

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

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Abstract

Agricultural economics needs more and better data. This paper proposes a new approach, social budgeting, for building and analyzing agricultural economic datasets. Social budgeting employs modern information technologies, such as social networks and cloud computing data centers, to collect and analyze economics data over the Internet. DevTreks, an open source, social budgeting web software program, is introduced as an example. A case study involving Midwestern corn soybean production uses DevTreks and the process-based Root Zone Water Quality Model (RZWQM) to illustrate how modern information technology can be used to study economic-pollutant tradeoffs. The paper recommends starting or joining open source, on-line efforts aimed at integrating the physical, chemical, biological, and economic aspects of agriculture.

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

1. Introduction - the Emperor has no clothes

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 Much of this data suffers from lack of standardization, antiquated software, single user bias, and outdated approaches to information technology (IT) data collection, dissemination, and analysis.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

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

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

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

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). 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/). *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. 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. Tegtmeier 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).

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

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

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

The comparisons were possible because the XML nodes of each budget document were tagged using a common tagging system (i.e. LSNT for Late Spring Nitrate Test). This tagging, or data classification, system is completely *ad hoc*: no standard agricultural classification system could be found. This lack of standards to 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

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

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

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

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

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

Conclusion and Recommendation

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

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

The principle information technology advance identified in this paper is the opportunity that social 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.

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

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

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

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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".

	Tillage	Rotation	N Amount kg N/ha	N Type	Season	Cover Crop	Plot Years
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	

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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

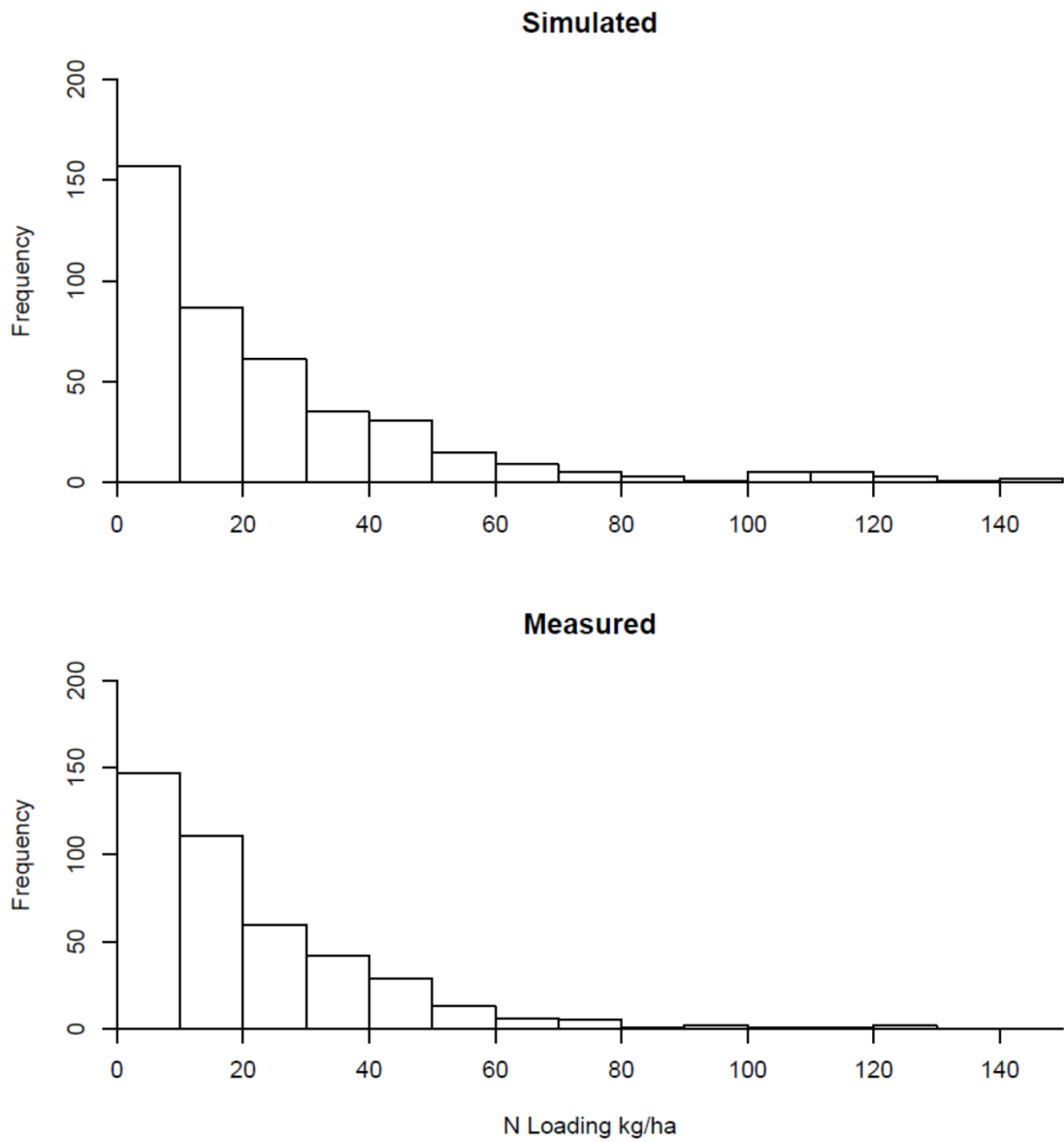
				budget data (i.e. inputs and outputs).
9. Other data (i.e. firm characteristics, soil quality) (2)	YES	YES	YES	NOT DIRECTLY. DevTreks allows economics data to be linked to structured XML data, such as surveys, natural resource attributes, or firm characteristics.
10. Excessive survey exposure and right to privacy exposure (2)	YES	NO	POSSIBLY	POSSIBLY. 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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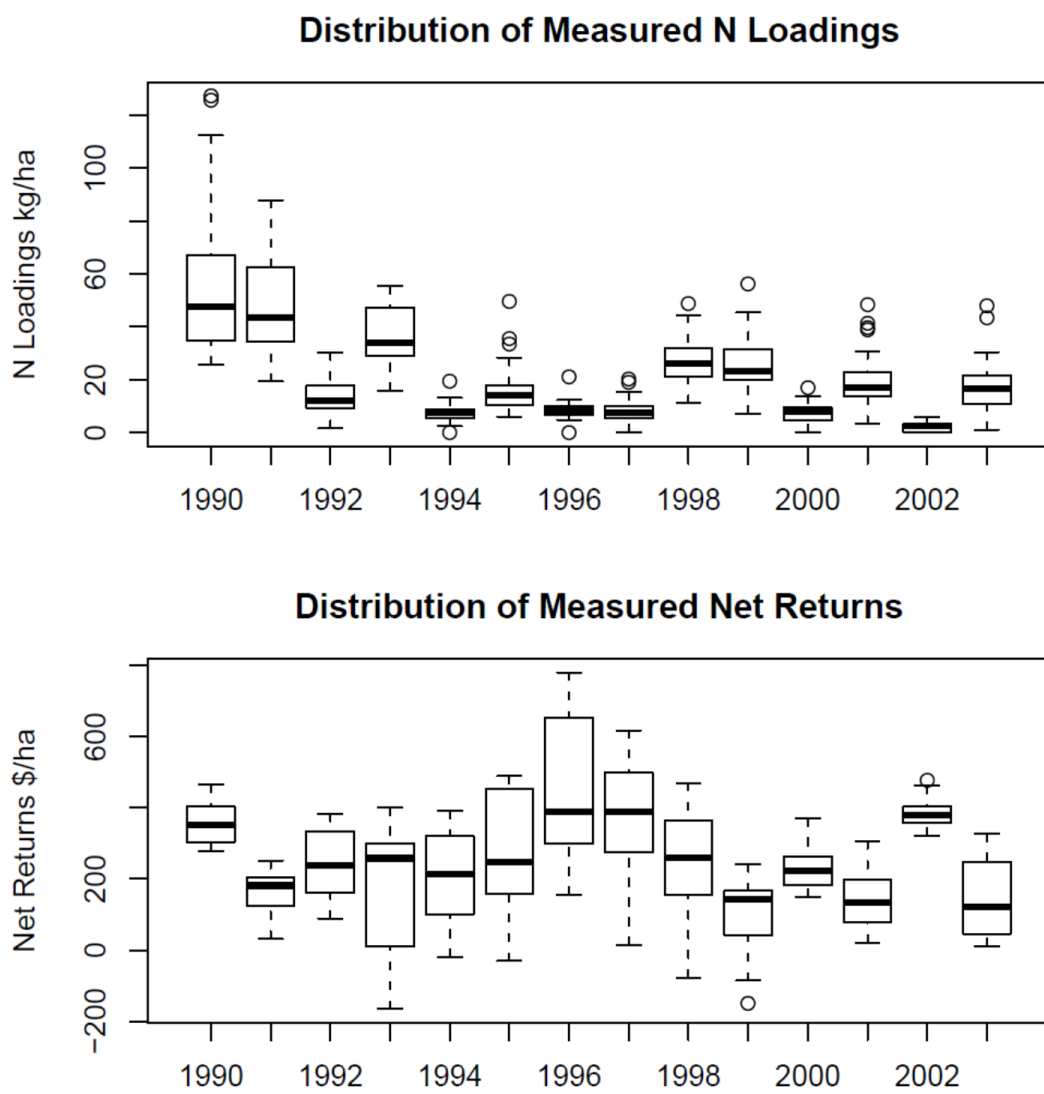
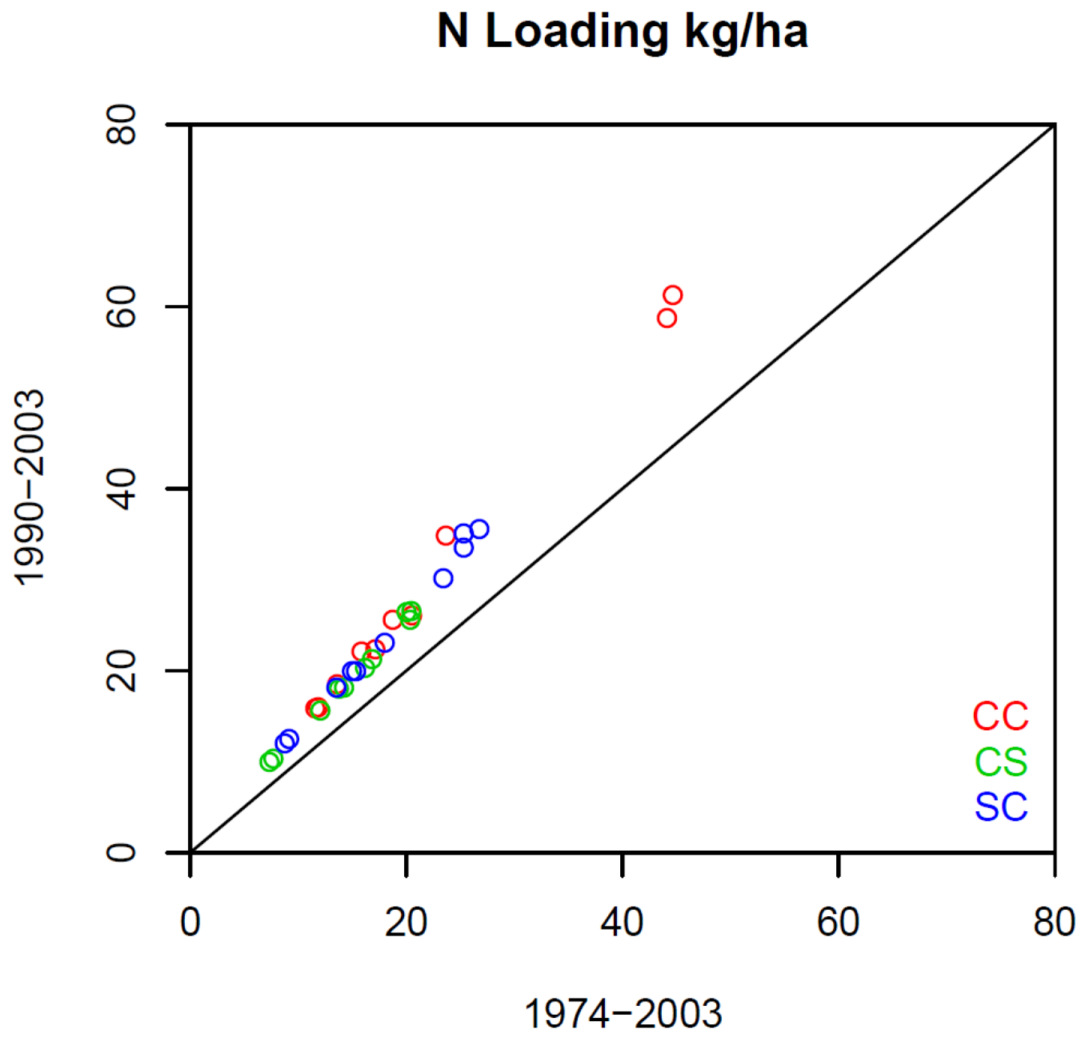


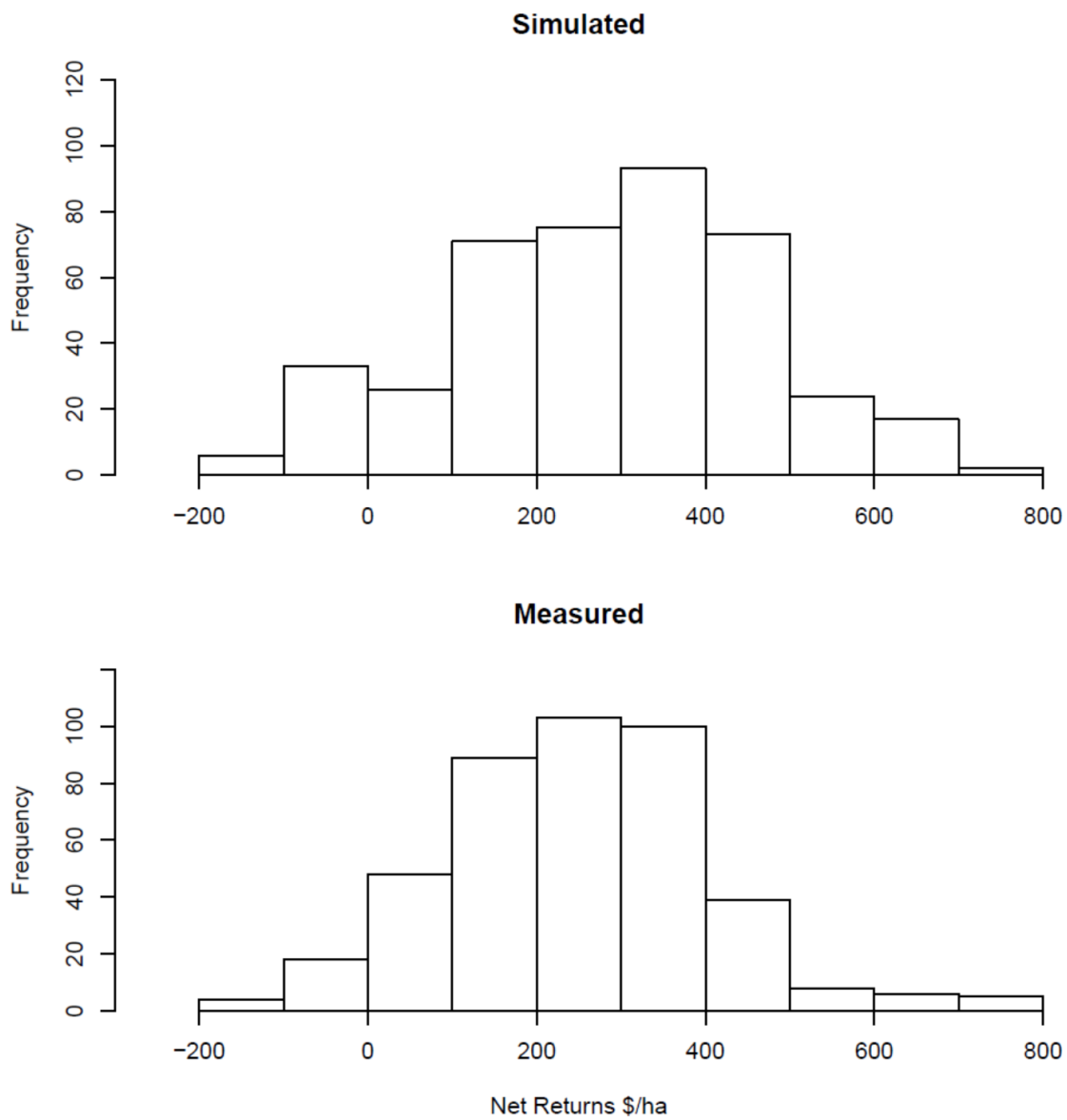
Figure 2



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801 Figure 3

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804 Figure 4

Net Returns for 10 Management Systems

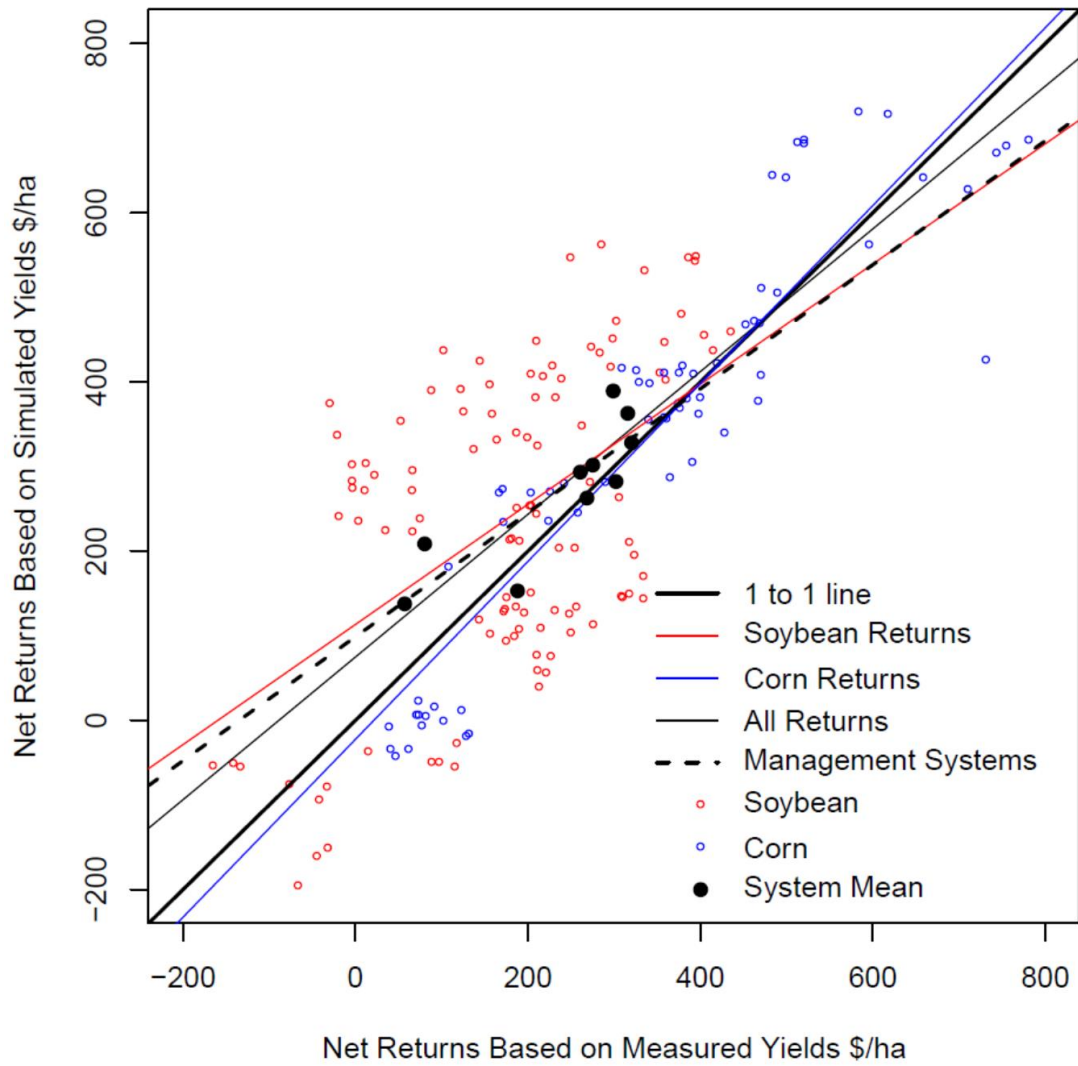


Figure 5

Trend in Chickasaw, IA County Crop Yields

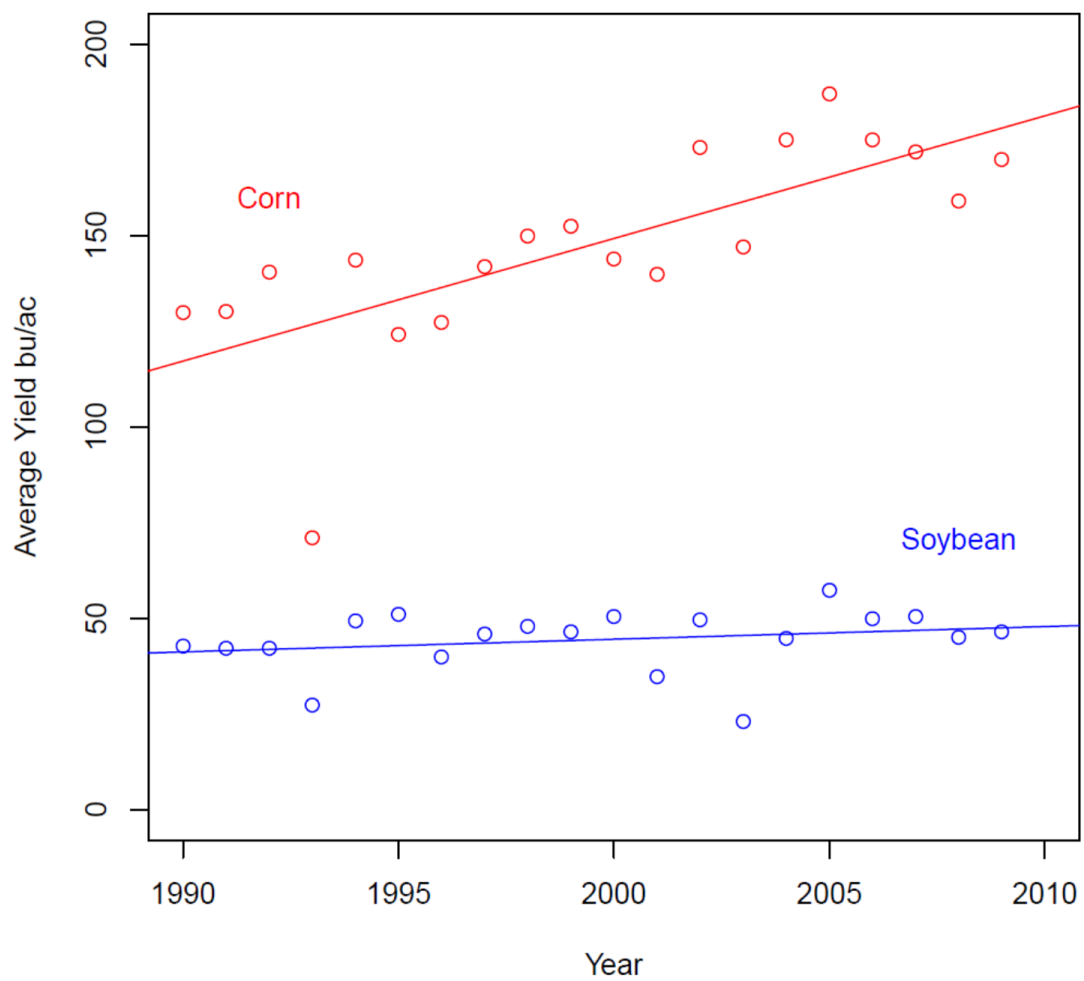
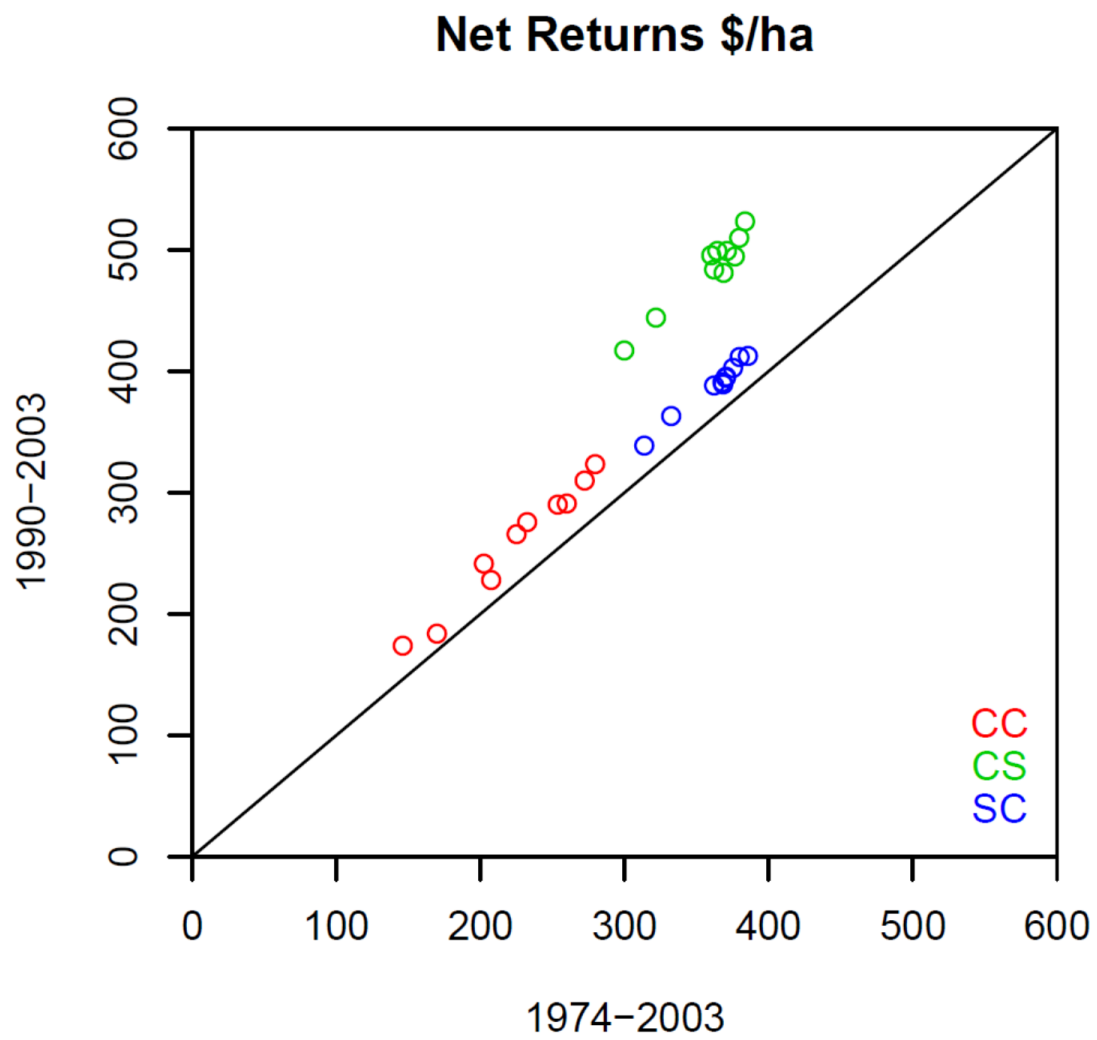


Figure 6



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812 Figure 7
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Simulated 30 Year Average Tradeoff

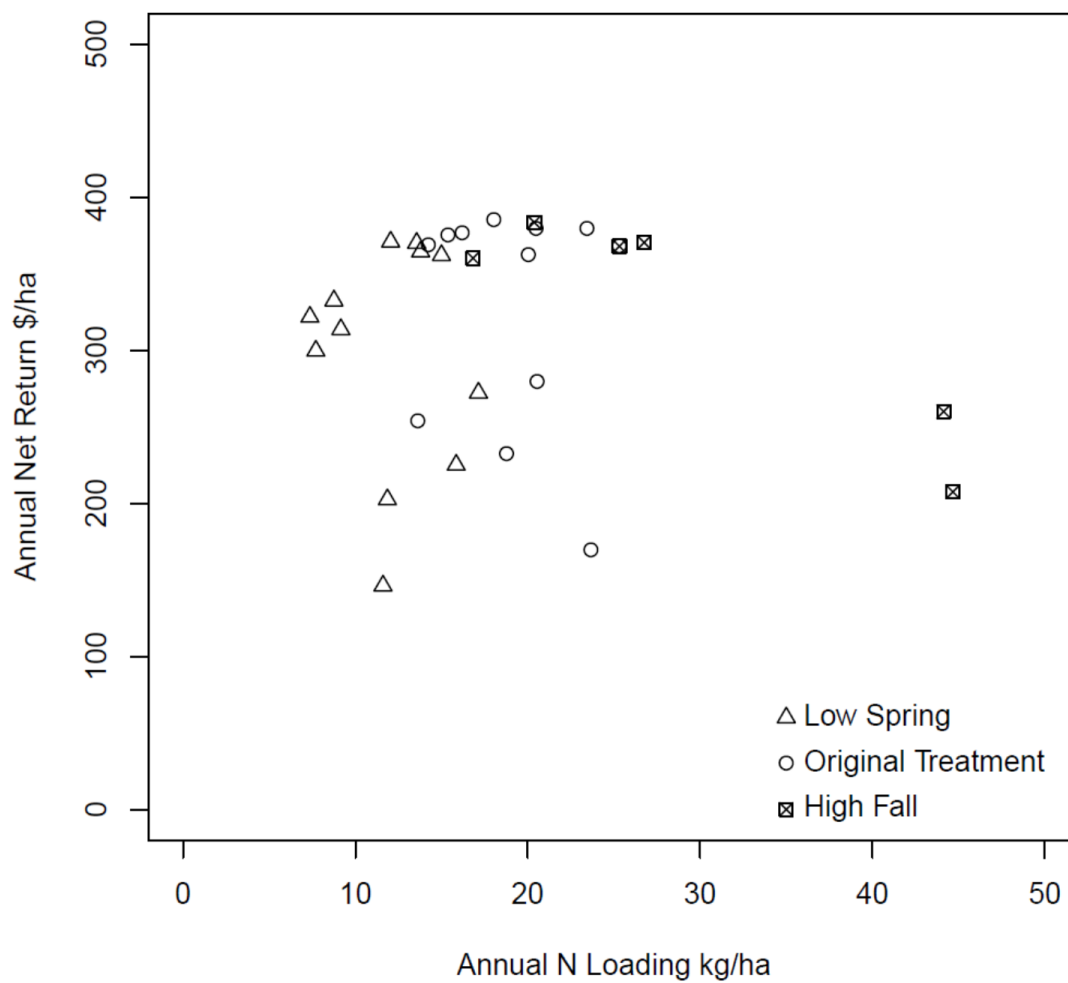


Figure 8

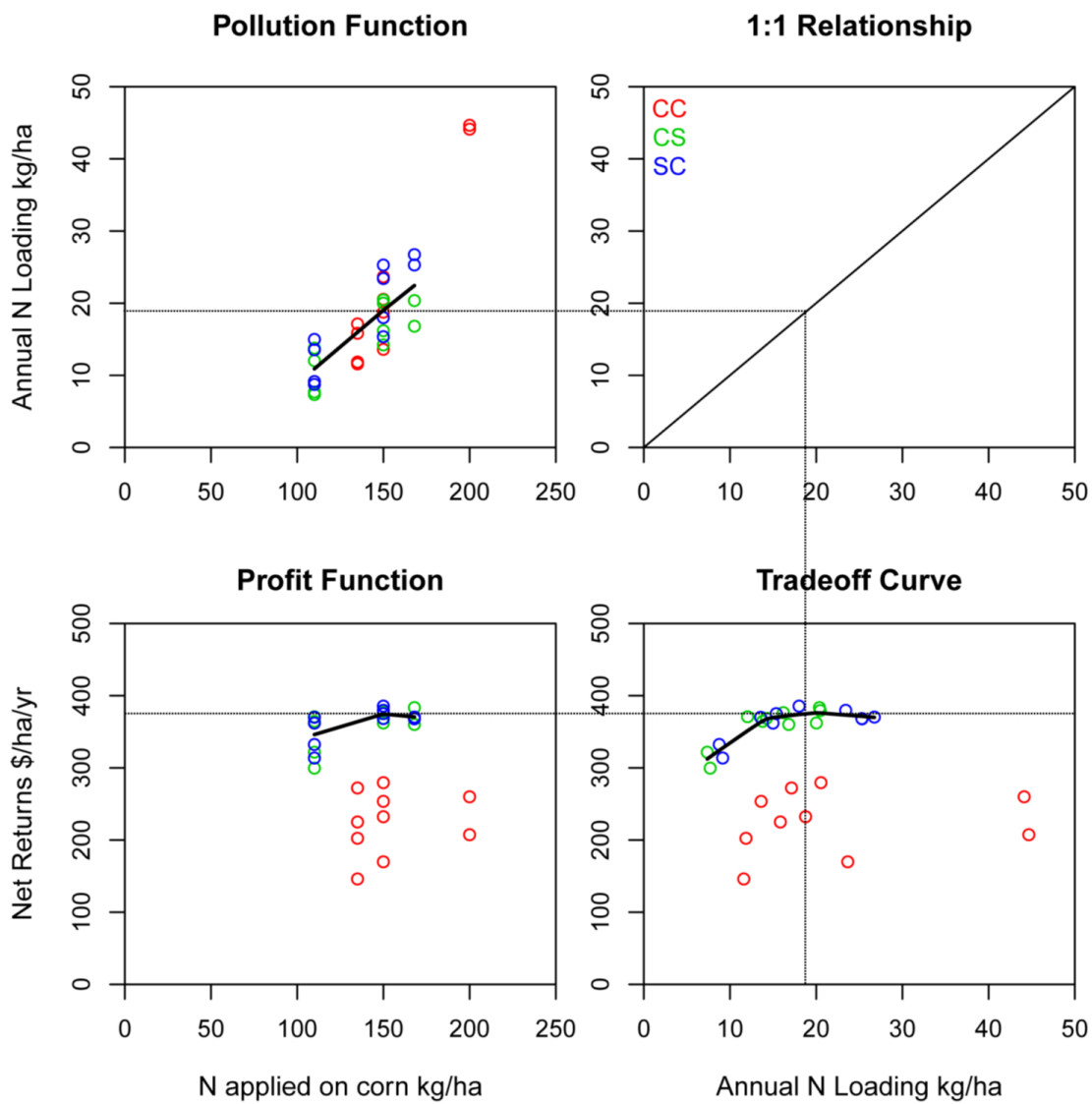


Figure 9