Chapter 18 The Role of Crop Systems Simulation in Agriculture

K. J. Boote University of Florida, USA

and Environment

J. W. Jones University of Florida, USA

G. Hoogenboom University of Georgia, USA

J. W. White USDA-ARS, Arizona, USA

ABSTRACT

Simulation of crop systems has evolved from a neophyte science into a robust and increasingly accepted discipline. Our vision is that crop systems simulation can serve important roles in agriculture and environment. Important roles and uses of crop systems simulation are in five primary areas: (1) basic research synthesis and integration, where simulation is used to synthesize our understanding of physiology, genetics, soil characteristics, management, and weather effects, (2) strategic tools for planning and policy to evaluate strategies and consequences of genetic improvement or resource management, (3) applications for management purposes, where crop systems simulations are used to evaluate impacts of weather and management on production, water use, nutrient use, nutrient leaching, and economics, (4) real time decision support to assist in management decisions (irrigation, fertilization, sowing date, harvest, yield forecast, pest management), and (5) education in class rooms and farms, to explain how crop systems function and are managed.

INTRODUCTION

Simulation of crop systems has advanced greatly over the past 30 to 40 years. From a neophyte science with inadequate computing power, the field has evolved into a robust and increasingly accepted science supported by improved software, languages, development tools, and computer capabilities, but the foundation continues to be scientific insights from plant physiology, soil

DOI: 10.4018/978-1-4666-0333-2.ch018

science, agroclimatology, and related fields. Crop system simulators contain mathematical equations describing basic flow and conversion processes of carbon, water, and nitrogen balance that are integrated daily or hourly by the computer program to predict the time course of crop growth, nutrient uptake, and water use, as well as to predict final yield and other plant traits and outputs. The goal of this paper is to give our vision on how crop systems simulation can serve important future roles in agriculture and environment and suggests how to prioritize research to better support these roles. The paper leads off with an historical overview describing how crop system models began, then discusses five primary roles and uses of crop systems simulation in agriculture and environment, concludes with a challenge for potential linkage of crop models with molecular biology-genetics, and suggests the need for continued improvement of the science in crop system models.

AN HISTORICAL OVERVIEW

The use of crop system models and simulation had its start in crop physiology, soil physics, and soil-crop-water processes. Early models focused mainly on the crop carbon (C) balance under optimum conditions, where only solar radiation and temperature were the driving variables. Simulation of crop canopy photosynthesis from leaf-level parameters was a primary focus (DeWit, 1965; Duncan, 1971), along with predicting crop development as described through their growth stages and examining strategies for increasing reproductive yield. These patriarchs of crop modeling soon advanced to developing simple whole crop models. Concurrently, the early agricultural engineers and soil physicists were developing soil-plantwater balance models that predicted daily crop evapotranspiration, crop water uptake, and water flow processes in soils (Whisler et al., 1986). See Whisler et al. (1986) for an overview and history of crop simulation models up to the mid-1980s,

including typical processes considered, data required, model testing, and applications. The crop aspects of many of the early soil-water-balance models were often fairly simple, estimating daily growth from light-interception and radiationuse-efficiency. The soil water balance models vary from tipping bucket one-dimensional water balance (Ritchie, 1985, 1998) to more complex Darcy-driven water flow with two dimensional flow such as 2-DSOIL (Ahuja, Ma, & Timlin, 2006) and RZWQM (Ma et al., 2003). The next improvement in crop system models came with the simulation of soil nitrogen (N) balance with a simple tipping bucket plug-flow of nitrate N to allow simulation of N leaching, but success was limited until improvement in two major components had occurred: first, the crop C balance routines needed to estimate crop N demand accurately and second, accurate routines to estimate soil organic matter mineralization are needed to estimate the supply of soil mineral N beyond that coming from applied fertilizer N. There are many published soil organic matter models (e.g., see Smith et al., 1997, who compared nine different soil organic matter models). The most frequently cited organic matter models are CENTURY (Parton, Stewart, & Cole, 1988) and ROTHC (Jenkinson & Rayner, 1977), and these models often serve as reference models for many studies (Traore, Bostick, Jones, Koo, Goita, & Bado, 2008). Each of these models has shortcomings, and there are many difficulties correctly simulating soil organic matter dynamics, even after 20-30 years of progress, because soils are so variable and soil organic matter is complex.

Over the past 10 to 20 years, crop system model developers have succeeded in linking good crop C balance (N demand) with good soil water balance and good soil-crop N balance. The DSSAT V3.5 models (Hoogenboom, Wilkens, Porter, Batchelor, & Hunt, 1999; Jones et al., 1998) were among the early models to succeed in this full linkage, but APSIM (Keating et al., 2003; McCown, Hammer, Hargreaves, Holzworth, & Freebairn, 1996) and other models are also at this stage of development.

12 more pages are available in the full version of this document, which may be purchased using the "Add to Cart" button on the publisher's webpage:

www.igi-global.com/chapter/role-crop-systems-simulation-agriculture/63769

Related Content

Data-Centric UML Profile for Wireless Sensors: Application to Smart Farming

Julian Eduardo Plazas, Sandro Bimonte, Gil De Sousaand Juan Carlos Corrales (2019). *International Journal of Agricultural and Environmental Information Systems (pp. 21-48).*

www.irma-international.org/article/data-centric-uml-profile-for-wireless-sensors/223868

Bioremediation of Industrial Waste Using Microbial Metabolic Diversity

Saurabh Gangola, Pankaj Bhatt, Parul Chaudhary, Priyanka Khati, Narendra Kumarand Anita Sharma (2018). *Microbial Biotechnology in Environmental Monitoring and Cleanup (pp. 1-27).*

www.irma-international.org/chapter/bioremediation-of-industrial-waste-using-microbial-metabolic-diversity/196789

Vertical Landscape Desgn

Gökçen Firdevs Yücel (2014). Green Technology Applications for Enterprise and Academic Innovation (pp. 276-292).

www.irma-international.org/chapter/vertical-landscape-desgn/109921

Modeling of current and future state of biodiversity in Central America using GLOBIO3 methodology

Denisse McLean R. (2011). Land Use, Climate Change and Biodiversity Modeling: Perspectives and Applications (pp. 349-375).

www.irma-international.org/chapter/modeling-current-future-state-biodiversity/53760

Environmental Reporting in the Public Interest

Hans-Knud Arndt, Mario Christand Oliver Gunther (2001). *Environmental Information Systems in Industry and Public Administration (pp. 347-354).*

www.irma-international.org/chapter/environmental-reporting-public-interest/18546