management scenarios were constructed: (1) a fertilizer application rate of $450 \text{ kg N} \text{ ha}^{-1} \text{ yr}^{-1}$, which is typical of agricultural production in this region, (2) a N fertilizer application rate of $200 \text{ kg N ha}^{-1} \text{ yr}^{-1}$, (3) an increase in the crop incorporation rate from the baseline (15%) to 90%, and (4) an increase in manure amendment rate from the baseline (0) to $2000 \text{ kg} \text{ C} \text{ ha}^{-1} \text{ yr}^{-1}$. The rest of the model's driving variables (i.e., climate, soil and farm management) were kept constant with the observed values and household survey data. The DNDC model was run for 100 yr with each of the scenarios with the climate data of 2010.

3.3.1 Long-term impacts on SOC

When the N fertilizer application rate was 1.5 times higher or 2/3 lower than the baseline scenarios, the SOC content decreased by 0.14% or increased by 0.60%, respectively (Fig. 8). This might be due to the parameter setting in the DNDC model. Because $188 \text{ kg N ha}^{-1} \text{ N}$ fertilizer is required to have maize reach maximum biomass, if more than $188 \text{ kg N ha}^{-1} \text{ N}$ fertilizer is applied, there will not be any further linear increase in yield, and the DNDC-modeled maize yield would no longer be sensitive to N fertilizer. In addition, excessive application of N fertilizer will decrease the ratio of C:N, and higher concentrations of N will increase the activity of soil microbes, which results in a higher decomposition rate, thus causing SOC to have a decreasing trend. Figure 8 also demonstrates that when the rate of crop residue incorporation or manure amendment was increased from 15 % to 90 % or 2000 kg C ha⁻¹ yr⁻¹ was added to the soil, the SOC content increased by 42% or 31%, respectively. Long-term simulation results showed that the increase of N fertilization rate alone was not effective in increasing the SOC content. The residue return rate and the manure amendment rate were the key factors governing the SOC content, and there was a positive correlation between them (Fig. 8).

3.3.2 Long-term impacts on N₂O fluxes

Compared with the baseline scenario, the simulation results explained that elevated manure amendment and N fertilizer application rates both increased the N₂O flux; decreased N fertilizer application rate decreased the N₂O flux. This was because the manure amendment increased SOC content, which provided more nitrification and denitrification substrate to stimulate the production of N₂O. An increased N fertilizer application rate could cause rapid increase of NO₃⁻ and NH₄⁺ in soil, thus promoting nitrification and denitrification; conversely, a decreased N fertilizer application rate could inhibit nitrification and denitrification. Returning straw to the soil reduces N_2O emission, perhaps because the high C/N ratio in the straw could accelerate biological nitrogen fixation and reduce nitrogen loss by denitrification. During straw decomposition, allelopathic substances that inhibit denitrification might be produced (Wang et al., 2006) (Fig. 9).

3.3.3 Long-term changes in net greenhouse gas emissions

The influence of different management practices on global warming was evaluated by calculating net GWP for the four scenarios. Figure 10 shows a rapid increase of net GWP in all four scenarios in the first 35 yr, and then it tends to level off. In comparison with the baseline results, larger amounts of N fertilizer use and manure amendment increases net GWP; smaller amounts of N fertilizer use decreases net GWP. Net GWP from the high residue return rate and the baseline rate are nearly equal after 40 yr. In general, increasing residue return rate would be a more effective measure than other scenarios for reducing net GWP emission.

4 Conclusions

Because agricultural ecosystems are one of the major sources of GHG emissions, reduction of GHG emissions in various agricultural management systems is a current focus of research. Assessing the impacts of various management practices on different crops and soil types through the explicit use of experimental data becomes very difficult. Biogeochemical models, such as the DNDC model, thus play an important role in such research. In the present study, N₂O and CO₂ emissions from agricultural soils were measured in a summer maize field in Zhangye City, China, in 2010. The effectiveness of the DNDC model was assessed by comparison with the field measurements. Sensitivity tests indicated that: (1) the most important factors governing N2O emissions were N fertilizer application rate, manure amendment and residue return rate; $(2) CO_2$ emissions increased with manure amendment, residue return and initial SOC; and (3) net GWP decreased with increasing manure amendment, residue return rate and precipitation and increased with N fertilizer application rate. During the simulated 100 yr, the DNDC model predicted four scenarios of long-term impacts on SOC, N2O and net GWP emissions. The results suggested that a high residue return rate was the most effective practice for maintaining sustainable development of agriculture in the study area.

Because the exact conditions of local agriculture must be considered in the incorporation of crop residue, low crop residue incorporation rate can be influenced by several factors: (1) maize straw is used not only as one of the primary sources of local fuel but also as livestock feed; (2) straw not thoroughly crushed cannot be completely mixed into the soil, thus affecting the germination of crop seed; and (3) because of a lack of mechanization in local agriculture, excessive manpower was required for crop residue incorporation. According to the results of this study, to decrease GHG emissions, the current method of agriculture management should be improved. The first change to be considered is to increase the crop residue incorporation rate, and the other factor to be considered could be a decrease in the amount of N fertilizer applied to a reasonable limit, thereby increasing the amount of organic fertilizer applied. In the real situation, the current status of local agriculture and social advancement must be considered in the application of these measurements. A potential method of increasing crop residue incorporation rate can be (1) seeking alternate source of fuel and livestock feed, instead of maize straw, to meet local needs, and (2) improving the process of crop residue incorporation and deploying a specific set of agricultural machinery for efficient use of manpower, as well as reducing the negative effect on the germination of crop seed. To optimize the amount of different fertilizers applied, a series of geobiochemical models based on conditions of the study area can be developed to provide guidance to local farmers in their application of fertilizer. Regulation and policy directions provided by the government can also help in the extension of management measurement optimized for reducing GHG emissions.

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