

1 **Title: Using Social Budgeting Web Software and Natural Resources Software Models to Improve**  
2 **Agricultural Economics Data Collection, Dissemination, and Analysis**

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25  
26 **Abstract**

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28 Agricultural economics needs more and better data. This paper proposes a new approach, social budgeting, for

29 building and analyzing agricultural economic datasets. Social budgeting employs modern information

30 technologies, such as social networks and cloud computing data centers, to collect and analyze economics data

31 over the Internet. DevTreks, an open source, social budgeting web software program, is introduced as an

32 example. A case study involving Midwestern corn soybean production uses DevTreks and the process-based

33 Root Zone Water Quality Model (RZWQM) to illustrate how modern information technology can be used to

34 study economic-pollutant tradeoffs. The paper recommends starting or joining open source, on-line efforts aimed

35 at integrating the physical, chemical, biological, and economic aspects of agriculture.

36  
37 **Keywords:** social budgeting, social networks, cloud computing, open source software, economic data, economic  
38 services, Nitrogen loading, tile drainage.

39 **1. Introduction - the Emperor has no clothes**

40 Agricultural economics data is hard to collect, disseminate, and use. Most practitioners rely on data that,  
41 to varying degrees, limits the full usefulness of their findings. Much of this data suffers from lack of  
42 standardization, antiquated software, single user bias, and outdated approaches to information technology (IT)  
43 data collection, dissemination, and analysis. Today, most data collected by different farm managers, researchers,  
44 or extension staffs cannot be used to support modern computing (the aggregation and analysis of data, by people  
45 and machines, in a consistent manner). The data difficulties are compounded by analysts' reliance upon  
46 disparate climatic, landscape, demographic, agronomic, and economics data sets.

47 The problem of inconsistent, incompatible, and missing economics data is not new. In 1991, the Applied  
48 and Agricultural Economics Association (AAEA) convened a national task force consisting of 68 economists  
49 (the National Task Force on Commodity Costs and Returns Measurement Methods) to deal with aspects of the  
50 problem. Their resultant publication, *Commodity Costs and Returns Estimation Handbook* addressed the Task  
51 Force's mission to "recommend standardized practices for generating costs and returns estimates ..." (Hallam et  
52 al., 1999).

53 Ten years later, Just and Pope (2001) reviewed the current state of agricultural production data  
54 availability and use. They highlighted the data and analytic challenges arising from agriculture's unique  
55 dependence on sequential production stages, weather, pests, diseases, lags between the application of inputs and  
56 harvest, crop rotations, soils, and farmer abilities. Just and Pope noted that "readily available data tends to  
57 consist of highly aggregated public data in which temporal detail (within a growing season) and spatial data  
58 (among plots, farms ...) is lost". They concluded, based largely on economists' use of poor quality data, that "the  
59 current state of empirical knowledge of agricultural production sums up to little more than an empty box." That  
60 conclusion led Just and Pope to the following proposition (p. 721):

61 We propose that a significant and complete data base for (...) agricultural production needs to be  
62 developed as an investment by/for the agricultural economics profession, ... Such a data base could  
63 facilitate investigation of many issues identified by this study as blocked by data unavailability. By  
64 comparison, the current proliferation of studies with uncommon data bases and incongruent maintained  
65 hypotheses has led to endless speculative explanations of differences in results with little comprehensive  
66 comparison.

67

68 No such complete agricultural economics database has been built in the intervening years. Quotes from four  
69 recent articles in agricultural economics publications show that the quality of economics data is still a major  
70 issue: “(E)stimates of net farm income derived from (ARMS [Agricultural Resource Management Survey]) (may  
71 be) biased downward ... it has been recognized that farm income is generally underreported to the IRS ...” (Key  
72 and Roberts, 2009) . “(O)f course, crop budgets and production costs are notoriously difficult to measure”  
73 (Goodwin, 2009), “(T)he (economics) evidence is mainly anecdotal ... (Coleman, 2009). “More research and  
74 data are needed ... More research and data are also needed ... More research and data are needed ... More  
75 research and data are also needed ...” (Bergstrom and Ready, 2009). It should be noted that just about any article  
76 in either of the two publications cited, the *American Journal of Agricultural Economics* and the *Review of*  
77 *Agricultural Economics*, could have been quoted to highlight this issue.

78 Resource and environmental economics analysis also relies on biophysical and economic data and may  
79 also be suffering from some “empty boxes”. Weyant (2008) notes “(the wide) uncertainty in mitigation cost  
80 projections ... is not surprising ... assumptions (must be made over long periods) about productivity growth, fuel  
81 prices, technology diffusion, the development of new technologies, and the interest rate/discount rate used in  
82 making intertemporal investment decisions”. A recent report on climate change (IPCC, 2007) found that “The  
83 literature on (climate change) adaptation costs and benefits remains quite limited, and fragmented in terms of  
84 sectoral and regional coverage.” A contributor to the Ecological Society of America's weblog (Hollister, 2007)  
85 wrote: “ecologists and environmental scientists must take a more active role in providing access to both data and  
86 the analytical techniques used to analyze those data. As our studies become increasingly broad, our analytical  
87 capabilities must also expand and, perhaps more importantly, we should be able to more easily share and  
88 reproduce complex analyses.”

89 Regardless of the reasons why individuals manage data poorly, many scientific disciplines, and  
90 economics in particular, have not been able to fully exploit modern information technology (IT). The authors  
91 find very few examples of the innovative use of modern information technology among economists, who often

92 manage individual datasets in spreadsheets or single files. The reasons why may not be as important as the  
93 likelihood for some solutions.

94 This paper investigates the contributions that modern IT can make to some of these issues by introducing  
95 DevTreks, an open source, social budgeting web software program. Social budgeting attempts to combine the  
96 advantages of social networking web sites with the analytic power of microeconomics. It relies upon on-line  
97 communities to build the site's content. In other words, social budgeting reduces the cost of, and provides a  
98 framework for, collaboration. The community of agriculturalists, by pooling their currently fragmented efforts,  
99 can then "divide and conquer" the problems associated with incompatible and inconsistent data.

100 The paper addresses the potential to use IT, social networks, and social budgeting to: 1. improve the  
101 collection and dissemination of basic economics data, 2. automate at least some economics analyses, and 3.  
102 improve how data is shared between biophysical and economics models. A case study, using two common  
103 agricultural science data sets (farm profits and nitrogen (N) loading from tile drained agriculture), and two  
104 modern software applications (RZWQM and DevTreks) illustrate the main points. Although the case study  
105 focuses on agricultural resource management data, the IT issues are the same for other types of economics and  
106 biophysical data.

107

## 108 **2 Modern IT - Clothing the Emperor anew**

### 109 **A. What is Social Budgeting**

110 Most definitions of social budgeting focus on how governments can spend money better by including citizens in  
111 government budget preparation. UNICEF (2007) describes the process as follows: “ In Kenya—where social  
112 budgeting was introduced in three districts and at the national level—the initiative seeks to strengthen  
113 (government budgeting) practices and make them more effective in delivering social development objectives.  
114 Specifically, the objective of the social budgeting exercise was to strengthen the (government budgeting process)  
115 in the five areas mentioned in the previous chapter: improving the focus on human development goals,  
116 increasing transparency and accountability, promoting wider partnerships, building capacity at the district level  
117 and improving the use of data and analysis”. Shah (2007) uses a related term, 'participatory budgeting' to

118 describe how: “(Participatory budgeting) offers citizens at large an opportunity to learn about government  
119 operations and to deliberate, debate, and influence the allocation of public resources ... the enhanced  
120 transparency and accountability that participatory budgeting creates can help reduce government inefficiency  
121 and curb clientelism, patronage, and corruption”.

122 In contrast, from DevTreks' perspective, social budgeting can benefit anyone who makes decisions based  
123 on budgets, whether governments, individuals, firms, or researchers. DevTreks describes social budgeting on its  
124 web site as follows:

125 “Social budgeting is the on-line sharing of social and economic knowledge among networks and groups  
126 of people for the purposes of improving lives and livelihoods. Families finances improve when neighbors show  
127 them how to budget their money, or energy use, better. Businesses improve by more easily discovering how their  
128 performance, be it profitability or carbon use, compares to their peers. Governments spend money better when  
129 project and program managers share information about the benefits, costs, and tradeoffs involved in the delivery  
130 of public goods and services.

131 Social budgeting employs the economics perspective that society, as a whole, is better off when citizens,  
132 firms, and governments use resources efficiently. Social budgeting, when done properly, shows people how to  
133 conserve scarce resources such as labor, capital, goods, services, know-how, talent, and natural resource assets. “

#### 134 **B. How Should Social Budgeting Be Carried Out?**

135 As with the definition of social budgeting, the nuts and bolts methods for carrying out social budgeting  
136 depend on the perspective taken. Most researchers describe the social budgeting process in terms of government  
137 accounting and budgeting. Scholz et al. (2000) identify the two major ingredients in social budgeting to be a  
138 “Social Accounting System” and a “Social Budget”. The first is a formal financial accounting system  
139 implemented using a national statistical reporting system. The second is a formal budget forecasting system that  
140 is a mathematical model of the accounting system. Both ingredients differ from traditional government  
141 accounting and budgeting systems by their focus on improving social sector spending. The authors describe this  
142 system in great detail and recommend that their book be used as a “How to build a Social Budget guidebook”.

143 In contrast to these financial accounting methods, DevTreks carries out social budgeting using traditional  
144 economic data such as input and output prices and operating and capital budgets, which are then aggregated and  
145 analyzed using economic analysis methods. The nuts and bolts collection and analysis of this economics data is  
146 accomplished through social networks who use online web software applications. How should these web  
147 applications deliver social budgeting services?

148 The authors' experience building DevTreks, and experience collecting agricultural economics production  
149 data (25+ years for the lead author), leads them to identify the following characteristics that a modern  
150 agricultural economics information system needs to carry out social budgeting:

151 **2.1 Searchable:** Data should be easy to find. A 10-10-10 fertilizer price series, an operating budget data set for  
152 Midwestern farm rotations, or a capital budget analysis of alternative dairy waste management investment  
153 returns, should be one or two clicks away. Google set the standard for easy, relevant searches. *An economics*  
154 *information system should quickly return relevant search results using an interface like Google.*

155 **2.2. Web based:** Users should be free to use whatever computer operating system they prefer. The Internet  
156 allows them this freedom. An economics information system should be web software that runs in all major  
157 browsers. Since cellphones now include web browsers, some economics information should be accessible using  
158 mobile web browsers. In addition, the Internet industry has established standards for making web based data  
159 accessible to machines. The standards fall under the general category of web services and include the use of  
160 Uniform Resource Identifiers (URIs), or web addresses, for every page of data. *An economics information*  
161 *system should be web based, supporting both desktop and mobile device browsers. It should allow users,*  
162 *including machines, to access data using simple Internet protocols, such as URIs.*

163 **2.3 Social:** Sites that attract the most visitors are social networks. Sites like "Facebook" or "YouTube" serve as  
164 information exchanges that improve how people with common interests can share data. Most of the content on  
165 these sites is developed by the users themselves. Niche sites, catering to professional audiences, are evolving  
166 from these general purpose social networking sites. Economics data would be easier to collect if thousands of  
167 groups of specialists, whether global warming cost estimators, water conservation investment analysts, or  
168 grocery store price watchers, joined together in a common social networking site to share their data. *An*

169 *economics information system should be a social networking site that targets professionals who have a need to*  
170 *collect, disseminate, and analyze economics data.*

171 **2.4 Open:** Scientists and researchers, and many young computer users, like to tinker with computers and  
172 software. Open source software has become an increasingly popular way to allow these tinkerers to collaborate,  
173 change, and improve software. The R project is an example of successful, open source statistical analysis project  
174 (R Development Core Team, 2009). Statisticians commonly contribute new statistics modules, or “packages”,  
175 that extend the capabilities of the R Project. Since economics data serves a social purpose -helping people to  
176 improve how society, not just individuals, allocate resources- it fits well with open, collaborative software. Open  
177 also implies more reproducible results, as other researchers can examine the calculations used in an analysis in  
178 detail. *An economics information system should be open-source and managed by competent open-source*  
179 *software managers.*

180 **2.5 Standards based:** A web based data application must meet a combination of professional, data, and  
181 information technology standards. Professional standards need to ensure that data is reliable, high quality, and  
182 follows 'best-of-class' practices, for calculations and analyses. Examples of these types of standards include the  
183 Costs and Returns Estimation Handbook (Hallam et al., 1999) and the American Society of Agricultural and  
184 Biological Engineers standards (ASABE, 2009). Information technology standards need to ensure that data can  
185 be collected, stored, and made accessible, in the best available manner to the widest possible audience. Examples  
186 of information technology standards include the W3C recommendations on XHTML (a web data format, see  
187 <http://www.w3.org/MarkUp/>), the Dublin Core metadata element set (data that describes data, see  
188 <http://dublincore.org/>), and the International Digital Publishing Forum's (IDPF) electronic publishing standards  
189 (ebooks, see [www.idpf.org](http://www.idpf.org)). Data standards need to ensure the data can be easily accessed, shared, aggregated  
190 and analyzed. Examples of data standards include the UNIFORMATII construction cost classifications (Charette  
191 and Marshall, 1999) and the Encyclopedia of Life's biological species classifications (see [www.eol.org/](http://www.eol.org/)). *An*  
192 *economics information system should closely follow the best available professional, data, and information*  
193 *technology standards. When these standards don't exist (i.e., agricultural resource management classification*  
194 *standards), teams should be established to develop the missing standards.*

195 **2.6 Linked:** The inventor of the world wide web, Tim Berners-Lee, recently gave a talk explaining his vision of  
196 how “linked data” (Internet data that is formally related to other Internet data) is a solution to many of the data  
197 shortfalls addressed in this paper (Berners-Lee, 2009). The need for good “linked data” is especially acute for  
198 agricultural economics data because of its dependencies upon climatic, landscape, and demographic data. This  
199 “linked data”, more formally known as the semantic web, is emerging as a new set of Internet technologies for  
200 making data more useful (Feigenbaum et al., 2007). It includes schemas for relating data (RDFs, or Resource  
201 Description Frameworks), taxonomies for classifying and defining data (ontologies), and inference engines for  
202 finding new relations among different ontologies. *An economics information system should return data that has  
203 been formally linked and related using semantic web technologies. The resultant data should support new, and  
204 better, forms of automated economic analyses.*

205 **2.7 Secure:** Once a research team toils for six months building a good data set, the data should not change  
206 unless the researchers explicitly grant permission to someone to edit the data. A farmer who agrees to share crop  
207 enterprise data, but desires to stay anonymous, should be ensured of their privacy. The Internet was not designed  
208 with data security in mind. An economics information system should continually evolve better ways to secure its  
209 data. *An economics information system should keep data private if it was provided with an understanding of  
210 confidentiality.*

211 **2.8 Long lived:** Many popular web sites appear juvenile and temporarily fashionable. In contrast, economics  
212 data tends to be serious and have historical importance. An economics web site needs to err on the side of  
213 gravitas rather than fashion. More importantly, it needs to be dependable over the long term. *An economics  
214 information system should be designed to store data for decades, if not centuries.*

215 **2.9 Easy to use:** Data should be easy to enter, navigate through, save, upload, package, calculate, analyze and  
216 download. Modern web applications are specifically designed to replace desktop applications. They make full  
217 use of modern Internet technologies such as asynchronous data loading (see AJAX,  
218 <http://en.wikipedia.org/wiki/AJAX>) and javascript libraries (see jQuery, <http://en.wikipedia.org/wiki/JQuery>).  
219 These technologies help to overcome common web application annoyances like page flickering, slow data

220 loading, getting lost while navigating, and poor user interaction. *An economics information system should use the*  
221 *latest web technologies to make the web applications as easy to use as many desktop software applications.*

222 **2.10 Rich:** Most web sites now commonly support the sharing of photos, videos, and other multimedia. Crop  
223 operating budgets would be easier to understand if photos and videos were included of the crop operations.  
224 Capital budgets could be assessed more meaningfully if blueprints were available for investment components.  
225 An economic analysis might be easier to understand if the author included a video explaining the how and why  
226 of the analysis. *An economics system should fully support the use of photos, videos and other multimedia to make*  
227 *economics data richer.*

228 **2.11 Understandable:** People with a potential interest in economic data vary from Indian farmers seeking  
229 current market price data, extension staffs wondering about the profitability of a prospective new crop  
230 introduction, microfinance groups trying to assess how well their group compares to other business  
231 organizations, and economic researchers assessing global warming abatement technologies. Practitioners and  
232 researchers throughout the world need data that they can understand. This means the data must be international  
233 and it must be explained. Data should be accompanied with stories explaining the context of the data. *An*  
234 *economics information system should be international in supporting multiple languages, measurement systems,*  
235 *and production technologies, and all data should be linked to explanatory stories.*

236 **2.12 Extensible:** Grameen Bank's App Lab (<http://www.grameenfoundation.applab.org/section/index> )  
237 demonstrates how a software application can be opened to software developers so that they can contribute their  
238 own software extensions. Economists have developed thousands of statistical models. Many of these models  
239 would make useful software extensions to an economics information system. To name a few, economic extension  
240 modules could analyze price indices, determine cost effectiveness, examine marginal costs, find an economic  
241 optimal solution, carry out what if scenario planning, and assess risks. *An economics information system should*  
242 *offer incentives to software developers to build "extensions" that extend the economics services offered by the*  
243 *system.*

244 **2.13 Accessible:** People with disabilities, including the blind and deaf, should be able to use and analyze the  
245 data. The standard row and column nature of price data, in particular, makes this feasible. The Internet industry

246 has recognized this need and developed sound standards for making data accessible (see NIMAS,  
247 <http://nimas.cast.org/>). *An economics information system should follow widely accepted standards for making*  
248 *data accessible to people with disabilities.*

### 249 **C. How Does DevTreks Carry Out Social Budgeting?**

250 DevTreks delivers social budgeting services through online social networks that focus on a general  
251 theme of importance to a social network's members, such as residential construction, agricultural conservation,  
252 family budgeting, or government cost benefit analysis. These social networks, which DevTreks refers to as  
253 “network groups” are divided further into networks, clubs and members. An agricultural network group might  
254 have individual networks devoted to crop production, organic farming, dairy, cattle ranching, or innovative water  
255 conservation technologies. Clubs within a cropping network might focus on corn and soybean production in  
256 Iowa, or cotton production in Texas. Each club's members might include extension workers, pest control  
257 advisers, carbon emission researchers, farmers, or ranchers. Each club offers their content, known as services, to  
258 other clubs through service agreements.

259 DevTreks web software allows clubs to build hierarchical data sets (i.e. DevTreks' services) holding  
260 input and output prices, operation and component costs, and operating and capital budgets. The web software  
261 includes calculators (i.e. to allocate the cost of agricultural machinery) and analyzers (i.e. mean and standard  
262 deviation statistics). It also includes basic story-telling (i.e. case studies, dictionary entries) and multimedia  
263 handling (i.e. photo gallery) services. All applications and services have independent search engines. The  
264 DevTreks web site includes videos demonstrating how the software works.

265 A DevTreks club uses standard web addresses, or URIs, to find whatever data needs to be edited or  
266 analyzed. These URIs take the form: [devtreks.org/agtreks/preview/crops/budget/Corn/12345](http://devtreks.org/agtreks/preview/crops/budget/Corn/12345). The parts of this  
267 URI can be interpreted as: host/networkgroup/webaction/network/node/commonname/id. This style of URI,  
268 and web content management, also allows machines to access data (this will be explained further in the  
269 conclusion). Members of clubs enter data using standard HTML forms which are posted to a DevTreks web site  
270 using standard HTTP methods (i.e. GET or POST). The web software uses a model-view-controller architecture  
271 to route the action needing to be taken by the server to the proper place. If the action needed is to edit data, web

272 software will edit an XML document associated with the URI and update an enterprise database. Once the edits  
273 are completed, the edited XML document is converted to XHTML using a stylesheet and the XHTML response  
274 is sent to the web browser that originated the action.

275           Using the 13 desirable social budgeting traits described above, DevTreks (version beta 0.8.7, January,  
276 2012) can identify many of its current pros and cons:

277 **2.1 Searchable.** Each web software application within DevTreks has an independent search engine. These  
278 applications include inputs, outputs, operations, components, operating budgets, capital budgets, media  
279 resources, linked views, custom documents, networks, members, clubs and services. The search engine includes  
280 several ways to filter data including keywords, categories, services and networks. Searches could be enhanced if  
281 DevTreks adopted some of the search algorithms employed by modern search engines.

282 **2.2 Web based.** All DevTreks content is delivered to all major web browsers using standard XHTML. Clients  
283 don't download any software, other than javascript libraries. DevTreks includes a sample project that  
284 demonstrates how machines can access all data found in a DevTreks database using standard web services (i.e.  
285 using REST, or Representational State Transfer, web services). A mobile phone version of DevTreks is in the  
286 planning stage.

287 **2.3 Social.** DevTreks' members organize themselves into one or more clubs which can join one or more  
288 networks. All content is owned by clubs, not members. Clubs can allow the public, or other clubs, to edit their  
289 data. DevTreks recognizes the need to adopt other types of social networking features, such as instant messaging,  
290 peer-to-peer feedback, media editing and streaming, and social networking analysis tools.

291 **2.4 Open.** DevTreks binaries and database are freely available through two open source software repositories  
292 (the DevTreks web site has links to the repositories). The source code is available to anyone planning to make a  
293 substantial contribution to the open source project. Although anyone can download and use the web software,  
294 DevTreks requests that users make voluntary donations to the project to defray the organization's expenses. As  
295 DevTreks becomes more popular, and gets the early kinks out, it plans to adopt more traditional open-source  
296 management practices such as allowing contributors to take responsibility for specific software features and  
297 using the source code review and testing services found on open source repository sites.

298 **2.5 Standards based.** All DevTreks data is delivered to web browsers using standard XHTML or XML  
299 documents. The web software has several calculators available that implement the recommendations made by the  
300 Commodity Cost and Returns Task Force (Hallam et al, 1997), including agricultural machinery, irrigation, and  
301 net present value calculators. The database has two sample economics data sets that demonstrate the use of  
302 classification standards for classifying economics data, including the construction industry's UNIFORMATII  
303 (Charette and Marshall, 1999) and an ad-hoc agricultural resource management system that was used in the case  
304 study below (section 3.3). DevTreks plans to continue adopting, or developing, improved data and information  
305 technology standards, in areas such as schema development, data definition documentation, review of economic  
306 calculation and analyses results, application programming interface development, and software testing and  
307 documentation.

308 **2.6 Linked.** DevTreks does not currently use semantic web technologies that allow Internet data to be formally  
309 linked and related. It does, however, have a specific web software application, named "Linked Views", that  
310 allow two unrelated pieces of data, such as an agricultural machinery calculator and a farm tractor input, to be  
311 easily linked (the semantic web relation between the two pieces of data is currently missing). DevTreks' business  
312 plan, which can be downloaded with the open source software, calls for devoting 15% of its time to semantic  
313 web data development in the future.

314 **2.7 Secure.** Although DevTreks makes use of standard Internet security protocols such as authorization and  
315 authentication, its security is still in beta status. It will undergo additional testing before it is fully secure.

316 **2.8 Long lived.** DevTreks is in beta testing and can't make any claims about being long lived. The current  
317 DevTreks business model, which derives most of its revenues from surcharges associated with clubs paying  
318 subscription fees to use other clubs' apps and data is untested. DevTreks plans to continue fine-tuning its  
319 business model as successes and failures become clearer.

320 **2.9 Easy to use.** The instructional videos on the DevTreks web site can give prospective users some idea about  
321 DevTreks' current ease of use. As DevTreks matures, it will adopt standard software industry practices for  
322 getting feedback from users about the software's ease of use.

323 **2.10 Rich and 2.11 Understandable.** The web software includes one application for managing media resources,  
324 such as images and videos. These images and videos can be linked to XML documents, using the “Linked  
325 Views” web software application, to produce structured “stories”, such as case studies, dictionary items, food  
326 recipes, or scientific articles. The “stories” can then be linked to all DevTreks economics content. In addition, all  
327 economics content can be localized (i.e. units, currencies, interest rates) by using DevTreks' calculators. Like any  
328 Internet company, DevTreks recognizes the need to enhance its media management capabilities by either hiring,  
329 or getting contributions from, html designers, graphic artists, video streaming experts, and other media  
330 management professionals.

331 **2.12 Extensible.** All calculators and analyzers, or “apps”, found in DevTreks make use of software patterns (i.e.  
332 Microsoft's Managed Extensibility Framework) that keep them completely separated from DevTreks core web  
333 software applications. A formal “app store” application programming interface, or API, that developers can  
334 follow to add additional apps to DevTreks, is in an early stage of development. Nevertheless, the existing  
335 extensions found in DevTreks demonstrate how to add additional apps to DevTreks. DevTreks business model,  
336 where clubs subscribe to, and pay for, the apps and data they like, is the only mechanism currently planned for  
337 ensuring the quality of these apps.

338 **2.13 Accessible.** DevTreks does not currently implement standards for making web data accessible to people  
339 with disabilities. This is a recognized shortfall that DevTreks will improve in future releases.

#### 340 **D. Can DevTreks be Used to Deliver Scientific Economics Data?**

341 Those researchers who define social budgeting in terms of better government budgeting, would probably  
342 prefer collecting and analyzing social budgeting data using an ideal government accounting and budgeting  
343 system (see Andrews (2007), What Would an Ideal Public Finance Management System Look Like?). In  
344 contrast, the DevTreks' perspective that anyone can benefit from better budgeting, prefers to use an “Ideal  
345 Economics Data Collection, Dissemination, and Analysis System”. What would such a system look like?

346 Both the National Task Force on Commodity Costs and Returns Measurement Methods (Hallam et al,  
347 1997) and Just and Pope (2001) identify the shortfalls in existing economics information systems and make  
348 recommendations about how to overcome these shortfalls. Both authors identify the need for national databases

349 of economics production data that has been collected using uniform, consistent, and scientifically valid  
350 approaches. They differ in their approach for carrying out the work. The Task Force (page 12-15) recommends a  
351 joint government survey-university farm record keeping approach. They present a table contrasting the  
352 characteristics of existing data collection systems with the proposed system (page 12-16). Just and Pope believe  
353 that neither government surveys nor state university departments are ideal for collecting and maintaining the  
354 needed national, or international, data. They think that a non-governmental organization may be a better solution.  
355 The Task Force presents a prototype of what such a system would look like (Appendix 13B) and identifies the  
356 Internet as being the preferred platform for such a system. Table 2 uses the authors' recommended characteristics  
357 for a scientifically valid agricultural economics information system to compare DevTreks with other economics  
358 information systems.

359         Table 2 shows that DevTreks compares favorably with other economics information systems – it allows  
360 clubs to: collect data anyway they prefer; collect highly detailed input, output, operating budget and capital  
361 budget data; collect longitudinal data; link to structured XML data, such as surveys, natural resource attributes,  
362 or firm characteristics; and use established practices for reducing the risk of excess survey exposure. The table  
363 points out that DevTreks comes with data sets, including a national data set of crop rotations that demonstrate  
364 most of these characteristics. The table shows that DevTreks can probably evolve to meet researchers need for  
365 national and international databases of “scientifically valid” agricultural economics production data.

366         Ultimately, clubs using DevTreks are responsible for delivering “scientifically valid” data. DevTreks  
367 provides the web software, the tools, the web site, and some sample data sets, but doesn't deliver economics  
368 content itself. The DevTreks business model envisions a market for economics data, where clubs subscribe to,  
369 and pay for, data from other clubs. Those clubs that deliver high quality data, and high quality analysis, are  
370 expected to generate higher demand for their data, and higher revenues, than other clubs.

#### 371 **Case Study - DevTreks and RZWQM in Iowa corn soybean plot data analysis**

372         Modern IT can help disentangle the complexity inherent in environmental problems in agriculture.  
373 Externalities from agricultural production are a significant problem that improved economic data could help  
374 address. Tegmeier and Duffy (2004) estimated that for the U.S. in 2002, externalities, or the cost of agricultural

375 production borne by someone other than producer, were between \$5.7 and \$16.9 billion. The externality  
376 addressed in this case study is N loading from tile-drained agriculture, a critical issue in addressing hypoxia in  
377 the Gulf of Mexico (Goolsby et al., 2000; Petrolia and Gowda, 2006). Weersink et al. point out that the textbook  
378 approach to agricultural pollution issues of setting a Pigouvian tax based on an optimal residual level may not be  
379 feasible given the high cost of quantifying the marginal benefits of reducing externalities. They propose instead  
380 that tradeoff curves “may actually be the absolute best method for the purpose of practical decision making”  
381 (2002, p. 126). Weersink et al. present a theoretical approach and do not present a case study, nor indeed, any  
382 economic or externality data.

383         The case study presented here illustrates how modern information technology can be used to improve  
384 economic data collection, dissemination, and analysis, culminating in the development of a tradeoff curve for N  
385 loading. Two software applications (Root Zone Water Quality Model, or RZWQM, and DevTreks) are used to  
386 study the economics of nitrate loading on Midwestern, tile-drained, corn soybean production. The characteristics  
387 of these software applications highlight common IT shortfalls and advances found in current modeling efforts  
388 and information systems development.

389         The data used in the case study comes from long-term plot experiments carried out by Iowa State  
390 University on the Northeast Iowa Experiment Station near Nashua, Iowa (Kanwar, 2006;  
391 <http://www.ag.iastate.edu/farms/northeast.html>, accessed 2/19/2010). The case study reflects a common need in  
392 studies of environmental problems caused by agriculture: many factors influence the quantity of pollutants  
393 released, such as climate, soils, slope, and crop rotation, tillage system, nutrient application methods, timing, and  
394 amount, and ancillary conservation practices. A similar list of factors affect the farmer’s net returns, including  
395 the producer’s knowledge, risk tolerance, equipment stocks, input and output prices, and government programs.  
396 The interactions of technical and economic factors jointly determine farm income and agricultural pollutant  
397 loading.

### 398 **3.1 RZWQM - Root Zone Water Quality Model**

399         The Root Zone Water Quality Model, RZWQM (Ahuja et al., 2000), simulates the physical, chemical,  
400 and biological processes occurring in one dimension from the bottom of the root zone to the top of the crop

401 plant. RZWQM has been extensively tested, with results published in over 200 papers. Crop yields are model  
402 outputs; RZWQM also simulates the flow of water out of the soil into tile drains and the associated concentration  
403 of nitrogen, providing a daily estimate of N loading throughout the year.

404 Measured variables used in the study come from field experiments performed between 1990 and 2003,  
405 and include weather records, corn (*Zea mays* L.) and soybean (*Glycine max* (L.) Merr.) yields, tile drainage  
406 volume, and N concentration in drainage. The data were collected from 36, 0.4 ha plots, in a randomized  
407 complete block design. Due to measurement and soil issues, data from only 30 plots were used for this study.  
408 Management treatments were studied in 6 year cycles.

409 Extensive testing of the RZWQM model with the Nashua dataset is documented in a special issue of  
410 *Geoderma* (Vol. 140, No. 3). In particular, Bakhsh et al. (2007) documented the effect of topographic variation  
411 on yield. Ma et al. (2007b) evaluated the ability of the model to simulate soil water, the water table, tile flow, and  
412 N in tile flow, as well as N uptake and mineralization. Ma et al. (2007a) compared crop yield and N losses for 2  
413 crop rotations and 4 tillage treatments. Malone et al. (2007a) simulated corn yield and N in tile flow from 5  
414 fertilizer and 5 swine manure treatments, as well as long-term effects from varying N application rates, timing,  
415 and sources, and the use of winter wheat as a cover crop. Lastly, Malone et al. (2007b) provide an empirical  
416 analysis of management effects on N loading in tile flow. Figure 1 summarizes the distribution of 420 RZWQM  
417 N loading simulations for 30 plots over 14 years compared to the measured values.

### 418 **3.2 RZWQM Stand Alone Analysis**

419 The parameterization of RZWQM for the *Geoderma* special issue was the starting point for this case  
420 study. Soil hydraulic conductivities and soil water retention curves were determined using soil samples collected  
421 in 2001 from a nearby field (Ma et al., 2007b). All soils were assumed to be the same as plot 18 at Nashua.

422 RZWQM was used to extend the observed results from the Nashua in two ways. First, each management  
423 system was simulated in RZWQM for 30 years. With a treatment cycle of 6 years, a two-year crop rotation  
424 implies only 3 observations of yield for each crop. As can be seen in Figure 2, weather in any particular year can  
425 have a significant effect on both N loading and crop yields, which affect net returns. The very high N loadings in  
426 the initial years are probably due to N applications during the very dry 1988 and 1989 years, which led to high

427 annual N loadings during the wet early 1990s, in combination with the use of moldboard plowing (Kanwar,  
428 2006). Weather data (solar radiation and daily rainfall) were derived from an on-site weather station and nearby  
429 cities (Saseendran et al., 2007) and a common climate input file from 1964-2003 was used, with the first 10  
430 years of simulation considered a warm-up period. Figure 3 compares N loading simulation results from the  
431 1990-2003 study period at Nashua with the full 30 year period 1974-2003. Average N loading is lower for all  
432 management systems for the 30 year simulations.

433         The second extension of the Nashua dataset is the consideration of management systems that have not  
434 been studied at Nashua. A total of 30 management systems were designed to explore combinations of crop  
435 rotation, tillage, N application amount, type, and timing, as well as a cover crop (Table 1). Twelve of the  
436 simulated management systems (Nashua Treatments) are very similar to those studied at Nashua and in common  
437 use in northeastern Iowa. Another 12 management systems (Low Spring) were designed to reduce N loading in  
438 tile flow by lowering the N application rate, ensuring N was applied only in the spring, and considering a winter  
439 cover crop. Lastly, 6 management systems (High Fall) reflect the fact that it simply is not feasible to apply N  
440 fertilizer on all corn acres in the spring, and that fall application rates are higher in anticipation of higher N  
441 losses because of the earlier N application. In all management systems N is applied only on the corn crop.

### 442 **3.3 DevTreks Data Stand Alone Analysis**

443         An early prototype of DevTreks was used to build two economics data sets: 196 rotational crop  
444 operating budgets stored as relational data in a database, and 504 budgets stored as whole XML documents in the  
445 same database. The 504 budgets represent a complete set of observations for all of the Nashua research plots (36,  
446 1 acre annual plots) for the period 1990 to 2003. These data sets can be found in the database that comes with  
447 the DevTreks open source software (although they have been used extensively for software testing). The budgets  
448 were built by combining National Agricultural Statistics Service (NASS) input and output prices with the  
449 experiment farm's input and output quantity data (USDA - NASS, 2003). The machinery is the experimental  
450 farm equipment actually used on the Nashua Farm. All machinery and crop budget calculations derive from  
451 guidelines recommended by Hallam et al. (1999).

452 DevTreks (the prototype had a different name) was deployed in an United States Department of  
453 Agriculture web farm (in 2004) and was configured using two web servers, one file system server, and one  
454 database server. Each server was deployed on a different physical computer. The prototype allowed two separate  
455 groups of researchers in separate states (i.e. Oregon and Arizona) to edit the budgets, at the same time, using two  
456 different web browsers (i.e. Internet Explorer and Firefox). Since all of the data delivered to the browsers was  
457 standard XHTML, neither group had to make any special adjustments to their browsers (other than allowing  
458 JAVASCRIPT to run). Both groups could instantly confirm that the edits were completed satisfactorily.  
459 DevTreks social networking features have evolved considerably since the prototype.

460 The budgets include sections for revenues, operating costs, allocated overhead costs, capital costs, and  
461 incentive-adjusted costs. Expenses were calculated for operating costs, such as materials, fuel, and repairs, as  
462 well as allocated overhead costs, like machinery capital recovery costs. Net income was derived by subtracting  
463 operating and allocated overhead costs from total revenues (crop yield \* crop price). These budgets are similar to  
464 published budgets for Iowa (Iowa State University, 2006), except the prices are for different time periods, there  
465 are some differences in the technologies represented, and there was no charge for land at Nashua.

466 Statistical comparisons, such as mean and standard deviation operating costs, were made of the crop  
467 treatments found in the 504 budget set using a DevTreks statistical analyzer. These analyzers have evolved  
468 considerably since the prototype. Current analyzers are built using self-contained software modules, or  
469 “extensions”, that can be developed independently, added to DevTreks, and then run without having to make any  
470 changes to the DevTreks software. These “extensions” are linked to specific content within DevTreks (i.e. an  
471 agricultural machinery calculator to a tractor input, a statistical analyzer to a group of crop budgets) using  
472 another software application within DevTreks.

473 The comparisons were possible because the XML nodes of each budget document were tagged using a  
474 common tagging system (i.e. LSNT for Late Spring Nitrate Test). This tagging, or data classification, system is  
475 completely *ad hoc*: no standard agricultural classification system could be found. This lack of standards to  
476 describe economic data remains a fundamental stumbling block to good data management and highlights a

477 significant, current, information technology shortfall. Data can't be aggregated, compared, and fully used for  
478 modern computing unless an agricultural classification system exists.

### 479 **3.4 RZWQM and DevTreks - An integrated analysis**

480         Perhaps the simplest way to address the hypoxia issue would be for farmers to apply less nitrogen, and a  
481 typical production economic analysis would hold everything else constant while varying the amount of N  
482 applied. An agronomic analysis on the other hand, would emphasize other management variables that can affect  
483 the amount of N that enters the tile lines, such as crop rotations, tillage systems, and N application methods and  
484 timing. The management systems studied here were designed using both perspectives. Net returns were  
485 estimated for each long-term simulation by creating individual budgets for one year's corn and soybean crop  
486 under each management system. Long-term returns were calculated on the assumption that the same series of  
487 operations were repeated within each crop rotation for 30 years. For simplicity, revenues were calculated as the  
488 crop yield for each year times the 2003 price for corn and soybeans.

489         Figure 6 shows histograms comparing the net returns calculated by DevTreks using the measured crop  
490 yields with those calculated using RZWQM simulated crop yields. A total of 173 plot years were considered for  
491 the 10 systems that were both observed at Nashua and on the list of 30 management systems for the case study  
492 (Nashua Treatment). There is an overestimation of crop yields based on the RZWQM simulations, especially for  
493 corn, as some of the factors reducing yields, such as hail and insect damage, are not simulated in RZWQM. As a  
494 consequence, mean simulated net returns of \$277/ha exceeds the mean of measured returns by \$37/ha or 15% of  
495 the observed net return. Figure 7 provides another view, showing higher net returns for the simulated than the  
496 observed for both individual years and averaged returns. For annual values, the  $r^2$  value is 0.56 and the RMSE is  
497 \$145, while for the results of averages of individual management systems, the  $r^2$  is 0.71 and the RMSE is \$64.

498         If simulated crop yields are too high, one option is to adjust yields by the difference between observed  
499 and simulated yields. For simplicity, rather than adjusting net returns downward, for long-term planning, the  
500 over-estimation of crop yield was assumed to be offset by the long-term increasing trend in crop yields, which  
501 was also ignored in the simulation. Figure 8 shows the long-term corn and soybean yield trend for Chickasaw

502 County, where Nashua is located. Corn yields are increasing at roughly 2% annually, and soybean yields at  
503 almost 1%.

504 Simulated net returns estimated for the 30 year period are consistently lower than for the 14 year period  
505 of observations at Nashua, as was the case for N loading. Interestingly, planting corn in even years as part of a  
506 corn soybean rotation seems to better than planting corn in odd years (Fig. 9). As would be expected, when  
507 comparing the net return and N loading averages for the 30 year simulations, the Low Spring systems tend to  
508 have the lowest N loading values, and the High Fall systems the highest (Fig. 10). All 3 groups of management  
509 systems exhibit a broad range of long term average net returns. There is an obvious upper bound on net returns at  
510 around \$370/ha, with a range of almost 10 to 30 kg/ha of annual average N loading, which contains systems  
511 from all 3 groups.

512 Lastly, Figure 11 shows the tradeoff of N delivered to the edge of the tile system and net returns for the  
513 simple case of varying N input amounts in the spirit of Figure 2 from Weersink et al. (2002). To emphasize the  
514 value of RZWQM in providing an agronomic foundation for the analysis, the variability in annual results from 5  
515 levels of N input across all 30 management systems is plotted, in addition to lines that show the relationships  
516 between N input and N loading, N input and returns, and N Loading and returns. The lines summarizing the  
517 tradeoffs were created using lowess relationships that ignored continuous corn systems (shown in red), as those  
518 returns were very low compared to rotations of corn and soybeans, at least based on 2003 prices. The same upper  
519 limit in net returns around \$370/ha as in Figure 10 is visible in the Tradeoff Curve panel, with corresponding  
520 annual N loading values in the range of 15 to 30 kg/ha.

521 Conservationists would want to encourage adoption of the systems at the left of the plateau, just under  
522 15 kg/ha of annual N loading, which are the Low Spring corn soybean rotations with 110 kg/ha of N applied on  
523 corn crops. Subsidies would be required for voluntary adoption of the Low Spring cover crop systems that could  
524 reduce long term simulated N loading further, to below 10 kg/ha. Although time constraints will require farmers  
525 to fall apply some N on corn crops, particularly for producers dedicated to swine production who need to dispose  
526 of swine manure, the improved returns to corn soybean rotations should be used to discourage the High Fall corn  
527 corn rotations which result in N loadings over 40 kg/ha.

528           The analysis presented in this case study could be improved in many respects. In fact, every associated  
529 profession could suggest useful improvements. A climate scientist might argue that “stationarity is dead” (Milly  
530 et al., 2008), so that climate scenarios other than the historical record should be considered. An agricultural  
531 engineer might argue for larger machinery or consideration of controlled drainage, at least on flatter areas. A soil  
532 conservationist might argue that a forward looking design would consider rotations that include crops for  
533 biomass to support cellulosic ethanol production. An economist would want to explore additional input and  
534 output price scenarios. Representatives of the Iowa Department of Natural Resources or the Environmental  
535 Protection Agency might want to focus on extreme events rather than average annual values, as well as  
536 expanding the study to consider a broader range of soils across the state. Farmers would want to know the costs  
537 on their particular farms, and they would appreciate a more dynamic decision support framework than the static  
538 plot shown in Figure 11. All of these potential improvements underscore the main point of this paper: to address  
539 agricultural economic problems systematically and flexibly, modern IT systems are needed to create economic  
540 datasets and easily link economic data to agronomic data.

#### 541 **Conclusion and Recommendation**

542           In Hans Christian Andersen's story about the Emperor with no clothes, only a child dares to state the  
543 obvious: the Emperor is naked. Economists readily admit that more and better data are needed to further their  
544 research. But not as many reach the conclusion drawn by Just and Pope (2001): the profession will produce a  
545 meager harvest if it continues to cultivate a data-limited field. Or, in Just and Pope's words, until economic  
546 datasets improve, empirical knowledge of agricultural production will remain “little more than an empty box”.

547           The need for progress in understanding the economics of natural resource issues has never been greater.  
548 As the world's population and income rise, it will be increasingly difficult to provide food at low prices while  
549 simultaneously reducing the external costs of agriculture, reducing water use, increasing production of biofuels,  
550 conserving natural ecosystems on land and at sea, managing carbon, and adapting to climate change. The  
551 potential payoff justifies a substantial effort at improving the datasets available to the profession.

552           The principle information technology advance identified in this paper is the opportunity that social  
553 budgeting offers to collect, disseminate, and analyze, basic economics data. The case study demonstrated that the

554 prototype DevTreks web software was capable of accomplishing this goal. The current software, when mature,  
555 will support the automated, online, sharing of economics data that will be so crucial in future information  
556 systems that are designed to improve agriculture.

557         The most serious information technology shortfall identified in this paper is that data, such as  
558 agricultural production data, is either not available, or cannot be easily accessed and shared. The case study  
559 highlighted this problem. RZWQM can't automatically and easily retrieve the profit data it needs from some  
560 other economics information system, such as DevTreks. DevTreks can't get the natural resources data it needs  
561 from a natural resources model, such as RZWQM. Can modern information technology alleviate this common  
562 problem?

563         At least three approaches stand out for integrating and automating natural resource and economics data  
564 sets. The first is to build either model as an independent “extension”, or pluggable software module, to the other  
565 (see the contributed packages in the R-Project, explore Grameen Bank's App Lab, or examine the  
566 DevTreks.Extensions projects in the open source software). This requires reasonably close collaboration among  
567 the model builders on computer platform choices, data models, and data integration interfaces. When both  
568 models belong to the same research lab, or Internet company, this is probably the best choice. Closely integrated  
569 models can do a very good job of automating data collection, dissemination and analysis and avoid problems  
570 associated with data repetition and incompatibility.

571         A second approach relies on standard Internet protocols, such as the HyperText Transfer Protocol  
572 (HTTP) and the Universal Resource Identifier (URI), to improve data sharing. The HTTP protocol defines a way  
573 to identify unique resources (URIs), such as images and documents, on the Internet and transfer them to web  
574 browsers. The term Representational State Transfer, or REST, web services refer to machines using these  
575 protocols to directly share data. The data sent from a URI to a web browser, or machine, “represents” the actual  
576 data being stored on a web server. HTTP messages, containing commands such as GET or POST, describe what  
577 to do with these representations.

578         To make this method practical, additional search parameters, sent as HTTP query or request parameters  
579 (i.e. “http://mysearchengine.org/search?keyword=corn”), are often needed. The resultant web service “APIs”

580 (Application Programming Interfaces - instructions describing how data can be mechanically accessed and used)  
581 are usually published on the web sites. An example of this type of web service is Amazon's data storage service  
582 at <http://aws.amazon.com/s3/>. Natural resource and economic models could access remote data by writing  
583 software instructions that use the web service API. DevTreks includes a sample project that uses Microsoft's  
584 Windows Communication Framework and OData data access technology to deliver REST web services.

585         The third method requires the most time and effort but is probably the best long term solution. Semantic  
586 web technologies are used to relate, classify, and analyze the economics and biophysical data. Berners-Lee  
587 (2009) suggests three basic rules to follow when first using this method:

588 1. Use HTTP names to make data available. This is the same as the RESTful web services mentioned in method  
589 two.

590 2. Define content: Formally define the content of the data that is returned from a URI. This can be done using  
591 XML schemas or some other data definition language. A particularly useful contribution along these lines would  
592 be a way to describe management operations that could be used to simultaneously parametrize a biophysical  
593 model and an economic budgeting tool.

594 3. Return linked data: Data that is returned from a URI should be linked data, not stand alone data. Using  
595 RZWQM and DevTreks as an example, links would be made between the farm budgets and the nitrate loading  
596 projections. When someone requests data at a standard DevTreks URI they would receive both the budgets and  
597 the links to the biophysical data. The W3C, an Internet standards setting organization, has a lot of semantic web  
598 technologies available for building these types of links.

599         The principle advantage to this last approach is that it forces the application developers to focus on the  
600 real data problems - data integration is not unique to their models - encompassing the whole fields of agricultural  
601 resource management and water quality science. The principle disadvantage is that collaboration may be needed  
602 across entire professions for building full blown semantic web features such as data classification and  
603 relationship systems. This can be time consuming and expensive, but one can imagine the benefits of being able  
604 to pass the same description of management operations to both a biophysical model and an economic model.

605 An example of this approach in the field of agriculture is The Agricultural Information Management  
606 Standards (AIMS) website (FAO, 2010). The stated goal of this web site to “improve coherence among  
607 agricultural information systems that will make such systems interoperable. The objectives of AIMS are to  
608 create a clearing house for information management standards, and to share and promote the use of common  
609 methodologies and tools”. Professional examples from other fields can be found among biologists  
610 (Encyclopedia of Life, [www.eol.org](http://www.eol.org)) and hydrologists (Consortium of Universities for the Advancement of  
611 Hydrologic Science - Hydrologic Information System, <http://his.cuahsi.org/>).

612 This paper makes three contributions. First, we propose that social budgeting is a feasible approach to  
613 building the economic datasets needed to address key agricultural issues and we describe the characteristics  
614 needed for social budgeting to be successful. Second, we demonstrate that an example social budgeting web  
615 software program, DevTreks, can be used, successfully, to build agricultural economics production datasets.  
616 Third, we present a case study linking biophysical and economic data to illustrate current information technology  
617 shortfalls (i.e. data couldn't be easily shared), advances (i.e. building online economics data sets), and planned  
618 improvements (i.e. using both RESTful web services and semantic web methods to link and share data in the  
619 future).

620 Open source, on-line data collection, dissemination, classification, and analysis efforts have been  
621 identified as important contributors to modern information systems development. Contribute to these projects by  
622 entering data sets, creating "extensions", and helping to build semantic and ontological systems. Open source  
623 efforts may hold the greatest promise for delivering the technical and economic information needed to improve  
624 agriculture.

625

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754 **Figure Captions**

755

756 Figure 1. Histograms of RZWQM simulated and measured annual N loadings into tileflow for 30 plots over 14  
757 years.

758

759 Figure 2. Box plots of measured N loading and net returns based on measured yields at Nashua for the 1990-  
760 2003 period.

761

762 Figure 3. Comparison of 14 year average simulation results by management system compared to 30 year results  
763 for N loading. CC is continuous corn; CS is corn in even years; SC is soybeans in even years.

764

765 Figure 4. Histograms of RZWQM simulated and measured annual net returns for 30 plots over 14 years.

766

767 Figure 5. Scatterplot of net returns based on simulated and observed crop yields showing both annual values and  
768 averaged net returns for 10 management systems. See Table 1 for management system descriptions and number  
769 of years compared for each system.

770

771 Figure 6. Time series plot showing increasing crop yield trends for Chickasaw County, IA.

772

773 Figure 7. Comparison of 14 year average simulation results by management system compared to 30 year results  
774 for net returns. CC is continuous corn; CS is corn in even years; SC is soybeans in even years.

775

776 Figure 8. A scatterplot comparing the net returns and N loading from 30 year simulation results for 3 groups of  
777 management systems.

778

779 Figure 9. Plot summarizing the tradeoff approach proposed by Weersink et al. (2002) for 30 year simulation  
780 results of 30 management options on N loading and net returns. CC is continuous corn; CS is corn in even years;  
781 SC is soybeans in even years. Locally smoothed lines ignore CC values.

782



784 Table 1. Management systems used in long-term simulations. CP="Chisel Plow", NT="No Till";  
 785 CC="Continuous Corn", CS="Corn-Soybean", SC="Soybean-Corn"; SM="Swine Manure", UAN=""  
 786 Urea Ammonium Nitrate".

	<b>Tillage</b>	<b>Rotation</b>	<b>N Amount kg N/ha</b>	<b>N Type</b>	<b>Season</b>	<b>Cover Crop</b>	<b>Plot Years</b>
Nashua							
Treatments	CP	CC	150	SM	Fall	No	18
	CP	CS	150	SM	Fall	No	20
	CP	SC	150	SM	Fall	No	22
	NT	CC	150	SM	Spring	No	
	NT	CS	150	SM	Spring	No	8
	NT	SC	150	SM	Spring	No	6
	CP	CC	150	UAN	Spring	No	18
	CP	CS	150	UAN	Spring	No	30
	CP	SC	150	UAN	Spring	No	27
	NT	CC	150	UAN	Spring	No	
	NT	CS	150	UAN	Spring	No	12
	NT	SC	150	UAN	Spring	No	12
Low Spring	CP	CC	135	UAN	Spring	No	
	CP	CC	135	UAN	Spring	Yes	
	NT	CC	135	UAN	Spring	No	
	NT	CC	135	UAN	Spring	Yes	
	CP	CS	110	UAN	Spring	No	
	CP	CS	110	UAN	Spring	Yes	
	NT	CS	110	UAN	Spring	No	
	NT	CS	110	UAN	Spring	Yes	
	CP	SC	110	UAN	Spring	No	
	CP	SC	110	UAN	Spring	Yes	
	NT	SC	110	UAN	Spring	No	
	NT	SC	110	UAN	Spring	Yes	
High Fall	CP	CC	200	Anhydrous	Fall	No	
	NT	CC	200	Anhydrous	Fall	No	
	CP	CS	168	Anhydrous	Fall	No	
	NT	CS	168	Anhydrous	Fall	No	
	CP	SC	168	Anhydrous	Fall	No	
	NT	SC	168	Anhydrous	Fall	No	

787

788

Characteristics of Data	Current USDA Survey	Current University Farm Record System	Integrated USDA and University System	DevTreks NGO System
1. Based on probability sampling (1)	YES	NO	YES	YES. Clubs can collect data anyway they prefer, including from probability surveys.
2. Consistent procedures used across states (1)	YES	NO	YES	YES (see #3)
3. Data accuracy (1) Accurate reporting Detailed information Close local scrutiny	MODERATE	MODERATE	MODERATE	YES. DevTreks includes a national data set of crop rotations for the USA, demonstrating how consistent, accurate, and detailed data can be collected using DevTreks. The data was collected from cooperative extension offices throughout the USA, who collected the base data using a great deal of local scrutiny.
4. Longitudinal data(1)	NO	YES	YES	YES. DevTreks includes a sample data set containing 14 years of experimental plot data for corn soybean rotations demonstrating how longitudinal data can be collected.
5. Cost (1)	HIGH	HIGH	HIGH	LESS HIGH. The social networking features of DevTreks may be able to decrease this cost substantially
6. Data availability (1)	Limited	Varies by State	YES	Fee-based or Free. DevTreks business model allows clubs to charge subscription fees for their data. Data will be available to wider audiences if clubs charge either low, or no, fee.
7. Highly detailed microeconomics data (2)	NO (input use by crop and application rate missing)	YES	MODERATE	YES (see #3 and #4)
8. Highly detailed investment data (2)	NO	Varies by State	MODERATE	YES. DevTreks includes a capital budgeting application with data that is as detailed as the operating

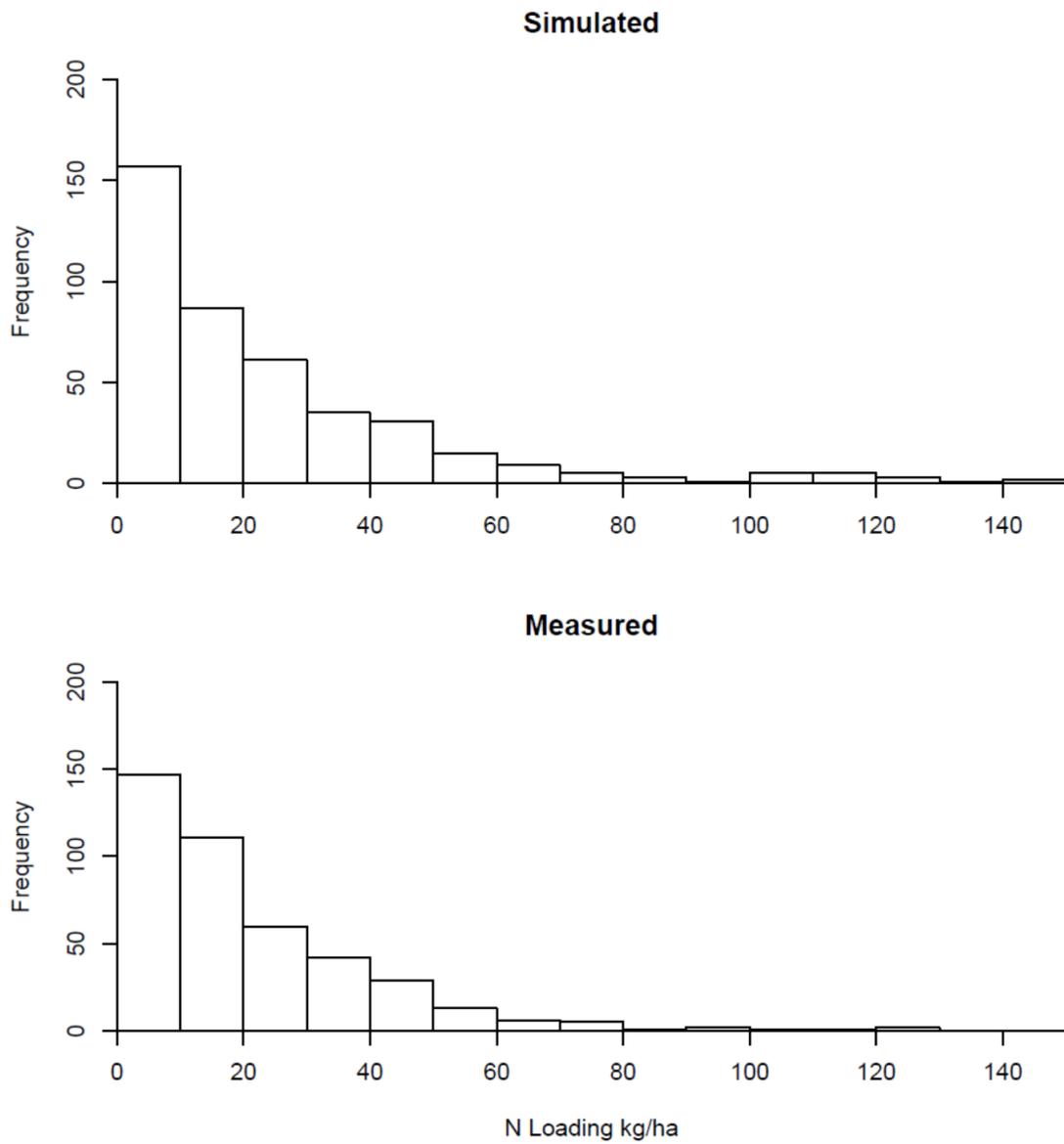
				budget data (i.e. inputs and outputs).
9. Other data (i.e. firm characteristics, soil quality) (2)	YES	YES	YES	<b>NOT DIRECTLY.</b> DevTreks allows economics data to be linked to structured XML data, such as surveys, natural resource attributes, or firm characteristics.
10. Excessive survey exposure and right to privacy exposure (2)	YES	NO	POSSIBLY	<b>POSSIBLY.</b> Clubs who collect their data using surveys risk the problem of excess survey exposure. All clubs collecting firm level data face right-to-privacy issues. These risks and issues have been mitigated by techniques developed by cooperative extension economists, such as collecting data from groups of farmers rather than individual farmers.

790 (1) Taken from Table 12.1, page 12-16, in Hallam et al, 1997

791 (2) Derived from Just and Pope (2001)

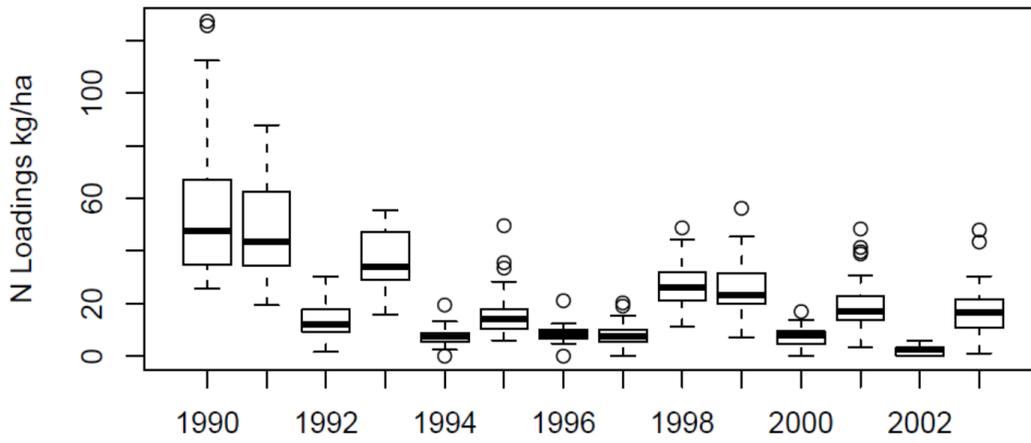
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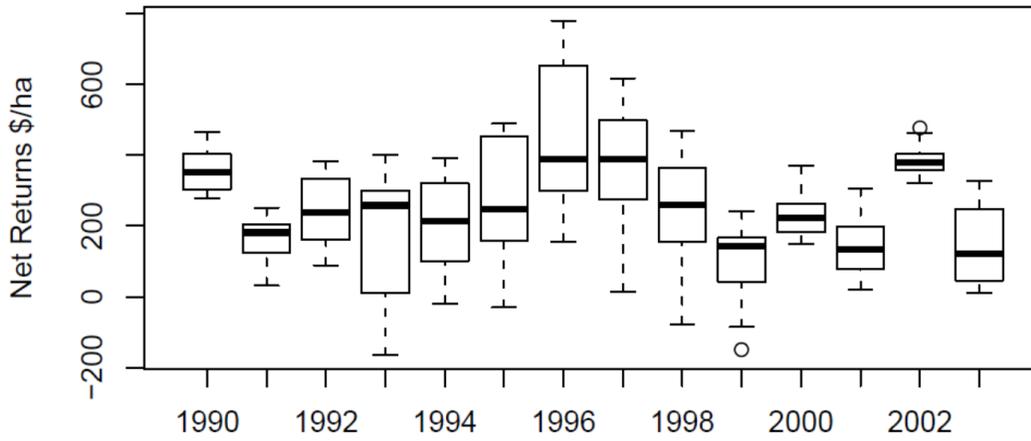


794   
 795 Figure 1  
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**Distribution of Measured N Loadings**



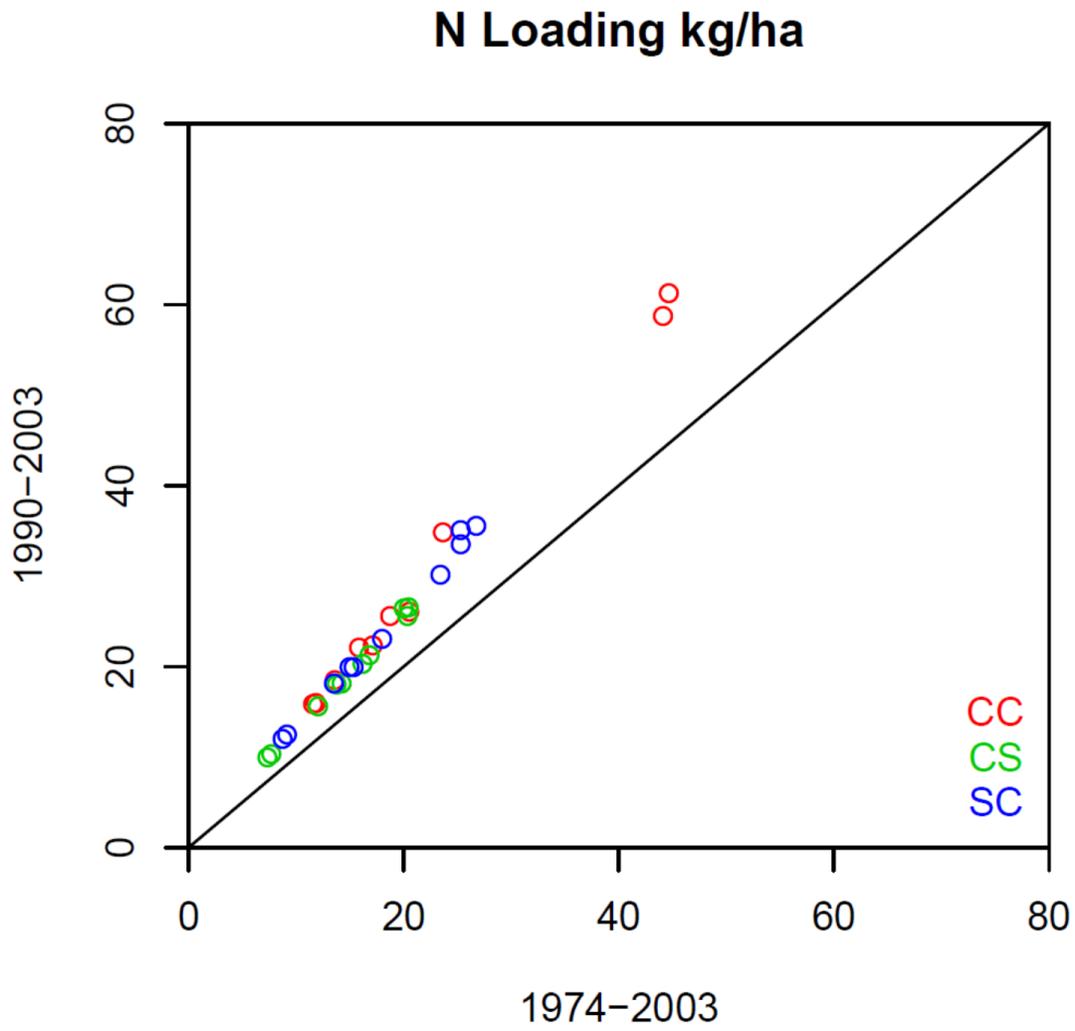
**Distribution of Measured Net Returns**



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798 Figure 2

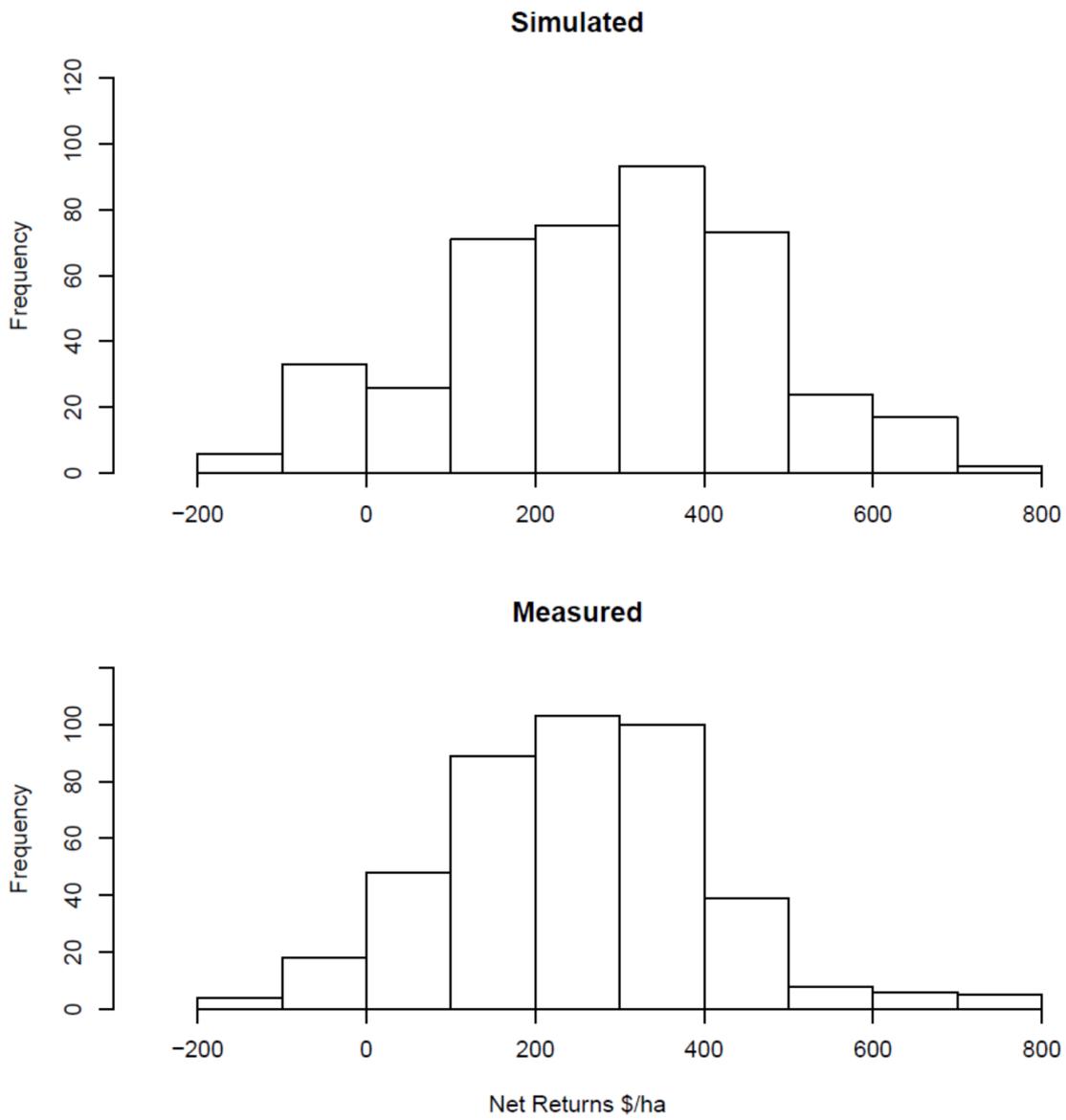
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801 Figure 3

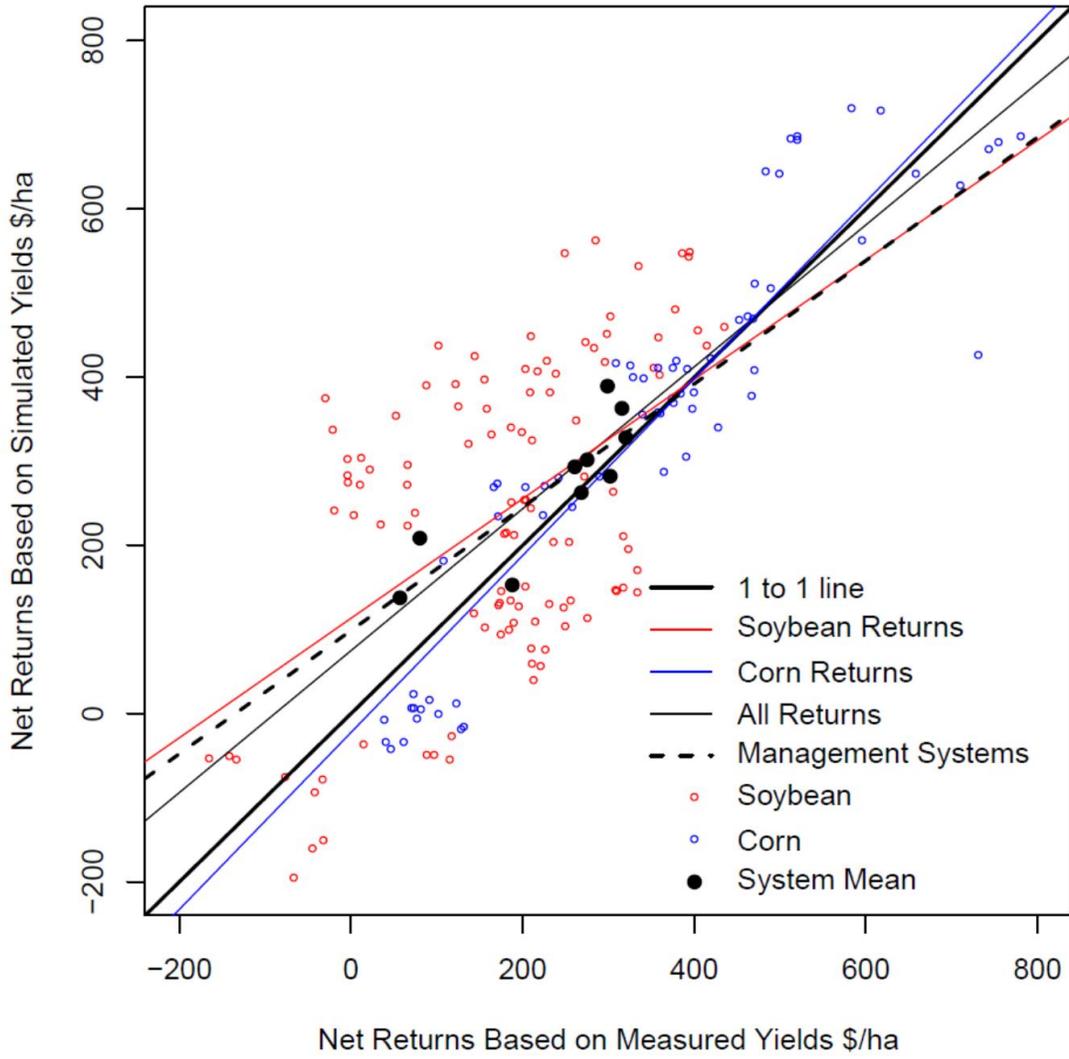
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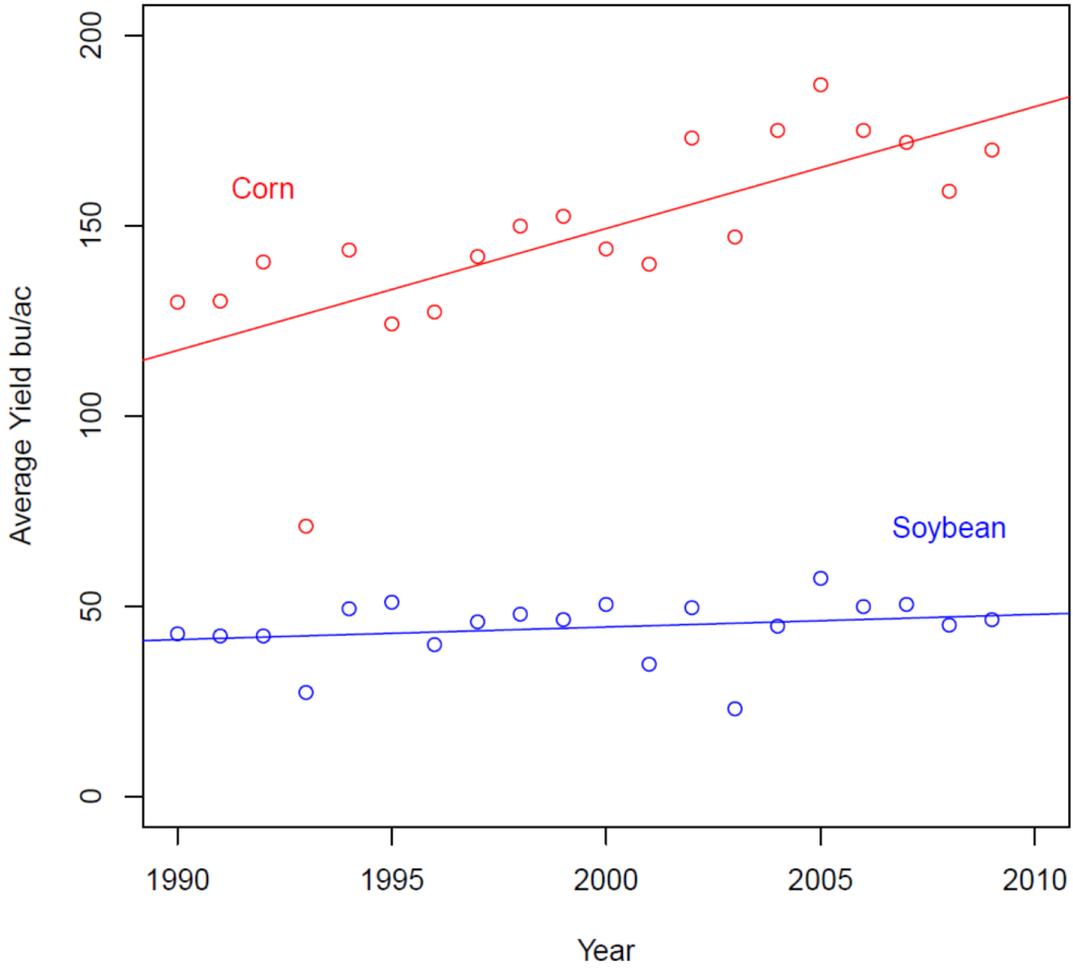
804 Figure 4

### Net Returns for 10 Management Systems

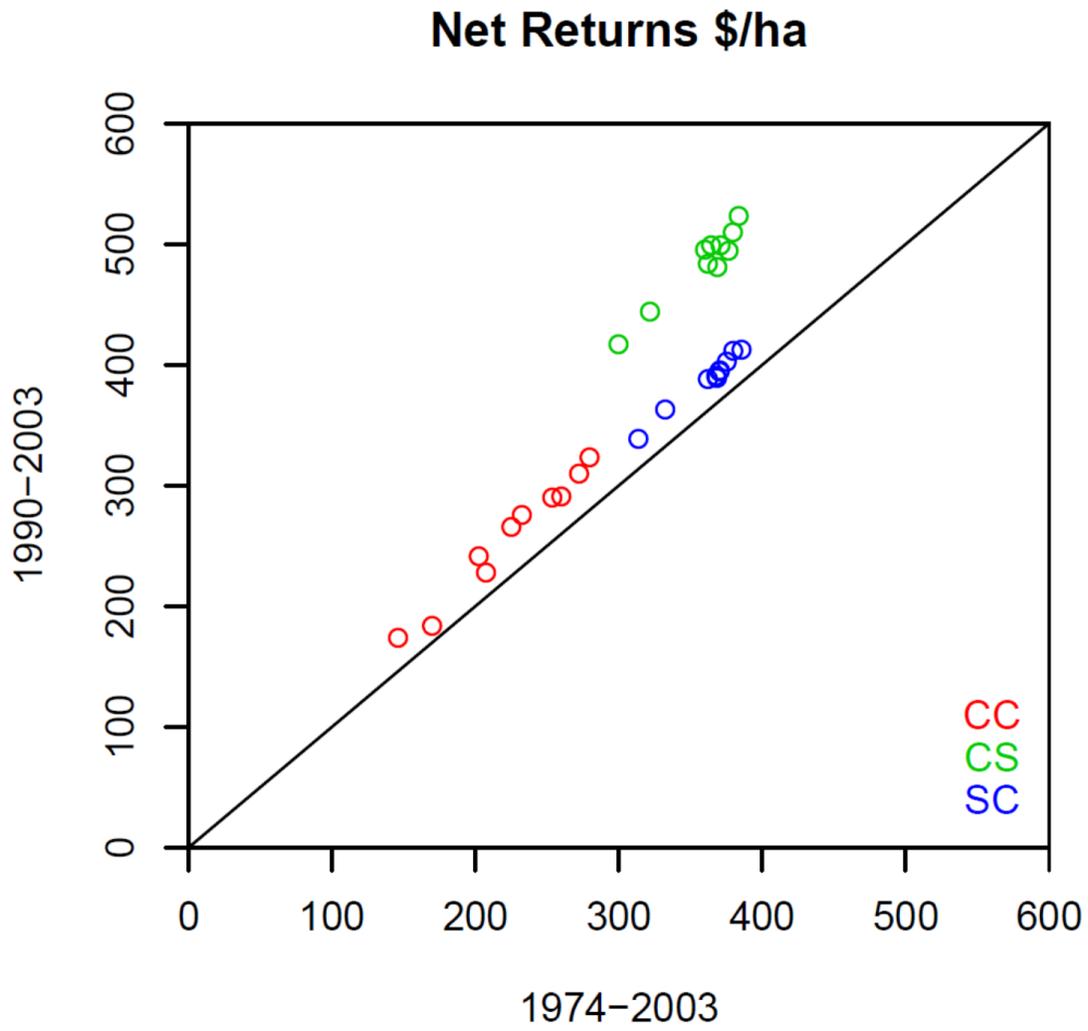


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806 Figure 5  
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### Trend in Chickasaw, IA County Crop Yields

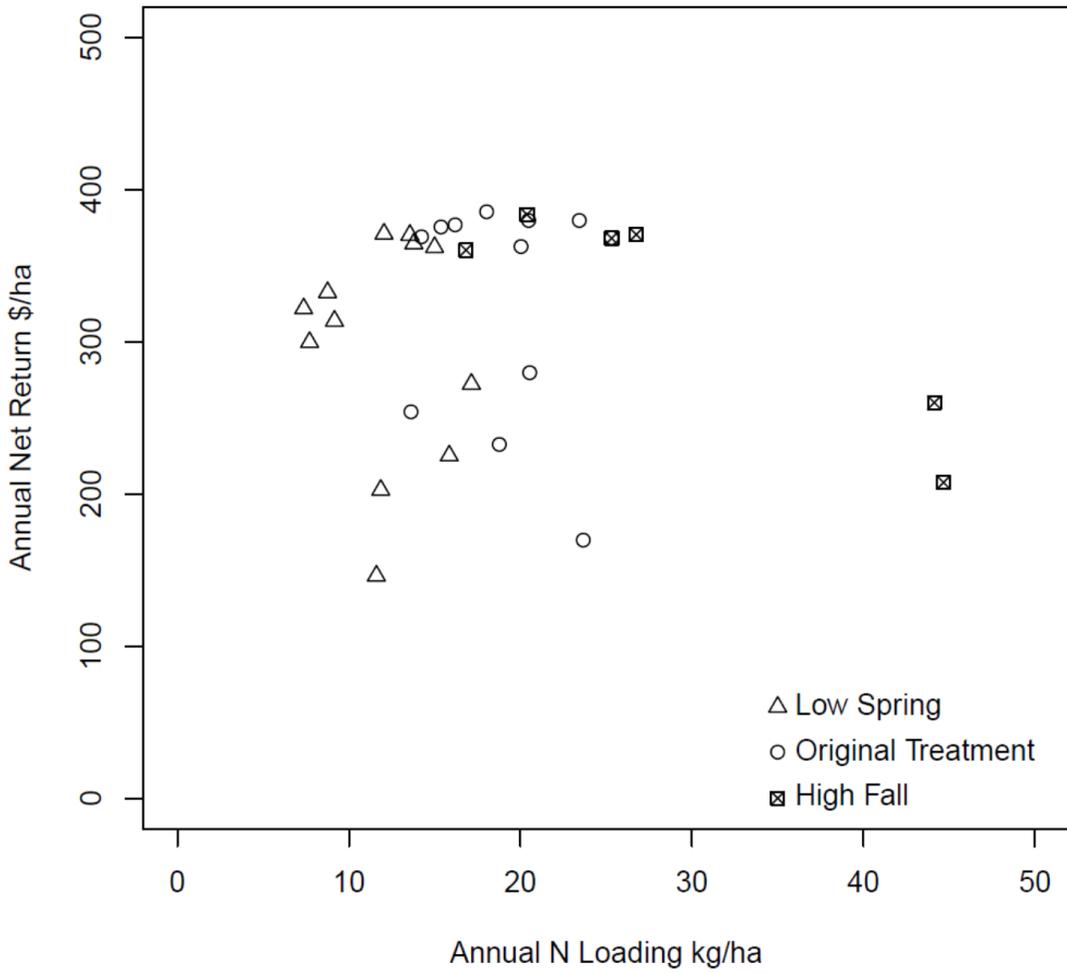


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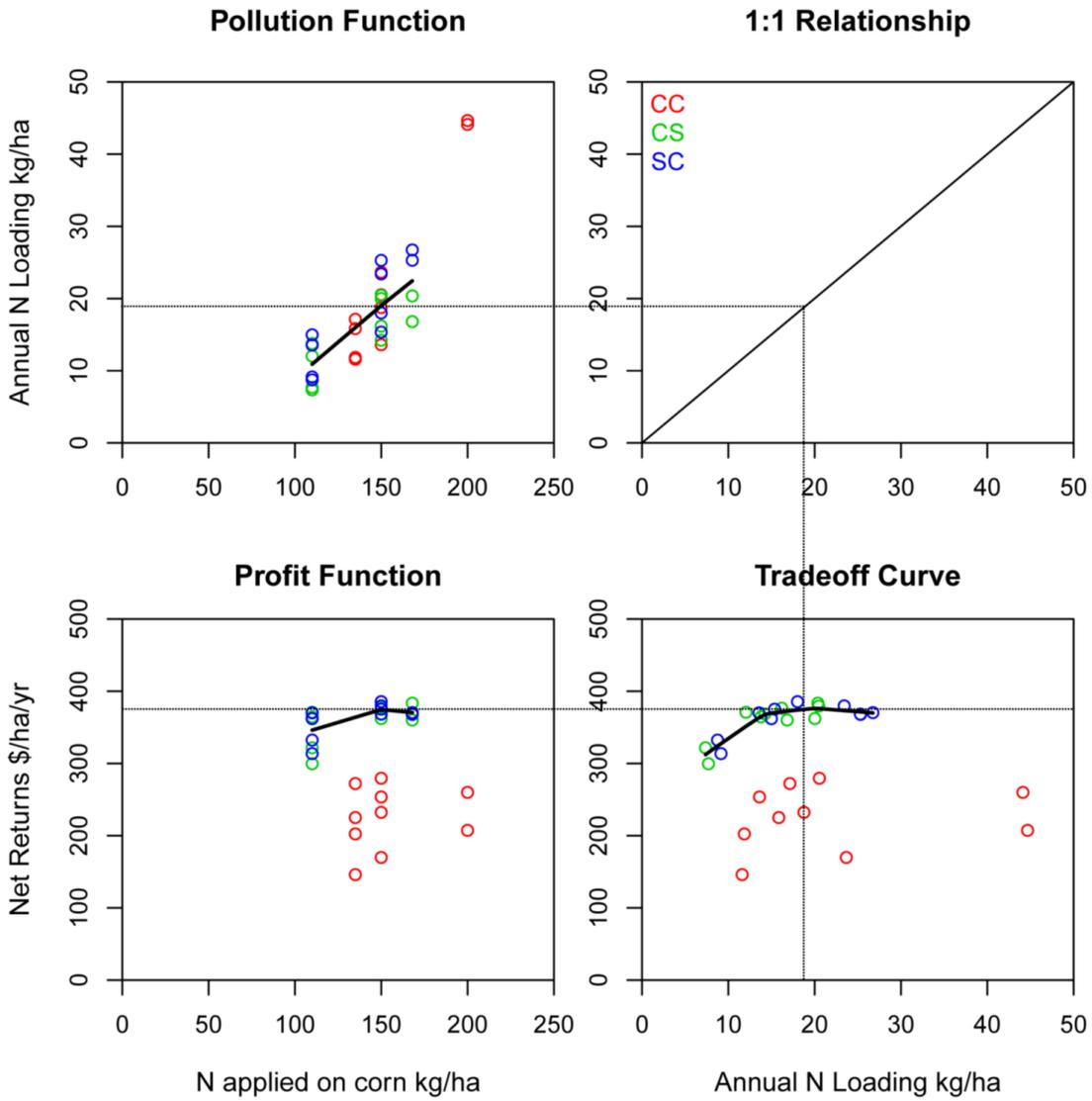


811  
812 Figure 7  
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### Simulated 30 Year Average Tradeoff



814  
815 Figure 8  
816



817  
818 Figure 9  
819