

Simulating maize yield in sub-tropical conditions of southern Brazil using Glam model

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Abstract – The objective of this work was to evaluate the feasibility of simulating maize yield in a sub-tropical region of southern Brazil using the general large area model (Glam). A 16-year time series of daily weather data were used. The model was adjusted and tested as an alternative for simulating maize yield at small and large spatial scales. Simulated and observed grain yields were highly correlated (r above 0.8; $p < 0.01$) at large scales (greater than 100,000 km²), with variable and mostly lower correlations (r from 0.65 to 0.87; $p < 0.1$) at small spatial scales (lower than 10,000 km²). Large area models can contribute to monitoring or forecasting regional patterns of variability in maize production in the region, providing a basis for agricultural decision making, and Glam-Maize is one of the alternatives.

Index terms: *Zea mays*, crop modeling, crop parameters, crop-weather relationships.

Simulação do rendimento de milho em condições subtropicais do Sul do Brasil por meio do modelo Glam

Resumo – O objetivo deste trabalho foi avaliar a viabilidade de se estimar a produção de milho numa região subtropical do Sul do Brasil por meio do “general large area model” (Glam). Foi utilizada uma série de 16 anos de dados meteorológicos diários. O modelo foi ajustado e testado como alternativa para simular rendimentos de milho em pequena e grande escala espacial. Os rendimentos de milho simulados e observados estiveram altamente correlacionados (R acima de 0,8; $p < 0,01$) em grande escala (mais de 100.000 km²), e apresentaram correlações variáveis e geralmente inferiores (R de 0,65 a 0,87; $p < 0,1$) em pequena escala (menos de 10.000 km²). Modelos de grande escala podem contribuir para monitorar ou prever padrões de variabilidade na produção de milho na região, o que fornece uma base para tomadas de decisão, e o Glam-Maize é uma alternativa.

Termos para indexação: *Zea mays*, modelagem, parâmetros de cultivo, relações clima-cultivo.

Introduction

The sub-tropical region of South America is responsible for the majority of soybean, wheat, maize, rice, coffee, and sugarcane production in Latin America. With the exception of rice, these crops are generally conducted in rainfed conditions, and the oscillation of productivity is usually intense due to irregular rainfall distribution. Therefore, simulation can be a useful tool for predicting crop productivity a season or more ahead (Coelho & Costa, 2010) and to investigate options for crop management (Challinor et al., 2005).

Many simulation studies of grain crop yield have been done in different sub-tropical regions of South

America. The crop models that integrate DSSAT (including Ceres for maize) have been largely tested for this issue (Cardoso et al., 2004; Travasso et al., 2006; Mercau et al., 2007; Tojo Soler et al., 2007). The Wang & Engel (1998) model was used by Streck et al. (2012) to simulate the developmental cycle of maize crops. In southern Brazil, the penalization model of Jensen (1968), based on empirical relationships between grain production and water conditions, was also tested for maize crops (Matzenauer et al., 1995; Mello et al., 2003). Doorenbos & Kassam (1979) crop yield models have been used to estimate potential and actual yields, as well as to test the sensitivity of maize genotypes to water deficits (Andriolli &

Sentelhas, 2009). Several observational studies on soil-plant-atmosphere relations have also been done in recent decades in southern Brazil (Müller et al., 2005; Bergamaschi et al., 2006). These modeling and observational researches have focused on spatial scales ranging from field (smaller than 10,000 km²) to larger areas (greater than 100,000 km²).

The general large area model (Glam) is a large area process-based crop model used for simulating yield of annual crops (Challinor et al., 2004). It has a low input data requirement for large-scale applications and can directly accept large-scale climate information from global and regional climate models. The model has been previously used to simulate groundnut (Challinor et al., 2004; Osborne et al., 2007) and wheat production (Sanai et al., 2007). The main hypothesis of this study is that the Glam model is robust enough to simulate grain yields at large areas, for different crops and climate conditions.

The objective of this work was to evaluate the feasibility of simulating maize yields in a sub-tropical region of Southern Brazil using Glam with a 16-year time series of daily weather data.

Materials and Methods

The study was carried out in the South of Brazil, in the state of Rio Grande do Sul (27.2° to 29.8°S and 51.2° to 56.0°W, at an altitude range of 0–1,380 m) (Figure 1). The state is one of the main maize producers in the country. During the study period (1998–2005), it was ranked as the third main producer, but changed to the sixth position in 2004–2006 (Companhia Nacional de Abastecimento, 2006). Maize (*Zea mays* L.) yield was simulated at three spatial scales: the main producer region (northern-northwestern zone of the state), which accounted for 83% of the state maize production during the studied period (Instituto Brasileiro de Geografia e Estatística, 2006); 11 micro-regions within the producer region; and one municipality within each micro-region (Figure 1; Table 1).

Maize yield production data from 1990 to 2005 (16 years) were collected from the Brazilian Institute of Geography and Statistics (Instituto Brasileiro de Geografia e Estatística, IBGE). Most maize crops in the state of Rio Grande do Sul are sown from September to the beginning of November and harvested from January to the beginning of March. The average sowing date for each municipality, obtained from

the data of the official extension service in the state (Emater, RS), represents the average date by which at least 50% of the maize crops were sown between 2002 and 2005. Once the sowing data was known, the mean date of tasseling, milky and physiological maturity was obtained from a set of experimental data (Matzenauer et al., 1995; Müller et al., 2005; Bergamaschi et al., 2006) for short season maize hybrids (Table 1).

Daily rainfall, global solar radiation, relative humidity, minimum and maximum air temperatures were taken in each of the 11 municipalities from 1990 to 2005. The same database was used for the municipality and the respective micro-region (Figure 1). Weather data for the main producer region was the mean value of the 11 weather data stations. Weather stations belonged to the following official networks: Fundação Estadual de Pesquisa Agropecuária, Fepagro (in the municipalities of Cruz Alta, Erechim, Ijuí, Júlio de Castilhos, Santa Rosa, São Borja, Taquari, and Veranópolis) and Instituto Nacional de Meteorologia, Inmet (in the municipalities of Iraí, Passo Fundo, and São Luiz Gonzaga). Water retention data were taken from soil surveys (Dedececk, 197; Beltrame et al., 1979) of spatial units normally larger than one municipality. The prevailing soil type was used as the representative soil for each micro-region. Average values of the 11 soil coefficients were calculated in order to provide a mean soil water condition of the whole producer region for: saturation, field capacity, and permanent wilting point.

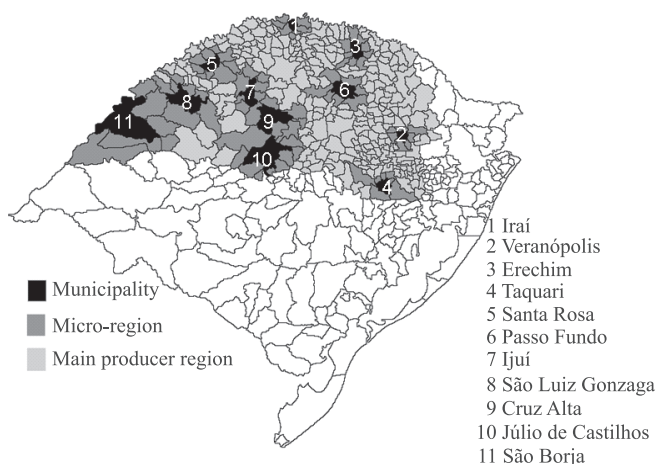


Figure 1. Map of the state of Rio Grande do Sul, Brazil, showing the main maize producer region, the 11 municipalities, and their respective micro-regions.

The Glam processes and methodology were kept similar to the original framework, described by Challinor et al. (2004). A field experimental dataset for maize was used to define new parameter values for Glam-Maize in the region (Table 2). Great part of the data was obtained from field experiments carried

Table 1. Area of maize cultivation, average dates of sowing and phenological stages for each municipality, micro-region, and main producer region.

Municipality	Area (km ²)		Sowing date ⁽¹⁾	Dates of the phenological stages ⁽²⁾			Period (days) from sowing to	
	Municipality	Micro-region		Tasseling	Milky maturity	Physiological maturity	Tasseling	Physiological maturity
São Borja	3,627	12,986	12 Sep.	30 Nov.	30 Dec.	30 Jan.	79	140
São Luiz Gonzaga	1,292	7,860	12 Sep.	30 Nov.	30 Dec.	30 Jan.	79	140
Santa Rosa	489	3,336	22 Sep.	3 Dec.	2 Jan.	31 Jan.	72	131
Ijuí	635	4,058	30 Sep.	9 Dec.	8 Jan.	8 Feb.	70	131
Iraí	200	1,393	20 Sep.	1 Dec.	31 Dec.	28 Jan.	72	130
Júlio de Castilhos	1,857	8,274	5 Oct.	10 Dec.	9 Jan.	10 Feb.	66	138
Cruz Alta	1,358	8,045	5 Oct.	10 Dec.	9 Jan.	10 Feb.	66	138
Passo Fundo	758	3,663	10 Oct.	17 Dec.	16 Jan.	7 Mar.	68	148
Erechim	426	2,349	10 Oct.	19 Dec.	18 Jan.	7 Mar.	70	148
Taquari	346	2,868	15 Oct.	22 Dec.	21 Jan.	22 Feb.	68	130
Veranópolis	276	1,469	21 Oct.	30 Dec.	29 Jan.	10 Mar.	70	140
Region	109,210	109,210	01 Oct.	14 Dec.	12 Jan.	12 Feb.	75	135

⁽¹⁾Of at least 50% of the maize crops, according to the official extension service in the state (Emater, RS). ⁽²⁾Mean date of tasseling, milky and physiological maturity for short season maize hybrids (Matzenauer et al., 1995; Müller et al., 2005; Bergamaschi et al., 2006).

Table 2. Parameters and specific values used in the Glam model for simulating maize growth and yield in sub-tropical conditions of southern Brazil.

Parameters	Values	Source
		Growth and development
Days from sowing to plant emergence	8 days	Müller et al. (2005)
Cardinal temperatures		
Base, optimum ⁽¹⁾ , and maximum	8°C, 32°C (28–34), and 40°C	Kiniry (1991), Birch et al. (2003), Streck et al. (2009, 2012)
Thermal requirement		
Stages 0, 1, 2, and 3	510, 580, 390, and 510 °C-day	Müller et al. (2005)
Leaf area index (L) rate (dL/dt)		
Stages 0, 1, 2, and 3	0.025, 0.173, -0.017, and -0.050 per day	Müller et al. (2005), Bergamaschi et al. (2006)
Root growth (dLv/dL)	1.0 km cm ⁻¹ m ⁻²	Adapted from Qin et al. (2006)
Root length density	0.38 cm cm ⁻³	Qin et al. (2006)
Harvest index (dH/dt)	0.0071 per day	Müller (2001), Müller & Bergamaschi (2005)
Soil water fraction threshold for L reduction	0.7	Bergamaschi et al (2006)
		Evaporation and transpiration
Critical L for radiation interception	2.7	Bergamaschi et al. (2001), Radin et al. (2003)
Maximum transpiration	0.63 cm per day	Bergamaschi et al. (2001), Radin et al. (2003)
Transpiration efficiency	3.33 Pa	Bergamaschi et al. (2001), Radin et al. (2003), Müller & Bergamaschi (2005)
Maximum transpiration efficiency	4.7 Pa	Bergamaschi et al. (2001), Radin et al. (2003), Müller & Bergamaschi (2005)
Light extinction coefficient	0.46	Birch et al. (2003), Müller & Bergamaschi (2005)
		Soil parameters and miscellaneous
Field capacity ⁽²⁾	0.45 (0.25–0.51)	
Permanent wilting point	0.30 (0.17–0.36)	Dedecek (1974), Beltrame et al. (1979)
Saturation	0.51 (0.33–0.59)	
Soil water extraction front velocity	1.66 cm per day	Müller (2001)
Soil water fraction for damage to anthesis	0.30	Bergamaschi et al. (2006)
Yield gap parameter	0.4–1.0 (varies spatially)	Estimated by minimizing RMSE after yield calculation

⁽¹⁾Data in parenthesis represent the range among authors. The parameter used in modeling was 32°C. ⁽²⁾Field capacity at matrix potential of -0.006MPa (72 hours of drainage after saturation). Data in parenthesis represent the variability among different soil types, in the region under study

out in the state of Rio Grande do Sul in the 1980s and 1990s. Other parameters were obtained from the literature, with a preference for the most suitable to the regional cropping system (Table 2). All parameter values were kept constant across the study region, except for sowing dates, yield gap parameter, and soil water storage capacities. Sowing dates are given in Table 1. The yield gap parameter (YGP) accounts for the impacts on yield due to factors other than weather (i.e., pests, diseases, and management factors, which reduce yield by an amount referred to as the yield gap). The YGP was calculated by minimizing the root mean square error found between simulation and observation dataset, as proposed by Challinor et al. (2004).

The maize crop cycle was divided into four stages (Table 2): stage 0, juvenile stage, from sowing to four fully expanded leaves; stage 1, rapid vegetative growth, from the end of stage 0 to tasseling; stage 2, flowering, from tasseling to the beginning of grain-filling; and stage 3, grain-filling, from the end of stage 2 to physiological maturity. Plant emergence was assumed as occurring eight days from sowing. The duration of all other stages was determined using thermal time (degree-days), calculated from the mean air temperature and cardinal temperatures (base, maximum, and optimum temperatures, Table 2). The thermal time chosen for each crop stage was in accordance with the experimental results for short season maize hybrids in the state. Those hybrids were assumed as photoperiod insensitive in the region.

The leaf area index growth rate (dL/dt) was based on a segmented linear model and varied for each stage (Table 2). The rate dL/dt increases from plant emergence to tasseling (stages 0 and 1) and decreases during stages 2 and 3. Those values were also taken from field experiments for short season maize hybrids, according to Müller et al. (2005).

Grain yield was calculated in Glam considering the above-ground biomass and the harvest index (HI). The latter was simulated using a mean rate of change in the harvest index (dHI/dt) during the reproductive period, from tasseling to physiological maturity (stages 2 and 3), and it was based on a six-year series of field experiments. Crop transpiration efficiency (TE) was calculated from the mean value of maximum evapotranspiration (ET_m), measured in a weighing lysimeter, in five crop cycles (Bergamaschi et al., 2001; Radin et al., 2003). The partitioning of ET_m to

transpiration and soil evaporation was obtained around the maximum leaf area index by measuring plant transpiration through sap flow using a heat pulse tracer (Santos et al., 2000).

Regression analysis of simulated yield against observations was done after any technology trend was removed, in order to test Glam-Maize performance in simulating grain yield at the scale of municipalities, micro-regions, and main producer region.

Results and Discussion

Variability indices of both grain yield and rainfall were calculated by dividing their anomalies (with respect to the historical mean) by the respective standard deviations (Figure 2). This normalized grain yield standardized anomaly tended to be larger for municipalities and micro-regions than for the main producer region (Table 2). Rainfall during crop cycle explained (R^2) more than one-half of the detrended yield variance, whereas the rainfall during the 0–30 days after tasseling explained around 70% of the yield variance.

The effect of severe drought on grain yield is clearly shown in the 2004/2005 crop season, which was very dry; consequently, the harvest was less than half of that of the previous crop season (1,250 against 2,700 kg ha⁻¹ for 2003/2004). Soil water availability is particularly important for maize crops during the tasseling and silking stages, when the leaf area index

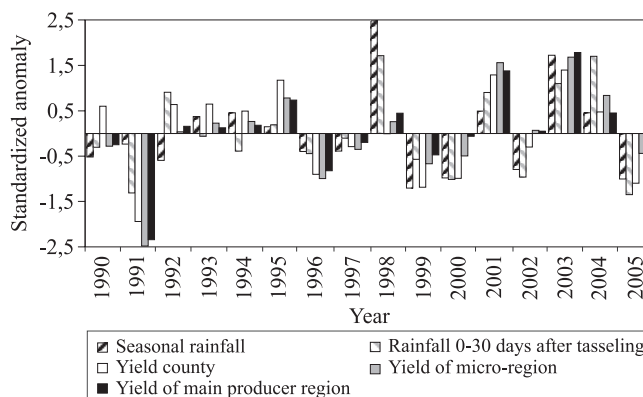


Figure 2. Variability of maize grain yield and rainfall at three spatial scales: municipality (county), micro-region, and main producer region, in the state of Rio Grande do Sul, Brazil. Standardized anomalies were calculated by dividing each value by the standard deviation in the time-series.

and the atmospheric evaporative demand are high (Bergamaschi et al., 2001, 2006; Radin et al., 2003). Tasseling occurred between December and January (Table 1). Therefore, short dry spells or isolated rainfall events during this period can cause high yield variability in maize crops in the region (Bergamaschi et al., 2006; Andriolli & Sentelhas, 2009). For example, the 1990/1991 season had a small rainfall anomaly for the entire crop cycle (i.e., standardized anomaly of -0.5, Figure 2), but a dry spell in the critical period of maize crops caused a strong negative effect in the 1990/1991 season (i.e., standardized anomaly of -1.25 for 0–30 days from tasseling). Since this short dry spell occurred in the critical period (tasseling to silking stages), a strong grain yield reduction was observed.

Some simulated and observed phenological events, as well as the leaf area index, are shown in Figure 3. The simulated data correspond to the municipality of Santa Rosa, whereas the observed data were measured at the Experimental Station of the Universidade Federal do Rio Grande do Sul (UFRGS), in the municipality of Eldorado do Sul, just outside of the main producer region, close to the Taquari micro-region (Figure 1). Those two places have similar thermal conditions and a difference of about two degrees in latitude. However, limitations caused by water deficits in maize crops are higher in Eldorado do Sul than in Santa Rosa.

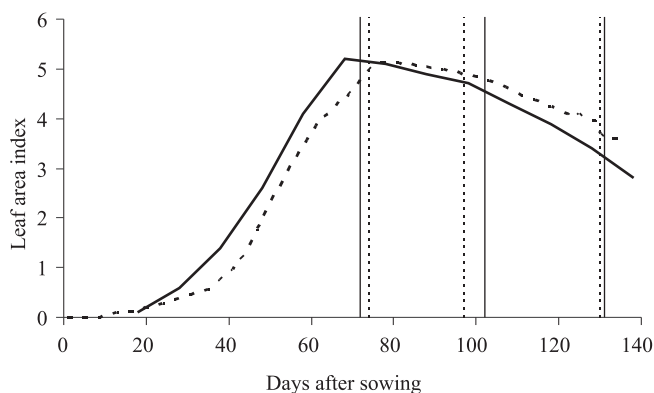


Figure 3. Maize leaf area index averaged over six cropping seasons (1993/1994 to 1998/1999) from experimental data in Eldorado do Sul (solid lines) and simulated by the Glam model for the municipality of Santa Rosa (dashed lines), in the state of Rio Grande do Sul, Brazil. Vertical lines show the observed (solid lines) and simulated (dashed lines) dates of maximum leaf area index, milk maturity, and physiological maturity.

Glam-Maize simulations of the time of the developmental stages were in good agreement with the observed data. Glam-Maize estimated maximum leaf area index at 74 days after sowing, which was two days earlier than the observed data. The occurrence of milky maturity (102 days) and physiological maturity (132 days) was predicted five and one day later, respectively, than the observed. The agreement in timing (days after sowing) between estimated and observed phenological stages can be explained by the high dependence of maize phenology on the degree-days accumulation, as observed by Streck et al. (2008). However, leaf area expansion showed a consistent lag between the simulated and observed data, which may be attributed to other factors. Since observed data were obtained from irrigated experiments, it is possible that water limitations in Santa Rosa (in rainfed conditions) promoted some delay in leaf area expansion. Because of the importance of the leaf area for solar radiation interception in crop modeling and of the complexity of the genotype x environment interactions on maize development in subtropical conditions (Streck et al., 2012), this aspect must be considered with extreme caution in adjusting and testing maize crops models.

Glam-Maize simulations of soil evaporation (E), crop evapotranspiration (ET) and transpiration (T) for the entire main producer region are presented in Figure 4. The total observed ET of irrigated maize measured with a weighing lysimeter, during five years,

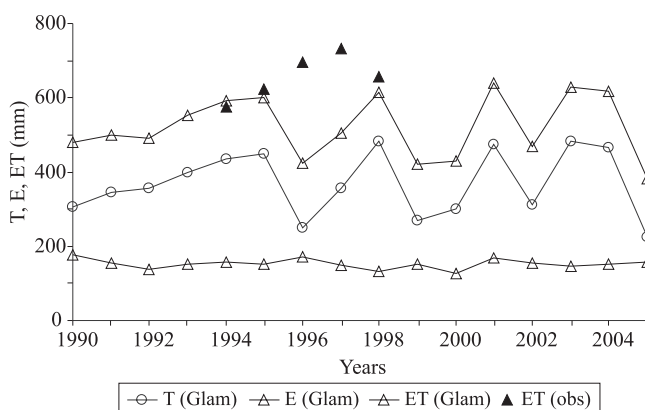


Figure 4. Actual evapotranspiration (ET), plant transpiration (T), and evaporation from the soil surface (E) estimated by the Glam model for maize crops over the main producer region of the state of Rio Grande do Sul, Brazil. Solid triangles are the maximum ET of maize measured in a weighing lysimeter in Eldorado do Sul, in the state of Rio Grande do Sul, Brazil (Bergamaschi et al., 2001; Radin et al., 2003).

is also shown (Bergamaschi et al., 2001; Radin et al., 2003). Estimated ET was in good agreement with the experimental results during the three rainiest seasons (1994, 1995, and 1998). In contrast, the observed crop ET was higher than the ones simulated for the two years with substantial dry periods (1996 and 1997). This difference between observed and simulated ET can be mainly associated to the irrigation supply, which was not considered in the simulation.

Simulated and observed maize yield time-series for the municipalities of Santa Rosa and Passo Fundo and their respective micro-regions are shown in Figure 5. These two micro-regions showed different correlations between observed yield and rainfall (Figure 3). Time-series of observed and simulated crop yield for the main maize producer region are presented in Figure 6. Glam-Maize simulated well the inter-annual variability in maize yield for these time-series, including those

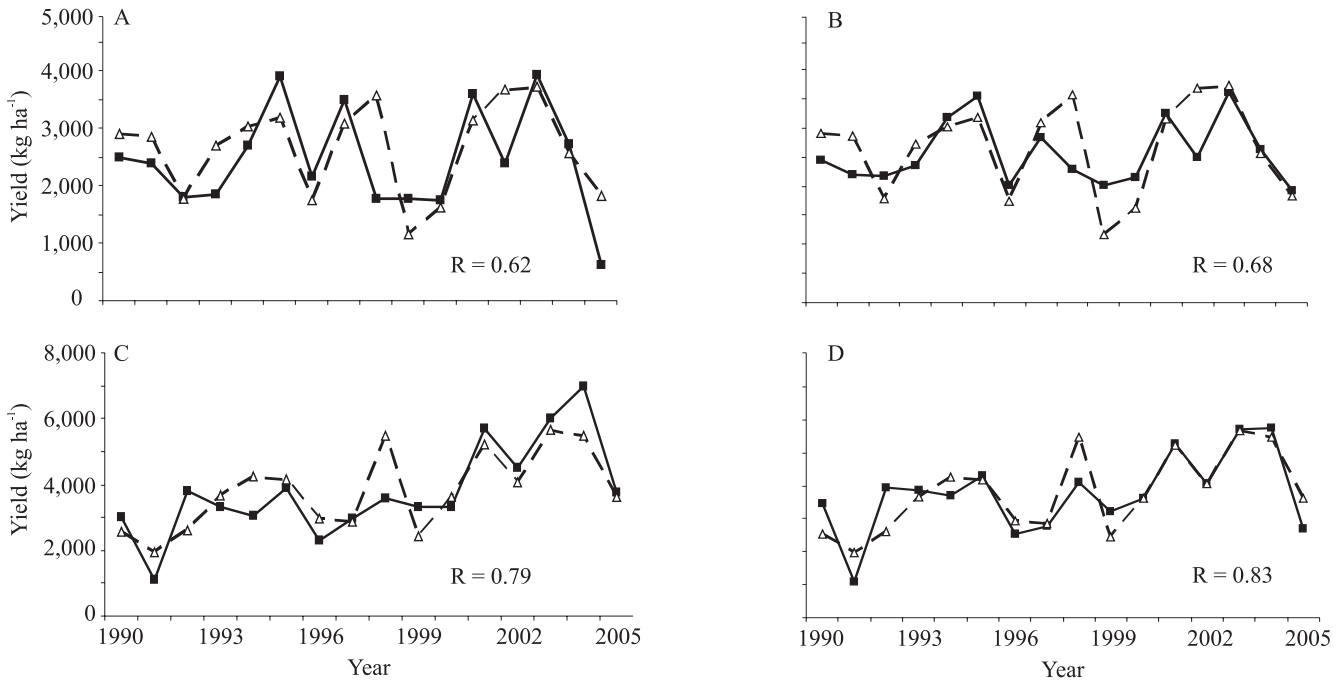


Figure 5. Observed (■) and simulated (Δ) maize grain yields for the municipalities of Santa Rosa (A) and Passo Fundo (C) and their respective micro-regions (B, D), in the state of Rio Grande do Sul, Brazil, from 1990 to 2005.

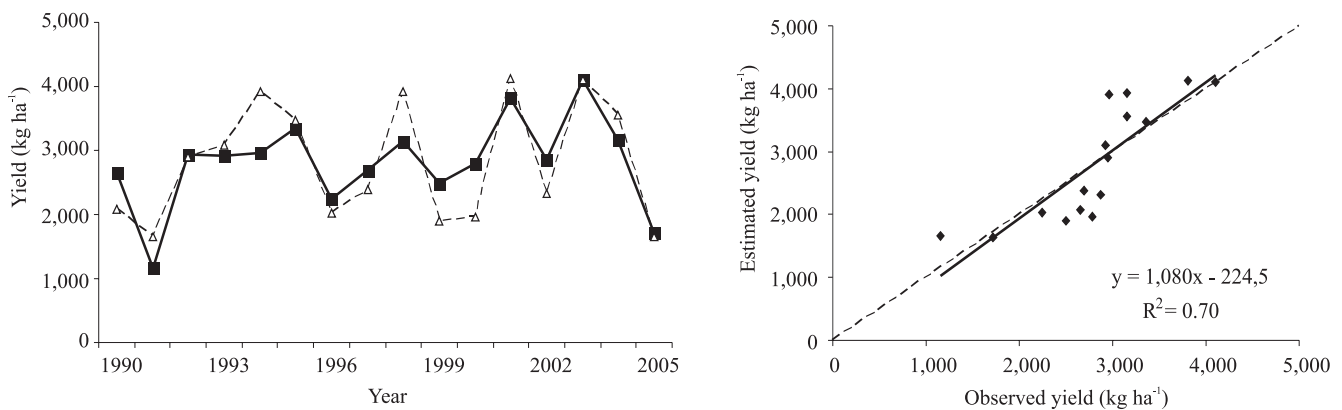


Figure 6. Relationships between the observed (■) and simulated (Δ) maize grain yields for the main maize producer region of the state of Rio Grande do Sul, Brazil, from 1990 to 2005.

years with long drought periods, such as 2004/2005. The agreement between simulated and observed yields tended to increase from municipality to micro-region, even though correlations between observed yield and rainfall did not show significant differences (Figure 2, Table 3). Glam-Maize performance in simulating yield varied when the sowing date was changed, but these differences were not statistically significant (Table 4). Glam performance was analyzed under different spatial scales, technological levels, soil types, and sowing dates. The parameterizations in Glam are relatively simple, but complex enough to capture the climatic component of yield variability. The non-climatic components include new plant genotypes, improvements in soil management and fertilization, and plant density (Berlato et al., 2005). Although possible variations in these factors were neglected (e.g., plant genotypes), Glam-Maize simulated grain yield variability to a good degree of accuracy.

Table 3. Correlation coefficients between rainfall and maize grain yields at three spatial scales: municipality, micro-region, and main producer region, in the state of Rio Grande do Sul, Brazil.

Spatial scale	Rainfall	
	Crop cycle (130–140 days)	0–30 days after tasseling
Municipality	0.52	0.67
Micro-region	0.53	0.69
Main production region	0.56	0.72

Table 4. Correlation coefficients between simulated and observed yields of maize, for small and large spatial scales, with different sowing dates around the basic sowing date (BSD), from the 1989/1990 to the 2004/2005 cropping season, in the state of Rio Grande do Sul, Brazil.

Spatial scale	Correlation coefficient ⁽¹⁾		
	BSD	10 days before BSD	10 days after BSD
Small spatial scale			
Santa Rosa municipality	0.69	0.65	0.73
Santa Rosa micro-region	0.77	0.77	0.78
Passo Fundo municipality	0.71	0.71	0.72
Passo Fundo micro-region	0.81	0.87	0.87
Large spatial scale			
Main producer region	0.77	0.69	0.85

⁽¹⁾The correlation coefficients were significant at 5% probability.

The comparison between observed and simulated yields is subject to constraints from both data and modeling. First, because a discrepancy (around 10%) between the two observed data sources (Conab and IBGE) for the same period (for the entire state) indicates a possible imprecision in the yield estimation methods of those official Brazilian agencies (Companhia Nacional de Abastecimento, 2006; Instituto Brasileiro de Geografia e Estatística, 2006). It is reasonable to expect that Glam-Maize performance could be better analyzed with more accurate yield observations. Second, the wide range of sowing dates adopted by farmers (Table 2) is problematic for large-area modeling. In the warmest zones, high thermal availability may sometimes permit two maize cycles in a single cropping season using very-short season hybrids. Similarly, maize is sometimes sown on small farms just after the harvest of other spring crops, such as dry beans and tobacco, that delay the sowing of maize and reduce its potentiality. However, neither of these processes has been simulated in the present study. Third, regional variability in soil and topography may influence the spatial variability of hydrological conditions, and some regions have higher spatial variability in soils than others. For instance, the Santa Rosa micro-region (site 5, Figure 1) has predominantly small farms and a mostly irregular pattern of soils and topography. In contrast, the Passo Fundo micro-region (site 6, Figure 1), located in the northern plateau of the state, consists of medium to large farms with more uniform soils, topography, and technological levels. In this context, improvements in simulating water dynamics in the soil-plant-atmosphere system can be performed by matching the soil-water storage to the observed conditions for each particular prevailing soil type.

Modeling maize yields in southern Brazil, even at large spatial scales, presents a great challenge, but can be very important for several applications, such as evaluating impacts of climatic changes on crops or estimating grain yield, taking seasonal climatic information as inputs. In this sense, Coelho & Costa (2010) observed a good performance of the Glam-Maize model (parametrized in the present study) in simulating maize yields in southern Brazil by using seasonal climate forecasts. This study also shows that a process-based model can simulate maize crop yield variability. A well-designed model

may produce accurate results when fitted and parametrized to regional cropping systems. For this purpose, a well-established experimental basis seems to be indispensable, considering the complexity of the interactions among environmental factors on maize crops, in particular under sub-tropical conditions (Streck et al., 2012), and the specificity in cropping systems. Further development of the Glam-Maize model may improve its performance by introducing new adjustments that take into account short-term water stresses around flowering time, when the number of grains is established (e.g., Ceres-Maize model).

Conclusions

1. The general large area model for annual crops (Glam) captures the high inter-annual variability of maize yield in the sub-tropical conditions of the state of Rio Grande do Sul, Brazil, when parametrized for regional conditions and cropping systems.

2. Simulations of grain yields by the Glam-Maize model are highly correlated with observed yields at large spatial scales, with variable correlations at smaller spatial scales.

3. Large area crop simulations by processed-based models can contribute to monitoring or forecasting regional patterns of variability in maize yield in sub-tropical conditions, providing a basis for agricultural decision making, and the Glam model is one of the alternatives.

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