Smallholder farms in east African tropical highlands have low soil greenhouse gas fluxes

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Abstract:

Few field studies examine greenhouse gas (GHG) emissions from African agricultural systems resulting in high uncertainty for national inventories. This lack of data is particularly noticeable in smallholder farms in sub-Saharan Africa, where low inputs are often correlated with low yields, often resulting in food insecurity as well. We provide here the most comprehensive study in Africa to date, examining annual soil CO$_2$, CH$_4$ and N$_2$O emissions from 59 smallholder plots, across different vegetation types, field types and land classes in western Kenya. The study area consists of a lowland area (approximately 1 200 m asl) rising approximately 600 m to a highland plateau. Cumulative annual fluxes ranged from 2.8 to 15.0 Mg CO$_2$-C ha$^{-1}$, -6.0 to 2.4 kg CH$_4$-C ha$^{-1}$ and -0.1 to 1.8 kg N$_2$O-N ha$^{-1}$. Management intensity of the plots did not result in differences in annual GHG fluxes measured ($P = 0.46$, 0.14 and 0.67 for CO$_2$, CH$_4$ and N$_2$O respectively). The similar emissions were likely related to low fertilizer input rates ($\leq 20$ kg N ha$^{-1}$). Grazing plots had the highest CO$_2$ fluxes ($P = 0.005$); treed plots (plantations) were a larger CH$_4$ sink than grazing plots ($P = 0.05$); while soil N$_2$O emissions were similar across vegetation types ($P = 0.59$). This study is likely representative for low fertilizer input, smallholder systems across sub-Saharan Africa, providing critical data for estimating regional or continental GHG inventories. Low crop yields, likely due to low fertilization inputs, resulted in high (up to 67 g N$_2$O-N kg$^{-1}$ aboveground N uptake) yield-scaled emissions. Improvement of crop production through better water and nutrient management might be therefore an important tool to increase food security in the region while reducing the climate footprint per unit of food produced.

1 Introduction:

Increased atmospheric concentrations of greenhouse gases (GHG: CO$_2$, N$_2$O and CH$_4$) over the last century have been correlated to increasing mean global temperature (IPCC, 2013), while N$_2$O is also the primary ozone-depleting anthropogenically emitted gas (Ravishankara et al., 2009). Globally, agriculture is directly responsible for approximately 14% of anthropogenic GHG emissions while indirect emissions...
due to conversion of natural landscapes to agricultural systems may contribute an additional 17% (Vermeulen et al., 2012). In developing countries however, agriculture can account for up to 66% of a country’s total GHG emission (Tubiello et al., 2014), with African GHG emissions from agriculture and other land uses estimated to be 61% of total continental GHG emissions (Valentini et al., 2014).

In parts of the developing world, such as Sub-Saharan Africa (SSA), smallholder farms (farm size < 10 ha) comprise almost 80% of farmland and up to 90% of the farms (Altieri and Koohafkan, 2008). Thus, it is likely that smallholder farms have a large effect on the GHG inventories of SSA. Unfortunately, there is a dearth of knowledge on agricultural soil GHG emissions from smallholder systems as only a handful of empirical studies (Table 1) have measured these (e.g. Baggs et al., 2006; Brümmer et al., 2008; Dick et al., 2006; Predotova et al., 2010). Previous studies in Africa were also limited in scope; measuring emissions from a low number of sites (generally less than 10) for a short time period (i.e. less than one year), often with low temporal resolution. This shortage of baseline data makes it impossible for many developing countries to accurately assess emissions from soils used for agriculture or to use Tier II methodology, which requires the development and documentation of country or regionally specific emission factors, to calculate GHG inventories (IPCC, 2006). Also, Tier 1 methodology assumes a linear response to fertilizer, which may not accurately reflect emissions in low input systems (Shcherbak et al., 2014). Finally, because most of the research behind the development of the Tier I methodology has been completed in temperate zones, the differences in climate, soils, farm management and nutrient balances (Vitousek et al., 2009) seem to result in consistent overestimates of GHG fluxes (Hickman et al., 2014; Rosenstock et al., 2013b). This likely translates to inflated national agricultural GHG inventories in SSA that may result in incorrect targeting and inefficient mitigation measures.

Soil GHG emission rates have been related to soil properties such as pH, organic carbon (SOC) content or texture (Khan et al., 2011, Chantigny et al., 2010, Rochette
et al., 2008, Stehfest and Bouwman, 2006), but also to vegetation (crop) type (Stehfest and Bouwman, 2006) and management operations e.g. tillage, fertilizer type or crop rotation, (Baggs et al., 2006; Drury et al., 2006; Grageda-Cabrera et al., 2004; Halvorson et al., 2008; Yamulki and Jarvis, 2002). In contrast to agricultural systems in most OECD (Organisation for Economic Co-operation and Development) states, smallholder farmers differentially allocate resources based on distance from homestead and perceived soil fertility, specifically manure and fertilizer applications, to their fields resulting in strong gradients in soil fertility (Tittonell et al., 2013). The differences in soil fertility can be predicted using remote sensing tools like “Normalized Difference Vegetation Index” (NDVI) to determine the magnitude and temporal variability of primary productivity (Paruelo et al., 2001). Differences in fertility can also be predicted using farmer questionnaires to determine how farmers allocate resources to the fields and then using this typology of farming activities (hereafter “field typology”) to estimate where soil GHG fluxes are more likely to be high. If strong correlations can be empirically observed, such fertility gradients may then be upscaled based on either the NDVI or farmer interviews, further allowing for effective landscape level predictions based on the field-scale GHG measurements.

The lack of good information on GHG fluxes related to agricultural activities in SSA and specifically on smallholder farming systems is a large data gap that needs to be addressed. The objectives of this study were to gather GHG flux data from smallholder farms of the western Kenyan Highlands that represent both the diversity in farming practices and the landscape heterogeneity typically found for many highland regions in East Africa. We hypothesized that a) in view of low rates of fertilizer applications by smallholders the GHG fluxes are generally at the low end of published fluxes from agricultural land, b) the seasonality of rainfall is mirrored by fluxes; and c) differences in land productivity as reflected by NDVI and field typology, as well as differences in vegetation can be used to explain spatial variability in field-scale soil GHG fluxes.
2 Materials and Methods

The study site was a 10 km x 10 km area in Kisumu county of Western Kenya (centered at 35.023°E, 0.315°S), north of the town of Sondu (Fig. 1), and ranged from a lowland area at approximately 1200 m asl to a highland plateau at approximately 1800 m asl. The site is one of the sentinel sites for the CGIAR Research Program on Climate Change, Agriculture and Food Security (CCAFS) and is described in detail in Sijmons et al. (2013). This site was selected as it was found to be broadly similar in terms of demographics (e.g. population density, income) and agro-ecological characteristics (e.g. elevation, temperature, precipitation) of other East African tropical highlands (Braun et al., 1997) allowing us to scale up the results to other countries in the region (Sijmons et al., 2013). Mean annual temperature is approximately 23°C and the average annual rainfall is 1150 mm (Köppen classification of a tropical savanna climate [AW]). Temperatures tend to be slightly cooler and precipitation slightly higher in the highlands compared to the lower regions of the study site. Precipitation patterns are bimodal with the “long rains” occurring from April to June (42% of annual precipitation) and the “short rains” occurring from October through December (26% of annual precipitation).

The site is primarily composed of smallholder rain-fed farms typically growing maize (Zea mays) and sorghum (Sorghum bicolor) during the long rains and beans during the short rains. Based on farmer interviews, approximately 27% of them applied fertilizers (i.e. manure or synthetic fertilizers) to their plots, with application rates being very low. For manure, application rates were approximately 200 kg manure ha⁻¹, which corresponds to approximately 95 kg of C and 5 kg N given typical N contents for cattle in this region (Pelster et al., 2016), while application rates for synthetic fertilizer (two farmers applied diammonium phosphate and one applied urea) were < 50 kg fertilizer ha⁻¹ (< 25 kg N ha⁻¹). These fertilizer rates are much lower than rates typical for industrial production where application rates often exceed 150 kg N ha⁻¹ for maize production.

Soil types in the study area are highly heterogeneous, ranging from well drained, acidic, Nitisols in the upper part of the landscape, to eutric and dystric Cambisols in
mid-altitude areas and poorly drained Planosols in the lower parts (IUSS Working Group WRB, 2015). Selected topsoil characteristics for the different land classes identified in the study region are provided in Table 2.

2.1 Landscape stratification
Differences in management intensity and vegetation were expected to affect GHG fluxes, and so the landscape was stratified to account for the expected variability. The stratification was based on a mixed method land use classification combining remote sensing and household surveys.

For the land classification we followed an approach based on vegetation functioning in terms of the magnitude and the temporal variability of primary productivity (Paruelo et al., 2001), assessed through the proxy variable “Normalized Difference Vegetation Index” (NDVI), which allows approximate but widespread characterizations of productivity across space and time and across different ecosystems (Lloyd, 1990; Xiao et al., 2004). We acquired 2001-2012 NDVI data from MODIS (Moderate Resolution Imaging Spectroradiometer). We selected only those NDVI values indicating good to excellent quality conditions (i.e. pixels not covered by clouds, and with a low to intermediate aerosol contamination). Then, we used the program TIMESAT v.3.1. to reconstruct temporal series (Jönsson and Eklundh, 2002).

From the reconstructed temporal series we assessed six functional metrics depicting the magnitude, seasonality and inter-annual variability of productivity. The metrics used were as follows: 1) the mean annual NDVI; 2) the minimum NDVI; 3) the browning rate (rate of NDVI decrease); 4) the peakness of the NDVI; 5) the intra-annual coefficient of variation (CV) of the NDVI; and 6) the inter-annual CV of the NDVI. These metrics allow us to differentiate between land cover types (e.g. cultivated vs. uncultivated) and between different cultivation management approaches (e.g. agroindustrial vs. subsistence) (Baldi et al., 2015). The different elevation bands and soil types resulted in different magnitudes, seasonality and inter-annual variability of productivity with the highlands generally having higher...
productivity due to the higher rainfall and more fertile soils. We then ran an ISODATA unsupervised classification algorithm (Jensen, 1996), and the resulting spectral classes were aggregated to create patches. After combining minor or sparsely-distributed patches, we ended up with five classes, characterized by the following features: 1) lowland subsistence farms with degradation signs (N = 7); 2) lower slopes, moderate sized mixed farms (N = 8); 3) mid slopes, moderate sized, primarily grazing / shrubland (N = 10); 4) upper slopes / highland plateau, mixed farms (N = 22); and 5) mid slopes, moderate sized mixed farms (N = 12).

We also stratified the plots by field typology using the following variables to define a field type score: 1) crop: this score is the sum of the crop types each household is cultivating in one plot; 2) fertilizer use: this score distinguishes organic and inorganic fertilizers; 3) number of subplots: which allows us to capture the spatial and temporal allocation of land to crops, crop mixtures, and combination of annual and perennial crops in intercropping, permanent and seasonal grazing land; 4) location of field: the assumption being that fields close to the homestead receive preferential land management (fertilization, addition of organic amendments, weeding etc) when compared to fields that are far away (Tittonell et al., 2013); and 5) signs of erosion: fields obtained a different score depending on the severity of the visible signs of erosion. Plots were scored based on the preceding information and those with a higher score were considered field type 1 (N = 17), those with a low score were considered field type 3 (N = 19) and those plots with intermediate scores were assigned a field type 2 (N = 23). It was assumed that field type 1 was the most intensively managed (i.e. more fertilizer /manure additions, etc) and field type 3 the least managed (i.e. none to very low fertilizer additions, degraded, low soil C, etc). For a more detailed description of the stratification process see Rufino et al (2015).

Finally, the plots were stratified by vegetation (cover) type: treed/bush (generally plantations of either Grevillia spp or Eucalyptus spp) (N = 7), perennial grasses/grazing (N = 15) and annual cropping (N = 37). Initially, the total number of
sample plots was 60 with the number per category based partly on the area covered
by each specific land classification/field type/vegetation type combination and
partly on logistical constraints (i.e. access). One plot however, was converted into a
construction site in late 2013, resulting in only 59 plots being measured for the full
year.

2.2 Field soil GHG flux survey
At the 59 field sites (see above and Fig 1) soil CO$_2$, N$_2$O and CH$_4$ fluxes were
measured weekly starting the week of 12 August 2013 through to 12 August 2014
(one full year including two growing seasons) using non-flowthrough, non-steady
state chambers (Rochette, 2011; Sapkota et al., 2014). Given the large number of
sites and the difficult access, this required four 2-person crews sampling 4 days per
week. Briefly, four rectangular (0.35 m x 0.25 m) hard plastic frames per site were
inserted 0.10 m into the ground. Fields planted with annual crops were ploughed,
either using an oxen-pulled plough or by hand, twice during this period, which
meant that the frames needed to be removed and then re-installed, however where
possible the chamber frames were left undisturbed for the entire period. For fields
planted with annual crops, the frames were installed between the rows and were
weeded the same week the farmers weeded the rest of the field. The chambers in
the grazing and treed plots would have included some vegetation (primarily
grasses), but these were kept short (<5 cm long) by the continual grazing. On each
sampling date, an opaque, vented and insulated lid (0.125 m height) covered with
reflective tape was tightly fitted to the base (Rochette, 2011). The lid was also fitted
with a small fan to ensure proper mixing of the headspace, and air samples (15 mL)
were collected from the headspace at 0, 15, 30 and 45 min after deployment, using a
syringe through a rubber septum. Even with the insulation and reflective tape on the
chambers, the air temperature inside the chambers still increased during
deployment (approximately 10°C on average), which may slightly affect microbial
and root activity in the soil underneath the chamber. The increase in temperature
inside the chamber headspace would also cause some bias in the calculation of
mixing ratios, which was estimated to be about 3%.
To increase the number of sites measured while still accounting for the representativeness of flux measurements in view of expected high spatial variability of fluxes at field scale samples were pooled from the four replicate chambers at each plot (Arias-Navarro et al., 2013) to form a composite air sample of 60 mL. This method has been found to provide flux estimates within 8% and 4% (for CO$_2$ and N$_2$O respectively) of the estimates calculated by separate sampling, although it is unclear which is the more accurate depiction of the true mean. Also, as noted by Arias-Navarro et al. (2013), this precludes the ability to examine within-site variability. However we believed that the trade-off between on-site variability and sampling a broader range of sites was worthwhile given our aims of characterizing emissions in a way that captured both the diversity in farming practices and landscape heterogeneity typically found for many highland regions in East Africa.

The first 40 mL of the sample was used to flush a 10 mL sealed glass vial through a rubber septum, while the final 20 mL was transferred into the vial to achieve an over-pressure to minimize the risk of contamination by ambient air. The gas samples were analyzed within 10 d of sample collection for CO$_2$, CH$_4$ and N$_2$O in an SRI 8610C gas chromatograph (9’ Hayesep D column) fitted with a $^{63}$Ni-electron capture detector for N$_2$O and a flame ionization detector for CH$_4$ and CO$_2$ (after passing the CO$_2$ through a methanizer). The flow rate for the carrier gas (N$_2$) was 20 mL min$^{-1}$. Every fifth sample analyzed on the gas chromatograph was a calibration gas (gases with known CO$_2$, CH$_4$ and N$_2$O concentrations in synthetic air) and the relation between the peak area from the calibration gas and its concentration was used to determine the CO$_2$, CH$_4$ and N$_2$O concentrations of the headspace samples.

### 2.3 Calculation of soil GHG fluxes

Soil GHG fluxes were calculated by the rate of change in concentration over time in the chamber headspace (corrected for mean chamber temperature and air pressure) after chamber deployment, as shown in Equation 1.

Equation 1. \[
F_{ghg} = \left(\frac{\partial c}{\partial t}\right) \left(\frac{M}{V_m}\right) (V/A)
\]
Where, $F_{GHG}$ is the flux of the GHG in question, $\partial c/\partial t$ is the change in concentration over time, $M$ is the molar mass of the element in question (N for N$_2$O and C for CO$_2$ and CH$_4$), $V_m$ is the molar volume of gas at the sampling temperature and atmospheric pressure, $V$ is the volume of the chamber headspace and $A$ is the area covered by the chamber.

For calculating the change in the GHG concentration over time, non-linear models are generally less biased than linear models; however, they also tend to be very sensitive to outliers (Rochette, 2011). Therefore, when there was a strong correlation for the non-linear model ($R^2 > 0.95$) we used a second-order polynomial; otherwise, we used a linear model. See Rochette and Bertrand (2008) for details on these models. If however the $R^2 < 0.95$ for the non-linear model and $<0.64$ for the linear model, we assumed there was no valid flux measurement and the data point was thrown out. We validated the data for each chamber measurement by examining the dynamics of the CO$_2$ concentrations over the 45-minute deployment period. Chambers that experienced a decrease in CO$_2$ greater than 10% between any of the measurement times were assumed to have a leak and all GHG fluxes were discarded unless the decrease occurred in the last measurement; in this latter case, the flux rate was calculated with the first three measurement points. In cases where the change in concentration was lower than the precision of the instrument, we assumed zero flux. The minimum flux detection limits (Parkin et al. 2012) were 3.61 and 12.46 µg N$_2$O-N m$^{-2}$ hr$^{-1}$ for the linear and non-linear models respectively and 0.015 and 0.051 mg CH$_4$-C m$^{-2}$ hr$^{-1}$ for the linear and non-linear models respectively. Also, negative fluxes for CO$_2$ were deleted, while negative CH$_4$ and N$_2$O fluxes were accepted, as uptake of either in upland soils is feasible. To minimize measurement error and uncertainty, we used methods that were ranked as either “good” or “very good” for 15 of the 16 criteria selected by Rochette and Eriksen-Hamel (2008), with only the deployment time exceeding the recommended time by about 10%. Cumulative annual fluxes were estimated for each plot using trapezoidal integration between sampling dates.
2.4 Soil analysis

At the beginning of the experiment and for each sampled site, five replicate soil samples were taken both at 0-5 cm and 5-20 cm depths with a stainless steel corer (40 mm inner diameter). Samples were individually placed in labelled zip-lock bags. All soil material was oven-dried at 40°C for a week with large clumps being progressively broken by hand. Carbon and N concentrations were determined on powdered samples using an elemental combustion system (Costech International S.p.A., Milano, Italy) fitted with a zero-blank auto-sampler. Soil pH was measured in a 2:1 water:soil solution. Soil texture was determined gravimetrically as described by van Reeuwijk (2002).

In addition, soil samples were collected periodically (every 2 months) for determination of inorganic N concentrations. Briefly, the topsoil (0-10 cm depth) was collected using a soil auger. Three samples from each plot were collected and pooled to form one composite sample. These were taken back to the lab and stored (4°C) for less than one week before extraction (1:5 soil:solution w:v ratio) with 2M KCl. Extracts were kept frozen until analyzed. Analysis for NO$_3$-N was done via reduction with vanadium, with development of color (540 nm) using sulfanilic acid and naphtylethylendiamin and measurement of light absorbance on an Epoch microplate spectrophotometer (BioTek, Winooski, VT, USA). The NH$_4$-N concentrations were measured using the green indophenol method (660 nm) using the same spectrophotometer (Bolleter et al., 1961).

2.5 Environmental data

Environmental data were collected at two sites, one in the uplands (35.056°E, 0.351°S, 1676 m a.s.l.) and the other in the lowlands (34.988°E, 0.308°S, 1226 m a.s.l.). Each of the two weather stations was installed at a farm where we also measured GHG emissions. Air temperature was measured using a Decagon ECT (Decagon Devices, Pullman, WA, USA) air temperature sensor (measurement every 5 minutes), while precipitation data were collected with a Decagon ECRN-100 high resolution, double-spoon tipping bucket rain gauge. Soil moisture and temperature were measured using a Decagon MPS-2 Water potential and temperature sensor.
Data were logged on a Decagon Em50 data collection system and downloaded periodically (typically monthly).

Air temperature, soil temperature and soil moisture (5 cm depth) were also measured at each site at the time of gas sampling using a ProCheck handheld datalogger outfitted with a GS3 sensor (Decagon Devices).

2.6 Plant production
To estimate crop yields and crop N content of annual crops in the study area, we randomly selected nine of the study plots including annual crops (four plots with maize, four with sorghum and one with green grams [Vigna radiate (L.) R.Wilczek]). In June 2013, all the plants within a 2.5 m x 2.5 m square near the center of the field (i.e. to avoid edge effects) were harvested and the grains were removed from the plant. Both the stover and grains were dried for 48 hours at 60°C and then weighed. A sub-sample of the grains was then ground and analyzed for C and N content on the same elemental combustion system described above for soil analysis. Yield-scaled \( \text{N}_2\text{O} \) emissions (g \( \text{N}_2\text{O} \cdot \text{N} \cdot \text{kg}^{-1} \) above ground N uptake) were calculated for each site by dividing the cumulative emissions of the growing season by the grain yields. The growing season lasted from mid-March to August, which corresponds to the period between preparation of fields for the long rains through harvest and up to the preparation of the fields for the following growing season. No estimate of crop yields (or yield-scaled emissions) was done for the second growing season.

2.7 Statistical analysis
Correlations between GHG fluxes and soil properties were tested using Pearson Correlation. The cumulative field fluxes for a 4-week period during the dry season were compared to cumulative fluxes for a 4-week period during the rainy season using ANOVA (AOV in RStudio v. 0.98.953), with the season, management practices (ploughed versus not ploughed for CO\(_2\) and fertilized versus not fertilized for N\(_2\)O) as fixed factors along with the two-way interaction terms. Cumulative field annual GHG fluxes were compared with ANOVA using an un-balanced design and cover type, land class and field type as fixed factors. In all cases, the distributions of flux
measurements were tested for normality using Shapiro-Wilks. Cumulative soil N$_2$O fluxes were not normally distributed and were transformed using the natural log.

3 Results

3.1 Field meteorological and site observations

The soils were slightly acidic to neutral, ranging in pH from 4.4 to 7.5 (mean = 6.0). Carbon and N contents ranged from 0.7 to 4.0% (mean = 2.2) and 0.07 to 0.33% (mean = 0.17) respectively (Table 2). The C/N ratio ranged from 7.7 to 18.1 (mean = 12.6) and the C and N contents in the top 20 cm of the soil were highly correlated with each other (R = 0.976; P < 0.0001).

Annual precipitation (15 August 2013 through 14 August 2014) in the lowlands was 1127 mm while there was 1417 mm of precipitation in the highlands, a 25% increase across the 450 m elevation difference between the two stations. The average minimum and maximum daily temperatures in the lowlands were 15.6 and 30.5°C while temperatures were slightly cooler in the highlands, with an average minimum of 12.6 and an average maximum of 26.9°C. Comparing the precipitation at the sites to a long-term 40-year (1960 to 2000) precipitation data set for the two nearby towns of Kisumu and Kericho (data available at africaopendata.org), we see that annual precipitation was within 10% of the long term average. The monthly rainfalls were generally similar to long-term trends, with the exception of the rainfall in December, which was 26% of the long-term average, and the rainfall in March, which was two-fold higher than the long-term mean.

3.2 Field scale soil GHG fluxes and ancillary information

Soil CO$_2$ fluxes during August 2013 ranged from 50 to 200 mg CO$_2$-C m$^{-2}$ h$^{-1}$, slowly decreased through to November and remained low (< 100 mg CO$_2$-C m$^{-2}$ h$^{-1}$) until the onset of the long rains during March/April 2014 (Fig. 2). The onset of the long rains increased the soil water content from an average of 0.09 m$^3$ m$^{-2}$ by beginning of March 2014 to an average of 0.31 m$^3$ m$^{-2}$ by mid March 2014. Within two weeks
of this increase in soil moisture, the CO\textsubscript{2} fluxes began to increase, reaching a maximum on 14 April 2014 (mean = 189 mg CO\textsubscript{2}-C m\textsuperscript{-2} h\textsuperscript{-1}; Fig. 2).

In general, soil CH\textsubscript{4} fluxes were negative. Uptake rates tended to stay between 0 and 100 µg CH\textsubscript{4}-C m\textsuperscript{-2} h\textsuperscript{-1} from August 2013 until April 2014, after which the variability decreased varying between 0 and 50 µg CH\textsubscript{4}-C m\textsuperscript{-2} h\textsuperscript{-1} (Fig. 2). Soil N\textsubscript{2}O fluxes were low (generally < 10 µg N\textsubscript{2}O-N m\textsuperscript{-2} h\textsuperscript{-1}) for most of the year; with fluxes increasing from a mean of 1.6 µg N\textsubscript{2}O-N m\textsuperscript{-2} h\textsuperscript{-1} for the period from October 2013 to March 2014 to a mean of 10.5 µg N\textsubscript{2}O-N m\textsuperscript{-2} h\textsuperscript{-1} for the 6-week period after soil re-wetting in March/April 2014. The inorganic N concentrations in the top 10 cm of soil (approximately 85% N-NO\textsubscript{3} and 15% N-NH\textsubscript{4}) generally remained below 20 mg N kg\textsuperscript{-1} soil, although concentrations did increase to around 30 mg N kg\textsuperscript{-1} soil in late December 2013 / early January 2014, shortly after the annual crops planted during the short rains were harvested but before the onset of the long rains in late March / early April 2014.

A comparison of the four-week cumulative fluxes from February (end of the dry season) to April (immediately following the start of the rainy season) shows greater cumulative CO\textsubscript{2} and N\textsubscript{2}O fluxes during the wet season, but no difference in CH\textsubscript{4} fluxes (Table 3). The increase in CO\textsubscript{2} and N\textsubscript{2}O fluxes coincided with farmers ploughing their fields and planting and fertilizing their annual crops. However, even though the increase in CO\textsubscript{2} and N\textsubscript{2}O fluxes was slightly larger in the managed plots (ploughed for CO\textsubscript{2} and fertilized for N\textsubscript{2}O comparisons), neither of these management interventions significantly altered emission rates (Table 3).

Cumulative annual GHG fluxes ranged from 2.8 to 15.0 Mg CO\textsubscript{2}-C ha\textsuperscript{-1}, -6.0 to 2.4 kg CH\textsubscript{4}-C ha\textsuperscript{-1} and -0.1 to 1.8 kg N\textsubscript{2}O-N ha\textsuperscript{-1}. There was no detectable effect on cumulative CO\textsubscript{2} fluxes by field type or land class (P = 0.46 and 0.19 respectively; Fig. 3); although grazed plots emitted more CO\textsubscript{2} overall than either annual cropland or treed plots (P = 0.005). Cumulative annual N\textsubscript{2}O fluxes also did not differ by either field type or vegetation type (P = 0.67 and 0.59 respectively; Fig. 3), however land class did significantly affect N\textsubscript{2}O fluxes (P = 0.09; Fig 3) with land class 3 (mid-slopes,
grazing) showing higher N$_2$O fluxes than land class 4 (upper slopes, mixed farms).

Cumulative annual CH$_4$ fluxes were predominately negative, and unlike N$_2$O and CO$_2$,
varied by land class ($P = 0.01$) and land cover type ($P = 0.01$), but not by field type ($P$
$= 0.16$; Fig. 3). Uptake of atmospheric CH$_4$ by soils in land class 2 (lower slopes,
degraded) was greater than in classes 1 (lowland farms with degraded soils) or 3
(mid-slopes grazing land; Fig. 3). Uptake was also almost three-fold greater in treed
plots than in those plots with grasses and or in those used for grazing (Fig. 3). The
difference seems to be primarily due to one grazing plot that was a CH$_4$ source for
14 of 24 sampling dates between 5 August 2013 and 10 February 2014. This same
plot also had the second highest cumulative N$_2$O fluxes (1.5 kg N$_2$O-N ha$^{-1}$ yr$^{-1}$),
however the CO$_2$ fluxes were average (7.2 Mg CO$_2$-C ha$^{-1}$ yr$^{-1}$) and the soil organic C
and N contents were relatively low (1.2 and 0.10% respectively) compared to the
rest of the plots (Table 2).

Both the soil C and N content were correlated with cumulative CO$_2$ fluxes ($r = 0.411$;
$P = 0.002$ and $r = 0.435$; $P < 0.001$, for C and N content respectively), but not with
either the cumulative N$_2$O fluxes ($P = 0.321$ and 0.365 for C and N respectively) or
the cumulative CH$_4$ fluxes ($P = 0.188$ and 0.312 for C and N respectively). The
cumulative CO$_2$ and N$_2$O fluxes were also not correlated ($P = 0.188$).

Many of the farmers within the study site complained that the annual crops planted
in March 2013 failed due to the poor timing of the rains. The crop yields ranged
from 100 to 300 kg ha$^{-1}$ for maize ($n = 4$), from 140 to 740 kg ha$^{-1}$ for sorghum ($n =$
4) and were approximately 20 kg ha$^{-1}$ for green grams ($n = 1$) during the long rain
season (March through June). The low yields resulted in yield-scaled soil N$_2$O fluxes
of up to 67 g N$_2$O-N kg$^{-1}$ aboveground N uptake.

4 Discussion

The soil CO$_2$ fluxes were seasonal, and it was thought that management events, such
as ploughing or fertilizer applications, would affect the GHG flux rates throughout
the year. However, during the commencement of the rainy season in March 2014,
which coincided with tilling, the ploughed fields did not show significant increases in soil respiration rates beyond the enhancement in soil CO$_2$ flux due to re-wetting that was also measured in un-tilled fields. Increased soil respiration due to ploughing however are short-term, usually lasting less than 24 hours (Ellert and Janzen, 1999; Reicosky et al., 2005), so because the chambers needed to be removed before ploughing and were not re-installed until sites were re-visited a week later, the ploughing-induced increase in soil respiration was probably not fully captured. Also, root respiration, which at seeding accounts for 0% of soil CO$_2$ fluxes but can increase to around 45% of fluxes (Rochette et al., 1999), may also result in greater CO$_2$ fluxes during the growing season for the annual cropping systems. However, the increase in soil CO$_2$ fluxes from dry to growing season in annual crops was similar to the increase experienced in the other vegetation types (Table 3; $P = 0.39$). It is therefore likely that the low yields for the annual crops corresponded with poor root growth and low root respiration rates.

Cumulative soil CO$_2$ fluxes, (2.7 to 14.0 Mg CO$_2$-C ha$^{-1}$ yr$^{-1}$), were well within the range of other African studies (Table 1) and were not related to land class or field type, although the higher soil respiration rates from grazing land was inconsistent with a previous study that found similar rates between perennial tropical grasslands, croplands and tree plantations (Mapanda et al., 2010). However, because we did not differentiate between root and microbial respiration components, we cannot exclude that the continual vegetation cover in the grazing plots enhanced the root respiration over the year to a higher extent than in the annual crops and treed plots. It is important to keep in mind though, that these CO$_2$ emissions were the result of root respiration and microbial decomposition of organic matter, since plants were purposely excluded (except for some short grasses, see methods). In order to obtain the full GHG balance, both photosynthesis and above-ground vegetation respiration should be considered.

Methane was generally taken up by these upland soils, with rates varying through the year (Fig. 2b). During August 2013, the soils were sinks for CH$_4$, however as the
soils dried, the emission / uptake rates became more erratic until the long rains started again in late March 2014. The CH$_4$ flux at the soil-atmosphere interface is the balance between simultaneous production and consumption of CH$_4$ in different microsites in the soil profile (Le Mer and Roger, 2001). Thus, the low rates of atmospheric CH$_4$ uptake during the long rains may be caused by greater soil CH$_4$ production due to higher soil moisture and anaerobic conditions at depth (e.g. Butterbach-Bahl and Papen, 2002) overriding the existing methanotropic activity; alternatively, the higher water content may have limited the CH$_4$ diffusion from the atmosphere into the soil.

The CH$_4$ uptakes observed in these sites were consistent with previous studies in upland agricultural soils and indicate that soils of smallholder farms are sinks for atmospheric CH$_4$ (Le Mer and Roger, 2001). There were no differences between field types; however there were differences between cover types and land classes, as the grazing plots took up less CH$_4$ than treed plots and land class 1 (small lowland mixed degraded farms) took up less than land class 2 (moderate sized farms with signs of degradation on lower slopes, see Fig. 3). The difference between cover types is consistent with previous studies that found that forest soils were greater CH$_4$ sinks than agricultural soils (MacDonald et al., 1996; Priemé and Christensen, 1999) and high degrees of degradation in land class 1 was likely responsible for reduced CH$_4$ oxidation rates.

The N$_2$O flux rates remained below 20 µg m$^{-2}$ h$^{-1}$, with the exception of the onset of the rainy season in March 2014 (Fig. 3). According to Linn and Doran (1984) maximum aerobic activity occurs at approximately 60% water filled pore space (approximately 40% WHC for our study,), above which anaerobic processes such as denitrification can occur. The soils in the study area were typically drier than this threshold suggesting that N$_2$O fluxes were limited by a lack of anaerobic conditions and that the increase in soil water content was responsible for the increases in N$_2$O fluxes during March 2014. However, soil moisture was greater than 35% WHC during September/October 2013 and March 2014, but it was only in the latter
period large increases in N\textsubscript{2}O fluxes were observed. The high soil moisture levels in March coincided with an increase in inorganic N likely caused by drying and rewetting (Birch, 1960), which can also stimulate N\textsubscript{2}O fluxes (Butterbach-Bahl et al., 2004; Davidson, 1992; Ruser et al., 2006). Commencement of the rainy season was also when farmers fertilized, although application rates were low (1-25 kg N ha\textsuperscript{-1}) and did not have a detectable effect on soil inorganic N concentrations, or N\textsubscript{2}O emissions (Table 3).

The inability to discern between fertilized and unfertilized plots suggests that the differences in soil fertility and primary productivity were too low to have a noticeable effect on the availability of substrate for microbial activity and the associated GHG emissions. Alternatively, it is possible that the sensitivity of the monitoring approach was not enough to catch differences between fields. For instance, the fixed sampling frequency may have caused us to miss some short-lasting emission peaks following fertilization, resulting in an underestimation of cumulative emissions. However, sampling during an emission pulse would result in an overestimate of emissions due to an extrapolation bias. Previous studies have found that weekly sampling resulted in an average uncertainty of ± 30% of the “best estimate” (Barton et al., 2015; Parkin, 2008) and that this uncertainty changes with the coefficient of variation in measured emission rates. However, the fertilizer was applied at a low rate (< 25 kg N ha\textsuperscript{-1}). Application of synthetic fertilizers up to 70 kg N ha\textsuperscript{-1} at planting in the region had no detectable effect on annual N\textsubscript{2}O emissions (Hickman et al., 2015), while another nearby study found no effect of N fertilization on annual N\textsubscript{2}O emissions (Rosenstock et al., 2016), suggesting that our weekly sampling did not miss relevant N\textsubscript{2}O/GHG pulses.

The large increase in N\textsubscript{2}O emission rates after soil re-wetting (April 2014) in land class 3 (mid-slopes, grazing land; Fig. 2) was primarily due to two (of 10) plots, both located on the same farm that emitted around four to six times more N\textsubscript{2}O than the rest of the land class 3 plots and 15 to 23 times more N\textsubscript{2}O than the average for all other plots. The reason for the much higher fluxes after the re-wetting compared to
other sites is not yet understood as the topsoil C and N contents were 1.45 and 0.12% respectively, well within the range of values for that land class (Table 2). The presence of livestock on these plots could have resulted in additions of N through either urine or manure deposition, causing these pulses of N₂O. However, the presence of N₂O emission hotspots in general is quite common as denitrification activity can vary dramatically across small scales (Parkin, 1987).

Annual N₂O fluxes were low (<0.6 kg N ha⁻¹ y⁻¹) compared with other tropical and sub-tropical studies, such as a fertilized field in Brazil (Piva et al., 2014) or China (Chen et al., 2000), with fluxes up to 4.3 kg N₂O-N ha⁻¹ y⁻¹. On the other hand, our results were similar to previous studies in low input African agro-ecosystems (Table 1). The low cumulative fluxes were most likely a result of low substrate (inorganic N) availability, in addition to low soil moisture limiting denitrification through much of the year. Similar to the CO₂ fluxes, the cumulative N₂O fluxes did not differ by cover type, field type or by land class. However, it is possible that differences between the classes could be too small to detect given the low cumulative N₂O fluxes, high microsite variability typical of N₂O fluxes (Parkin, 1987) and weekly sampling (Barton et al., 2015; Parkin, 2008).

As shown in the supplementary material, maximum N₂O and CO₂ fluxes (i.e. flux potentials) from 5 cm soil cores differed by land class (Fig. S1), suggesting that there is the potential for differences in field emissions as well. However, these potentials in the field appeared to be limited by climatic conditions (i.e. lack of precipitation). Also, the maximum N₂O flux rates observed within the soil core study were correlated (Spearman Rank test) with the cumulative annual fluxes at the field sites (ρ = 0.399, P = 0.040), while CO₂ fluxes followed a similar trend (ρ = 0.349, P = 0.075). The CH₄ fluxes from the soil cores however, were not correlated with measured flux at the field sites (ρ = -0.145, P = 0.471; see Supplementary material). Therefore although incubations should not be used to predict baseline emissions in the field they may be used as a quick and relatively inexpensive method to identify locations with potential for high soil N₂O and CO₂ fluxes (i.e. emission hotspots).
There are additional sources of uncertainty associated with the sampling methods (chamber architecture, instrumentation sensitivity, etc). According to Levy et al. (2011), the uncertainty of the methods then would be about 20%, which when combined with the uncertainty around the weekly sampling would be about 50%. Although this may sound high, this is similar to the majority of other studies (e.g. see Helgason et al. (2005)) measuring GHG emissions and better than many of the studies so far in Africa (Table 1).

Our study showed no detectable differences in GHG fluxes between the different field types, contrary to our expectations. We had anticipated differences in GHG fluxes because of differences among field types in input use, food production, partial N and C balances and soil fertility as previously reported in the region (Tittonell et al., 2013); and these variables often affect soil GHG fluxes (Buchkina et al., 2012; Jäger et al., 2011). We further hypothesized that land class and cover type would also have significant effects on soil GHG fluxes since a significant amount of the variability in soil CO$_2$ fluxes in agro-ecosystems can be explained by NDVI (Sánchez et al., 2003) and cover type (Mapanda et al., 2010). We found however no clear effect of field or land type on soil GHG fluxes. Tittonell et al. (2013) reported important differences between field types only at each farm individually (Tittonell et al., 2013), which in our case might have resulted in greater within-type variation that masked differences between the field types. Moreover, the small differences in the degree of inputs and labor may have not been enough to provoke distinct GHG fluxes, because the whole region/study site is characterized by low nutrient availability. For example, manure inputs have previously been found to increase soil C content (Maillard and Angers, 2014), but the inputs in our study area were very low (4-6 wheelbarrow loads or approximately 95 kg C ha$^{-1}$) and probably not enough to cause field-level differences. Further, considering that a previous study found that N is being rapidly mined from soils in the Lake Victoria basin (Zhou et al., 2014), it is likely that soil C is also being lost across the landscape. As most of this area has been converted from natural forests, and forests generally have higher SOC stocks than croplands (Guo and Gifford, 2002), time since conversion could play a
larger part in determining the SOC content, which could mask any effects that
management activities have on soil respiration rates in these low input systems.

Crop yields from the annual cropping systems (100 – 750 kg ha\(^{-1}\) for one growing
season) were at the lower end of the range for rain-fed smallholder farms (600 to
3740 kg ha\(^{-1}\)) previously reported across SSA (Adamtey et al., 2016, Sanchez et al.,
2009, Tittonell et al., 2008). The farmers in our study complained of poor timing of
the rains that caused low yields. However poor timing of the rains tend to be
common in east Africa with estimations that 80% of growing seasons have critical
water shortages during flowering and grain filling, further resulting in low yields
(Barron et al., 2003). These studies therefore suggest that low yields are common
within this region. Increased nutrient inputs and improved management such as
rainwater harvesting (Lebel et al., 2015) are required to increase yields (Quiñones
et al., 1997), which may also result in increased GHG fluxes. However, previous
studies have found that increases in GHG fluxes tend to be lower than the
 corresponding increase in crop yields following addition of nutrients (Dick et al.,
2008), resulting in lower GHG intensities particularly at lower application rates
(Shcherbak et al., 2015). Another study in western Kenya found that fertilizer
applications up to 100 kg N ha\(^{-1}\) provoked no detectable increase in soil N\(_2\)O
emissions but did increase grain N contents (Hickman et al., 2014). The mean yield
scaled fluxes calculated for the eight maize and sorghum sub-samples was 12.9 g
N\(_2\)O-N kg\(^{-1}\) above-ground N uptake (range = 1.1 to 41.6), approximately 54% higher
than the 8.4 g N\(_2\)O-N kg\(^{-1}\) above-ground N uptake for plots fertilized at 180 – 190 kg
N ha\(^{-1}\) in a European meta-analysis (van Groenigen et al., 2010). These data further
suggest that improved agronomic performance through better soil, nutrient and
water management in East Africa has potential to potentially lower or at least
maintain yield-scaled fluxes while increasing food production from smallholder
farms in SSA.
CONCLUSION

This study indicates that soil GHG fluxes from low-input, rain-fed agriculture in western Kenya are lower than GHG fluxes from other tropical or sub-tropical agricultural systems with greater management intensities (e.g. China and Latin America). The input intensity for these farming systems is currently low, and so GHG fluxes were not related to management activities at the farm level. Given that this type of smallholder, low-input farming is very common across SSA, it is likely that our findings are valid at a much wider scale, although additional studies are required to confirm this hypothesis. Given that GHG emissions are often associated with soil moisture and that much of East Africa is drier than the climate at this study site, baseline emissions of GHG across East Africa may be extremely low. However, even though absolute emissions were low, high yield-scaled GHG fluxes in western Kenya could be reduced through interventions to increase yields (e.g. increased fertilizer, improved soil and water harvesting). As far as we know, this study provides the most comprehensive estimate of GHG emissions from smallholder African farms, in terms of number of sites, monitoring duration and temporal frequency of the measurements. However, more studies are needed to capture interannual variability as well as examining baseline emissions in other regions of the continent. These baseline studies are required to compare with proposed low emission development strategies to ensure that improvements in agricultural production continue to minimize GHG emissions, while also examining how intensification affects yields and soil GHG fluxes.

ACKNOWLEDGEMENTS

We thank the CGIAR Research Program on Climate Change, Agriculture, and Food Security (CCAFS) and its Standard Assessment of Mitigation Potential and Livelihoods in Smallholder Systems (SAMPLES) program for technical support and its financial support of scientists and laboratories working on this program. We acknowledge the CGIAR Fund Council, Australia (ACIAR), Irish Aid, European Union, International Fund for Agricultural Development (IFAD), Netherlands, New
Zealand, Switzerland, UK, USAID and Thailand for funding to the CGIAR Research Program on Climate Change, Agriculture and Food Security (CCAFS). The data used for this manuscript will be made available on the CCAFS website: ccafs.cgiar.org/.

This research was also funded by the German BMBF (Bundesministerium für Bildung und Forschung) through the IRADIATE project (grant number 01DG13012).

We would also like to thank David Musuya and Bernadette Nangira for their help collecting the field samples and Benard Goga for his lab work.
REFERENCES:

Adamtey, N., Musyoka, M. W., Zundel, C., Cobo, J. G., Karanja, E., Fiaboe, K. K. M.,
Muriuki, A., Mucheru-Muna, M., Vanlauwe, B., Berset, E., Messmer, M. M.,
Gättinger, A., Bhullar, G. S., Cadisch, G., Fliessbach, A., Mäder, P., Niggli, U., and
Foster, D.: Productivity, profitability and partial nutrient balance in maize-
based conventional and organic farming systems in Kenya, Agriculture,

Altieri, M. A. and Koohafkan, P.: Enduring farms: Climate change, smallholders and
traditional farming communities, 2008.

Arias-Navarro, C., Díaz-Pinés, E., Kiese, R., Rosenstock, T. S., Rufino, M. C., Stern, D.,
Neufeldt, H., Verchot, L. V., and Butterbach-Bahl, K.: Gas pooling: A sampling
technique to overcome spatial heterogeneity of soil carbon dioxide and nitrous

emissions following tillage and no-tillage of agroforestry residues in western

Baldi, G., Houspanossian, J., Murray, F., Rosales, A. A., Rueda, C. V., and Jobbágy, E.
G.: Cultivating the dry forests of South America: Diversity of land users and
imprints on ecosystem functioning, Journal of Arid Environments, doi:
http://dx.doi.org/10.1016/j.jaridenv.2014.05.027, 2015.

Barron, J., Rockström, J., Gichuki, F., and Hatibu, N.: Dry spell analysis and maize
yields for two semi-arid locations in east Africa, Agricultural and Forest

Barton, L., Wolf, B., Rowlings, D., Scheer, C., Kiese, R., Grace, P., Stefanova, K., and
Butterbach-Bahl, K.: Sampling frequency affects estimates of annual nitrous
oxide fluxes, Scientific Reports, 5, 15912, 2015.

Birch, H. F.: Nitrification in soils after different periods of dryness, Plant Soil, 12,

Bolte, W. T., Bushman, C. J., and Tidwell, P. W.: Spectrophotometric

Braun, A. R., Smaling, E. M. A., Muchugu, E. I., Shepherd, K. D., and Corbett, J. D.: Maintenance and improvement of soil productivity in the highlands of Ethiopia, Kenya, Madagascar and Uganda: an inventory of spatial and non-spatial survey and research data on natural resources and land productivity,

Brümmer, C., Brüggemann, N., Butterbach-Bahl, K., Falk, U., Szarzynski, J., Vielhauer,


Butterbach-Bahl, K. and Papen, H.: Four years continuous record of CH₄ exchange between the atmosphere and untreated and limed soil of a N-saturated spruce and beech forest ecosystem in Germany, Plant and Soil, 240, 77-90, 2002.


Jäger, N., Stange, C., Ludwig, B., and Flessa, H.: Emission rates of N\textsubscript{2}O and CO\textsubscript{2} from soils with different organic matter content from three long-term fertilization experiments—a laboratory study, Biol. Fertil. Soils, 47, 483-494, 2011.


Khan, S., Clough, T. J., Goh, K. M., and Sherlock, R. R.: Influence of soil pH on NO\textsubscript{x} and N\textsubscript{2}O emissions from bovine urine applied to soil columns, New Zealand Journal of Agricultural Research, 54, 285-301, 2011.


pool due to integrated crop-livestock in a subtropical Ferralsol, Agriculture, Ecosystems & Environment, 190, 87-93, 2014.


Rochette, P.: Towards a standard non-steady-state chamber methodology for measuring soil N₂O emissions, Animal Feed Science and Technology, 166-167, 141-146, 2011.


Tubio, F. N., Salvatore, M., Condor, R., Ferrara, A., Rossi, S., Federici, S., Jacobs, H., and Flammini, A.: Agriculture, forestry and other land use emissions by


Table 1. List of *in situ* empirical studies of greenhouse gas fluxes from agricultural systems in sub-Saharan Africa

<table>
<thead>
<tr>
<th>Reference</th>
<th>Location (&amp; crop type / treatment)</th>
<th>Sites</th>
<th>Time of measurement</th>
<th>Sampling frequency</th>
<th>Flux rates[^1]</th>
</tr>
</thead>
<tbody>
<tr>
<td>(Brümmer et al., 2008; Brümmer et al., 2009)</td>
<td>Burkina Faso (sorghum, cotton or peanut)</td>
<td>4</td>
<td>June – Sept 2005 April – Sept 2006</td>
<td>1 – 3X per week</td>
<td>N₂O: 0.19 – 0.67 kg ha⁻¹ y⁻¹ CO₂: 2.5 – 4.1 Mg ha⁻¹ y⁻¹ CH₄: -0.67 – -0.7 kg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>(Dick et al., 2008)</td>
<td>Mali (pearl millet with / without legume intercropping)</td>
<td>3</td>
<td>Jan 2004 – Feb 2005</td>
<td>Monthly</td>
<td>N₂O: 0.9 – 1.5 kg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>(Hickman et al., 2015)</td>
<td>Kenya (maize)</td>
<td>1</td>
<td>Mar 2011 – July 2011 Apr 2012 – Jan 2013 July 2005 – Sept 2006</td>
<td>Daily to weekly</td>
<td>N₂O: 0.1 – 0.3 kg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>(Koerber et al., 2009)</td>
<td>Uganda (vegetables)</td>
<td>24</td>
<td>Jan 2006 – Dec 2008</td>
<td>Monthly</td>
<td>CO₂: 30.3 – 38.5 Mg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>(Lompo et al., 2012)</td>
<td>Burkina Faso (urban gardens)</td>
<td>2</td>
<td>Mar 2008 – Mar 2009</td>
<td>2X per day (&quot;several&quot; times per cropping period) Weekly</td>
<td>CO₂: 22-36 Mg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>(Predotova et al., 2010)</td>
<td>Malawi (maize with agroforestry)</td>
<td>1</td>
<td>Oct 2001 – Apr 2002</td>
<td>Weekly</td>
<td>CO₂: 2.6 – 7.8 Mg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>(Makumba et al., 2007)</td>
<td>Niger (urban and peri-urban gardens)</td>
<td>3</td>
<td>Apr 2006 – Feb 2007</td>
<td>2X per day for 6 days (repeated 8 - 9X per year) 1 – 2X per month</td>
<td>N₂O: 48 – 92 kg ha⁻¹ y⁻¹ CO₂: 20 – 30 Mg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>(Predotova et al., 2010)</td>
<td>Tanzania (maize, with / without residue)</td>
<td>2</td>
<td>Mar 2007 – June 2010</td>
<td>Weekly</td>
<td>CO₂: 0.9 – 4.0 Mg ha⁻¹ y⁻¹</td>
</tr>
<tr>
<td>This study</td>
<td>Kenya (annual crops, grazing land, woodlots, fodder)</td>
<td>59</td>
<td>Aug 2013 – Aug 2014</td>
<td>Weekly</td>
<td>N₂O: -0.13 – 1.83 kg ha⁻¹ y⁻¹ CO₂: 2.8 – 15.0 Mg ha⁻¹ y⁻¹ CH₄: -5.99 – 2.44 kg ha⁻¹ y⁻¹</td>
</tr>
</tbody>
</table>
### Seasonal Flux Estimates

<table>
<thead>
<tr>
<th>Location</th>
<th>Methods</th>
<th>Dates</th>
<th>Frequency</th>
<th>N$_2$O</th>
<th>CO$_2$</th>
<th>CH$_4$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Kenya (maize with agroforestry, till/no till)</td>
<td>Baggs et al., 2006</td>
<td>Feb – June 2002 (Rainy Season)</td>
<td>Weekly</td>
<td>N$_2$O: 0.2 – 0.6 kg ha$^{-1}$</td>
<td>CO$_2$: 1.8 – 2.3 Mg ha$^{-1}$</td>
<td>CH$_4$: 0.1 – 0.3 kg ha$^{-1}$</td>
</tr>
<tr>
<td>Madagascar (maize with soybean)</td>
<td>Chapuis-Lardy et al., 2009</td>
<td>Nov 2006 – April 2007 (Rainy Season)</td>
<td>Weekly</td>
<td>N$_2$O: 0.3 kg ha$^{-1}$</td>
<td>CO$_2$: 1.8 – 2.3 Mg ha$^{-1}$</td>
<td>CH$_4$: 0.1 – 0.3 kg ha$^{-1}$</td>
</tr>
<tr>
<td>Zimbabwe (maize/improved fallow)</td>
<td>Chikowo et al., 2004</td>
<td>Dec 2000 – Feb 2001 (Rainy Season)</td>
<td>Weekly</td>
<td>N$_2$O: 0.1 – 0.3 kg ha$^{-1}$</td>
<td>CO$_2$: 0.7 – 1.6 Mg ha$^{-1}$</td>
<td>CH$_4$: 0.1 – 0.3 kg ha$^{-1}$</td>
</tr>
<tr>
<td>Zimbabwe (maize, with different fertilizer rates and types)</td>
<td>Mapanda et al., 2011$^2$</td>
<td>Nov 2006 – Jan 2007 (Rainy Season)</td>
<td>1X per 2 months</td>
<td>N$_2$O: 0.1 – 0.5 kg ha$^{-1}$</td>
<td>CO$_2$: 0.7 – 1.6 Mg ha$^{-1}$</td>
<td>CH$_4$: -2.6 – +5.8 kg ha$^{-1}$</td>
</tr>
<tr>
<td>Kenya (maize with regular and improved fallow)</td>
<td>Millar et al., 2004</td>
<td>Sep 1999 – Dec 1999 (Rainy Seasons)</td>
<td>1 – 2X per week</td>
<td>N$_2$O: 0.1 – 4.1 kg ha$^{-1}$</td>
<td>CO$_2$: 0.7 – 1.7 Mg ha$^{-1}$</td>
<td>CH$_4$: -2.6 – +5.8 kg ha$^{-1}$</td>
</tr>
</tbody>
</table>

### Mean Flux Rates from Short Duration Studies

<table>
<thead>
<tr>
<th>Location</th>
<th>Methods</th>
<th>Dates</th>
<th>Frequency</th>
<th>N$_2$O</th>
<th>CO$_2$</th>
<th>CH$_4$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Kenya (maize)</td>
<td>Kimetu et al., 2007</td>
<td>Mar 2000 – June 2000 (Rainy Season)</td>
<td>3X per month</td>
<td>N$_2$O: 1.3 – 12.3 µg m$^{-2}$ h$^{-1}$</td>
<td>CO$_2$: 42.5 – 46.8 mg m$^{-2}$ h$^{-1}$</td>
<td>CH$_4$: -9.4 – +6.9 µg m$^{-2}$ h$^{-1}$</td>
</tr>
<tr>
<td>Zimbabwe (grassland/grazing, tree plantations and maize)</td>
<td>Mapanda et al., 2010$^2$</td>
<td>Nov 2006 – Mar 2007 (Rainy Season)</td>
<td>2X per month to 1X per 2 months</td>
<td>N$_2$O: 1.0 – 4.7 µg m$^{-2}$ h$^{-1}$</td>
<td>CO$_2$: 22.5 – 46.8 mg m$^{-2}$ h$^{-1}$</td>
<td>CH$_4$: -9.4 – +6.9 µg m$^{-2}$ h$^{-1}$</td>
</tr>
<tr>
<td>Botswana (grazing)</td>
<td>Thomas, 2012</td>
<td>Feb, April, July, Nov 2010 (Both Rainy and Dry Season)</td>
<td>7X per day; 12 separate days only</td>
<td>CO$_2$: 1.1 – 42.1 mg m$^{-2}$ h$^{-1}$</td>
<td>CO$_2$: 1.1 – 42.1 mg m$^{-2}$ h$^{-1}$</td>
<td></td>
</tr>
</tbody>
</table>

1 Study includes fertilization up to 200 kg N ha$^{-1}$
2 Sampling is too infrequent for accurate estimates of cumulative fluxes (Barton et al., 2015)
Uses photoacoustic spectroscopy, which has recently had questions raised about its accuracy (Rosenstock et al., 2013a); also, these studies used exceptionally high N application rates, from 473 to approximately 4000 kg N ha\(^{-1}\) y\(^{-1}\).

Note: flux rates are given as the range of values from the various replicates used in the studies (i.e. the spatial variability and where available [Mapanda et al. 2011 and Thomas 2012], the temporal variability as well), and are reported as N- \(\text{N}_2\text{O}\), C- \(\text{CO}_2\) and C- \(\text{CH}_4\); Please also note units: where possible, annual cumulative fluxes are presented, however in cases with insufficient data to estimate cumulative annual fluxes, we present either mean flux rates, or the cumulative for the given period.
Table 2: Soil properties (± 1 SEM) for 0 to 20 cm depth, sampled immediately before initiation of gas sampling for the different land classes

<table>
<thead>
<tr>
<th>Land class</th>
<th>C$^2$ content (%)</th>
<th>N content (%)</th>
<th>CN ratio</th>
<th>pH</th>
<th>Bulk Density (g cm$^{-3}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1) Lowland small (&lt;2 ha) mixed farms with degradation$^1$ signs (n = 7)</td>
<td>1.38 ± 0.13</td>
<td>0.10 ± 0.01</td>
<td>13.18 ± 0.51</td>
<td>6.61 ± 0.09</td>
<td>0.86 ± 0.03</td>
</tr>
<tr>
<td>(2) Lower slopes$^3$, moderate (2-5 ha) sized mixed farms with degradation signs (n = 8)</td>
<td>1.18 ± 0.14</td>
<td>0.10 ± 0.01</td>
<td>11.60 ± 0.58</td>
<td>6.58 ± 0.16</td>
<td>1.14 ± 0.08</td>
</tr>
<tr>
<td>(3) Mid-slopes, moderate sized grazing land (n = 10)</td>
<td>2.27 ± 0.37</td>
<td>0.18 ± 0.03</td>
<td>12.16 ± 0.42</td>
<td>6.02 ± 0.21</td>
<td>0.98 ± 0.07</td>
</tr>
<tr>
<td>(4) Upper slopes/highland plateau, mixed farms (n = 22)</td>
<td>2.67 ± 0.17</td>
<td>0.21 ± 0.02</td>
<td>12.69 ± 0.52</td>
<td>5.46 ± 0.24</td>
<td>0.80 ± 0.06</td>
</tr>
<tr>
<td>(5) Mid-slopes, isolated moderate sized farms (n = 12)</td>
<td>2.83 ± 0.36</td>
<td>0.24 ± 0.02</td>
<td>13.02 ± 0.81</td>
<td>5.84 ± 0.20</td>
<td>0.71 ± 0.04</td>
</tr>
</tbody>
</table>

$^1$ degradation signs were bare soil and evidence of erosion visible on MODIS images.
$^2$ due to lack of carbonates, total C equals organic C
$^3$ Sloped areas went from the lowlands (approx. 1200 masl) up to the highlands (approx. 1800 masl) ranging from 10 – 30%.
Table 3: Comparison of mean (± 1 SEM) cumulative soil CO₂-C, CH₄-C and N₂O-N fluxes for four weeks during the dry season (February 2014) and rainy season (April 2014) for differently managed sites in western Kenya.

<table>
<thead>
<tr>
<th>GHG</th>
<th>Dry Season</th>
<th>Wet Season</th>
<th>P values</th>
<th>Management¹</th>
<th>Interaction</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Annual Crop (n = 42)</td>
<td>Other (n = 17)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>CO₂-C (g m⁻²)</td>
<td>19.4 ± 2.8</td>
<td>20.0 ± 3.8</td>
<td>76.6 ± 5.0</td>
<td>62.7 ± 5.7</td>
<td>&lt; 0.0001</td>
</tr>
<tr>
<td>CH₄-C (mg m⁻²)</td>
<td>-7.4 ± 4.4</td>
<td>2.2 ± 6.7</td>
<td>-3.7 ± 3.6</td>
<td>-15.0 ± 3.5</td>
<td>0.610</td>
</tr>
<tr>
<td></td>
<td>Fertilized (n = 16)</td>
<td>Not Fertilized (n = 43)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N₂O-N (mg m⁻²)</td>
<td>0.52 ± 0.23</td>
<td>1.44 ± 0.40</td>
<td>9.87 ± 4.23</td>
<td>5.35 ± 1.14</td>
<td>&lt; 0.0001</td>
</tr>
</tbody>
</table>

¹ The term management refers to plowing versus no plowing for the CO₂ and CH₄ fluxes and to fertilized versus no fertilizer application for the N₂O fluxes
Figures:

Fig. 1. Map of study area showing the sampling location by the different vegetation cover types

Fig. 2. CO\textsubscript{2} (mg C\textperiodcentered CO\textsubscript{2} m\textsuperscript{-2} h\textsuperscript{-1}), CH\textsubscript{4} (\mu g C\textperiodcentered CH\textsubscript{4} m\textsuperscript{-2} h\textsuperscript{-1}), and N\textsubscript{2}O (\mu g N\textsubscript{2}O-N m\textsuperscript{-2} h\textsuperscript{-1}) fluxes over 1 year (August 2013 through July 2014), as well as precipitation (mm), soil water content (SWC) at 5 cm depth (m\textsuperscript{3} m\textsuperscript{-3}) and inorganic N (IN = NO\textsubscript{3} + NH\textsubscript{4}) soil concentrations for 59 different fields in western Kenya by land class as well as soil temperature (°C) by topography. Note: Vertical dotted lines correspond to planting and vertical dashed lines correspond to harvesting of annual crops. (Land class 1 = degraded lowland farms; class 2 = degraded farms, lower slopes; class 3 = mid slopes, grazing; class 4 = upper slopes/plateau, mixed farms; and class 5 = mid slopes moderate sized farms). SEM for the various gases ranged from 2.1 to 57.4 for CO\textsubscript{2} flux, 0.0 to 106.6 for CH\textsubscript{4} flux and 0.2 to 45.6 for N\textsubscript{2}O flux with the highest variability occurring between 20 and 27 March 2014 for CO\textsubscript{2} and N\textsubscript{2}O while the highest variability in CH\textsubscript{4} flux occurred during the week of 4 August 2013. For all gases the greatest variability occurred in landclass 3 (n = 10).

Fig. 3. Box and whisker plots of cumulative annual fluxes of CO\textsubscript{2} (Mg CO\textsubscript{2}-C ha\textsuperscript{-1} year\textsuperscript{-1}), CH\textsubscript{4} (kg CH\textsubscript{4}-C ha\textsuperscript{-1} year\textsuperscript{-1}) and N\textsubscript{2}O (kg N\textsubscript{2}O-N ha\textsuperscript{-1} year\textsuperscript{-1}) from 59 different fields in western Kenya split by land class, field type or vegetation type. (Land class 1 = degraded lowland farms; class 2 = degraded farms, lower slopes; class 3 = mid slopes, grazing; class 4 = upper slopes/plateau, mixed farms; and class 5 = mid slopes moderate sized farms); Field type is based on Rufino et al (2015), with Field Type 1 being the most highly managed and Type 3 being the least managed plots. Different lower case letters indicate significant differences between treatments, while a lack of letters indicate no difference between any of the treatments.
Fig. 2
Fig. 3