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## 5 Simulating the effects of elevated CO<sub>2</sub> on crops: approaches 6 and applications for climate change

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### 10 Abstract

11  
12 Several crop models may be used to simulate the effects of elevated CO<sub>2</sub> on crop productivity. Yet no summary exists  
13 in the literature attempting to describe differences among models and how simulations might differ under climate  
14 change conditions. We provide an introductory review focusing on simulating the impacts of elevated CO<sub>2</sub> on crops. We  
15 describe and discuss modeling approaches, component modules, applications to climate change and model validation  
16 and inter-comparison studies. By searching the recent peer-reviewed literature from 1995 to present, we found that  
17 about 20% of published crop modeling studies have focused on climate change impacts. About half of these studies  
18 explicitly analyzed the effects of elevated CO<sub>2</sub> on crop growth and yield. Our analysis further suggested that the crop  
19 models that have been used the most in climate change assessments are also those that have been evaluated the least  
20 using available data from elevated CO<sub>2</sub> experiments. Based on our review, we identify a set of recommendations aimed  
21 at improving our confidence in predictions of production under elevated CO<sub>2</sub> and climate change conditions. These  
22 include continued model evaluation with existing field experiment data; increased focus on limiting factors such as pest,  
23 weeds, and disease; and attention to temporal and spatial scaling issues.

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25 *Keywords:* Elevated CO<sub>2</sub>; Crops; Agriculture; Modeling; Climate change

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### 26 1. Introduction

27 Atmospheric CO<sub>2</sub> concentration is today at 375  
28 μl/l, or 30% higher than during pre-industrial  
29 times, and is increasing at about 0.5% per annum.  
30 Trends in global energy and land use suggest that  
31 anthropogenic emissions of CO<sub>2</sub> and of other  
32 greenhouse gases will continue to be substantial

33 for many decades. As a result, atmospheric CO<sub>2</sub> 33  
34 concentration is projected to be in the range of 34  
35 550–750 μl/l by the end of this century, the lower 35  
36 range depending on whether or not climate policy 36  
37 agreements such as the Kyoto Protocol are soon 37  
38 put in place (e.g., [www.unfccc.org](http://www.unfccc.org), IPCC, 2001). 38  
39 Increasing concentrations of CO<sub>2</sub> in the atmo- 39  
40 sphere are linked to a high probability of climate 40  
41 change, characterized by increased surface tem- 41  
42 peratures, by changed global and regional patterns 42  
43 of precipitation, and by climatic shifts in both 43

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44 mean and variability that could threaten ecosystems  
45 functions and human welfare.

46 In particular, elevated CO<sub>2</sub> and associated  
47 climate change may greatly affect agricultural  
48 production worldwide (IPCC, 1995). It is thus  
49 not surprising that a large body of work has been  
50 devoted to analyzing potential impacts on future  
51 local, regional and global crop production (e.g.,  
52 Rosenberg, 1993; Rosenzweig and Parry, 1994;  
53 Rosenzweig et al., 1995; Jones et al., 2000; Fischer  
54 et al., 2001; Reilly et al., 2001). In the majority of  
55 such studies, crop models were employed to assess  
56 the simultaneous effects on crop growth and yield  
57 of future elevated CO<sub>2</sub>, regional climate change,  
58 and crop management (with or without adapta-  
59 tion). It is in fact well-recognized that CO<sub>2</sub>  
60 concentration and management factors will inter-  
61 act in complex ways to determine the ultimate  
62 impacts of climate change on crop production.  
63 While elevated CO<sub>2</sub> alone tends to increase growth  
64 and yield of most agricultural plants (Kimball,  
65 1983; Cure and Acock, 1986; Allen et al., 1997;  
66 Kimball et al., 2002), warmer temperatures and  
67 changed precipitation regimes may either benefit  
68 or damage agricultural systems (e.g., Rosenzweig  
69 and Hillel, 1998). Water and fertilizer application  
70 regimes will further modify crop responses to  
71 elevated CO<sub>2</sub> (e.g., Reilly et al., 2001).

72 Because of such multi-factor interactions, the  
73 simulated increase in productivity under elevated  
74 CO<sub>2</sub> often determines not only the magnitude, but  
75 even the sign, of the overall changes in crop yields  
76 in global warming scenarios (Tubiello et al., 2002).  
77 The dependence of agricultural assessment studies  
78 on the simulated CO<sub>2</sub> response makes it thus  
79 pertinent to ask the following questions: (i) How  
80 are key plant processes, and their responses to  
81 elevated CO<sub>2</sub>, implemented in current agricultural  
82 crop models? (ii) How have different approaches  
83 been evaluated and/or compared? (iii) How have  
84 the models been used?

85 The modeling state-of-the art with respect to  
86 plant responses to elevated CO<sub>2</sub> has already been  
87 reviewed (e.g., Boote and Loomis, 1991; Allen et  
88 al., 1997). However, there is little overview about  
89 the variety of crop models and incorporated  
90 approaches to modeling the effects of elevated  
91 CO<sub>2</sub>. Although many of these models are quite

92 routinely used in climate change studies, we do not  
93 fully understand the impacts that different meth-  
94 ods may have on projected results.

95 Clearly, most agricultural crop models used in  
96 climate change studies were not originally devel-  
97 oped with the intention to model plant responses  
98 to elevated CO<sub>2</sub> under climate change conditions,  
99 but rather to provide, under current climate: (a)  
100 decision support to farmers, regional or national  
101 authorities; (b) insight into specific physiological  
102 processes; (c) agronomic relationships among  
103 crops and cropping systems; or (d) analyses of  
104 environmental effects at various spatial scales (e.g.,  
105 plot, field, ecosystem, regional or even global).  
106 Therefore, while a few models already contained  
107 relationships that accounted for effects of elevated  
108 CO<sub>2</sub> on individual processes, many had to be  
109 specifically modified.

110 In the following sections, we analyze differences  
111 among crop models widely used in climate change  
112 studies. In detail, we describe approaches to  
113 modeling effects of CO<sub>2</sub> on crops, we perform a  
114 literature search on model applications and ana-  
115 lyze studies on models testing and inter-compar-  
116 ison that have used high-quality data from  
117 elevated CO<sub>2</sub> experiments. Finally, we identify  
118 specific modeling and methodological issues that  
119 require attention in order to increase confidence in  
120 the simulations of the effects of elevated CO<sub>2</sub> and  
121 climate change conditions on crop production.

## 122 2. Crop response to elevated CO<sub>2</sub>: background

123 Irrespective of climate change issues, the posi-  
124 tive impacts of elevated CO<sub>2</sub> on plant growth and  
125 yield were well understood and put into practice  
126 by greenhouse vegetable growers since the 1930s,  
127 leading to elevated CO<sub>2</sub> large-scale commercial  
128 operations within a few decades (Nederhoff, 1994).  
129 The scientific recognition and rudimentary theo-  
130 retical understanding of the positive role of CO<sub>2</sub> on  
131 plant photosynthesis had of course developed  
132 much earlier (1770–1850), while the first plant  
133 growth controlled experiments with enriched CO<sub>2</sub>  
134 were performed at the beginning of the 20th  
135 century (Browne and Escombe, 1902). Progress  
136 in biochemistry and plant physiology subsequently

led to the discovery of the C3 carbon fixation pathway by Calvin in the 1940s; of the interactions of CO<sub>2</sub> and transpiration via effects on leaf stomata by Gaastra (1959) in the 1950s; of C4 and CAM photosynthesis pathways in the 1960s (see, e.g., Hatch, 1992). Improved measurements of O<sub>2</sub> evolution and isotope discrimination characterized progress in the 1970s and 1980s, leading to measurements of photosynthetic quantum yield and increased understanding of stomatal dynamics (Koh and Kumura, 1971; McCree, 1972; Farquhar et al., 1982; Bjorkman and Demmig, 1987).

In the last decade, great effort was devoted to further understand the effects of elevated CO<sub>2</sub> on growth and yield of most agricultural plants (Kimball, 1983; Cure and Acock, 1986; Bowes, 1993; Allen et al., 1997; Kimball et al., 2002). Mechanisms regulating the interactions of CO<sub>2</sub> with other environmental conditions, e.g., light, temperature, soil quality, soil water status, nutrient supply, exposure to air pollutants, weeds and pests, have been investigated extensively (Allen, 1990; Lawlor and Mitchell, 1991; Idso and Idso, 1994; Morison and Lawlor, 1999). Recent research has focused greatly on the effects of elevated CO<sub>2</sub> on key plant and ecosystem processes such as community-level carbon assimilation and respiration; stomatal conductance and transpiration; partitioning to grain and fruit; above- and below-ground partitioning; phenological development; root and soil processes; and on the potential for acclimation of individual processes to elevated CO<sub>2</sub> conditions (e.g., Amthor and Loomis, 1996; Allen et al., 1997; Norby et al., 2001).

### 3. Modeling approaches of crop responses to CO<sub>2</sub>

#### 3.1. Development in crop modeling

Plant models were historically developed following the accumulation of knowledge and the progressive availability of experimental data. Leaf-level models of photosynthesis were developed early last century (Blackman, 1919) to describe photosynthesis-light response curves. It was not until the early 1950s and 1960s, however,

that models computing canopy-level light interception and carbon assimilation rates were built (Monsi and Saeki, 1953; de Wit, 1965; Duncan et al., 1967; Hesketh and Baker, 1967). The simplifying concept of crop radiation-use efficiency (RUE) was developed and applied to agronomic crop modeling in the 1970s (Sinclair et al., 1976; Monteith, 1977; Norman, 1979); during the same period, maintenance respiration was quantified and implemented in plant growth models (Penning de Vries, 1975; de Wit, 1978). The first crop photosynthesis models to include CO<sub>2</sub> as an explicit variable were built in the 1970s and early 1980s, and included rectangular hyperbolae describing leaf photosynthesis dependence on light and CO<sub>2</sub> concentration, scaled to canopy level (Acock et al., 1971; Thornley, 1976; Acock et al., 1978; Charles-Edwards, 1981; Acock and Allen, 1985; Goudriaan et al., 1985). Finally, leaf-level biochemical models of photosynthesis including direct CO<sub>2</sub> effects on photosynthesis were published in the early 1980s (Charles-Edwards, 1981; Farquhar et al., 1980; Farquhar and von Caemmerer, 1982; Ball et al., 1987).

More recent developments in crop modeling have aimed at harmonizing and improving these various approaches, from better scaling routines from leaf to canopy (or even from cell to canopy) to the introduction of leaf nitrogen distributions affecting photosynthetic capacity; to refining temperature–CO<sub>2</sub> and water–CO<sub>2</sub> interactions (e.g., see for further details: Long, 1991; Boote and Loomis, 1991; Norman, 1993; Boote et al., 1997).

#### 3.2. Types of models and scale issues

Integration into crop models of experimental knowledge on the effects of elevated CO<sub>2</sub> on plant growth and yield is required for predicting crop productivity under scenarios of global change. Many crop models have been modified to this end, yet there is no summary in the literature documenting today's state-of-the art approaches, or discussing how model performances compare across models and against experimental data. One problem for compiling such a summary is certainly represented by the large number of models—and

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Table 1

Example of studies in which crop simulation models were used to predict effects of elevated CO<sub>2</sub> on wheat

Study/Model	Country/Region <sup>a</sup>	Reference
<i>Impact assessment studies</i>		
AFRCWHEAT2	United Kingdom, France	Semenov et al. (1993)
CropSyst	Italy	Tubiello et al. (2000)
CERES	Argentina	Magrin et al. (1997)
	Bangladesh	Karim et al. (1996)
	Bulgaria	Alexandrov and Hoogenboom (2000)
	Canada	Brklacich and Stewart (1995)
		El Maayar et al. (1997)
	China	Shi et al. (2001)
	Commonwealth of Independent States	Menzhulin et al. (1995)
	India	Lal et al. (1998)
	France	Delecalle et al. (1995)
	Romania	Cuculeanu et al. (1999)
	Uruguay, Argentina	Baethgen and Magrin (1995)
	USA	Adams et al. (1990)
	Tubiello et al. (2002)	
EPIC	USA	Easterling et al. (1992a)
		Easterling et al. (1992b)
		McKenney et al. (1992)
		Easterling et al. (1993)
		Brown and Rosenberg (1999)
		Brown et al. (2000)
		Easterling et al. (2001)
EuroWheat	Europe	Harrison and Butterfield (1996)
GAEZ	Global	Fischer et al. (2001)

Table 1 (Continued)

Study/Model	Country/Region <sup>a</sup>	Reference
SUCROS87	Europe	Nonhebel (1996)
WOFOST	Europe	Wolf (1993)
APSIM	Australia	Reyenga et al. (1999)
<i>Other studies<sup>b</sup></i>		
AFRCWHEAT2, LINTULCC2, Sirius	Spain	Ewert et al. (2002)
AFRCWHEAT2,	United Kingdom	Porter et al. (1995)
AFRCWHEAT3S		
AFRCWHEAT2, CERES, NWHEAT, Sirius, SOILN	United Kingdom, Spain	Semenov et al. (1996)
		Wolf et al. (1996)
AFRCWHEAT2-O3, LINTLCC	Europe	van Oijen and Ewert (1999)
CENTURY	USA	Paustian et al. (1996)
CERES	USA	Rosenzweig and Tubiello (1996)
CERES, EPIC, Stewart and Sinclair models	Canada	Touré et al. (1995)
CLIMCROP	Denmark	Olesen et al. (2000)
EPIC	USA	Stöckle et al. (1992)
	USA	Brown and Rosenberg (1997)
SOIL/SOILN	Sweden	Eckersten et al. (2001)
Wheat model	Australia	Wang and Connor (1996)

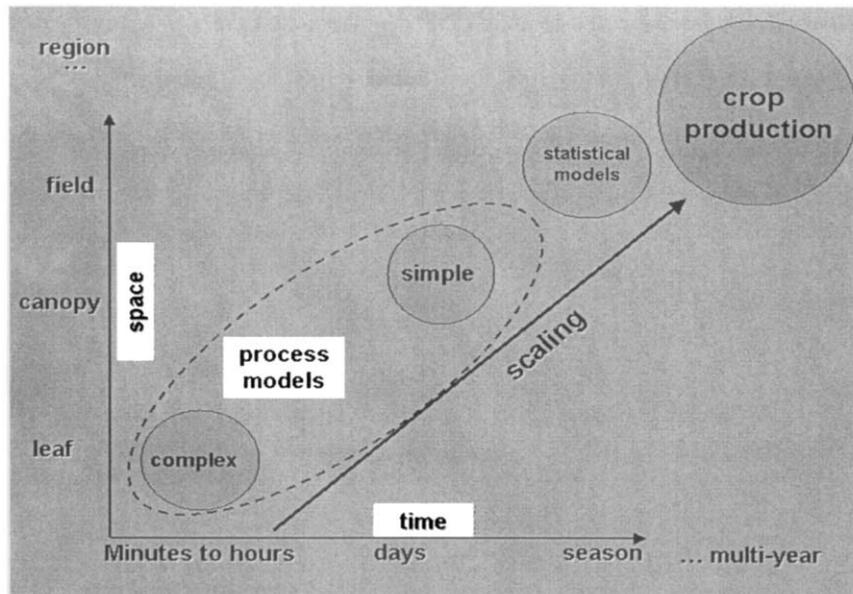
<sup>a</sup> Simulations were performed either for selected sites or entire regions, or countries.

<sup>b</sup> Studies with more focus on sensitivity analysis (e.g., model components, input data), model inter-comparison etc.

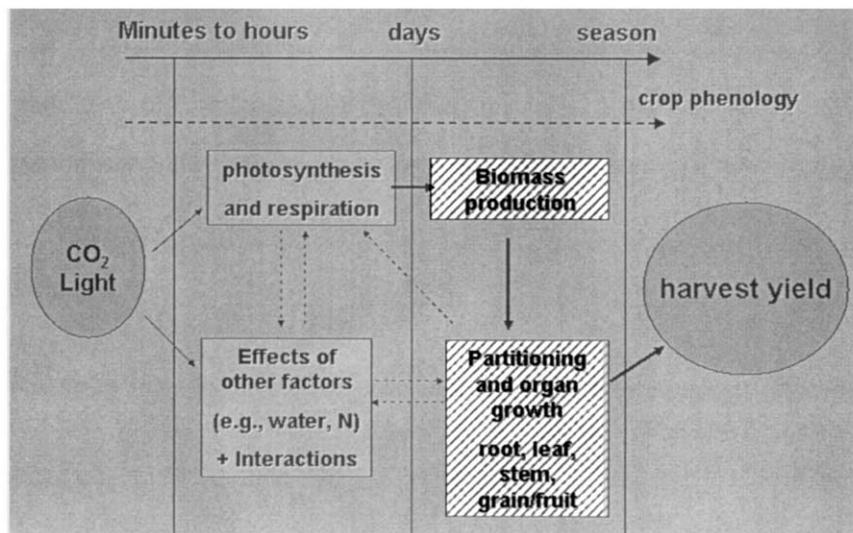
model versions—used in assessment studies. An example for wheat models is given in Table 1. Particularly, the existence of many different modeling approaches, including differences in models structure and modeling detail[fe1] make any such comparison difficult.

As shown in Fig. 1a, two kinds of models can be generally identified: statistical models, used to empirically predict large-scale (county to region) agricultural yields from regression analyses based on monthly or annual variables; and process-oriented models, further referred to as process

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(a)



(b)

Fig. 1. Schematic illustrations of crop modeling approaches of (a) types of models in relation to levels of spatial (leaf, canopy, field) and temporal (minutes to hour, days, seasons) scales and (b)  $\text{CO}_2$  effects on different processes. Note that processes are simulated with different time-steps, e.g. leaf level photosynthesis in minute to hour intervals, crop biomass production in days and harvest yield in season or multi-year steps (see text for further explanation). The bold arrow lines among boxes indicate the flow of carbon from leaf to canopy production to harvest yield. Important feedbacks (dotted lines among boxes) link many of these processes across timescales.

275 models, used to compute crop dynamics at smaller  
 276 spatial scales (leaf to canopy and/or field levels),  
 277 based on deterministic equations and simulation of  
 278 underlying processes at timescales of minutes to

279 days. Process models can be further grouped into  
 280 'complex' and 'simple'. Complex models compute  
 281 processes at the level of organs or lower; for  
 282 example, the dynamics of carbon and water

283 calculated at the leaf-level, requiring time-steps  
284 ranging from minutes to hours. Simple models are  
285 more holistic and compute canopy-level dynamics  
286 directly, using empirical relationships without  
287 consideration of underlying processes, typically  
288 using daily time-steps (for a discussion on simple  
289 vs. complex modeling issues, see [Passioura, 1979](#);  
290 [Thornley, 1980](#); [Charles-Edwards et al., 1986](#);  
291 [Sinclair and Seligman, 2000](#)).

292 In general, both statistical and process models  
293 adequately predict agronomic yields at given  
294 scales. Statistical models were intrinsically de-  
295 signed to operate at the multi-seasonal, regional  
296 scale, and are thus best suited for analyzing inter-  
297 annual variability of regional production. Process  
298 crop models were developed to simulate crop  
299 responses to environmental conditions at the plot  
300 and field level and can be used to analyze inter-  
301 seasonal dynamics of field-level crops. [fe2]Many  
302 assessment studies have employed process models  
303 to project the impacts of climate change and  
304 elevated CO<sub>2</sub> on crops from field-scale to regional  
305 and even global levels. Yet, no clearly defined  
306 methodologies exist for extending field-level yields  
307 computed with process models to large regions.

### 308 3.3. Model components and responses to CO<sub>2</sub>

309 We focus on process crop models, as these  
310 [fe3]capture the dynamics of crop response to  
311 elevated CO<sub>2</sub>, and because they have widely been  
312 used in climate change studies. These models have  
313 different components that can be simplistically  
314 grouped into those computing: plant phenology as  
315 a function of accumulated temperature and day-  
316 length; photosynthesis and respiration; water bal-  
317 ance, N-uptake and distribution and effects of  
318 other factors; partitioning, biomass accumulation  
319 and organ growth ([Fig. 1b](#)). These components  
320 may operate at different timescales. For instance,  
321 photosynthesis and water exchange are resolved at  
322 timescales from minutes to hours (complex process  
323 models) to days (simple process models). Biomass  
324 production and partitioning, and ultimately yield,  
325 are generally computed at daily (process models)  
326 to seasonal (statistical models) time-steps. Thus,  
327 linkages among model components are often  
328 across timescales. For instance, ‘long-term’ pat-

terns of root partitioning may affect soil-water 329  
dynamics, which in turn may modify ‘short-term’ 330  
stomatal dynamics and photosynthesis; patterns of 331  
biomass accumulation and growth of reproductive 332  
organs may trigger source-sink relations, also 333  
capable of modifying leaf photosynthetic rates, 334  
etc. 335

336 Simulating the effects of elevated CO<sub>2</sub> on crop  
337 growth and yield within process models involves  
338 the introduction and/or modification of specific  
339 components. Previous reviews have focused on  
340 theoretical aspects of such modifications, describ-  
341 ing in detail model equations, especially focusing  
342 on leaf and canopy photosynthesis (e.g., [Boote et  
343 al., 1997](#); [Long, 1991](#)). We provide herein a  
344 summary of modeling solutions implemented in  
345 current crop models, following the simplified list  
346 of components illustrated in [Fig. 1b](#) (see also [Table  
347 2](#)).

#### 348 3.3.1. Light interception and photosynthesis

349 Process models have some component to simu-  
350 late light interception of the canopy. The approach  
351 taken largely depends on the concept used to  
352 simulate carbon assimilation or biomass produc-  
353 tion; e.g. whether a big-leaf model ([Boote and  
354 Loomis, 1991](#)), a sunlit/sunshade two-box model  
355 ([Boote and Loomis, 1991](#)) or a multiple-layered  
356 model is used, and whether or not leaf-angular  
357 distributions and crop geometries and other fac-  
358 tors are accounted for (i.e., [Spitters, 1986](#)). Radia-  
359 tion absorption is often computed separately for  
360 direct and diffuse radiation.

361 Many crop models employ equations written at  
362 leaf or canopy level to explicitly simulate gross  
363 photosynthesis rates from absorbed light. The  
364 photosynthesis response to light is often calculated  
365 using exponential or rectangular hyperbolic func-  
366 tions with parameters representing quantum effi-  
367 ciency and light saturated rate of photosynthesis;  
368 examples are AFRCWHEAT2 ([Weir et al., 1984](#);  
369 [Porter, 1993](#)), CROPGRO ([Hoogenboom et al.,  
370 1992](#)), SUCROS ([Goudriaan and van Laar, 1994](#))  
371 and WOFOST ([Boogard et al., 1998](#)). Effects of  
372 atmospheric CO<sub>2</sub>, temperature and other factors,  
373 depending on the model, on quantum efficiency  
374 and light saturated rate of photosynthesis are  
375 realized via empirical relationships. Few crop

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Table 2

Summary of modeling approaches of CO<sub>2</sub> effects on plant processes considered in crop models. Note, that implementation of modeling approaches differ among models

Processes	CO <sub>2</sub> effect	Modeling Approaches
Assimilation	Increase	Direct effect using: (a) Biochemical model of leaf photosynthesis (b) Photosynthesis-light response curve (c) RUE which empirically increases with CO <sub>2</sub>
Respiration <sup>a</sup>	Increase	Not considered <sup>b</sup>
Stomatal conductance	Decrease	(a) Direct effect via empirical reduction of stomatal conductance (b) Indirect effect via coupling of models for photosynthesis and stomatal conductance
Partitioning	Variable responses	Indirect effect via source-sink relationships
Organ growth	Increase	Indirect effect via increase in assimilation
Phenological development	Variable responses (Mostly acceleration)	Indirect effect via increase in canopy temperature
<i>Soil water balance</i>		
(a) Transpiration <sup>c</sup>	Decrease	(a) Direct effect via empirical reduction in transpiration with CO <sub>2</sub> (b) Indirect effect via reduction in stomatal conductance
(b) Water uptake	Decrease <sup>d</sup>	Indirect effect via reduction in stomatal conductance and transpiration <sup>c</sup> , and acceleration of crop development
(c) Water use efficiency	Increase	Indirect effect via reduction in transpiration
<i>Nitrogen dynamics</i>		
N-concentration in biomass	Decrease	Indirect effect via increase in biomass
N-uptake	Increase	Indirect effect via increase in N demand

See text for explanation.

<sup>a</sup> Respiration rate per unit dry weight.

<sup>b</sup> Respiration may increase via increase in canopy temperature under elevated CO<sub>2</sub>.

<sup>c</sup> Transpiration per unit leaf area.

<sup>d</sup> Water uptake may also increase via increase in canopy size and root growth at elevated CO<sub>2</sub>. The net result will depend on the specific environmental conditions.

376 models use more detailed, biochemical equations,  
377 such as the ones described by Farquhar et al.  
378 (1980); examples are DEMETER (Kartschall et  
379 al., 1995) and LINTULCC2 (Rodriguez et al.,  
380 2001). However, these equations have many coef-  
381 ficients that must be derived from leaf measure-  
382 ments and parameterization of such equations  
383 remains difficult (Boote et al., 1997).

384 In these models, gross photosynthesis is often  
385 computed at minute to hourly intervals and scaled  
386 to canopy levels. In addition, respiration losses  
387 and conversion units are calculated and used to  
388 finally compute daily biomass accumulation. By  
389 contrast, simple models often calculate net bio-  
390 mass production directly, typically in daily time-  
391 steps, by multiplying the light intercepted by the

392 crop's RUE. In many such models RUE is a  
393 constant that is empirically derived by comparing  
394 seasonal data of crop biomass accumulation  
395 versus light totals (e.g., Sinclair and Horie, 1989).  
396 However, in some models RUE may be dependent  
397 on light intensity and plant age (Ritchie and Otter-  
398 Nacke, 1985). The effects of elevated CO<sub>2</sub> on RUE  
399 are modeled empirically, using simple linear or  
400 curvilinear multipliers. Examples of simple ap-  
401 proaches are the modeling systems CERES  
402 (Ritchie and Otter-Nacke, 1985; Tsuji et al.,  
403 1994), EPIC (Williams et al., 1989), APSIM  
404 (Reyenga et al., 1999) or the wheat model Sirius  
405 (Jamieson et al., 2000).

406 RUE models can hardly be evaluated against  
407 leaf-level data, so that validation of the RUE

408 response to CO<sub>2</sub> remains difficult. By contrast,  
409 complex model predictions of leaf or canopy-level  
410 instantaneous photosynthesis rates can be shown  
411 to perform well against a range of environments.  
412 Nonetheless, as we discuss in a following section,  
413 several simple and complex models alike, evalu-  
414 ated using above-ground biomass and yield data  
415 under ambient and elevated CO<sub>2</sub>, were found in  
416 agreement with observed data.

### 417 3.3.2. Respiration

418 Complex models that simulate photosynthesis  
419 also compute maintenance and growth respiration.  
420 Respiration is not explicitly considered in the more  
421 simple RUE models. None of the models we  
422 considered included direct effects of elevated CO<sub>2</sub>  
423 on respiration, although some experiments have  
424 indicated the importance of such effects (Amthor,  
425 1997). Indirect increases in simulated maintenance  
426 and growth respiration rates under elevated CO<sub>2</sub>  
427 are computed in complex models only as a  
428 consequence of larger standing biomass and higher  
429 growth rates, from which respiration rates typi-  
430 cally depend.

### 431 3.3.3. Water balance

432 In some complex models simulations of photo-  
433 synthetic carbon uptake are linked with calcula-  
434 tions of leaf stomatal conductance. Based on  
435 mechanisms that optimize carbon fixation and  
436 water loss, leaf stomates can close under water  
437 stress, automatically reducing the flow of CO<sub>2</sub> into  
438 the leaf/canopy, and limiting photosynthetic rates  
439 (e.g., Ball et al., 1987; Leuning, 1995). Alterna-  
440 tively, increasing CO<sub>2</sub> concentration may also  
441 induce stomatal closure and, thus, reduce water  
442 loss through transpiration. These dynamics require  
443 computation of leaf or canopy level energy bal-  
444 ances and simulation time-steps are from minutes  
445 to hours, substantially increasing the number of  
446 calculations per model run. Examples of models  
447 implementing such an approach are CROPGRO  
448 (Hoogenboom et al., 1992), LINTULCC2 (Rodri-  
449 guez et al., 2001) and DEMETER (Kartschall et  
450 al., 1995; Grossman-Clarke et al., 2001). A simpler  
451 approach used in a number of crop models is to  
452 simulate effects of water stress on photosynthesis  
453 via empirically calculated factors. Using daily

454 time-steps, these models first compute canopy  
455 potential transpiration. Based on the assumption  
456 that stomatal closure is controlled by the balance  
457 between available plant root water uptake and  
458 potential transpiration demand (Tubiello et al.,  
459 1995; Ewert et al., 2002), a working assumption is  
460 made that, if actual transpiration—at most equal  
461 to available root water uptake—is less than the  
462 potential demand, ‘stomates will have adjusted  
463 over the course of a day to account for that  
464 imbalance’ (Ritchie and Otter-Nacke, 1985). This  
465 implies that the actual reduction of daily biomass  
466 production depending on that water stress must be  
467 proportional to the ratio of actual to potential  
468 evapotranspiration (Ritchie, 1972). Some models  
469 may further limit biomass production under  
470 water-limiting conditions by means of a transpira-  
471 tion efficiency coefficient, TE, dependent on air  
472 relative humidity, multiplied by daily total canopy  
473 transpiration to obtain daily total biomass accu-  
474 mulation. In these models, actual biomass produc-  
475 tion is computed as the minimum between RUE-  
476 and TE-dependent quantities (e.g., Stöckle et al.,  
477 1992; Reyenga et al., 1999).

478 The adjustment for elevated CO<sub>2</sub> conditions in  
479 these models is made empirically, by reducing  
480 potential transpiration demand via a multiplier,  
481 representing reduction of maximum stomatal con-  
482 ductance as a function of CO<sub>2</sub>. Examples of crop  
483 models following such approach include the  
484 DSSAT-CERES and EPIC family of models  
485 (e.g., Peart et al., 1989; Stöckle et al., 1992).

486 As in the case of photosynthesis modeling,  
487 simple approaches to transpiration cannot be  
488 evaluated against leaf-level data. Only few data  
489 exist of crop canopy gas exchange measurements  
490 (e.g., Rodriguez et al. 2001) so that it is difficult  
491 to validate modeled CO<sub>2</sub> impacts on transpiration  
492 and photosynthesis. However, many authors have  
493 often incorrectly assumed that the carbon–water  
494 relations observed under elevated CO<sub>2</sub> could only  
495 be captured by complex modeling (e.g., Grant et  
496 al., 1995). As shown in a later section, compar-  
497 isons with observed data for biomass and yield  
498 have rather shown that complex and simple  
499 approaches alike could well reproduce these dy-  
500 namics.

### 3.3.4. Phenological development

Elevated CO<sub>2</sub> may affect crop development via effects on leaf temperature via CO<sub>2</sub>-induced stomatal closure. Complex models compute such an effect via energy balance calculations, affecting leaf temperature and enhancing plant senescence (e.g., DEMETER, ecosys). Simple models do not include such an effect, which is nonetheless thought to be small in most environments.

### 3.3.5. Biomass partitioning and yield

The degree of complexity of process crop models relative to partitioning of biomass depends on their ability to dynamically allocate carbon among roots, stems, leaves, and grain or fruit, as a function of resource status. Under elevated CO<sub>2</sub>, feedbacks between photosynthesis rates and organ growth/size, known as source-sink relations, may significantly modify photosynthesis rates, partitioning and biomass accumulation over time (Boote et al., 1997; Grace, 1997). Most crop models used in current studies do not simulate source-sink relations, lacking dynamic partitioning rules. Constant allocation fractions for allocating carbon among organ groups are used instead. In some models allocation fractions change empirically with crop development. Among the reviewed crop models, the simplest computed harvest yield from final above-ground biomass, via a harvest index coefficient that could be reduced as a function of water stress accumulated during the growing season. Examples are EPIC (Williams et al., 1989); GAEZ (Fischer et al., 2001); and Cropsyst (Stöckle et al., 1994). More complex approaches were those computing partitioning to roots, leaves, stem and grain or fruit, via coefficients that depended on phenology and, in some cases, on water stress. Examples were the DSSAT models. With these models, simulations under elevated CO<sub>2</sub> were capable of generating dynamic feedbacks between root systems, water uptake, and biomass accumulation (e.g., Tubiello et al., 1995).

### 3.3.6. Interaction with other conditions

The effects of CO<sub>2</sub> on crop growth and yield can greatly vary depending on other environmental and management factors. The interactive effects of

CO<sub>2</sub> and air temperature on crop photosynthesis are one notable example. Elevated CO<sub>2</sub> levels tend to shift the leaf-level photosynthetic optimum towards higher temperatures (Long, 1991). Simulating such interactions can best be resolved by complex models that account for bio-chemical relationships of leaf photosynthesis and for feedbacks among photosynthesis, water and energy balance (see photosynthesis and water balance sections above). Models such as CROPGRO and DEMETER capture the complexity of these effects, while simple approaches such as those implemented into DSSAT and EPIC do not. However, comparison of simulations from complex and simple models showed that such interactions between CO<sub>2</sub> and temperature on leaf photosynthesis may be small when averaged over the whole growing season (Boote et al., 1997).

Water and nutrients also affect crop responses to CO<sub>2</sub>. Experimental data suggest that the relative effects of elevated CO<sub>2</sub> on crop growth and yield is more pronounced under water-limited as compared to well-watered conditions (Chaudhuri et al., 1990; Kimball et al., 1995). The contrary is true for nitrogen: well-fertilised crops respond more positively to CO<sub>2</sub> than less fertilised ones (Sionit et al., 1981; Mitchell et al., 1993). Such effects have been included in crop models indirectly, via effects on stomatal closure and/or transpiration and canopy development. As discussed in the next section, both simple and complex crop models have shown some ability to mimic the interactions of elevated CO<sub>2</sub> with water and N.

Effects of CO<sub>2</sub> on crops also vary depending on presents of air pollutants, such as tropospheric ozone, which may limit CO<sub>2</sub> effects by reducing stomatal conductance and/or by decreasing biochemical activity due to cell damage. Such effects have been included into a few mechanistic models (e.g., Ewert et al., 1999). Current crop models do not yet include other important factors that could limit crop response to CO<sub>2</sub> in the field, such as soil quality, competition with weeds, pests and disease.

Finally, biochemical acclimation of plant photosynthesis to elevated CO<sub>2</sub> was not implemented in the crop models considered.

#### 593 4. Model applications: elevated CO<sub>2</sub> and climate 594 change assessments

595 How are the crop models modified for elevated  
596 CO<sub>2</sub> used? We have used the science portal Scirus  
597 ([www.scirus.com](http://www.scirus.com)) for searching the peer-reviewed  
598 literature published since 1995, in order to derive  
599 information on the number of studies assessing the  
600 effects of elevated CO<sub>2</sub> and climate change on crop  
601 production. A preliminary search indicated about  
602 1000 published articles that used crop modeling to  
603 investigate a variety of issues, e.g., field-level to  
604 regional crop production, soil and water quality,  
605 land-use and economic assessments. Of the total  
606 number of references found, roughly 20% were  
607 climate change assessment studies. Those con-  
608 cerned with elevated CO<sub>2</sub> were 8% of the total,  
609 irrespective on whether or not climate change  
610 issues were considered (Fig. 2).

611 Our search also indicated that the crop models  
612 that have been most widely used both in general  
613 applications as well as to predict future crop yields  
614 under climate change and elevated CO<sub>2</sub> concen-  
615 trations, were decision support systems. These  
616 modeling environments cover many crop types,  
617 often including the major cereal (wheat, maize,  
618 rice, barley, sorghum, millet), soybean, potato, oil  
619 crops (peanut, rape oil) and some vegetable crops

(tomato). Our web search showed that three 620  
decision support models, DSSAT, EPIC, and 621  
SUCROS–WOFOST, were those most widely 622  
used in the literature. The percent of climate 623  
change and/or elevated CO<sub>2</sub> studies performed 624  
with these three models was higher than the 625  
general average. Roughly 35% of all crop model- 626  
ing studies published with these three models were 627  
climate change assessments. The percent of studies 628  
involving the effects of elevated CO<sub>2</sub> was 14%. 629

630 We next searched for single-crop modeling  
631 studies. As also shown in Fig. 2, we found that  
632 60% of all studies published in the last 6 years  
633 involved analyses of wheat, maize, soybean, and  
634 rice. Of the remaining, potato, sorghum, tomato,  
635 sunflower, millet were among the crops most  
636 simulated. Within each crop, the percentage of  
637 climate change assessment studies with or without  
638 CO<sub>2</sub> effects was close to that found overall.

639 A more detailed analysis on model applications  
640 was performed for wheat. Our web search found a  
641 group of nine models as the most cited over the  
642 last 6 years, of which CERES-Wheat, EPIC and  
643 SUCROS represented three-quarters of all crop  
644 modeling studies in wheat. AFRCWHEAT2, Sir-  
645 ius, Cropsyst, SWHEAT, LINTUL, APSIM were  
646 among the models most used in the remaining  
647 group. The differences among models in the  
648 frequency of usage were the same for climate  
649 change studies. When refining our search by  
650 considering CO<sub>2</sub> effects some sharp asymmetries  
651 emerged among models. The percentage of the  
652 CO<sub>2</sub> studies that were also performed in conjunc-  
653 tion with climate change impact analyses was close  
654 to 100% for EPIC, and about 40% for CERES-  
655 Wheat. By contrast, SUCROS, SWHEAT, Sirius,  
656 and AFRCWHEAT2 were by far more frequently  
657 used in studies investigating the effects of elevated  
658 CO<sub>2</sub> alone and evaluating models with experimen-  
659 tal data and without connections to climate change  
660 assessments. These differences plainly indicate that  
661 some of the wheat models most widely used in  
662 climate impact studies are not those evaluated  
663 under elevated CO<sub>2</sub>! It has been argued that  
664 successful validation of sub-routines or model  
665 components, as extensively done for leaf photo-  
666 synthesis, provide sufficient evidence for crop  
667 model validity and justify its application in climate

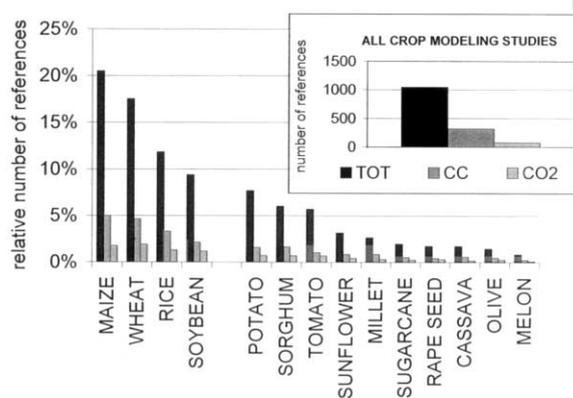


Fig. 2. Number of references from the peer-reviewed literature since 1995 dealing with crop modeling obtained from a web search using Scirus ([www.scirus.com](http://www.scirus.com)). Results refer to total number of crop modeling studies found (TOT); those explicitly concerned with climate change impacts (CC); and those considering effects of elevated CO<sub>2</sub> (CO<sub>2</sub>) for all crops (inserted graph) and individual crops (large graph).

change studies. However, effects of elevated CO<sub>2</sub> on crop growth and yield are complex, including responses of different growth and development processes (e.g., Kimball et al., 2002). Consequently, crop models require testing against a range of data from experiments in which effects of elevated CO<sub>2</sub> on crops' growth and yield were investigated.

In the following section we focus on the availability and suitability of experimental data for model evaluation and on published results on model testing and inter-comparison. Again, we restricted our analysis to wheat because it has been investigated and modeled most extensively.

## 5. Model testing and inter-comparison

About a decade ago, many authors had recognized that experimental data of the effects of elevated CO<sub>2</sub> on crop plants, particularly in combination with other factors, was needed in order to advance model development (see e.g., Lawlor and Mitchell, 1991). Since then a number of studies were performed using different experimental approaches (Table 2). A large number of experiments were performed within controlled environment chambers (e.g., Mitchell et al., 2001); under semi-controlled conditions using greenhouses (e.g., Gifford and Morison, 1993; Lawlor and Mitchell, 1993); inside temperature gradient tunnels (e.g., Conroy et al., 1994; Rawson, 1995); or within open-top chambers (e.g., Mulholland et al., 1998; Hertstein et al., 1999). More recently, elevated CO<sub>2</sub> experiments were designed under more realistic field conditions, using free-air carbon dioxide enrichment facilities (e.g., Kimball et al., 1995; Norby et al., 2001; Bindu et al., 2001; Miglietta et al., 2001; Weigel and Dämmgen, 2000).

Several attempts have been made to use data from controlled (e.g., Mitchell et al., 1995) or semi-controlled conditions for model testing in wheat grown under elevated CO<sub>2</sub> (e.g., Ewert et al., 1999; van Oijen and Ewert, 1999; Ewert and Porter, 2000; Ewert et al., 2002). Modification of growing conditions in controlled or semi-controlled environments were, however, reported to limit the

applicability of these data for model testing (Ewert et al., 2002). A number of factors related to the microclimatic conditions with chambers and restricted rooting volume, caused additional variation in growth and yield which could not be reproduced by models (van Oijen and Ewert, 1999; Ewert et al., 2002).

Further data have become available from FACE experiments during the last 6 years and have been used most extensively for model testing and inter-comparison under conditions of elevated CO<sub>2</sub>, together with the interactions of either water (e.g., ecosys, Grant et al., 1995; DEMETER, Kartschall et al., 1995; Wechsung et al., 1999; mC-Wheat, Tubiello et al., 1999; AFRC-WHEAT2, Sirius, LINTULCC2, Ewert et al., 2002) or N (AFRCWHEAT2, Sirius, and FASSET, Jamieson et al., 2000). The crop models tested using FACE data had different approaches to the modeling of both biophysical and agronomic variables, but all showed good agreement, under both ambient and elevated CO<sub>2</sub>, with a large number of observed variables such as time courses of phenology, above-ground biomass, LAI, in addition to final above-ground biomass and grain yield (Table 3). Both simple and complex approaches were able to capture the observed interactions of elevated CO<sub>2</sub> with both water and N (e.g., Jamieson et al., 2000; Ewert et al., 2002). For example, Fig. 3 shows that several models well captured the observed increase in the relative CO<sub>2</sub> effect on grain yield underwater-limited compared to well-watered conditions, provided some effects of elevated CO<sub>2</sub> on stomates and/or transpiration were included. Furthermore, as shown in Fig. 4, model inter-comparison studies underlined that differences among tested models were often larger than those computed between single models and observations.

A number of issues were identified in these studies in order to improve model evaluation and inter-comparison with data from elevated CO<sub>2</sub> studies (e.g., Ewert et al., 2002). Firstly, even in the case of high-quality datasets such as those developed under FACE, there was often a disconnection between the nature of collected data and the format ideally required for model testing. For example, while models need exact dates of pheno-

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Table 3

Selected experiments and treatments used to test crop model simulations of wheat responses to elevated CO<sub>2</sub>

Treatments		Model	Reference
CO <sub>2</sub>	Others		
<i>Free-air carbon dioxide experiments</i>			
Ambient, ambient × 1.5	Water (well-watered, water-stressed), 2 seasons	mC-Wheat	Tubiello et al. (1999)
		AFRCWHEAT2, LINTULCC2, Sirius	Ewert et al. (2002)
		DEMETER	Kartschall et al. (1995), Grossman-Clarke et al. (2001)
		ecosys	Grant, et al., (1995, 1999)
Ambient, ambient × 1.5	N (optimal, limited), 2 seasons	Sirius, AFRCWHEAT2, FASSET	Jamieson et al. (2000)
<i>Open-top chamber experiments</i>			
Ambient, ambient × 2	Ozone (ambient, ambient × 1.5), 3 seasons, 8 sites (across Europe)	AFRCWHEAT2-O3, LINTULCC	Ewert et al. (1999), Ewert and Porter (2000), van Oijen and Ewert (1999)
Ambient, ambient × 2	Water (well-watered, water-stressed), 2 seasons	AFRCWHEAT2, LINTULCC2	Ewert et al. (2002), Rodriguez et al. (2001)
<i>Controlled environment experiments</i>			
Ambient, ambient × 2	Temperature (ambient, ambient + 4 °C)	ARCWHEAT1	Mitchell et al. (1995)

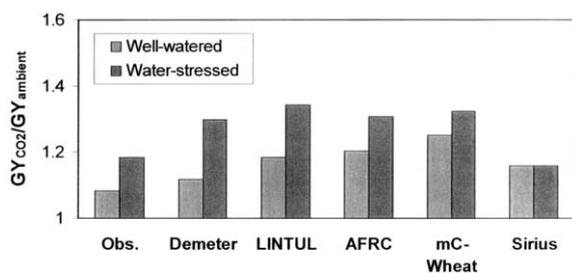


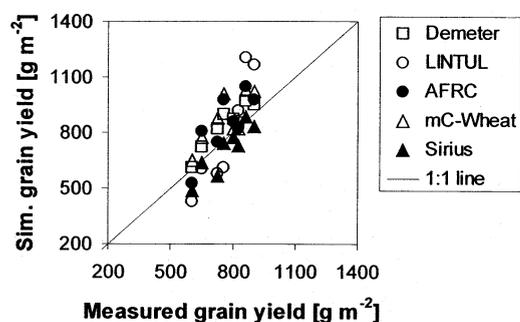
Fig. 3. Comparison of simulated relative CO<sub>2</sub> effects on grain yield (GY) from different crop models with observations (Obs.) from free-air CO<sub>2</sub> enrichment experiments in Maricopa, Arizona in 1992/93 and 1993/94. Relative effects were calculated from the data presented in Fig. 4.

prior to actual harvest. Secondly, differences in model structure and linkages among sub-components made it inherently difficult to thoroughly compare overall model performance. Thirdly, the existence of only a few field experiments available to date for evaluation studies and the use of a narrow range of elevated CO<sub>2</sub> concentrations (~ 550 μl/l), may have contributed to hide larger model differences than observed, ones that could become more apparent at elevated CO<sub>2</sub> concentrations and for a wider range of conditions. A few recent efforts have indeed attempted to extend the range of model development using observed datasets, elevated CO<sub>2</sub> and/or climate change scenarios (e.g., Paustian et al., 2000; Reilly et al., 2001; Ewert et al. 2002).

## 6. Recommendations and conclusions

Based on the reviewed material, we elaborate a set of recommendations of importance to the use of crop models for studies that involve elevated

logical development, pre-scheduled measurement campaigns were such that Zadoks growth stages had to be estimated in the field often in between key growth stages, and then interpolated to date phenological events. Additionally, most wheat models predicted exact grain maturity, while data collected in the field included mass decrease after maturity, due to physical and respiratory losses,



	RMSD (g m <sup>-2</sup> )				
	Measured	Demeter	LINTUL	AFRC	mC-Wheat
Demeter	82	-	-	-	-
LINTUL	185	191	-	-	-
AFRC	123	82	186	-	-
mC-Wheat	135	62	219	69	-
Sirius	84	141	187	151	188

Fig. 4. Comparison of simulated grain yield from different crop models with observed data from free-air CO<sub>2</sub> enrichment experiments in Maricopa, Arizona in 1992/93 and 1993/94. Simulations were taken from Tubiello et al. (1999), Grossman-Clarke et al. (2001), Ewert et al. (2002). Data refer to two [CO<sub>2</sub>] (ambient and 1.5 × ambient CO<sub>2</sub>) and two drought (well-watered and water-stressed) treatments. Models simulate [CO<sub>2</sub>] effects on assimilation with different detail in the order of DEMETER and LINTULCC2 (most detailed), AFRC-WHEAT2, mC-Wheat and Sirius. Simulations were compared using root mean squared deviated (RMSD) between observed and simulated and between simulated data. Note that RMSDs are often larger between models than between simulated and observed data.

CO<sub>2</sub> and/or climate change conditions. Firstly, those models that have been used extensively in climate change assessment studies but have not yet been sufficiently tested using the available field experimental data need to undergo renewed evaluation. Secondly, successful model testing under a restricted set of CO<sub>2</sub>, climate and management conditions does not warrant unlimited ability to perform equally well under an extended range of climate change and management conditions. To this end, model evaluation studies should also include, in addition to model testing against standard sets of experimental data, sensitivity simulations with a range of climate scenarios, management, and CO<sub>2</sub> concentrations. Thirdly, current field experiments on crop responses to CO<sub>2</sub>

should also address the effects of factors that are known to determine farm-level yield variability. Such experiments will provide better understanding of effects of soil quality, competition with weeds and pests and diseases on crop responses to CO<sub>2</sub> and will improve our ability to assess CO<sub>2</sub> impacts at larger scales.

However, not only spatial but temporal issues of model predictions require attention, such as reproducing inter-annual variability of yield over historical periods. For instance, point-level simulations from process-oriented crop models show larger inter-annual yield variations than evident from reported regional production data. Aside from key socio-economic issues that shape year-to-year agronomic decisions, factors like geographic heterogeneity, pests and diseases, and even mathematical methods used to calculate yield averages affect inter-annual yield variations in ways that models presently do not account for. Clearly, crop models need to be evaluated with multi-year datasets and the implication of models performances for predicting CO<sub>2</sub> effects under climate change needs to be assessed. In this respect, developments in agro-technology that have been a major determinant of changes in crop yields in the last century (e.g., Amthor, 1998), but have largely been ignored in current crop modeling, require particular attention.

Previous review studies have focused on describing specific aspects, mostly photosynthesis, important to modeling the effects of elevated CO<sub>2</sub> on crops. This review focused on process-level agronomic models capable of predicting harvest yield as a function of environmental and management factors. Our conclusions can be summarized as follows:

- i) Increase model testing under field conditions and elevated CO<sub>2</sub>;
- ii) Increase model inter-comparison, including sensitivity studies employing ranges of CO<sub>2</sub> concentration and climate change scenarios;
- iii) Focus on issues of temporal and spatial scale;
- iv) Clearly indicate limits of crop models used in climate change assessments.

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