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## Maize yield potential; critical processes and simulation modeling in a high-yielding environment

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### Abstract

Understanding processes of maize (*Zea mays* L.) growth and production of grain in high-yielding, irrigated conditions offers hope to understand yield potential in many other environments. In this study we investigated such processes at the plant level, and attempted to simulate maize yields at the field level and county level in the high yielding region of the High Plains of Texas. In addition, we used the normalized difference vegetation index (NDVI) from satellite data of year 2000 to update leaf area index for yield simulation in three counties. In the field study, we measured maize leaf area index (LAI), the fraction of photosynthetically active radiation intercepted (FIPAR), and the harvest index (HI) in irrigated plots near Dumas, Texas. The light extinction coefficient ( $k$ ) for Beer's law was calculated with the FIPAR and the LAI. The radiation use efficiency (RUE) was determined with sequential measurements of the fraction of photosynthetically active radiation (PAR) intercepted and biomass harvests. The RUE was 3.98 g of above-ground biomass per MJ of intercepted PAR in 1999 and 3.41 in 2000 for three sampling dates prior to silking. These values are 106 and 93% of the expected RUE values at the measured vapor pressure deficits, using a previously published response function. The mean  $k$  value was  $-0.46$  in 1999 and  $-0.47$  in 2000, similar to the expected value of  $-0.43$  reported in the literature for this row spacing. The mean HI measured in 2000 was 0.52, similar to values of 0.53 and 0.54 in the literature. Application of these parameters to maize simulation with the Agriculture Land Management Alternatives with Numerical Assess-

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31 ment Criteria (ALMANAC) model for 13 center pivot irrigated fields near Dumas in 1999  
32 provided simulations within  $1.0 \text{ Mg ha}^{-1}$  with a mean error of  $0.03 \text{ Mg ha}^{-1}$  and a mean  
33 square error of 0.10. For five years of grain yields reported for each of four counties in this  
34 region of Texas, ALMANAC simulations were within 5% of the mean measured yields. Intro-  
35 duction of PAR interception, based on the satellite-derived normalized difference vegetation  
36 index (NDVI), into ALMANAC resulted in slight increases in accuracy of yield prediction  
37 for two counties and a slight decrease in accuracy in one county for year 2000. Consistency  
38 in values of RUE,  $k$ , and HI in this study as compared with values reported in the literature  
39 will aid modelers simulating maize growth and grain yields in similar high-yielding, irrigated  
40 conditions.

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42 *Abbreviations:* ALMANAC, Agriculture Land Management Alternatives with Numerical Assessment  
43 Criteria; FIPAR, fraction of photosynthetically active radiation intercepted by plants; HI, harvest index;  
44 IPAR, photosynthetically active radiation intercepted by plants,  $\text{MJ plant}^{-1} \text{ d}^{-1}$ ;  $k$ , light extinction  
45 coefficient for Beer's law; LAI, leaf area index; NASS, National Agricultural Statistical Service; NDVI,  
46 normalized difference vegetation index; PAR, photosynthetically active radiation,  $\text{MJ m}^{-2} \text{ d}^{-1}$ ; RUE,  
47 radiation use efficiency, g per MJ intercepted photosynthetically active radiation

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## 48 1. Introduction

49 Accurate grain yield prediction by crop models requires realistic simulation of  
50 yield potential and realistic simulation of environmental reductions in yield. Accu-  
51 racy for yield potential requires accuracy in processes such as leaf area growth, light  
52 interception, biomass production, and partitioning of biomass into grain yield. The  
53 High Plains of Texas is an excellent region for such high-yielding maize research.  
54 County maize grain yield estimates there are the largest reported in the US (National  
55 Agricultural Statistical Service (NASS, 2001). Of the county grain yields for 1995–  
56 1999, maximum values for this region were  $9.9 \text{ Mg ha}^{-1}$  in Castro and Hansford  
57 Counties,  $10.7 \text{ Mg ha}^{-1}$  in Moore County, and  $11.2 \text{ Mg ha}^{-1}$  in Dallam County  
58 ( $186\text{--}212 \text{ bu acre}^{-1}$ ). Throughout this paper, grain yields in  $\text{Mg ha}^{-1}$  are dry weight  
59 and grain yields in  $\text{bu acre}^{-1}$  are with 15.5% moisture. For comparison, the highest  
60 county values for maize in 1995–1999 were  $8.4 \text{ Mg ha}^{-1}$  for Ohio,  $8.5 \text{ Mg ha}^{-1}$  for  
61 Indiana and Wisconsin,  $9.0 \text{ Mg ha}^{-1}$  for Idaho,  $9.1 \text{ Mg ha}^{-1}$  for Iowa, and  $9.4 \text{ Mg}$   
62  $\text{ha}^{-1}$  for Nebraska. Except for the Nebraska value, which was only for irrigated  
63 maize, these were for all the maize in a county.

64 Yield trials in four recent years in Moore County Texas had grain yield values  
65 even larger (Pietsch et al., 1996, 1997, 1998, 1999). The mean yield of the five highest  
66 yielding hybrids each year was  $13.3 \text{ Mg ha}^{-1}$  ( $250 \text{ bu acre}^{-1}$ ). This is similar to the  
67 mean of  $13.0 \text{ Mg ha}^{-1}$  ( $244 \text{ bu acre}^{-1}$ ) for the four highest yielding treatments in  
68 an irrigated experiment in Argentina in 1991/1992 (Otegui et al., 1995). In this Ar-  
69 gentine experiment, measured grain yields were larger in the previous year, with a  
70 mean of  $16.0 \text{ Mg ha}^{-1}$  ( $302 \text{ bu acre}^{-1}$ ) for the highest four treatments. These higher  
71 yields were attributed to higher incident solar radiation after silking in that year.

72 Studies with high-yielding, irrigated maize in this region of Texas have raised  
73 questions which have made field measurements of crop development in such condi-  
74 tions invaluable. Attempts at simulating irrigated maize in the High Plains near Du-  
75 mas, Texas using five years of measured yield trial results showed mean simulations  
76 with the ALMANAC model (Kiniry et al., 1992) only 90% of mean measured grain  
77 yields (Kiniry and Bockholt, 1998). Field determination of the maximum leaf area  
78 index (LAI), the light extinction coefficient ( $k$ ) for Beer's law (Monsi and Saeki,  
79 1953), radiation use efficiency (RUE), and the harvest index (HI) under such condi-  
80 tions could greatly improve efforts to simulate maize yield potential in such environ-  
81 ments.

82 The National Oceanic and Atmospheric Administration (NOAA) satellite pro-  
83 vides real-time information that can augment simulation models by improving their  
84 simulation of interception of PAR. The advanced very high-resolution radiometer  
85 (AVHRR) sensor onboard the NOAA-14 satellite provides daily data at 1 km $\times$ 1  
86 km spatial resolution. The NDVI from AVHRR data is often correlated with the  
87 photosynthetically active radiation (PAR) interception during vegetation growth  
88 (Coops et al., 1998; Sakai et al., 1997; Baez-Gonzalez et al., 2002). The intercepted  
89 PAR is strongly related to the accumulation of biomass or primary production (Ki-  
90 niry et al., 1989).

91 Our first objective of the present study was to investigate whether previously pub-  
92 lished values for LAI,  $k$ , RUE, and HI need to be revised for such a high-yielding  
93 environment. We derived values for LAI and  $k$  on different dates and RUE during  
94 the season for two maize growing seasons in the high-yielding conditions of the High  
95 Plains near Dumas, Texas. In addition, HI values were measured in the second year.  
96 The ALMANAC model was then applied for one year to 13 center pivot irrigated  
97 fields with differential irrigation amounts on the same farm as where the measure-  
98 ments were taken. Next, the ALMANAC model was tested using five years of irri-  
99 gated maize grain yield averages for four counties in this region using plant  
100 parameters derived from this field study. Finally, the ALMANAC model, with up-  
101 dated LAI based on NDVI values of year 2000, was applied to three counties in  
102 the region. The objective of using satellite image data was to apply weekly maximum  
103 NDVI values to estimate daily fraction of intercepted PAR for irrigated maize fields  
104 in three counties in this region and to assess the accuracy of maize yield simulation  
105 with such inputs.

## 106 2. Materials and methods

### 107 2.1. Dumas field study

108 Pioneer maize hybrid 31A13 (Pioneer Hi-Bred Int., Inc., Amarillo, TX) was  
109 planted on 13 May, 1999 and 15 April, 2000 on a Sherm silty clay loam (fine, mixed,  
110 mesic Torric Paleustolls) on the farm of Mr. Harold Grail near Dumas, Texas  
111 (35°52'N, 101°58'W; 1114 m above sea level) (Fig. 1). This soil is 2.0 m deep and  
112 has 300 mm of plant-available-water at field capacity. For several years Texas

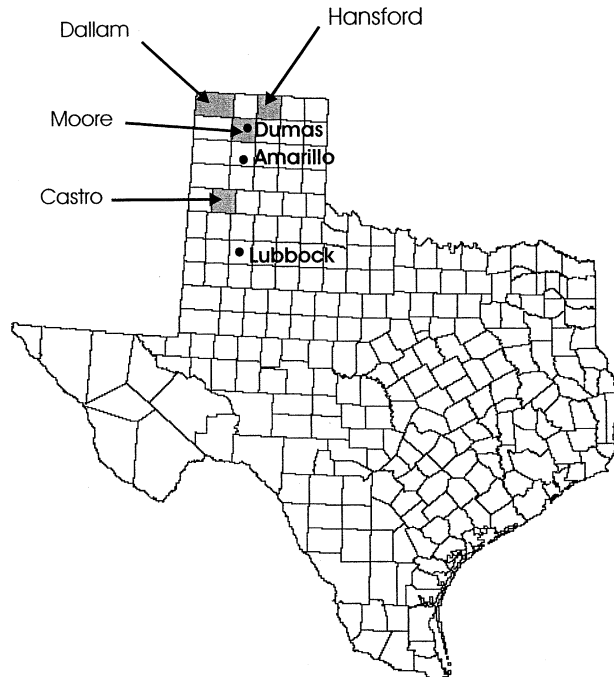


Fig. 1. Map of the region for field measurements and simulations of high yielding, irrigated maize.

113 A&M University has conducted yield trials for maize and sorghum (*Sorghum bicolor*  
114 L. (Moench)) on this farm. Measured planting density in both years of our study was  
115 84,000 plants ha<sup>-1</sup> in 0.76 m rows. Plots were fertilized with 308 kg N ha<sup>-1</sup>, 90 kg  
116 P<sub>2</sub>O<sub>5</sub> ha<sup>-1</sup> and 2 kg K<sub>2</sub>O ha<sup>-1</sup> each year before planting. In both years the plots were  
117 sufficiently irrigated to insure no drought limitations to crop growth, using a center  
118 pivot irrigation system.

119 On four dates prior to grain filling each year, measurements of fraction of PAR  
120 intercepted, LAI, and plant dry weights were taken. We measured PAR interception  
121 during each growing season with a 0.8-m-long Sunfleck Ceptometer (Decagon, Pull-  
122 man, Washington, US). In each of 10 areas of the field, we took a series of measure-  
123 ments in rapid succession, consisting of 10 PAR measurements above the canopy, 10  
124 below the canopy, and 10 more above the canopy. The fraction of PAR intercepted  
125 was calculated with the mean of the above-canopy measurements and the mean of  
126 the below-canopy measurements. While taking the readings below the canopy, the  
127 light meter was moved across the plant rows. Measurements were taken between  
128 1020 and 1200 h during times with relative stable incident solar radiation (without  
129 intermittent clouds).

130 Destructive samples consisted of all the plants in five, 1.0 m lengths of row. All the  
131 plants of a sample and one representative plant from each sample were weighed  
132 fresh. Each representative plant was then overnight-mailed to Temple, Texas where

133 leaf area was measured with a LiCor LI-3100 leaf area meter (LiCor Inc., Lincoln,  
134 Nebraska), the plant dried, and the plant weighed again. Weights of the plants were  
135 measured after drying in a forced-air drying oven at 70 °C until the weight stabilized.  
136 Leaf area of the entire sample was calculated from the leaf area of the one plant and  
137 the ratio of the total fresh weight of all plants divided by the fresh weight of the one  
138 plant. Similarly, dry weights of an entire 1.0 m sample was calculated by the dry  
139 weight of one plant and the ratio described above.

140 With these techniques, we derived values for LAI, above-ground dry weight per  
141 square meter ground area, and intercepted PAR. The light extinction coefficient  
142 ( $k$ ) for Beer's law (Monsi and Saeki, 1953) was calculated from the fraction of  
143 PAR intercepted (FIPAR) and the LAI. Values for  $k$  were calculated for each har-  
144 vest date as

$$k = [\ln(1 - \text{FIPAR})]/\text{LAI}. \quad (1)$$

146 Daily intercepted PAR was determined by linearly interpolating fraction inter-  
147 cepted PAR between measurement dates, using the incident total solar radiation  
148 each day. PAR was assumed to be 45% of the total solar radiation (Monteith,  
149 1965; Meek et al., 1984). Daily intercepted PAR was summed during the season.  
150 RUE was calculated from the slope of the above-ground dry weight ( $\text{g m}^{-2}$ ) as a  
151 function of summed intercepted PAR ( $\text{MJ m}^{-2}$ ). Thus, the slope was in units of g  
152 of dry weight per MJ of intercepted PAR.

153 At maturity, on 4 September 2000, plants were harvested, mailed to Temple as de-  
154 scribed above, dried as before, and weighed. The harvest index was the dry weight of  
155 the grain divided by the dry weight of the entire above-ground plant.

## 156 2.2. Description of the ALMANAC crop model

157 The model used in this project was ALMANAC (Kiniry et al., 1992). This model  
158 simulates the soil water balance, the soil and plant nutrient balance, and interception  
159 of solar radiation. This model includes subroutines and function from the erosion-  
160 productivity impact calculator (EPIC) model (Williams et al., 1984, 1989) with added  
161 details for plant growth. The model has a daily time step. It simulates plant growth  
162 for a wide range of species and is implemented easily.

163 ALMANAC simulates plant growth using LAI. The model simulates population  
164 density by adjusting potential LAI. Maximum potential simulated LAI at high plant-  
165 ing densities is 6.0 for maize.

166 The model uses a modified HI approach to simulating grain yield. For maize, HI  
167 reported for temperate regions usually were between 0.46 and 0.58 (Kiniry et al.,  
168 1997). In a study with several sites in Texas, Kiniry and Bockholt (1998) used a  
169 HI value of 0.53.

## 170 2.3. Dumas 1999 simulation study

171 On the same farm as where the field measurements were made, we used the AL-  
172 MANAC model to simulate maize grain yield of 13 irrigated fields. The fields varied

173 in size from 48 to 198 ha. The runoff curve number was set to 60 for these fields and  
174 for simulations in the following section to avoid runoff of irrigation water. Amounts  
175 of irrigation each field received during the growing season varied from 430 to 809  
176 mm. All fields were irrigated with center pivot systems.

177 We assumed all fields had the same soil as described above. The recorded irriga-  
178 tion dates and amounts for each field were used as inputs. Measured grain yields  
179 were compared with simulated yields.

#### 180 2.4. Four county yield simulation study

181 Four counties in the Texas High Plains were simulated for 1995–1999 (Fig. 1).  
182 Measured maize grain yields were the yield averages for each county for each year  
183 as reported by NASS (2001). These counties had large areas of irrigated maize  
184 and differed in grain yield potential. Counties and their associated weather stations  
185 were Hansford County with Morse, Texas (36°04'N, 101°29'W; 969 m above sea le-  
186 vel); Moore County with Etter, Texas (35°52'N, 101°58'W; 1114 m above sea level);  
187 Castro County with Dimmitt, Texas (34°36'N, 102°19'W; 1173 m above sea level);  
188 and Dallam County with Dalhart, Texas (36°05'N, 102°29'W; 1220 m above sea le-  
189 vel).

190 In each case, we simulated the major soil for irrigated production in the county.  
191 Hansford and Castro Counties were simulated with Pullman clay loam (fine, mixed,  
192 thermic Torrertic Paleustoll), Moore County was simulated with a Sherm silty clay  
193 loam (fine, mixed, mesic Torrertic Paleustoll), and Dallam County was simulated  
194 with a Dallam fine sandy loam (fine-loamy, mixed, mesic Ardic Paleustalf).

#### 195 2.5. Using NDVI data to update intercepted PAR in ALMANAC

196 AVHRR high-resolution picture transmission (HRPT) data of the maize growing  
197 season between April and September 2000 were downlinked daily from the NOAA-  
198 14 satellite to the receiving station located at the Blackland Research and Extension  
199 Center in Temple, Texas. The raw visible (VIS) and near-infrared (NIR) data were  
200 calibrated to top-of-atmosphere (TOA) reflectance factor. A simplified method for  
201 atmosphere correction (SMAC) developed by Rahman and Dedieu (1994) was ap-  
202 plied to compute surface reflectance. The NDVI values were computed from surface  
203 reflectance (after atmosphere correction). A cloud detection method developed by  
204 Chen et al. (2002) for the state of Texas was applied for cloud removal. Cloud-con-  
205 taminated pixels were assigned a specific value to differentiate them from clear-sky  
206 pixels. Weekly NDVI composites were built from cloud-screened data based on  
207 the method of maximum value compositing (Holben, 1986 and Chen et al., 2003).

208 FIPAR was measured on 9 June, 22 June, 6 July, and 20 July, 2000 in the study  
209 site near Dumas. Maize silking occurred near 6 July, 2000. FIPAR decreased grad-  
210 ually after silking, because leaves started withering and turning yellow. Therefore,  
211 the weekly maximum NDVI values of the first three sampled dates were correlated  
212 to the sampled FIPAR. The weekly FIPAR values were simulated using weekly max-  
213 imum NDVI values based on the sampled correlation. The simulated weekly FIPAR

214 values were determined as the FIPAR of the end of week date. The daily FIPAR val-  
 215 ues in a week were linearly interpolated. The daily FIPAR values based on the NDVI  
 216 replaced the daily values simulated by ALMANAC. In addition, the sampled corre-  
 217 lation was applied to the other high-yield maize sites in Castro County, Dallam  
 218 County and Moore County to estimate their daily FIPAR and to further simulate  
 219 maize yield using the ALMANAC model. Hansford County was not included be-  
 220 cause it lacked sufficiently large, uniform irrigated areas where NDVI values could  
 221 be calculated correctly.

222 **3. Results**

223 *3.1. Dumas field study*

224 Values for LAI and  $k$  were similar to potential LAI of 6.0 used to simulate this site  
 225 previously (Kiniry and Bockholt, 1998) and the  $k$  value of  $-0.43$  predicted for a 0.76  
 226 m row spacing of maize (Flénet et al., 1996). Maximum LAI values were 7.9 in 1999  
 227 and 5.2 in 2000 (Figs. 2 and 3). Values for  $k$  varied somewhat during the season but  
 228 the annual means were within 10% of the expected value. The mean  $k$  was  $-0.46$  in  
 229 1999 and was  $-0.47$  in 2000.

230 RUE values calculated with the first three sample dates showed values similar to  
 231 RUE predicted for the vapor pressure deficit (VPD) conditions (Figs. 2 and 3). Val-  
 232 ues for RUE using these sampling dates were 6–7% different from expected values  
 233 from previous studies (Stockle and Kiniry, 1990; Kiniry et al., 1998). The differences  
 234 from expected were much less than the SEs of the RUE values. The mean VPD dur-

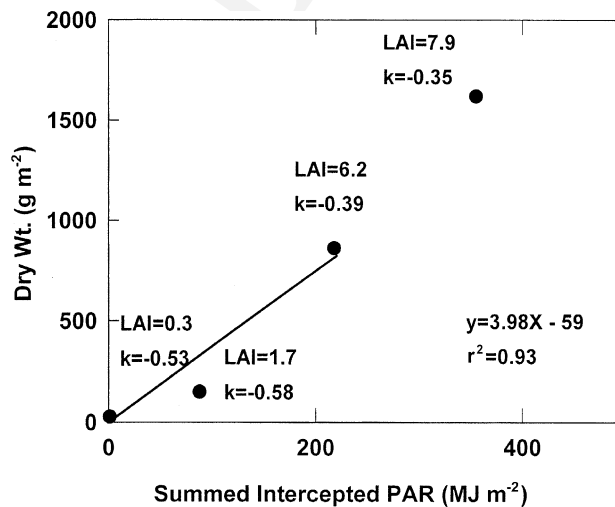


Fig. 2. In 1999, the relation between above-ground dry weight and cumulative intercepted photosynthetically active radiation (IPAR) for maize near Dumas, Texas.

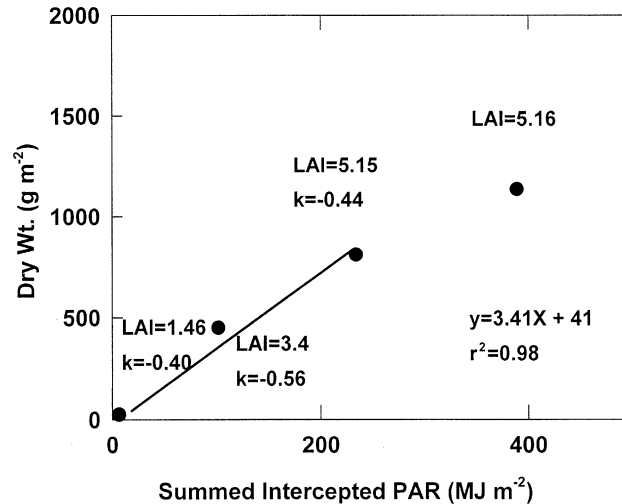


Fig. 3. In 2000, the relation between above-ground dry weight and cumulative intercepted photosynthetically active radiation (IPAR) for maize near Dumas, Texas.

235 ing the measurements in 1999 was 1.10 kPa. The expected RUE at this VPD is 3.75 g  
 236 of biomass per MJ of intercepted PAR. The measured value of 3.98 ( $\pm$ SE of 1.11)  
 237 was 6% greater than this. The mean VPD during the period of the first three sam-  
 238 plings in 2000 was 1.26. Expected RUE at this VPD is 3.65. The measured RUE  
 239 was  $3.41 \pm 0.48$  g MJ<sup>-1</sup>, 7% less than expected. Thus, these data support the previ-  
 240 ously reported values of RUE at such low VPD values.

241 Using data from all four sampling dates showed RUE values more divergent from  
 242 expected values. In 1999, the fourth sampling was at silking, whereas in 2000, silking  
 243 occurred near the third sampling. Mean VPD values over the interval for the four  
 244 samplings were 1.11 kPa in 1999 and 1.44 in 2000. Predicted RUE for these values  
 245 are 3.75 in 1999 and 3.53 in 2000. Values from the measured data for the four dates  
 246 were  $4.69 \pm 0.57$  in 1999 (125% of predicted) and  $2.83 \pm 0.34$  in 2000 (80% of ex-  
 247 pected).

248 Grain yields measured in 2000 were high, as expected. The mean was 12.3 Mg  
 249 ha<sup>-1</sup> with a SD of 1.9 ( $232 \pm 36$  bu acre<sup>-1</sup>). The mean harvest index was 0.52.

### 250 3.2. Dumas 1999 simulation study

251 The ALMANAC model realistically simulated the variability in grain yields  
 252 among fields with different irrigation treatments (Fig. 4). The mean error (simulated  
 253 minus measured) was 0.03 Mg ha<sup>-1</sup> and the root mean square error was 0.32 Mg  
 254 ha<sup>-1</sup>. The model realistically simulated the two highest yielding fields, the two lowest  
 255 yielding fields, and the intermediate fields. The regression slope was 0.94 ( $\pm$ SE of  
 256 0.15), the  $y$ -intercept was 0.72, and the  $r^2$  was 0.78. The regression line was close  
 257 to the 1:1 line. Thus, the model shows promise as a tool for effectively simulating  
 258 grain yield responses to different center pivot irrigation amounts.



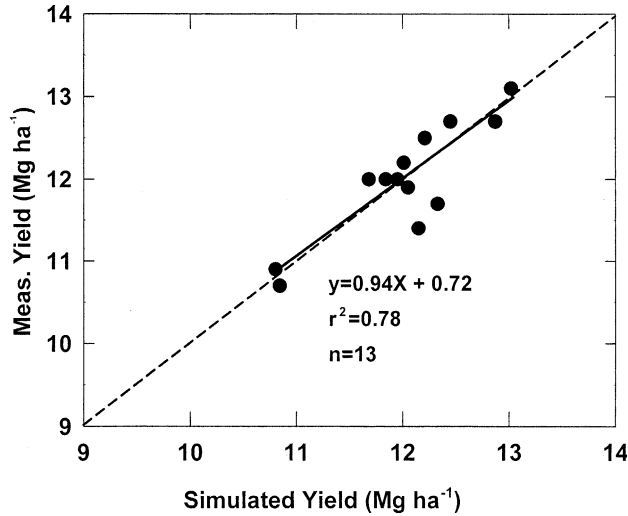


Fig. 4. Maize simulations with ALMANAC on fourteen fields in 1999 near Dumas, Texas. The darker solid line is the regression line with all data, the lighter solid line is the regression line, and the dashed line is the 1:1 line through the origin.

259 3.3. Four county yield simulation study

260 The yield trend across locations was reasonably simulated by ALMANAC  
261 (Fig. 5). While ALMANAC's mean simulated grain yields for the four locations were

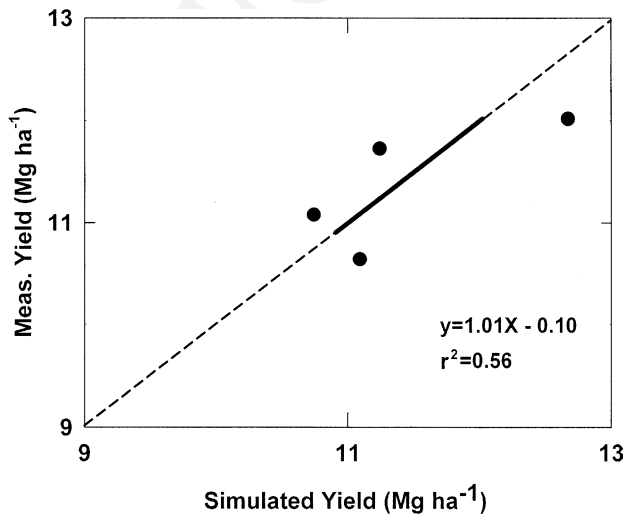


Fig. 5. Mean maize simulated grain yields with ALMANAC on four counties in the Texas High Plains for 1995–1999. Measured grain yields are from the county yield estimates for the years. The solid line is the regression line and the dashed line is the 1:1 line through the origin.

262 not significantly ( $\alpha = 0.05$ ) related to the mean measured grain yields, the slope was  
263 close to 1.0 and the  $y$ -intercept was small. The slope of 1.01 ( $\pm$ SE of 0.64) was not  
264 significantly different from 1.0 and the  $y$ -intercept of  $-0.10$  ( $\pm$ SE of 0.69) was not sig-  
265 nificantly different from 0.0. The mean simulated grain yields for the five years were  
266 5% too large for the highest yielding site (Dalhart), 4% too large for the lowest yield-  
267 ing site (Morse), 4% too small for Etter, and 3% too small for Dimmitt.

268 The model did not significantly account for year-to-year variation in grain yields  
269 within each of these locations, as indicated by slopes that were not significant. For  
270 Morse, the slope was 0.42 and the  $r^2$  was 0.07. For Etter, the slope was  $-0.3$  and  
271  $r^2$  was 0.10. For Dimmitt, the slope was 0.12 and  $r^2$  was 0.002. Finally, for Dalhart  
272 the slope was  $-0.3$  and  $r^2$  was 0.33. Thus the mean model simulations for each site  
273 were within 5% of the mean reported grain yield for the county, but something not  
274 simulated by the model was causing the variability among years in the reported  
275 county grain yields.

#### 276 3.4. Using NDVI data to update LAI in ALMANAC

277 ALMANAC simulations in 2000 were improved with the updated fraction inter-  
278 cepted PAR based on satellite data for two of the three counties considered. County  
279 yields reported in 2000 were 10.44 Mg ha<sup>-1</sup> for Dallam County, 10.51 for Moore  
280 County, and 10.80 for Castro County. The simulations without NDVI data were  
281 12.01 (115% of measured) for Dallam, 11.72 (111% of measured) for Moore, and  
282 12.90 (119% of measured) for Castro. After including the NDVI data to improve in-  
283 tercepted PAR values, simulated yields were 12.19 (117% of measured), 10.84 (103%  
284 of measured) and 12.41 (115% of measured), respectively. Thus, using satellite data  
285 to update ALMANAC's simulation in intercepted PAR provided a more accurate  
286 simulation for two of the three counties for yield simulation.

#### 287 4. Discussion

288 To accurately simulate maize grain yields in diverse environments, crop models  
289 must realistically simulate key processes of plant growth and of yield production un-  
290 der non-limiting conditions. The region of this study arguably produces the highest  
291 maize grain yields in the US. Previously described values for LAI,  $k$  for Beer's law,  
292 RUE, and HI appeared reasonable for these conditions and offer promise for accu-  
293 rate grain yield simulation in high yielding conditions.

294 The measured values for LAI,  $k$ , RUE, and HI reported herein support previously  
295 published values and provide guidance for future simulation modeling in high-yield-  
296 ing environments. The values for LAI were similar to those used by Kiniry and  
297 Bockholt (1998) when simulating grain yields near Dumas. The values of  $k$  were sim-  
298 ilar to those reported by Flénet et al. (1996) for the row spacing used. The harvest  
299 index value was also similar to the HI used by Kiniry and Bockholt (1998).

300 The measured radiation use efficiency values add credence to the VPD response  
301 published by Stockle and Kiniry (1990). This function predicted RUE values in these

302 high yielding conditions that were within 7% of measured, just as it had provided ac-  
303 curate maize simulations under drought stress conditions in central and southern  
304 Texas (Xie et al., 2001). The function improves the simulation of biomass and grain  
305 yields over a wide range of environments.

306 By using parameter values derived in this study, the ALMANAC model realisti-  
307 cally simulated grain yields with different amounts of irrigation applied by center  
308 pivots. Thus the model, with these parameters, can be a valuable tool for simulating  
309 grain yield with variable irrigation amounts.

310 The ALMANAC model was capable of simulating differences among counties in  
311 mean county yields, but could not account for factors causing year to year variability  
312 in each county. Accurate simulation of the five-year means illustrated the model's  
313 ability to simulate maize yield potential. However, the model's inability to simulate  
314 individual year yields needs further study. Under such irrigated conditions, it ap-  
315 peared that factors contributing to year to year variability are not adequately de-  
316 scribed by the model. Future research could involve investigation of these factors  
317 in the region. It appears that such a model based on incident solar radiation, with  
318 yield reductions due to soil moisture deficiency and soil nutrient deficiency, may  
319 not be sensitive to some yield determining factors in these high yielding conditions.

320 The improvements in ALMANAC's simulated yields as a result of the NDVI in-  
321 formation offer some hope for improving large area yield simulations. As more such  
322 satellite data becomes available, further model testing will be valuable. It will be es-  
323 pecially important to test this model with independent data for several years.

### 324 Acknowledgements

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329 Bushland, Texas for providing weather data.

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