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Potential production and environmental effects of switchgrass and traditional crops under current and greenhouse-altered climate in the central United States: a simulation study

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Abstract

If, as many climate change analysts* speculate, industrial and other emissions of CO₂ can be offset by substitution of biofuels, large areas of land, including agricultural land, may be converted to the production of biomass feedstocks. This paper explores the feasibility for the Missouri–Iowa–Nebraska–Kansas (MINK) region of the US of converting some agricultural land to the production of switchgrass (*Panicum virgatum* L.), a perennial warm season grass, as a biomass energy crop. The erosion productivity impact calculator (EPIC) crop growth model simulated production of corn (*Zea mays* L.), sorghum (*Sorghum bicolor* (L.) Moench), soybean (*Glycine max* L.), winter wheat (*Triticum aestivum* L.) and switchgrass at 302 sites within the MINK region. The analysis is done for both current climatic conditions and a regional climate model-based scenario of possible climate change. Daily climate records from 1983 to 1993 served as baseline and the NCAR-RegCM2 model (RegCM hereafter) nested within the CSIRO general circulation model (GCM) provided the climate change scenario. Crop production was simulated at two atmospheric CO₂ concentrations ([CO₂]) at 365 and 560 ppm to consider the CO₂-fertilization effect. Simulated yields of the perennial switchgrass increased at all sites with a mean yield increase of 5.0 Mg ha⁻¹ under the RegCM climate change scenario. Switchgrass yields benefited from temperature increases of 3.0–8.0°C, which extended the growing season and reduced the incidence of cold stress. Conversely, the higher temperatures under the RegCM scenario decreased yields of corn, soybean, sorghum and winter wheat due to increased heat stress and a speeding of crop maturity. With no CO₂-fertilization effect, EPIC simulated maximum decreases from baseline of 1.5 Mg ha⁻¹ for corn, 1.0 Mg ha⁻¹ for sorghum, 0.8 Mg ha⁻¹ for soybean and 0.5 Mg ha⁻¹ for winter wheat. Simulated yields increased for all crops under the RegCM scenario with CO₂ set to 560 ppm. Yields increased above baseline for 34% of the soybean and 37% of the winter wheat farms under RegCM/[CO₂] = 560 ppm scenario. Water use increased for all crops under the higher temperatures of the CSIRO scenario. Precipitation increases resulted in greater runoff from the traditional crops but not from switchgrass due to the crop's increased growth and longer growing season. Simulated soil erosion rates under switchgrass and wheat cultivation

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were less severe than under corn management. However, simulated erosion under switchgrass was considerable in eastern Iowa during the period of crop establishment because of strong winds at that time. ©2000 Elsevier Science B.V. All rights reserved.

Keywords: Biomass energy; Climate change; Central US; Switchgrass; EPIC

1. Introduction

Biomass-based energy systems have been proposed as a means of reducing greenhouse gas emissions (e.g. Hall et al., 1990; Graham et al., 1992; Hall et al., 1993; Wright and Hohenstein, 1994) when substituted for traditional fossil fuels. Such systems are carbon neutral since, when combusted, biomass returns to the atmosphere CO₂ recently captured by the feedstock. There are, of course, CO₂ costs embedded in the inputs (e.g., fertilizer, diesel fuel) required in biomass production but there are also savings in the sequestration of carbon in the soil. Biomass energy can be produced from short rotation woody crops, herbaceous energy crops, halophytes, some annual crops and oilseeds.

The economic and environmental consequences of large-scale biomass production are viewed as positive or manageable by a number of analysts. Sampson et al. (1993) conclude that dedicated energy crops could reduce carbon emissions to the atmosphere from 0.2 to 1.0 Gt year⁻¹. Hall (1997) projects a biomass contribution of 17–30% of global energy requirement up to 2050. Graham et al. (1996) applied an analytical assessment framework to two Tennessee regions and concluded that dedicated energy crops in place of conventional crops would significantly reduce erosion and chemical runoff. While biomass crops offer potential benefits, the conversion of large areas of productive agricultural lands to this use has possible downsides as well. The most serious of these is the competition for land that large-scale biomass culture will create. Rosenberg and Scott (1994) surveyed data that bear on this question. Some examples: an area of 465 million hectares — almost as large as the trans-Mississippi west — would have to be re- or afforested to offset the annual increase in atmospheric CO₂ emissions (Sedjo and Solomon, 1989); an area of 90.7 million hectares of new forests or 48% of all US croplands would be required to offset 50% of the 1.27 Gt year⁻¹ annual US emissions of carbon (Moulton and Richards, 1990). Giampietro et al. (1997) believe that large-scale

biomass production will greatly stress available supplies of land and water everywhere and reverse economic progress in developing countries.

It is evident, then, that biomass production on a scale that will permit a significant substitution of fossil fuels cannot be accomplished on marginal lands alone but will require large areas of prime agricultural land and the substitution of biomass crops for crops currently grown in some regions. Whether or not a particular biomass crop, switchgrass, can substitute for traditional crops in an important agricultural region, the Missouri–Iowa–Nebraska–Kansas (MINK) region of the central US, and whether or not climate change will favor such a substitution is evaluated here. Results are presented on a simulation study comparing the yield potential of switchgrass, its water use and effects on soil erosion with those of the region's dominant agricultural crops — corn, sorghum, soybeans and winter wheat. The comparison is made for current climatic conditions (baseline) and for a scenario of greenhouse-forced climate change with and without a CO₂-fertilization effect.

2. Methods and materials

The erosion productivity impact calculator (EPIC) (Williams, 1995), the crop growth simulator used in this study, is described in this section as are the input datasets that it requires. Brown et al. (1998) provides a more detailed description of EPIC and representative farms used in this study.

2.1. EPIC model description

EPIC (ver. 5320) is a process-based model that simulates soil erosion and crop productivity. The model runs on a daily time step at the scale of a single field. EPIC requires soil properties (e.g., bulk density, wilting point, water holding capacity) and crop management (e.g., cultivars, fertilization, tillage) as inputs.

Daily values of precipitation, minimum/maximum air temperature, solar radiation, windspeed and relative humidity drive EPIC. The model simulates bio-physical processes encompassing an agricultural system such as photosynthesis, evapotranspiration, runoff, water erosion and wind erosion.

EPIC models a wide range of crops using parameters specific to each crop. The crop growth model uses radiation use efficiency to calculate the potential daily photosynthetic production of biomass. The daily potential is decreased by stresses caused by shortages of radiation, water and nutrients, by temperature extremes and by inadequate soil aeration in proportion to the severity of the most severe stress for that day. Atmospheric CO₂ concentration ([CO₂]) affects stomatal conductance. Stomatal conductance affects radiation use efficiency and water use efficiency and, as a result, photosynthesis and evapotranspiration. Stockle et al. (1992a, b) adapted EPIC to simulate the CO₂-fertilization effect on plant photosynthesis and transpiration. Crop maturity is determined by the accumulation of heat units during the growing season. Crop yields are estimated by multiplying above-ground biomass at maturity by a harvest index (proportion of total biomass in harvested organs) for the particular crop. For additional details on EPIC as it is used in climate change impact simulations, see Brown and Rosenberg (1997, 1999).

2.2. The study area and climate scenarios

A high resolution climate change scenario for the US (Mearns et al., 1996) provided the climate change scenario. Giorgi et al. (1998) employed the NCAR-RegCM2 regional climate model (RegCM) nested within a GCM and run at a 50 km grid point spacing. The RegCM receives lateral and initial boundary conditions from a GCM. The CSIRO model (Watterson et al., 1995) was chosen in this application because of its ability, relative to other available GCMs, to reproduce historic climate over the central United States. The RegCM network encompasses a network of 302 50 × 50 km grid boxes (Fig. 1). This fine-scale grid allows simulation of switchgrass and the traditional crops in sufficient detail to explore variability due to geography, climate, soils and farm

management. A 10-year daily climate record was created for both the control and the climate change scenario. The latter is based on the assumption of an effective doubling of the preindustrial atmospheric CO₂ concentration (2 × CO₂ climate). Differences between the control and 2 × CO₂ climates were then applied to the baseline of each grid box to create its climate change scenario.

2.3. EPIC representative farms

A set of ‘representative farms’ was developed to describe agricultural practices and environmental conditions in each of the 302 grid boxes. The representative farm design stems from an earlier study of the MINK region (Easterling et al., 1992; Rosenberg, 1993) and subsequent follow-up analyses (Easterling et al., 1998). A subset of eight farms identified in Fig. 1 is used to provide details on the mechanisms that determine crop response under current and changed climate. Table 1 presents the location of these farms, their soils and some climatic data.

Daily climate records from the National Weather Service Cooperative Climate Network provided baseline precipitation and minimum/maximum temperature data from 1983 to 1993. Daily solar radiation, relative humidity and windspeed records were assembled with the EPIC weather generator (WXGEN) of

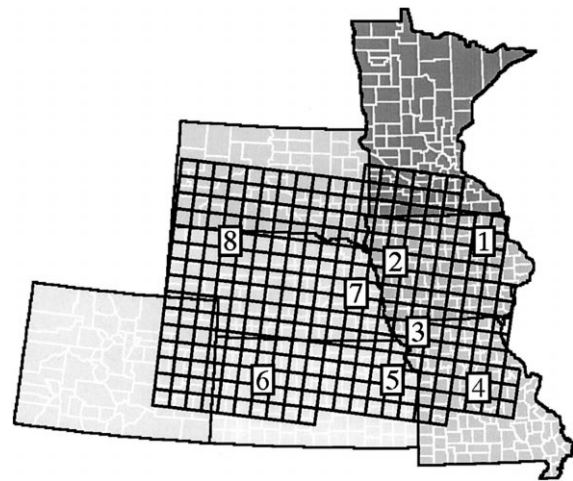


Fig. 1. The RegCM grid network and its orientation with respect to the MINK region. The numbered boxes identify the subset of representative farms discussed and described in Section 3.

Table 1
Location, climate and soil series of the subset of representative farms identified in Fig. 1

Farm #	State	Climate station	Soil identification	Soil		North latitude (°)	West longitude (°)	Annual precipitation (mm) (total)	Annual maximum temperature (°C)	Annual minimum temperature (°C)
				Series name	Soil taxonomy					
1	Iowa	Fayette	Clyde	Typic Haplaquolls	Mollic Gleysol	42.8	92.2	883	13.8	2.5
2	Iowa	Castana	Napier	Cumelic Hapludolls	Pachic Phaeozom	42.3	95.5	811	15.5	3.6
3	Missouri	Conception	Colo	Cumelic Endoaquolls	Mollic Gleysol	40.3	94.6	882	17.2	4.7
4	Missouri	Columbia	Hartwell	Typic Argialbolls	Mollic Planosol	38.6	92.5	1064	19.2	7.1
5	Kansas	Oskaloosa	Martin	Aquertic Argiudolls	Vertic Phaeozom	40.0	95.5	935	18.8	6.8
6	Kansas	Quinter	Ulysses	Aridic Haplustolls	Calcic Chernozem	38.9	100.3	569	18.6	4.3
7	Nebraska	Clarkston	Moody	Udic Haplustolls	Haplic Kastanozem	41.5	96.9	765	16.3	3.7
8	Nebraska	Merriman	Valentine	Typic Ustipsamments	Eutric Arenosol	42.7	101.8	465	16.4	0.9

Richardson and Nicks (1990) using monthly averages from the nearest National Weather Service (NWS) first order station. Soil properties were derived from the STATSGO (USDA, 1992) and EPIC (Williams et al., 1990) databases. Tillage practices employed with corn, sorghum, soybean and winter wheat were derived from the agricultural sector model (ASM) crop budgets for 1996 (McCarl et al., 1998). A series of switchgrass field experiments (Anderson, 1989; Anderson and Shapiro, 1990; Mitchell et al., 1994) provided a single switchgrass management budget for the entire region. Only one mid-summer harvest was simulated for switchgrass, although two harvests may be more appropriate in southern Missouri and Kansas.

3. Results

3.1. Validation of EPIC

3.1.1. Traditional crop simulations

EPIC simulations of crop yield have been extensively validated against actual crop yields. Kiniry et al. (1990) found agreement between simulated and actual yields of corn, wheat, rice, soybean, barley and sunflower under a variety of management systems and climatic conditions. Rosenberg et al. (1992) compared EPIC-simulated yields with historic county yields, yields from agronomic experiments and yield estimates by local agricultural experts and concluded that EPIC is suitable for simulating agricultural

production in the MINK region. Easterling et al. (1998) reported that EPIC simulations of representative farms with soils and climate data on a 0.5° grid scale explained 65% of the annual variation in eastern Iowa corn and 54% in western Kansas wheat yields.

The validation exercises described above employed earlier versions of EPIC, and the crop budgets, soils and climates of the representative farms differed slightly from those used here. Given the previous work, the goal of this validation was modest: to determine if EPICs algorithms and the representative farm descriptions generally agree with regional historic yields and accurately predict regional yield patterns.

Results of the yield simulations for dryland corn, sorghum, soybean and winter wheat on a subset of 13 representative farms (8 of which are identified in Fig. 1) are compared in Fig. 2 to USDA-NASS county mean crop yields (yield/harvested acre) for the period 1983–1993 (USDA-NASS, 1983–1993). EPIC yields account for 78% of the variability in NASS yields for all crops considered. Wheat and soybean yields are estimated reasonably well by EPIC. Corn and sorghum yields are more frequently overestimated than underestimated. Overestimation is to be expected since EPIC assumes crop management at uniformly high levels of technology and harvest efficiency. No attempt has been made in these simulations to account for the effects of episodic climatic events such as severe frosts, hail and windstorms or for pest damage — all of which lower real agricultural yields.

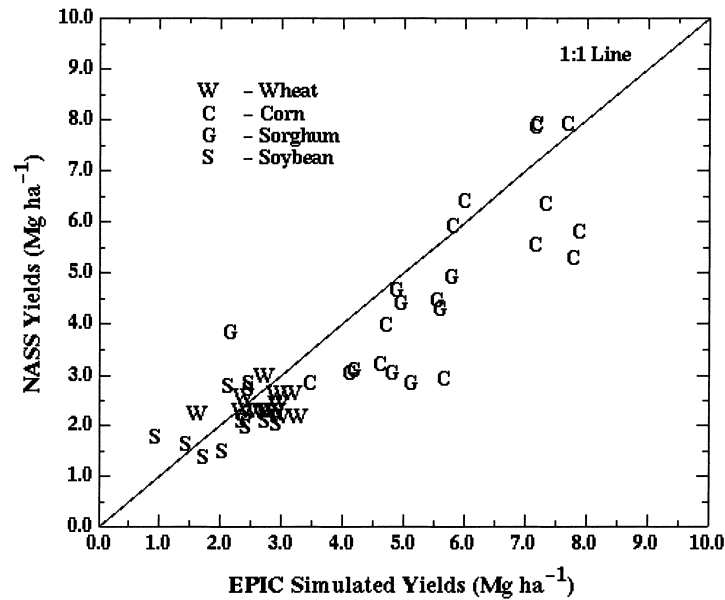


Fig. 2. EPIC-simulated yields compared with historic yields (from USDA-NASS county crops database) for grain crops on 13 representative farms in the MINK region.

3.1.2. Switchgrass simulations

Switchgrass has not been cultivated extensively in the MINK region. Thus, no long-term records of historic yields exist against which to validate the EPIC switchgrass results. Recent switchgrass yield trials conducted in the MINK region (Vogel, 1996a) present an opportunity for validation. Rosenberg et al. (1992) argue that EPIC, in any event, is best compared with experimental yields, since both utilize optimal management. Kiniry et al. (1996) compared switchgrass yields simulated by ALMANAC — a model similar to EPIC — with yields from agronomic trials in Texas and found that simulated yields explained 79% of the variability in measured yields.

Switchgrass simulations were compared against data from factorial experiments which tested the effects of nitrogen fertilization rates and harvest schedule on switchgrass yield between 1993 and 1995 at Ames, IA (42.00 N lat., 93.50 W long., \approx 725 mm annual precipitation, soil series Webster) and Mead, NE (41.30 N lat., 96.60 W long., \approx 675 mm annual precipitation, soil series Sharpsburg) (Vogel, 1996a). Switchgrass was planted in 1993 to establish the crop. Fertilization rates were varied from 0 to 300 kg ha⁻¹ of N in 60 kg ha⁻¹ increments, and the crop was

harvested at weekly intervals from late June into September. A final harvest was made at the end of the growing season in November. EPIC datasets were constructed to approximate the environmental and agronomic conditions of the experiment and then used to model switchgrass production over a 30-year period, repeating the climate records of 1993–1995 and using simulated mean yields for comparison.

Table 2 compares the mean experimental and EPIC simulated yields for switchgrass at both sites. Mean experimental yields were calculated by averaging all data for 1994–1995. Experimental yields existed for both summer and fall cutting, allowing us to compare EPIC's ability to match annual yield and to model switchgrass growth and development through the growing season. On average, EPIC underestimated the summer harvest and overestimated the fall harvest. Agreement was better for total yields. Fig. 3 compares the response of the observed and simulated yields to differing levels of N fertilization rates. Agreement between observed and simulated yields is excellent for the Ames, IA site up to the N application rate of 180 kg ha⁻¹. At greater rates the simulated yields continue to rise linearly while the experimental yields show a diminishing response. EPIC simulated yields

Table 2

Mean yields, standard deviations and root mean square error (RMSE) for switchgrass experimental yields and EPIC simulated yields by location and harvest

Site		Measured yields (Mg ha^{-1})		Simulated yields (Mg ha^{-1})		RMSE ^b
		Mean	Stdev. ^a	Mean	Stdev. ^a	
Ames	Summer cutting	9.9	2.4	9.1	1.8	2.6
Mead		9.3	1.9	7.5	2.2	2.9
Ames	Fall cutting	1.9	0.6	3.7	0.9	2.0
Mead		1.5	0.4	2.4	0.6	1.0
Ames	Annual yield ^c	11.7	2.3	12.8	2.5	3.0
Mead		10.8	2.6	9.8	2.6	2.7

^a Standard deviation.

^b Root mean square error.

^c Annual yield is calculated by summing yields from the summer and fall cutting.

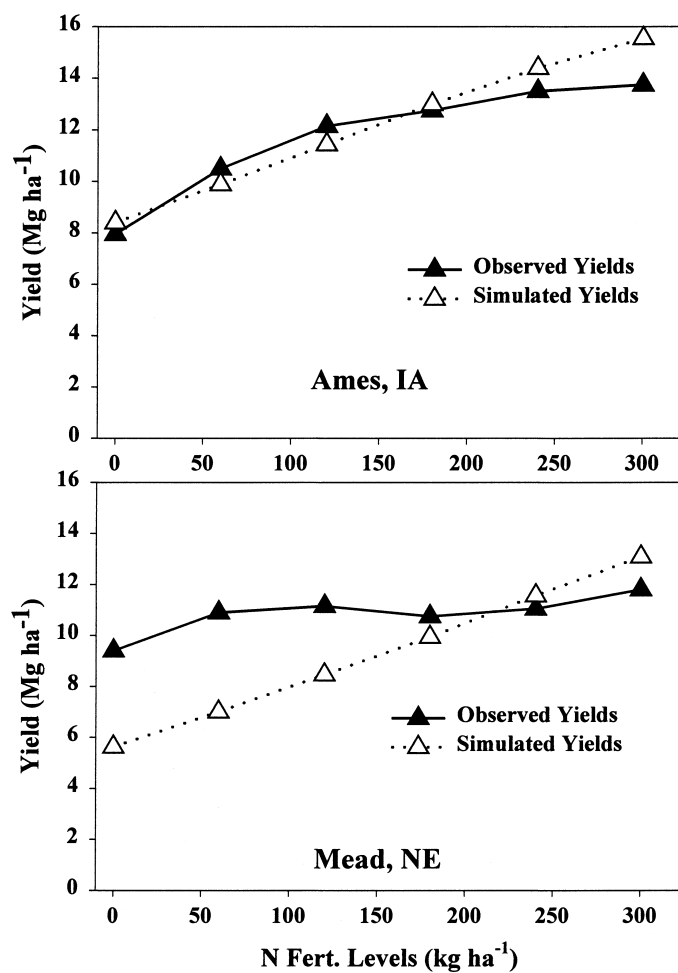


Fig. 3. Comparison of experimental and simulated switchgrass yield response to differing nitrogen fertilization rates at Ames, IA and Mead, NE.

Table 3

Mean seasonal baseline temperatures, seasonal precipitation totals and deviations from baseline under the RegCM climate change scenario for eight representative farms within the MINK region

Farm # ^a	Winter (D,J,F)			Spring (M,A,M)			Summer (J,J,A)			Fall (S,O,N)		
	Baseline	RegCM ^b	Change	Baseline	RegCM	Change	Baseline	RegCM	Change	Baseline	Regcm	Change
<i>Mean temperature (°C)</i>												
1	-6.7	0.1	6.8	8.7	16.2	7.5	21.5	26.2	4.6	9.0	13.8	4.8
2	-4.5	1.5	6.0	10.0	16.3	6.3	22.6	27.2	4.7	10.1	14.9	4.8
3	-2.4	2.7	5.1	11.2	16.8	5.6	23.5	28.4	4.9	11.6	16.2	4.6
4	1.0	5.7	4.7	13.3	18.8	5.5	24.4	29.1	4.7	13.9	18.4	4.5
5	0.0	4.2	4.3	12.9	18.0	5.1	24.8	29.4	4.6	13.5	18.1	4.6
6	-0.9	2.9	3.7	11.0	15.3	4.3	23.8	28.1	4.3	11.8	16.2	4.4
7	-3.5	1.5	5.0	10.2	15.8	5.6	23.2	27.6	4.5	10.3	15.1	4.8
8	-3.7	0.9	4.6	8.1	13.2	5.1	21.4	25.7	4.2	8.8	13.0	4.2
<i>Total precipitation (mm)</i>												
1	84	113	29	254	266	12	298	343	44	247	361	115
2	56	67	11	271	311	40	299	316	18	186	273	87
3	80	91	11	254	360	107	321	322	1	227	275	48
4	168	196	28	305	359	54	286	314	28	304	400	96
5	93	106	13	296	376	81	289	302	13	258	321	63
6	48	49	2	208	330	122	217	269	52	96	115	19
7	56	63	7	245	291	47	278	325	47	186	231	45
8	31	38	7	139	218	79	215	246	30	80	93	13

^a Farm locations are identified in Fig. 1 and summarized in Table 1.

^b NCAR RegCM2 model simulation of 2 × CO₂ climate change scenario.

from the Mead, NE, site also show a linear increase to N fertilization; however, agreement with observed yields is poor, with the observed yields virtually unresponsive to fertilizer rates.

EPIC reasonably matches mean annual switchgrass yields from agronomic trials at two sites in the MINK region. The validation results are comparable with previous EPIC model validations (Rosenberg et al., 1992; Kiniry et al., 1996; Brown and Rosenberg, 1999). However, the results also show disagreement between observed and simulated data with regard to the effect of yield to N application rates. While these flaws are not severe enough to dismiss the modeling results, they do introduce a level of uncertainty and highlight areas where the model needs to be improved to correctly mimic switchgrass growth and development.

3.2. Climate scenarios

The climate scenarios used in this study are reported in full by Mearns et al. (1999a, 1999b). Seasonal means of temperature and precipitation for the baseline climate and deviations from baseline under

the RegCM scenario are presented in Table 3. Mean winter temperatures range from 1.6°C in the south to -8.8°C in the most northerly regions. Spring temperatures range from 13.8 to 5.8°C; summer from 26.1 to 18.5°C; fall from 14.4 to 6.6°C. For fall and summer, the RegCM scenario increases temperatures from 4 to 5°C for the entire study region (summer, Fig. 4(b)). For winter and spring, the RegCM scenario raises temperature from 6 to 8°C in Iowa and northeastern Nebraska (farms 1–2) from southwest to northeast (winter, Fig. 4(a)). Giorgi et al. (1998) attribute these dramatic increases in the region's winter and spring temperatures to the loss of permanent snow cover, resulting in a snow albedo feedback.

The study region experiences increased precipitation under the RegCM scenario. Increases in winter (Fig. 4(c)) are modest ranging from 0 to 50 mm, but as a percentage change comparable to the other seasons. Spring increases range from 0 to 200 mm with the largest increases concentrated in north central Missouri, south central Iowa and central Kansas. Changes in precipitation are more randomly distributed in summer than other seasons. Summer (Fig. 4(d)) increases range from 0 to 200 mm, however,

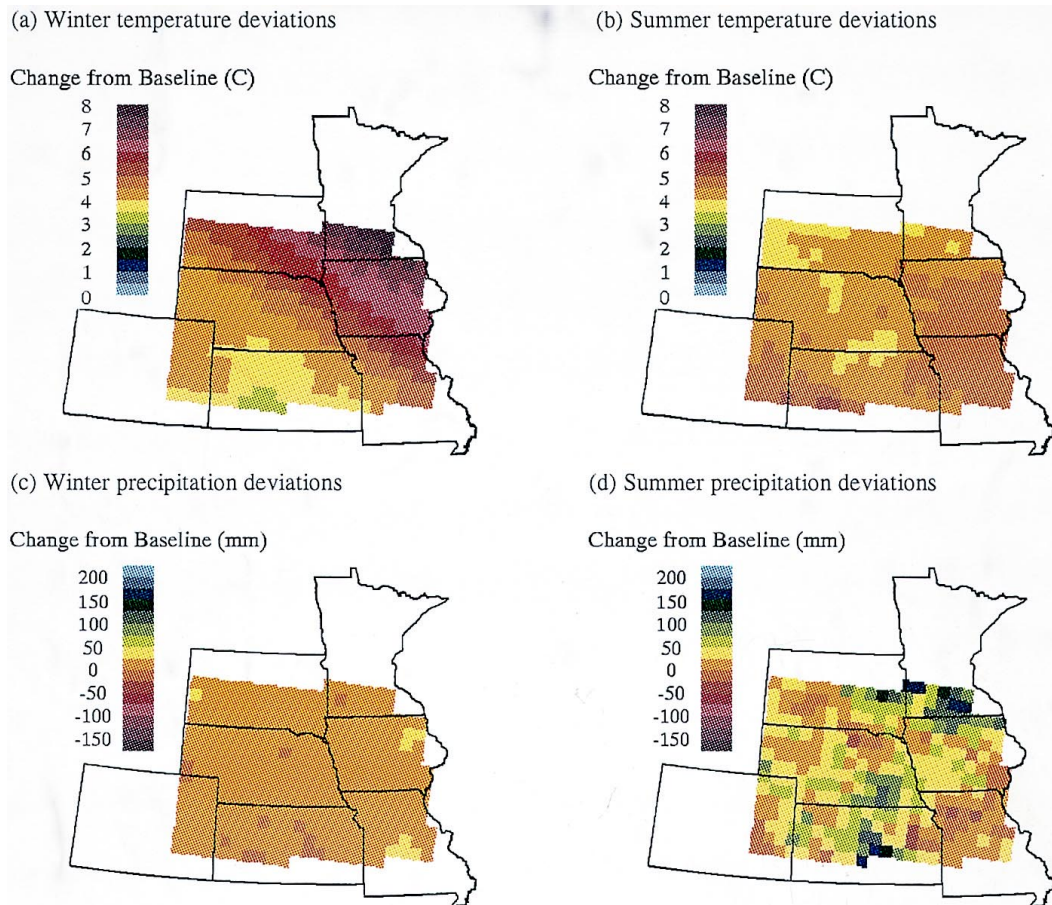


Fig. 4. Temperature and precipitation deviations from baseline under the RegCM climate change scenario for winter (December–January–February) and summer (June–July–August).

precipitation decreases in 12% of the grids. Deviations ranging up to 200 mm occur in fall, increasing in amount from west to east.

3.3. Baseline and RegCM simulated yields

3.3.1. Traditional crops

Table 4 summarizes yield results from the EPIC simulations and Table 5 presents EPIC recorded stress day data simulations for both the traditional crops and switchgrass under baseline and the RegCM climate change scenario. Table 6 presents simulated yields and stress data for eight representative farms (identified in Fig. 1 and described in Table 1) under corn, winter wheat and switchgrass cultivation.

Under baseline climate, simulated crop yields are highest in Iowa and Missouri with yields declining in western Nebraska and Kansas. The high corn and soybean yields in Iowa and Missouri coincide with minimal water and nutrient stress. The highest sorghum yields occurred in eastern Nebraska. Interestingly, sorghum yields in Iowa are reduced from the maximum due to cold temperature stress. Water stress decreases corn and soybean yields in central Nebraska and Kansas. Sorghum and winter wheat, which are relatively insensitive to precipitation shortages, maintain yields at levels approaching those on the eastern farms. Winter wheat yields showed reasonable potential in the western zone, exceeding 1.0 Mg ha^{-1} in some cases.

Table 4
Simulated yield means, standard deviations and range for the 302 representative farms sorted by crop and climate scenario

Crop	Baseline (Mg ha ⁻¹)			RegCM ^a (Mg ha ⁻¹)			RegCM + CO ₂ ^b (Mg ha ⁻¹)		
	Mean	Stdev.	Range	Mean	Stdev.	Range	Mean	Stdev.	Range
Corn	5.2	2.0	0.6–7.9	5.3	1.3	0.4–7.2	5.7	1.2	0.9–7.5
Sorghum	4.0	1.3	0.7–6.1	3.9	1.1	0.8–5.5	4.2	1.1	0.9–5.8
Soybean	1.8	0.7	0.4–3.0	1.7	0.5	0.4–2.4	2.0	0.5	0.6–2.9
Winter wheat	2.5	0.5	1.1–3.6	2.5	0.4	1.6–3.8	2.8	0.5	1.6–4.1
Switchgrass	6.8	1.9	2.2–12.0	11.8	3.2	3.8–17.5	13.4	3.6	4.0–19.7

^a NCAR RegCM2 model simulation of 2 × CO₂ climate change scenario with [CO₂] = 365 ppm.

^b NCAR RegCM2 model simulation of 2 × CO₂ climate change scenario with [CO₂] = 560 ppm.

Temperature increases under the RegCM climate decrease crop yields in the eastern and central portion of the study region. The higher temperatures induce a more rapid accumulation of heat units, speeding phenological development, shortening the growing season and reducing crop yields. Yield decreases ranged from 0.6 to 1.1 Mg ha⁻¹ for corn, for sorghum from 0.2 to 0.6 Mg ha⁻¹, for soybean from 0.1 to 0.6 Mg ha⁻¹ in the eastern and central zones. In the case of winter wheat, yield decreases were associated with increases in water stress (e.g., farm 1). Increased temperatures reduced the frequency of cold temperature stress suffered by winter wheat, which

benefited yields. A decrease in temperature stress is responsible for the increased yield of winter wheat on farm 3. This effect is also evident for sorghum and soybean in the northern zone (farms 1, 2). In the case of corn, the higher temperatures, while no doubt decreasing cold stress, also increase heat stress, resulting in a net increase in temperature stress days. Precipitation increases benefited crop yields by reducing water stress. Yield response was most evident in western Nebraska and Kansas. Corn, soybean and sorghum are most sensitive with yields increasing above baseline levels (e.g., farms 6, 8 for corn; 6 for soybean).

Table 5
Simulated stress day mean, median and range for the 302 representative farms by crop and climate scenario

Crop	Scenario	Water stress (days)			Temperature stress (days)			Nitrogen stress (days)			Phosphorus stress (days)		
		Mean	Median	Range	Mean	Median	Range	Mean	Median	Range	Mean	Median	Range
Corn	Baseline	17	9	1–66	15	14	4–62	6	1	0–30	2	0	0–22
	RegCM ^a	9	3	1–39	19	17	6–37	5	3	0–29	2	1	0–22
	RegCM + CO ₂ ^b	6	2	1–35	18	17	5–37	8	8	0–36	3	1	0–27
Sorghum	Baseline	17	12	1–56	20	19	3–70	3	0	0–31	1	0	0–29
	RegCM	8	4	1–37	14	14	3–34	3	1	0–31	1	0	0–24
	RegCM + CO ₂	5	2	1–34	14	14	3–34	4	3	0–36	1	1	0–27
Soybean	Baseline	21	18	2–62	18	16	4–66	0	0	0–0	2	0	0–24
	RegCM	9	6	1–37	13	12	3–27	0	0	0–0	1	0	0–11
	RegCM + CO ₂	7	4	1–35	13	12	4–27	0	0	0–0	2	1	0–22
Winter wheat	Baseline	31	27	9–78	120	119	64–161	10	5	0–74	2	0	0–33
	RegCM	26	26	5–73	68	69	28–90	19	17	0–75	6	4	0–48
	RegCM + CO ₂	20	20	4–66	79	81	40–105	30	30	0–87	7	5	0–59
Switchgrass	Baseline	30	28	1–75	193	194	156–245	37	41	0–70	32	35	0–74
	RegCM	44	47	0–90	150	154	106–200	2	0	0–66	45	51	0–89
	RegCM + CO ₂	53	49	8–137	148	152	104–200	3	0	0–75	70	76	1–108

^a NCAR RegCM2 simulation of 2 × CO₂ climate change scenario with [CO₂] = 365 ppm.

^b NCAR RegCM2 simulation of 2 × CO₂ climate change scenario with [CO₂] = 560 ppm.

Table 6

EPIC simulated mean yields, water stress days, temperature stress days and growing season under baseline climate, RegCM climate and RegCM climate with increased CO₂ for eight representative farms under corn, winter wheat and switchgrass cultivation

Farm # ^a	Yield (Mg ha ⁻¹)			Water stress (days)			Temperature stress (days)			Growing season (days)		
	Baseline	RegCM ^b RegCM ^b	RegCM + CO ₂ ^c	Baseline	RegCM	RegCM + CO ₂	Baseline	RegCM	RegCM + CO ₂	Baseline	RegCM	RegCM + CO ₂
<i>Corn</i>												
1	7.2	6.3	6.7	3	3	2	21	19	18	137	118	118
2	7.7	6.6	7.0	6	1	1	12	20	20	139	121	121
3	7.9	6.9	7.4	5	2	1	13	18	18	155	126	126
4	7.8	7.2	7.5	1	1	1	8	13	11	155	128	128
5	5.8	5.2	5.3	3	1	1	6	8	7	158	126	126
6	4.6	4.7	4.8	17	3	1	13	10	9	169	132	132
7	6.0	5.8	6.0	3	1	1	10	16	16	162	128	128
8	3.5	3.7	4.1	36	22	15	30	31	31	138	127	127
<i>Winter wheat</i>												
1	3.0	2.5	2.9	26	33	33	144	76	84	268	230	230
2	2.9	2.9	3.4	28	25	24	131	74	82	262	230	230
3	3.0	3.2	3.5	26	15	15	112	71	73	245	216	216
4	3.2	2.9	3.1	14	7	6	84	54	52	233	204	204
5	2.7	2.7	2.8	13	6	5	96	65	63	251	224	224
6	2.4	2.4	2.5	26	14	11	110	79	77	259	235	235
7	2.8	2.8	2.9	24	22	17	129	67	86	277	249	249
8	1.6	1.7	1.7	35	39	35	129	68	89	287	258	258
<i>Switchgrass</i>												
1	6.2	15.9	17.7	4	21	13	203	159	158	81	112	112
2	9.0	16.2	18.8	10	43	28	191	153	152	94	111	111
3	8.8	15.8	18.0	17	48	37	183	139	138	99	110	110
4	10.8	15.4	18.0	19	49	36	160	114	114	108	118	118
5	9.4	13.0	15.2	31	59	45	166	123	123	103	117	117
6	6.3	10.3	11.2	51	89	70	181	137	137	99	108	108
7	6.8	13.0	15.1	24	52	40	191	146	145	99	112	112
8	4.3	8.7	9.6	37	69	59	212	170	168	80	109	109

^a Farm locations are identified in Fig. 1 and summarized in Table 1.

^b NCAR RegCM2 simulation of 2 × CO₂ climate change scenario with [CO₂] = 365 ppm.

^c NCAR RegCM2 simulation of 2 × CO₂ climate change scenario with [CO₂] = 560 ppm.

A pattern develops in which yields improve in the western and northern portions of the study region while those in the southern and eastern portions decline. This pattern is most evident in corn (Fig. 5(a)), soybean (Fig. 5(b)) and sorghum (not shown). For the study region, yield increases and decreases tend to cancel one another and regional mean yields of the traditional crops are not greatly affected (Table 4). These results for corn, wheat and soybean are comparable to those of Mearns et al. (1999b).

3.3.2. Switchgrass

Since switchgrass is a perennial forage crop, direct comparison of yields, growing season length and stress day frequencies with those of the grain crops can be misleading. Switchgrass yields differ from grain yields since all of its aboveground biomass is harvested. Stress days are more frequent for switchgrass since its growing season includes the period after summer harvest until dormancy.

Under baseline climate, simulated switchgrass yields range from 2.2 to 12.0 Mg ha⁻¹ across the study

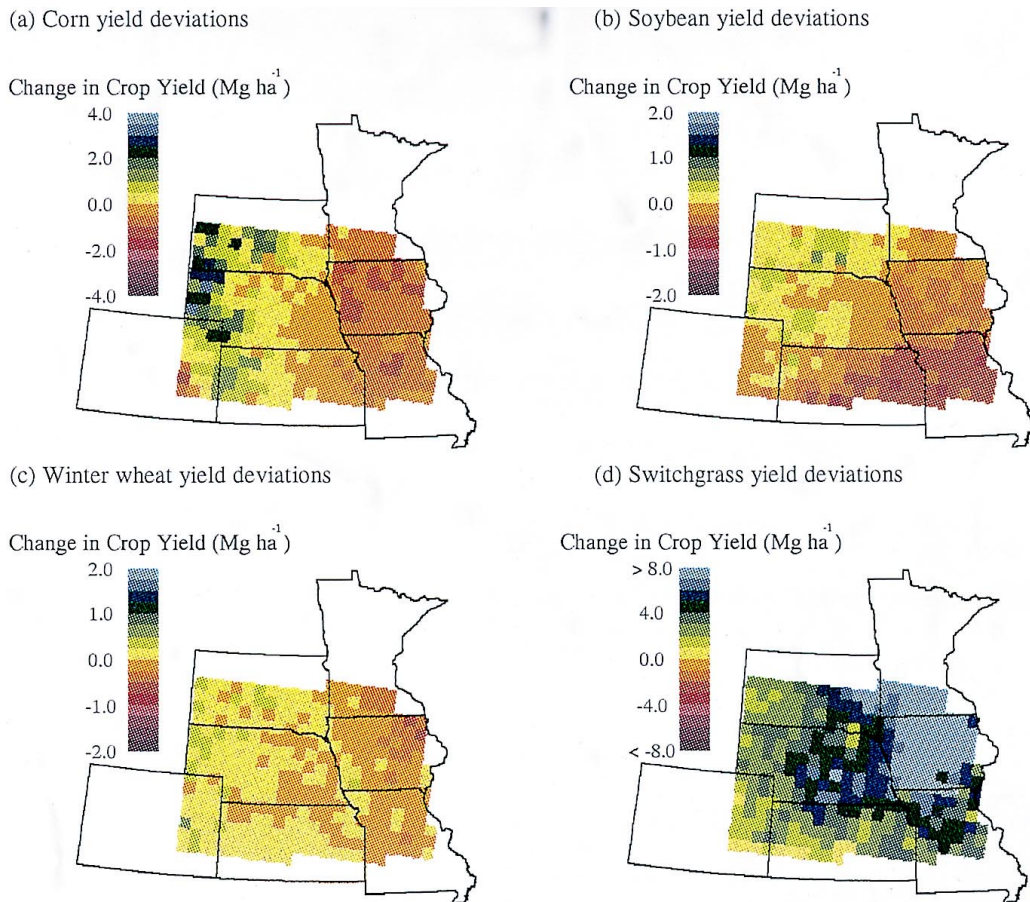


Fig. 5. Yield deviations from baseline for corn, soybean, winter wheat and switchgrass under the RegCM climate scenario with $[\text{CO}_2] = 365$ ppm.

region with the highest yields in southern Kansas and Missouri and the lowest in western Nebraska. Low temperatures stress switchgrass and limit the growing season in the entire study region. Temperature stress in the northern zone exceeds 200 days in some cases (winter days are included in this number). Nitrogen stress also limits switchgrass yields and is primarily recorded during spring emergence. Water stress was evident in the western zone where yields ranged from 4.0 to 6.0 Mg ha^{-1} . However, as compared to corn, switchgrass yields were insensitive to water shortages.

Notably, switchgrass yields (Fig. 5(d)) were favored by the RegCM climate change scenario with yields increasing from baseline on all 302 representative farms. Increases exceeding 8.0 Mg ha^{-1} are simulated in Iowa and eastern Nebraska. Switchgrass

benefits from temperature increases of 6.0 to 8.0°C which reduce cold temperature stress and lengthen the growing season. Yield increases of 2.0 to 5.0 Mg ha^{-1} prevail in the western and central portions of Kansas and Nebraska. The lesser yield increases are associated with more modest increases in spring and winter temperatures and an increased frequency of water stress. This occurs despite increased precipitation and is most evident in southern Kansas.

Nutrient stress is virtually eliminated on all farms under switchgrass cultivation as the higher soil temperatures speed nitrate formation and increase crop available N. This effect is most important in late winter/early spring when switchgrass is emerging from dormancy. Since EPIC calculates N demand from crop biomass, switchgrass consumes the extra N available

for crop uptake. Thus, volatilization and denitrification rates do not increase appreciably. This differs from the traditional summer crops that are planted after N becomes available in early spring.

3.4. Simulated evapotranspiration, runoff and soil erosion

Table 7 presents data on simulated water use and soil erosion under corn, winter wheat and switchgrass cultivation for baseline and the RegCM climate change for the eight representative farms identified in Fig. 1. Under baseline climate, the wheat crop ET is lowest, ranging from 372 to 644 mm while switch-

grass ET is greatest ranging from 414 to 901 mm. The high switchgrass ET rates are due to the crop's longer growing season and ability to extend roots deeper into the soil profile, both factors which are included in EPICs water use algorithms. Water use increases for all crops under the RegCM climate with switchgrass ET increasing significantly and ET of the traditional crops more moderately (Table 7). ET increases are due both to higher temperatures which raise evaporative demand and the greater precipitation, which increases water supply. The dramatic increases for switchgrass ET is due to the multiplicative effects of a lengthened growing season and increased plant biomass.

Table 7

EPIC simulated annual mean evapotranspiration, surface runoff and erosion under baseline climate, RegCM climate and RegCM climate with increased CO₂ for eight representative farms under corn, winter wheat and switchgrass cultivation

Farm # ^a	Evapotranspiration (ET) (mm year ⁻¹)			Surface runoff (Q) (mm year ⁻¹)			Soil erosion ^d (Mg ha ⁻¹ year ⁻¹)		
	Baseline	RegCM ^b	RegCM + CO ₂ ^c	Baseline	RegCM	RegCM + CO ₂	Baseline	RegCM	RegCM + CO ₂
<i>Corn</i>									
1	533	578	545	1	12	14	21	53	50
2	618	655	617	32	82	90	19	38	38
3	627	658	617	12	35	36	10	21	20
4	736	767	726	19	64	66	5	13	13
5	682	693	659	73	155	160	58	123	124
6	558	659	621	5	18	20	159	170	168
7	599	630	589	7	25	26	26	39	37
8	394	437	428	2	15	16	422	305	305
<i>Winter wheat</i>									
1	494	535	508	1	15	15	9	5	5
2	542	597	568	39	93	96	2	2	2
3	538	598	561	14	35	36	2	1	1
4	644	712	671	24	64	65	0	1	1
5	636	728	677	79	159	167	8	16	14
6	514	624	578	9	21	22	29	18	16
7	531	632	583	10	27	28	4	5	4
8	372	414	386	4	18	18	33	28	28
<i>Switchgrass</i>									
1	627	874	825	1	12	13	28	17	17
2	719	865	851	27	68	72	11	12	12
3	758	927	908	12	34	35	1	1	1
4	901	1034	1018	20	57	59	0	1	1
5	802	935	925	66	122	129	1	4	4
6	561	694	693	4	14	15	28	5	5
7	666	817	804	7	19	21	13	11	10
8	414	509	503	2	13	14	43	47	47

^a Farm locations are identified in Fig. 1 and summarized in Table 1.

^b NCAR RegCM2 simulation of 2 × CO₂ climate change scenario with [CO₂] = 365 ppm.

^c NCAR RegCM2 simulation of 2 × CO₂ climate change scenario with [CO₂] = 560 ppm.

^d Soil erosion data includes both simulated wind and water erosion estimates.

Simulated annual runoff was lower under switchgrass cultivation than under the traditional crops. Under baseline climate, for example, runoff on Farm 5 in Kansas ranged from 79 mm under winter wheat to 66 mm under switchgrass cultivation. Low baseline precipitation limits runoff in the western zone regardless of crop cover. Greater precipitation under the RegCM climate change increases runoff. Farms growing corn and wheat show more sensitivity to the changed climate than farms growing switchgrass. Increases from baseline runoff ranged from 11 to 82 mm with corn, from 13 to 80 mm with wheat and from 10 to 56 mm with switchgrass as cover. While the greatest absolute increases in runoff occurred on farms in the eastern and central zone of the study region, the relative changes in the western zone are also notable.

EPIC-simulated rates of erosion are generally highest on farms growing corn due to more frequent tillage of this crop. The high rates of erosion on corn farms 6 and 8 are due to steep topography. At these sites, cultivation of winter wheat and switchgrass reduces erosion significantly. Annual average erosion on farm 1 is slightly greater under switchgrass cultivation (28 Mg ha^{-1}) than under corn or wheat cultivation. This higher value is an artifact explained by the incidence of strong wind erosion during the establishment of the switchgrass in the first year of its 10-year rotation. More than 95% of all the erosion occurs in that year. Ranney and Mann (1994) present experimental data showing similar erosion results. They conclude that most erosion in short rotation woody crops grown for biomass occurs during the establishment period and suggest that proper management of crop residues can mitigate much of this erosion.

The pattern of simulated erosion rates under the RegCM scenario differs by crop. In general, erosion under switchgrass cultivation decreases, it increases under corn and remains insensitive under winter wheat cultivation (Table 7). The higher erosion rates for corn are due to the increased precipitation and runoff which prevail under the RegCM climate. The increased growth for switchgrass reduces the erosion risk by providing a continuous cover throughout the year. These results are similar to those of Phillips et al. (1993) where increases in storm frequency and intensity under a set of GCM climate change scenarios

increased the national potential erosion amounts on cropland by 2–16%.

3.5. The CO_2 -fertilization effect

The effects of the RegCM climate change scenario on EPIC-simulated yields are modified when the simulations reported above were repeated with $[\text{CO}_2] = 560 \text{ ppm}$. Tables 4–6 present simulated yield and stress day data and Table 7 presents simulated water use and soil erosion data RegCM climate change with increased CO_2 .

For all crops, increased $[\text{CO}_2]$ decreases water stress in the central and western zones (farms 6, 8) where water supply is inadequate even under the climate change scenario. Corn and sorghum yields are most responsive to decreased water stress, while winter wheat is least affected. Elevated $[\text{CO}_2]$ does not affect temperature stress. N stress, on the other hand, is increased by elevated $[\text{CO}_2]$ in all crops but soybean. Increases in N stress are due to increased production of biomass which increases crop demand for it.

Relative yield increases under CO_2 -fertilization are greatest for soybean and winter wheat, C3 crops which benefit from increased water use efficiency and greater photosynthetic efficiency. Yield gains for winter wheat are greatest in Iowa and Missouri with maximum increases from baseline of 0.5 Mg ha^{-1} . Elevated $[\text{CO}_2]$ raised yields only slightly on wheat farms in central and western Kansas and Nebraska (farms 6, 8). However, these increases were large enough to return yields to or above baseline. For corn and sorghum, relative yield increases are smaller than in soybean and winter wheat and inversely correlated with water stress. EPIC is programmed to simulate only small responses in photosynthetic efficiency for C4 crops.

Switchgrass increases yields up to 2.6 Mg ha^{-1} above those under climate change alone. As with the other C4 crops, the yield increases at higher $[\text{CO}_2]$ are explained by improved water use efficiency. Switchgrass is the only crop in which the simulations show increased P stress under elevated $[\text{CO}_2]$ — by as many as 38 days more than baseline on farm 8. Increased P stress is explained by two factors: the rapid growth rate of switchgrass may exhaust crop available P and the diminished number of temperature, water and N stress days allows P stress to register even if the severity of that stress does not change appreciably.

This is an artifact of EPIC which allots a stress day to only that stress which is most severe on that day. That yields, the true measure of how well a plant adapts and grows in a given environment, increase despite the high number of P stress days supports this interpretation.

Since the CO₂-fertilization effect increases water use efficiency, crop ET is decreased in all cases under the RegCM climate when [CO₂] = 560 ppm. The greatest decreases in switchgrass ET occur in northeastern Iowa (farms 1, 2, 3), while corn and wheat ET decreases across the entire study region. Climate change with elevated [CO₂] increases runoff slightly regardless of crop cover while elevated [CO₂] has minimal impact on soil erosion with small changes evident for all crops.

4. Discussion

In general, temperature increases under the RegCM climate are most important in determining yield response for both the traditional crops and switchgrass. Temperature increases lowered yields of the traditional crops, while improving those of switchgrass. Higher temperatures speed maturity of the grain crops which, all else equal, decreases yields. This finding is consistent with results obtained by Easterling et al. (1992), Rosenzweig et al. (1994), Brown and Rosenberg (1999) and Mearns et al. (1999a) which also show the negative impact of rising temperatures and shortened growing season on grain yields. The increases in precipitation benefited crop yields, but only slightly. Yields of corn, soybean and sorghum increased above baseline levels in western Kansas and Nebraska. It is unclear whether such yield increases would be sufficient to permit dryland production in the western MINK region. Such a question can be answered only by linking these results to an integrated economics and human activities modeling effort.

The simulation results presented with regard to changes in water use, runoff and soil erosion apply to the individual sites and depend on the specific soil, climate and management regime used. However, the results do show that large-scale conversion to switchgrass in the MINK region may affect water resources through decreased surface runoff. To fully address the potential implications of climate change and biomass

energy crops for the region's water resources will require integration of the farm level results with a regional hydrologic assessment of impacts on water quality and quantity.

Past analyses have addressed or commented on the tradeoffs among biomass energy production, traditional energy sources and agriculture (Lowenberg-DeBoer and Cherney, 1989; Reese et al., 1993; Vogel, 1996b) but have not considered the implications of greenhouse-forced climate change on biomass production. Other analyses have examined the impact of climate change on agriculture in the US Midwest using crop growth simulation models (Rosenzweig, 1989; Rosenberg, 1993; Rosenzweig et al., 1994; Brown and Rosenberg, 1997). Certain of these assessments consider farm level adaptations to climate change, but biomass cropping has not been among the adaptations considered. This simulation exercise has compared the yield response of switchgrass with those of corn, soybeans, sorghum and wheat to changes in climate and concluded that switchgrass can not only compete, in agronomic terms, with traditional crops under baseline climate, but will have a comparative advantage over these crops in a warmer climate. These findings here are consistent with those of an earlier study (Brown and Rosenberg, 1996) and taken together lead us to conclude that switchgrass could be the basis of a biomass energy industry in the MINK region. These findings also demonstrate the possible usefulness of switchgrass, a species native to the Great Plains, as an option for agricultural adaptation to climate change. Further, it will be easier for farmers to adopt switchgrass rather than woody species for biomass production since the same farm machinery and equipment as is used for grain production in the region can be employed. This analysis addresses only the potential production of switchgrass at the farm level. A full economic analysis of the crop's potential that considers production costs, transportation costs and market demand is needed before the practicality of a switchgrass agriculture for the MINK region can be established.

5. Conclusions

Finding presented in this paper point to the following conclusions:

1. Switchgrass yields are comparable, in agronomic terms, with yields of traditional annual grain crops under the current climate of the MINK region.
2. Under the RegCM climate change scenario employed here and characterized by increases in both temperature and precipitation, switchgrass yields increase over baseline levels for the entire MINK region. Yields of the traditional agricultural crops generally decrease in the eastern and southern portions of the region because of higher temperatures. Increases in precipitation in the western and central Great Plains improve yields of both switchgrass and the traditional crops over baseline and may, should they actually occur, increase this region's ability to sustain dryland agriculture.
3. Increases in atmospheric CO₂ increase yields for all crops with soybean and winter wheat exhibiting the greatest response due to improvements in photosynthetic and water use efficiency. Because of improved water use efficiency, the C4 crops — corn, switchgrass and sorghum — are more responsive to increased atmospheric CO₂ in the water scarce regions.
4. Switchgrass generally consumes more water than do the traditional crops under all climate conditions and also reduces runoff. Soil erosion rates under switchgrass cultivation are comparable to those under winter wheat and are generally lower than those under corn cultivation.
5. Simulated switchgrass yields for the MINK region were sufficiently good under both baseline and the RegCM climate change to suggest its promise as a biomass energy feedstock. Because switchgrass thrives in warmer temperatures, the crop may serve as an agricultural adaptation to climate change as well.

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