

1 Final Report

2 Project Title: Assessing the Effects of Conservation Practices and Fertilizer Application Methods
3 on Nitrogen and Phosphorus Loss from Farm Fields – A Meta Analysis

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6 Introduction

7 Fertilizers increase crop yields but have consequences for society in the United States
8 because they may be lost from agricultural fields and become pollutants to receiving water
9 bodies (Sharpley 1995, Vagstad et al. 1997, Fuhrer 1999, Crain et al. 2006). Many models track
10 the transport of nutrients in a watershed, including agricultural inputs (Robertson and Saad
11 2011). Detailed descriptions of site characteristics allow these models to account for factors
12 such as soil type, crop type, and precipitation that can influence nutrient transport (Robertson and
13 Saad 2011). However, incomplete descriptions often limit applying a model to novel locations.
14 Additionally, site characteristics that differ from field to field are seldom taken into
15 consideration.

16 The Measured Annual Nutrient loads from AGricultural Environments (MANAGE) database
17 was created by the USDA to provide field scale information on nutrient losses, load, and
18 concentration data from agricultural lands (Harmel et al. 2006, Harmel et al. 2008). This
19 database provides support for models developed to address the nonpoint source transport of
20 nutrients (Harmel et al. 2006). It includes land uses, rainfall quantities, soil loss data, and other
21 site characteristics. This database is a compilation of data from several publications that contain
22 field scale nutrient load and concentration data (Harmel et al. 2006). Data from each publication

23 is input into all applicable tables that comprise the MANAGE database. In the most recent
24 edition of MANAGE there are four different tables available – agricultural load, agricultural
25 concentration, forest load, and forest data. The focus of this project is on the agricultural load
26 table in the MANAGE database.

27 After the start of this project an addition table became available. Christianson and Harmel
28 (2015) created an additional table, the MANAGE “Drain Load” database, to be added to the
29 tables in the MANAGE database and published a summary of their findings based on the “Drain
30 Load” table.

31 Since its creation, there have been 6 updates to the database and 4 articles have been
32 published describing these updates and subsequent areas needing further research (Harmel et al.
33 2006, Harmel et al. 2008). However, Harmel *et al.* (2006) outlined criteria for adding papers to
34 the MANAGE database which included only results published in peer-reviewed journals. This
35 criterion creates a publication bias in the MANAGE database by omitting any other sources of
36 information, including relevant agency reports (Easterbrook et al. 1991).

37 Harmel *et al.* (2006) found no significant decrease of nutrient loss with the implementation
38 of conservation practices, but did not include important confounding factors such as land use and
39 soil type (Harmel et al. 2006, Harmel et al. 2008). For example, the primary goals of
40 conservation practices are to reduce runoff and nutrients leaving the field, thus we expect the
41 practices to reduce nutrient loss (Duriancik et al. 2008). Other characteristics that influence
42 nutrient loss include fertilizer application timing and method. Fertilizers are important to
43 consider because they are applied with the intent to increase crop yields, but nutrients not taken
44 up by crops can leach from fields and become pollutants (Smith et al. 2007).

45 This project aims to quantify the effects of conservation and fertilizer application
46 practices on nutrient losses and crop yields. The October 2014 edition of the MANAGE
47 database, created by the USDA, is used as the starting point for our statistical analyses because it
48 contains the most information available in any edition of MANAGE up to this point. The
49 database was created to provide field scale information on nutrient loss from agricultural lands
50 and to assist farmers and managers to determine the most effective field management practices
51 (Harmel et al. 2006, Harmel et al. 2008). Currently, summary statistics of the October 2014
52 edition and analysis of variables that have low data availability have been performed. This
53 project and the funding provided by The Fertilizer Institute are the beginning of a student's
54 Master's thesis that began in Fall 2014. As such, the final results of this project will be published
55 in a student's thesis after the thesis defense. This is projected to occur in Spring 2016. The
56 results included in this final report are the findings that have been accomplished thus far. In this
57 report we summarize (1) the efforts of updating the MANAGE database, (2) the development of
58 statistical methods for the proposed meta-analysis and (3) current results from the proposed
59 methods that estimate the effect that conservation practices have on the reduction of nutrient loss
60 from agricultural fields.

61 **Methods**

62 **Updating MANAGE**

63 To understand the data availability and limitations of the October 2014 edition of
64 MANAGE, we performed summary statistics using the statistical program R (R Core Team
65 2014). Summary statistics conducted included sum, mean, median, and aggregation of various
66 attributes by other attributes. These summary statistics were chosen to compare the October 2014
67 edition to previous editions of MANAGE and to convey what is available to users. The statistical

68 package reshape2 (Wickham 2007) in R was employed to assist with these calculations. In
69 addition, arcGIS 10.2 (ESRI 2013) was used to map various attributes by state and create color
70 gradient maps such as Figure 1. This map was created to characterize the geographic distribution
71 of the entries in the October 2014 edition of MANAGE for users. Overall, the methods employed
72 to summarize the efforts for updating the current edition of MANAGE, the first objective of this
73 paper, include summary statistics using R and mapping using arcGIS 10.2.

74 **Meta-Analysis Methods**

75 We used the 2007 edition of the MANAGE database to test the proposed statistical methods
76 – the propensity score matching analysis and the multilevel modeling approach. Both statistical
77 methods are common for observational data in the social science field and are used to take
78 confounding factors into consideration. Different field characteristics in MANAGE can act as
79 confounding factors because they play a large role in determining whether a treatment is applied
80 to a field and can influence the dependent variable being measured (Gelman and Hill 2007).
81 We used these two methods to estimate the effects of conservation practices in reducing P loss
82 from fields using the 2007 edition of MANAGE. In addition propensity score methods have
83 been applied to the October 2014 update to MANAGE to estimate the effect of conservation
84 practices in reducing total phosphorus (TP).

85 The first statistical method, propensity scores, will be used to find the average causal
86 effect of a treatment (conservation practice) by comparing the treatment to a control (no
87 conservation practices) (Rosenbaum and Rubin 1983, Gelman and Hill 2007). This method uses
88 a logistic regression to create one number that will show the probability that the field in question
89 will receive a treatment when all confounding variables are considered (Gelman and Hill 2007).
90 Each confounding variable contributes to the likelihood of the field receiving a treatment and the

91 propensity score quantifies each confounding variable's likelihood into one number (Gelman and
92 Hill 2007). A field with the treatment will then be matched to a control field with the closest
93 propensity score. These propensity scores will allow for controls (fields without a conservation
94 practice) and treatments (fields with a conservation practice) to be compared while the averages
95 of all other confounding variables are similar, analogous to what is done in lab experiments
96 (Rosenbaum and Rubin 1983, Gelman and Hill 2007). Using the propensity score method is ideal
97 because observational data differs from randomized experiments in that confounding variables
98 cannot be normalized in observational data, while effects of confounding variables can be
99 removed in randomized experiments (Gelman and Hill 2007).

100 The resulting scores will then be used to match a treatment field with a control field and
101 create the two subsets that are, on average, more similar than the unmatched data. After sub-
102 setting with propensity score, we will run a two sample *t*-test on the control and treatment groups
103 to determine if the nutrient loss (average TP) is significantly different from each other, thus
104 showing us if conservation practices have an effect on nutrient loss. Specifically we will be
105 looking to see if the mean nutrient loss in the treatment group, fields with a conservation
106 practice, is less than that of the control group. Next we will conduct regressions to look at the
107 differences in slope of applied phosphorus (in fertilizer) between the treatment and control group
108 to understand the effect of conservation practices on nutrient loss from a field. This regression
109 will be used to determine the percent increase in TP leaving a field for every 1% increase in
110 phosphorus from fertilizer application. Overall, the *t*-test and regression on our matched data will
111 show the effect of conservation practices on nutrient loss.

112 The second proposed method, multilevel modeling, stratifies the data into groups with similar
113 attributes and a regression model is then run on each group (Gelman and Hill 2007). This method

114 is increasingly used for ecological data with interactions at difference scales because outcomes at
115 one scale can be affected by events at a different scale (Qian et al. 2010). An example of this in
116 MANAGE is how the effect of a conservation practice may vary by land use. The multilevel
117 models will be run with fertilizer application and conservation practices as inputs when looking
118 at the outcomes of total phosphorus. By running multilevel models I aim to quantify the effects
119 that the inputs have on the outcome variables while considering confounding factors.

120 These proposed statistical analyses (multilevel modeling and propensity scores) will address
121 the confounding factors noted in past studies, including crop type, study regions, soil
122 characteristics, and runoff. Additionally, these analyses will determine the causal effect of the
123 inputs (conservation practices and fertilizer application methods) on the outcome variables
124 (nutrient loss). Analyzing the data with two different methods will increase the strength of the
125 results if both tests determine the same overall average.

126 **Results**

127 **Updating MANAGE**

128 The October 2014 edition of MANAGE has the most available data for analysis out of
129 any edition thus far. This edition is comprised of the 55 publications in the 2007 database as well
130 as 10 additional publications, all of which equate to 330 entries and 1,980 watershed years. The
131 watershed year variable is the total number of years monitored for each individual entry – for
132 example a field monitored for 9 years would have 9 watershed years. Conversely, an entry in the
133 MANAGE database represents an observation from a single field. Thus, one entry is a single
134 field while the watershed years for that field is the number of years the field was observed for. A
135 publication in MANAGE may contribute more than one entry because the study may have

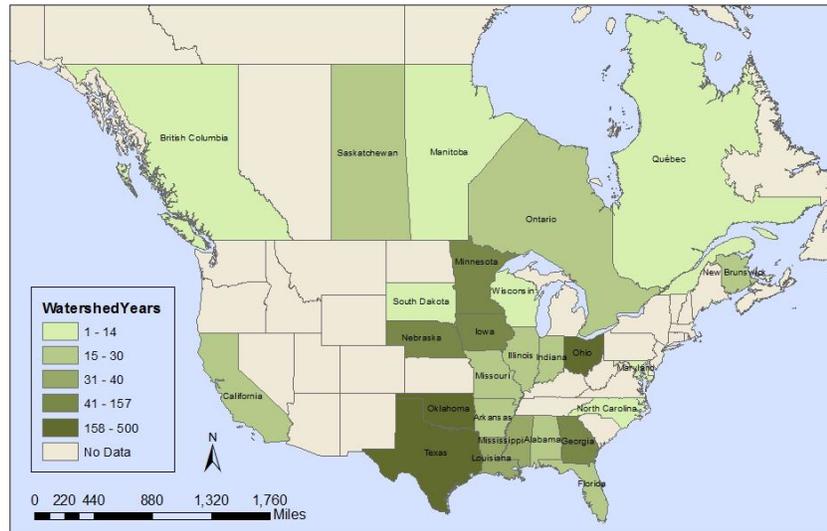
136 monitored more than one field. With these additional papers, the October 2014 edition of
137 MANAGE has more data available to the public than previous editions.

138 To understand the nutrient information in this edition of MANAGE and the average load
139 leaving the field, the area weighted average of phosphorus and nitrogen loading was calculated.
140 The area-weighted average takes account of the size of the individual fields in the database. The
141 area-weighted average of total nitrogen loading is 12.8 kg ha⁻¹. For the total phosphorus load, the
142 area-weighted average is 2.1 kg ha⁻¹. Both calculations were taken from the entries with either
143 phosphorus or nitrogen and watershed size data available. The area-weighted average was then
144 calculated using the 2007 edition of MANAGE to allow comparisons between both editions of
145 MANAGE. For total nitrogen the area-weighted average loading was 14.2 kg ha⁻¹ and for total
146 phosphorus 2.2 kg ha⁻¹ was the area-weighted average loading (Harmel et al. 2006). Both values
147 are similar enough to suggest that calculations were performed correctly and that the differences
148 in the values can be attributed to the addition of 10 studies to the database.

149 The geographic balance of all editions of MANAGE, including the October 2014 edition,
150 is not uniform and thus does not include all regions of the United States. The data is concentrated
151 in the central region of the United States, where a large portion of the agriculture industry is
152 located. In the original edition of MANAGE, entries mainly come from Oklahoma and Texas
153 and equate to 42 percent of the watershed years (Harmel et al. 2006). The 2008 publication
154 regarding MANAGE featured data that was largely from Oklahoma, accounting for 30 percent of
155 the watershed years (Harmel et al. 2008). Following Oklahoma, Texas and Ohio featured the
156 greatest number of watershed years with 16 percent and 15 percent of the total watershed years,
157 respectively (Harmel et al. 2008). This trend is also consistent with the October 2014 update of
158 MANAGE. In this edition, Oklahoma again has the most watershed years with 25 percent of

159 watershed years in the database. Texas and Ohio also follow trends of past editions by occupying
 160 16 percent and 14 percent of the total watershed years. The spatial aggregation of the 2014
 161 edition of MANAGE can be seen in Figure 1.

Watershed Years per State



162 **Figure 1.** The number of watershed years per state in the October, 2014 edition of
 163 MANAGE.

164 To understand the availability of data in the 2014 edition of MANAGE, we have
 165 calculated the number of missing entries and watershed years for several variables of interest
 166 (Table 1). These calculations show variables that have a suitable sample size for analysis and
 167 others that may act to limit analyses. Two variables with limited sample size that can raise
 168 concerns when using them in analyses include conservation practices and crop yields.

Variable	Number of Missing Entries	Percent of Missing Entries	Number of Missing Watershed Years	Percent of Missing Watershed Years
Fertilizer Application Method 1	127	38.5%	803	40.6%
Fertilizer Application Timing 1	136	41.2%	868	43.8%
Conservation Practice 1	268	81.2%	1640	82.8%
Crop Yield	277	83.9%	1673	84.5%
Land Use	0	0.0%	0	0.0%

Tillage	3	0.9%	9	0.5%
Average Nitrogen Applied	114	34.5%	842	42.5%
Average Phosphorus Applied	59	17.9%	466	23.5%

169
170 **Table 1.** Number and percentage of missing entries and the associated watershed years for
171 eight variables of interest.

172 The number of entries containing crop yield data in the MANAGE database are limited
173 and can limit the sample size when using crop yields in analyses. This information is important
174 when considering agricultural economics and nutrient uptake by plants (Vagstad et al. 1997). To
175 better understand the crop yield information that is available the average crop yield for the
176 primary crop on each field was calculated (Table 2). Of the entries with crop yield data, potato
177 has the highest average crop yield. The yield data are difficult to compare among different
178 crops. Furthermore, when comparing individual crops, sample size for each comparison will be
179 too small for a meaningful statistical analysis.

Crop Type	Average Crop Yield (Mg/ha)
Alfalfa	12.4
Coastal bermudagrass	11.4
Corn	7.151538
Cotton	2.045
Pasture	5
Potato	27.485714
Soybeans	3.392
Wheat	2.64

180
181 **Table 2.** Average crop yield by crop type in the October 2014 edition of MANAGE.

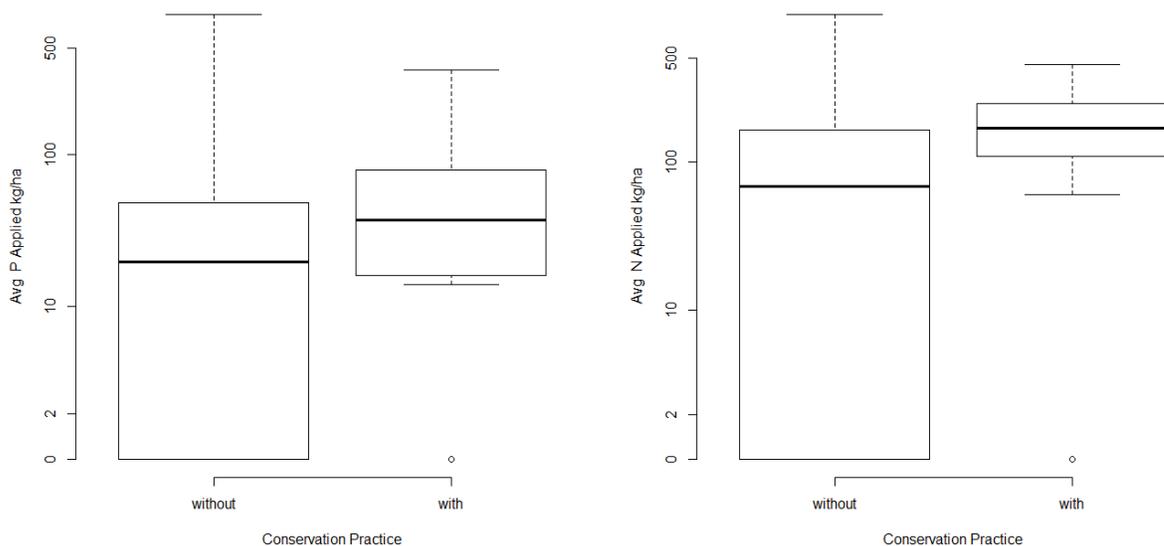
182 Considering the limited data available for crop yields, we propose to extend our study by
183 exploring new modeling approaches and transforming the yield data into % of expected yield so
184 that a meaningful comparison can be carried out. This part of the research will be conducted in
185 the Fall of 2015 with additional funding from NOAA.

186 To understand the methods and timing of fertilizer applications included in the database,
187 the most common method and timing were identified. Fertilizer application type and timing can
188 impact nitrogen and phosphorus losses from agricultural fields. Precipitation and irrigation
189 events carry nutrients from the field when they drain and make it important to consider the
190 method and the timing of application. Surface applied fertilizers were the most common type of
191 fertilizer application method included in the database, accounting for 30.5 percent of the data.
192 With regard to fertilizer timing, the most common time of fertilizer application was “grass in
193 growing season.” Both the most common fertilizer application method and timing came from
194 those entries in MANAGE that had data available. The percent of missing data for fertilizer
195 application method was 40.6 percent and 43.8 percent for fertilizer application method (Smith et
196 al. 2007). While the amount of missing data for fertilizer information was about 40 percent, this
197 is considerably less than that of crop yields.

198 Data regarding conservation practices are also limited and can thus limit sample size
199 when performing analyses that include conservation practice. Conservation practices can impact
200 the nutrient loads that leave a field in runoff. The primary conservation practices considered in
201 the MANAGE database include contour farming, filter strip, terrace, and grassed waterways.
202 These conservation practices account for 17 percent of watershed years or 19 percent of entries
203 in MANAGE. The other 83 percent of watershed years or 81 percent of entries do not have a
204 conservation practice listed. This means one of two things – 1) the entries do not have a
205 conservation practice applied to the field or 2) the publication containing the data does not list if
206 there is or is not a conservation practice implemented. These two meanings of the missing data
207 can have very different impacts when analyzing the database. The second meaning (the
208 conservation practice data was not included in the publication) should be considered missing

209 data while the first meaning (no conservation practice applied) is a separate category that should
 210 be included in analyses (Duriancik et al. 2008).

211 Current and past editions of the MANAGE database have provided field scale
 212 information on agricultural fields to public, but a number of the variables within the database are
 213 correlated with each other and can influence analyses (Harmel et al. 2006, Harmel et al. 2008).
 214 One particular of variable that exemplifies the trend is the amount of phosphorus or nitrogen
 215 applied to a field as it relates to whether the field has a conservation practice. Fields tend to have
 216 a conservation practice applied when there are high amounts of fertilizers applied, shown in
 217 Figure 2, thus these two variables are not independent of one another. For both phosphorus and
 218 nitrogen, the mean amount of fertilizers applied in Figure 2 is greater for fields with conservation
 219 practices than with fields without conservation practices. This relationship means that subsequent
 220 analyses will be impacted if this correlation is not taken into consideration.



221
 222 **Figure 2.** Side by Side box plots of fertilizer application (Phosphorus Applied and
 223 Nitrogen Applied) for fields without a conservation practice and fields with a

224 conservation practice. Both the left plot (Phosphorus Applied) and right plot (Nitrogen
225 Applied) show a high mean fertilizer application in fields with conservation practices.
226 The current edition has additional attributes, including crop yield, that have a low
227 percentage of available data and can thus limit the sample size available for analyses.
228 Additionally, current and past editions have variables that need to be considered carefully
229 because they can act as confounding factors. Although this is the case, the most recent update
230 also has the highest number of publications and watershed years included in any edition of
231 MANAGE thus far.

232 **Methods Development**

233 One objective of our project is to develop statistical meta-analysis methods for estimating
234 the effect of agricultural conservation practices on reducing nutrient loss. The difficulty of the
235 task lies in the nature of data available for such analysis. Almost all available data are
236 observational data, which can be confounded by differing crop types and differing management
237 practices. As we may not have the full knowledge of these confounding factors, conventional
238 statistical meta-analysis methods are often ineffective. In this proposal, we discuss the use of two
239 statistical causal analysis methods (propensity score and multilevel modeling) for quantifying the
240 effects of water and soil conservation practices in reducing phosphorus loss from agricultural
241 fields. With the propensity score method, a subset of the data was used to form a treatment group
242 and a control group with similar distributions of confounding factors. With the multilevel
243 modeling approach, the data were stratified based on important confounding factors and the
244 conservation practice effect was evaluated for each stratum.

245 We applied both methods to the 2007 version of MANAGE database and estimated the
246 conservation practice effect in reducing TP loss. Prior to applying our methods, we evaluated the

247 effects of four different conservation practices (grassed waterways, contour farming, terraces,
248 and riparian forest buffers/filter strips) (Qian and Harmel, 2015). This analysis suggested that the
249 effect of individual conservation practices was similar to the effect of all conservation practices
250 grouped together (Qian and Harmel, 2015). When applying our methods, the 18 percent of data
251 with conservation practices listed were considered our treatment group – based on our previous
252 analysis – and the 82 percent without a practice listed were used as our control group of “no
253 conservation practice”. Our preliminary results are summarized in a manuscript submitted to
254 *Journal of American Water Resources Association* (Qian and Harmel, 2015). The manuscript
255 reported an average reduction of P loss of approximately 70 percent. In addition, both methods
256 show evidence of conservation practices reducing the incremental increase in TP export per unit
257 increase in fertilizer application.

258 The propensity score method has also been applied to the October 2014 edition of
259 MANAGE. The linear model created to calculate propensity score includes average phosphorus
260 applied, average nitrogen applied, average runoff, average soil loss, dominant soil type,
261 hydrologic soil group, land use, tillage, fertilizer application methods and an interaction term for
262 fertilizer application methods and land use as covariant for the calculation of each score.
263 Similarly as in Qian and Harmel (2015) values with a total phosphorus load of 0 were replaced
264 with 0.002, half of the lowest reported value, to allow these entries to be kept for analyses. In the
265 development of propensity score, conservation practices were combined to increase the sample
266 size of the treatment group and the control group included all entries without a conservation
267 practice listed. This allows for the two groups to have similar variance and maximizes the sample
268 size after matching. The model produced a subset with 58 observations – 29 for the control and
269 29 for the treatment. The sub-set data shows that the two groups are more similar on other

270 covariates than they were prior to matching. Figure 3, below, shows two overlaid histograms of
 271 the distribution of runoff, a confounding variable. The figure shows that the treatment and
 272 control groups are less skewed from each other with the matched data than with the original data.
 273 The increased similarity of the two groups after matching achieves what we aimed to do with the
 274 propensity score because the treatment and control groups are closer to having the same
 275 distribution like we would find in a laboratory experiment.

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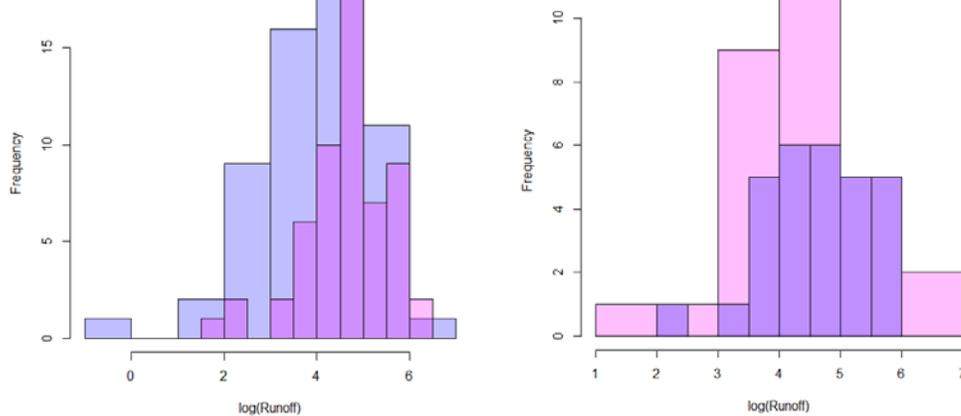
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284 **Figure 3.** Overlaid histogram of runoff before (left) and after (right) matching. Pink
 285 represents no conservation practice and blue represents with conservation practice.

286 After matching the data using propensity score as described above, we then conducted a
 287 two sample t-test to compare the log-mean nutrient losses of the treatment group versus the
 288 control group. The results of this t-test showed a significant decrease ($p < 0.01$) between the mean
 289 of the control group, no conservation practice, and treatment group, having a conservation
 290 practice. Taking this understanding a step further, we used the difference in the log means to
 291 calculate the percent reduction in total phosphorus loads. The estimated effect was calculated as
 292 $\beta = -0.7861$ ($p < 0.01$). To calculate the multiplicative effect we used $e^{-0.7861} = 0.4556$. This
 293 equates to a 54 percent reduction in the total phosphorus load leaving the field. These results

294 show that conservation practices did reduce the amount of total phosphorus leaving a field when
295 accounting for the confounding variables. Our results also support the findings of Qian and
296 Harmel (2015) – that conservation practices made a significant reduction in the amount of total
297 phosphorus leaving a field (Qian and Harmel 2015).

298 **Discussion**

299 The various editions of the MANAGE database provide accessible data to the public that
300 can be used to support nonpoint source models and aid decision makers (Harmel et al. 2006,
301 Harmel et al. 2008). The most recent edition – edited in October 2014 – provides data on
302 fertilizer application, conservation practices, rainfall, runoff, land use, and nutrient loads leaving
303 a field. This edition of MANAGE includes 10 more papers than the previous edition equating to
304 a total of 65 papers. The data from these 65 papers provide 330 entries or 1980 watershed years
305 as data for analysis by users.

306 Past analyses of MANAGE have yielded interesting results that can be partially attributed
307 to confounding factors that were not accounted for (Harmel et al. 2006, Harmel et al. 2008).
308 These confounding factors, which are factors that are not the independent variable but influence
309 the outcome of the dependent variable, include field characteristics such as land use, soil
310 characteristics, precipitation, and seasonal variations. In addition to confounding factors not
311 considered previously, there are a number of attributes in the MANAGE database that have
312 limited data available. As noted by previous publications, there is a definite gap in the amount of
313 concentration data versus the quantity of nutrient load data (Harmel et al. 2008). Additionally,
314 there is a limited number of entries that have crop yield information. Only 16 percent of the data
315 in the October 2014 edition of MANAGE possess crop yield quantities. This imbalance of data
316 availability can cause difficulties when using MANAGE to find trends involving these limited

317 variables, especially when taking the confounding factors into consideration. To address this
318 limited data availability, we are reviewing the current entries for crop yield information that may
319 not have been included, as well as ensure that additional entries contain this information.

320 The propensity score results from both Qian and Harmel (2015) and current analysis on
321 the October 2014 database have shown significant reductions in TP loads leaving a field when
322 conservation practices were implemented. Qian and Harmel (2015) showed a 70 percent
323 reduction in the amount of TP leaving a field using the 2007 version of MANAGE while current
324 analysis of the October 2014 edition showed a 54 percent reduction in TP leaving the field.
325 These reductions both signify that the application conservation practices to a field reduce the
326 amount of nutrient loss leaving a field.

327 The next steps of this project include further statistical analyses to find the causal effect of
328 conservation practices and fertilizer application on nutrient loss. These analyses will be
329 conducted by using multilevel modeling in addition to the current application of propensity
330 score. Intermediate goals proposed have been met to achieve the overall objective to quantify the
331 effects of conservation practices and fertilizer application practices on nutrient loss. These
332 intermediate goals include incorporating changes to MANAGE and adding grey literature and
333 government research to the next edition of MANAGE. Adding grey literature will help to
334 address limited sample sizes and any publication bias that exists. Since the 2007 edition 10
335 additional studies have been added as well as variables such as fertilizer formula and fertilizer
336 application timing. The summary statistics performed in 2014 highlighted additional tasks that
337 need to be addressed so the most accurate analysis of the effect of the two agricultural practices
338 on nutrient losses and crop yield can be quantified.

339 **Interpretive Summary**

340 This project has made notable advancements towards the objective of quantifying the effect
341 of conservation and fertilizer application practices on nutrient loss and crop yield, listed in the
342 project proposal. Initial summary statistics were used to find attributes with limited sample size
343 and overview the data in the October, 2014 edition of the MANAGE database. Crop yield and
344 conservation practice – important variables for achieving the project outcomes – were both found
345 to be variables with limited sample size. Additionally, initial statistics showed that the entries are
346 spatially aggregated toward the central region of the United States. Ohio, Oklahoma, and Texas
347 account for 1,092 of the 1,980 watershed years in the database. To address attributes with limited
348 sample size, plans have been made to review the 65 papers included in the database. In addition,
349 we plan to add grey literature to MANAGE to address any publication bias. The project has
350 achieved the intermediate objective to update the MANAGE database and is in the process of
351 achieving the objective of finding the effects of the two noted agricultural practices on nutrient
352 loss. The first proposed statistical method of propensity score has been implemented on both the
353 2007 edition and the October 2014 edition of MANAGE and has yielded significant reductions
354 in the amount of TP leaving a field when conservation practices were implemented. These results
355 will then be compared to multilevel modeling results to ensure the validity of both statistical
356 methods.

357 This project supports one graduate student (Ms. Stephanie Nummer) in the Department of
358 Environmental Sciences, The University of Toledo. Part of the research is used in developing
359 Ms. Nummer's master's thesis. Initial results summarized in the paper by Qian and Harmel
360 (2015) were presented at the Annual Conference of the American Water Resources Association

361 in Washington D.C. (November 2014). In addition, results from this research allowed us to apply
362 for additional support from the Federal government to expand the scope of our initial proposal.

363

364

365 **References**

- 366 Christianson, L., and R. Harmel. 2015. The MANAGE Drain Load database: Review and
367 compilation of more than fifty years of North American drainage nutrient studies.
368 *Agricultural Water Management* **159**:277-289.
- 369 Crain, A. S., K. D. o. Agriculture, and G. Survey. 2006. Concentrations, and Estimated Loads
370 and Yields of Nutrients and Suspended Sediment in the Little River Basin, Kentucky,
371 2003-04. U.S. Geological Survey.
- 372 Duriancik, L. F., D. Bucks, J. P. Dobrowolski, T. Drewes, S. D. Eckles, L. Jolley, R. L. Kellogg,
373 D. Lund, J. R. Makuch, M. P. O'Neill, C. A. Rewa, M. R. Walbridge, R. Parry, and M. A.
374 Weltz. 2008. The first five years of the Conservation Effects Assessment Project. *Journal*
375 *of Soil and Water Conservation* **63**:185A-197A.
- 376 Easterbrook, P. J., R. Gopalan, J. Berlin, and D. R. Matthews. 1991. Publication bias in clinical
377 research. *The Lancet* **337**:867-872.
- 378 ESRI 2013. ArcGIS Desktop: Release 10.2. Redlands, CA: Environmental Systems Research
379 Institute.
- 380 Hadley Wickham (2007). Reshaping Data with the reshape Package. *Journal of Statistical*
381 *Software*, 21(12), 1-20. URL <http://www.jstatsoft.org/v21/i12/>.
- 382 Harmel, D., S. Potter, P. Casebolt, K. Reckhow, C. Green, and R. Haney. 2006. Compilation of
383 Measured Nutrient Load Data for Agricultural Land Uses in the United States JAWRA
384 *Journal of the American Water Resources Association* **42**:1163-1178.
- 385 Harmel, D., S. Qian, K. Reckhow, and P. Casebolt. 2008. The MANAGE Database: Nutrient
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390 **37**:2403-2406.
- 391 Qian, S. S., T. F. Cuffney, I. Alameddine, G. McMahon, and K. H. Reckhow. 2010. On the
392 application of multilevel modeling in environmental and ecological studies. *Ecology*
393 **91**:355-361.
- 394 Qian, S.S. and Harmel, D. 2015. Applying statistical causal analysis to agricultural conservation:
395 a case study examining P loss impacts. Revision submitted to *Journal of American Water*
396 *Resources Association*.
- 397 R Core Team (2014). *R: A language and environment for statistical computing*. R Foundation for
398 *Statistical Computing*, Vienna, Austria. URL <http://www.R-project.org/>.
- 399 Robertson, D. M., and D. A. Saad. 2011. Nutrient Inputs to the Laurentian Great Lakes by
400 *Source and Watershed Estimated Using SPARROW Watershed Models*. *JAWRA*
401 *Journal of the American Water Resources Association* **47**:1011-1033.
- 402 Sharpley, A. 1995. Identifying Sites Vulnerable to Phosphorus Loss in Agricultural Runoff. *J.*
403 *Environ. Qual.* **24**:947-951.
- 404 Smith, D. R., P. R. Owens, A. B. Leytem, and E. A. Warnemuende. 2007. Nutrient losses from
405 *manure and fertilizer applications as impacted by time to first runoff event.*
406 *Environmental Pollution* **147**:131-137.
- 407 Vagstad, N., H. O. Eggestad, and T. R. Høyås. 1997. Mineral Nitrogen in Agricultural Soils and
408 *Nitrogen Losses: Relation to Soil Properties, Weather Conditions, and Farm Practices.*
409 *Ambio* **26**:266-272.