

Evaluating the impact of legacy P and agricultural conservation practices on nutrient loads from the Maumee River Watershed

*Rebecca Logsdon Muenich**, *Margaret Kalcic*, and *Donald Scavia*

Graham Sustainability Institute, University of Michigan, 625 E. Liberty St., Suite #300, Ann Arbor, MI 48104

*Corresponding author: 625 E. Liberty St., Suite #300, Ann Arbor, MI 48104; 734-647-2495 (ph); rlogsdon@umich.edu

KEYWORDS. Maumee River Watershed (MRW), Western Lake Erie, Harmful Algal Blooms (HABs), Soil and Water Assessment Tool (SWAT), Agricultural Management

ABSTRACT. The recent resurgence of hypoxia and harmful algal blooms in Lake Erie, driven substantially by phosphorus loads from agriculture, have led the U.S. and Canada to begin developing plans to meet new phosphorus load targets. To provide insight into which agricultural management options could help reach these targets, we tested alternative agricultural land use and land management scenarios on phosphorus loads to Lake Erie. These scenarios highlight certain constraints on phosphorus load reductions from changes in the Maumee River Watershed (MRW), which contributes roughly half of the phosphorus load to the lake's western basin. We evaluate the effects on phosphorus loads under nutrient management strategies, reduction of fertilizer applications, employing vegetative buffers, and implementing widespread cover crops and alternative cropping changes. Results indicate that even if fertilizer application ceased, it

21 may take years to see desired decreases in phosphorus loads, especially if we experience greater
22 spring precipitation or snowmelt. Scenarios also indicate that widespread conversions to
23 perennial crops that may be used for biofuel production are capable of substantially reducing
24 phosphorus loads. This work demonstrates that a combination of legacy phosphorus, land
25 management, land use, and climate should all be considered when seeking phosphorus loading
26 solutions.

27 INTRODUCTION

28 Recent resurgence of harmful algal blooms (HABs) in Lake Erie's western basin and depleted
29 oxygen (hypoxia) in its central basin¹⁻² threaten human and animal health as well as ecosystem
30 integrity. While all loads to the western and central basins contribute to the evolution of hypoxia,
31 the main HAB driver has been shown to be elevated phosphorus (P) loads coming from the
32 watersheds that drain into the western basin³, particularly from the Maumee River Watershed
33 (MRW).⁴⁻⁵ The 2012 Great Lakes Water Quality Agreement (GLWQA)⁶ calls for revision of the
34 1978 Lake Erie P loading targets,⁷ and the US and Canada have recently approved new March-
35 to-July P loading targets⁸ for the Maumee River Watershed of 186 metric tonnes (~0.109 kg/ha)
36 of dissolved reactive phosphorus (DRP) and 860 metric tonnes (~0.506 kg/ha) of total
37 phosphorus (TP).

38 Significant lake modeling effort was used to inform the target P loads for Lake Erie,^{2,4,5,9-10} but
39 less research has been done to identify policy and management actions needed to achieve them.
40 Bosch et al.¹² suggested that extensive implementation of multiple conservation practices would
41 be needed to address nutrient and sediment load problems in Lake Erie and Daloglu et al.¹³
42 demonstrated the negative interactive effects of changes in conservation practices and recent

43 changes in precipitation patterns on DRP loads. Knowledge and models of P transport within
44 watersheds has continued to improve since these studies were published,¹⁴⁻¹⁵ and there is a need
45 to revisit and expand the potential strategies, especially in light of the new target loads.

46 The goal of the present work is to explore the potential outcomes of land management and land
47 use changes in the MRW to determine the bounds of what might be expected from extensive
48 changes in agricultural conservation practices. Our main objectives were to evaluate the impact
49 of legacy P in the watershed and simulate the effects of hypothetical, rather extensive
50 management strategies, including reducing farm fertilizer applications, employing vegetative
51 buffers on all farm fields, and implementing winter cover crops and alternative cropping systems
52 across all farm fields.

53 MATERIALS AND METHODS

54 **Study Area & Model Development** - We focused on the MRW (Figure 1) because it
55 contributes the largest P load of any Lake Erie tributary and has been shown to be a primary
56 driver of Lake Erie's HABs⁴⁻⁵. The watershed occupies over 17,000 km² and extends from
57 northwest Ohio into Indiana and Michigan. Its land use is about 70% row crop agriculture,
58 dominated by rotations of soybeans, corn, and winter wheat. The land is characterized by low
59 slopes and heavy, clayey soils with poor natural drainage, with the majority of cropland
60 artificially drained by subsurface tile drains. This watershed is also a perfect test bed for these
61 analyses because the river P loads, measured near the outlet of the Maumee River, have been
62 monitored daily for over 40 years by Heidelberg University's National Center for Water Quality
63 Research (NCWQR),¹⁶ thus providing stringent constraints on model performance.

64 [FIGURE 1]

65 We used the Soil and Water Assessment Tool (SWAT), a semi-distributed, hydrologic model
66 that takes into account land use, land management, soils, slopes, and climate information to
67 simulate watershed processes such as crop growth, streamflow, and nutrient and sediment
68 dynamics across the landscape and in streams and rivers.¹⁷ SWAT has been used for a variety of
69 climates and land uses, and has been shown to work well in agricultural landscapes.¹⁸⁻¹⁹ Our
70 model was set up, calibrated, and validated for the MRW as described in Kalcic et al.²⁰ for
71 streamflow and loadings of TP, DRP, sediment, and total nitrogen. The model's baseline
72 cropping system consisted primarily of corn and soybean rotations, with 45% of cropland having
73 some winter wheat in rotation. Fertilizer application rates were estimated based on fertilizer
74 sales, manure applications were estimated based on numbers of animals, and tillage was
75 estimated based on the Conservation Technology Innovation Center database. Tile flow
76 accounted for 38-42% of the streamflow, and 42-48% of the DRP (8-10% of TP) loads to the
77 river. The model performance for common evaluation criterion was deemed to be very good.^{20,21}
78 For more details of the SWAT model set-up, calibration, and validation see supporting
79 information (SI) and Kalcic et al.²⁰

80 In the next sections, we first evaluated the impact of legacy P (see “Legacy Phosphorus
81 Impacts”) and then significant implementation of agricultural conservation practices (see
82 “Alternative Phosphorus Reduction Strategies”). All simulations are theoretical in nature, but
83 provide insight on reaching target loads.

84 **Legacy Phosphorus Impacts** - Before running agricultural conservation scenarios, we evaluated
85 the watershed's response to eliminating fertilizer applications to assess the impact of legacy P.
86 The baseline model was run for twelve years under observed climate and current management

87 regimes as a warm up period, followed by eighty years of the same weather with one of four
88 management scenarios:

89 (1) Business as usual (BAU): same fertilization and management used in baseline model

90 (2) No fertilizers (No Fert): ceasing all inorganic and organic nitrogen and P fertilizer
91 applications

92 (3) No P fertilizers (No P Fert): ceasing all inorganic and organic P fertilizer applications

93 (4) No inorganic P fertilizers (No Inorganic P Fert): ceasing inorganic P fertilizer
94 applications but continuing manure applications

95 For each case, we repeated a single year's temperature and precipitation patterns over the 80 year
96 simulation to isolate the impact of legacy P. The other three weather variables used by SWAT
97 (relative humidity, solar radiation, and wind speed) were generated within SWAT by the
98 WXGEN weather generator which, combined with agricultural rotations, causes some year to
99 year variability in outputs. Given that precipitation is a main driver of P losses via surface and
100 subsurface pathways, we also wanted to consider the potential impacts of weather, so we selected
101 average, wet, and dry years from the 1981-2010 record as follows. Because the target P loadings
102 are based on spring (March-to-July) loads, we extracted precipitation, simulated streamflow, and
103 P loads for each year (SI Figure S1) from the 1981-2010 simulation.²⁰ Two representative 'wet'
104 years were selected for analyses: 1981, which had high spring precipitation, and 1982, which had
105 high spring streamflow (driven, in this case, by snowmelt), because high rainfall and high
106 streamflow can have different impacts on P loading. For example, the high rainfall year (1981)
107 has greater simulated DRP load in the spring whereas the high streamflow spring (1982) had

108 higher simulated TP loading. The year with the lowest overall values for all four variables (1988)
109 was selected as the representative ‘dry’ year, and a year with values close to the average for all
110 four variables (2004) was selected as the representative ‘average’ year.

111 **Alternative Phosphorus Reduction Strategies** - The baseline SWAT model was also used to
112 test the impact of agricultural management and land use changes on P loads and crop yields
113 under the observed 1981-2010 weather conditions. Table 1 briefly describes the scenarios run
114 under these conditions; further details are provided in the SI. Two types of in-field strategies
115 were tested: the reduction of P fertilizer application rates and addition of cereal rye winter cover
116 crops. The rate of application was tested because this practice could be a cost-effective strategy
117 to reduce P loads. While we tested a range of reductions, we want to emphasize that this study is
118 watershed-scale; therefore, while the reductions were applied across all fields, the interpretation
119 of the results would be that some farmers who are over-applying P would have larger reductions
120 in rates, while others already applying correct amounts of P would have no change in rates. The
121 addition of winter cover crops (cereal rye that is killed, not harvested, before cash crop plantings)
122 was tested because they are commonly used in the Midwest and have been shown to provide
123 multiple water quality benefits by providing ground cover for soil typically exposed during the
124 winter months.²³⁻²⁴ Recent surveys also suggest that there is currently little adoption of cover
125 crops in the watershed, so there is great potential impact for addition of this practice.²²

126 Vegetated filter strips were also implemented to test the impact of an edge-of-field practice on P
127 loads. Vegetated filter strips can be planted alongside open waterways to intercept sediment and
128 nutrients from adjacent farmland²⁵⁻²⁶ and are a commonly recommended practice. In this
129 watershed, around 31% of acres have an edge-of-field trapping practice;²² therefore, there is still
130 potential impact with further implementation. Filter strips were simulated with varying

131 effectiveness using the newest filter strip routine in SWAT that is based on empirical
132 relationships derived from simulations of the Vegetative Filter Strip MODel.²⁷ To vary the
133 effectiveness of filter strips, the two parameters controlling filter strip effectiveness
134 (FILTER_CH: the fraction of flow through the most heavily loaded 10% of filter strip which is
135 fully channelized and FILTER_CON: the fraction of field drained by most heavily loaded 10%
136 of filter strip) were changed along with the ratio of field area to filter strip area; see SI for more
137 details.

138 Two agricultural land use changes were also evaluated: alternative row crops (sunflower and
139 lentils) and cellulosic biofuel crops (switchgrass and miscanthus). Sunflowers and lentils are
140 examples of row crops with potential to replace corn and soybeans in terms of ease of adopting
141 existing equipment and providing similar food and feed products.²⁸⁻²⁹ Although it is unlikely that
142 they may be grown continuously or in rotation due to problems such as pests, these crops were simulated
143 solely to test the impact of row crop variety on P loads. Switchgrass and miscanthus are two cellulosic
144 biofuel crops capable of addressing the United States Energy Security and Independence Act of
145 2007 fuel mandates while potentially improving water quality.³⁰⁻³² These land use changes are
146 examples of alternative farming pathways that allow farmers to produce crops; however, both
147 options require policy, economic, and technological advances to be effective and feasible to
148 producers.

149 [TABLE 1]

150 **RESULTS AND DISCUSSION**

151 **Legacy P Impacts** - When fertilizer applications were stopped under average weather
152 conditions, spring DRP loads decreased to at or below the targets within 3-5 years, and continued

153 to decline over time in contrast with the BAU scenario which remained steadily above the target
154 (Figure 2; gray line). This highlights the potential importance of phosphorus applied in a given
155 year. TP loads, however, took longer to reach the targets depending on weather, perhaps due to
156 lower mobility of the particulate P fraction. When only inorganic P fertilizer was stopped and
157 manure applications continued, DRP and TP loads hovered around the targets longer than in the
158 no P fertilizer scenarios (Figures 2 & 3; gray lines). Such time lags in the watershed's response
159 to cessation of fertilizer applications demonstrate how much legacy P may currently be in the
160 MRW. This is discussed further below.

161 [FIGURE 2]

162 [FIGURE 3]

163 These results, similar to Daloglu et al.,¹³ also show that weather variability can influence the
164 timescale of the watershed's response. Under BAU management, the DRP targets are actually
165 met in the driest years, but not in average or wet years (Figure 2) possibly due to the high
166 mobility of dissolved phosphorus. However, that target was eventually met in every fertilizer
167 reduction scenario under all weather conditions, *albeit* after 30-40 years in some cases. In years
168 of high spring streamflow (and thus high runoff) and high spring precipitation it is unlikely that
169 TP targets will be met, even under this extreme case of no fertilizer application (Figure 3). Under
170 dry and average conditions, however, targets were met fairly quickly in the no fertilizer and no P
171 fertilizer scenarios. However, when manure application continued, it added approximately ten
172 years to the length of time to reach the target compared with the no P fertilizer scenario. While
173 our approach begins to evaluate the impact of legacy P by putting "bounds" on the time it could
174 take to reach target loads after cessation of P fertilizers, given varying weather conditions, in

175 reality the time to reach targets would likely occur sometime between the results for dry and wet
176 year simulations. Additionally, given that it is unreasonable to assume that all P fertilizer
177 applications would cease, the legacy P in the soils and streams in the MRW along with continued
178 applications could prolong reaching the targets unless other agricultural conservation practices
179 are employed. This simulation also suggests that the legacy P in the watershed will need to be
180 addressed along with other conservation practice implementation.

181 We also evaluated the response time of corn and soybean yields after eliminating all fertilizer
182 applications under average weather conditions (Figure 4). As expected, the impacts on corn
183 yields were seen almost immediately when both nitrogen and P applications were stopped.
184 However, while eliminating only P applications decreased yield only slightly at first, significant
185 declines did not occur for almost 25 years. Soybeans were not impacted by eliminating nitrogen
186 fertilizer because they can fix nitrogen; however, they too saw significant impacts from cessation
187 of P applications after 25 years.

188 This time lag between eliminating P applications and decreasing corn and soybean yields
189 suggests significant excess or legacy P stored in the watershed's soils. Previous studies of
190 intensively managed croplands have shown that legacy P in the soil can maintain crop growth for
191 8-20 years after P fertilizer applications cease.³³⁻³⁵ The level of legacy P in soils may be
192 substantial in the MRW. Soil test P is often used to guide application rates³⁶ because higher soil
193 test P levels correspond to greater plant available P; higher soil test P can also lead to greater
194 dissolved P in runoff.³⁷ Only 17% of fields surveyed in a nearby Lake Erie watershed had a
195 Mehlich-3 Phosphorus (M3P) soil test P value of less than 21 ppm, whereas 48% had values
196 between 21-43 ppm, 25% had values between 43-71 ppm, and 11% had values greater than 71
197 ppm.³⁸ The Tri-State Fertilizer Recommendations (for Michigan, Ohio, and Indiana)³⁹ suggest

198 that a Bray-P1 soil test P value of 15 ppm (~ 27.7 ppm M3P)⁴⁰ is the critical P level for corn and
199 soybeans, thus, in the soils reported³⁸, a large portion likely did not need additional P fertilizer.
200 Yet, that study indicated that even for fields surveyed with greater than 71 ppm M3P soil test
201 values, additional P fertilization was still being recommended for 75% of fields. Above the
202 critical level, soil can supply P needed for the crop to grow, and below this level P will likely
203 limit crop growth. For soil at critical soil test P levels, the Tri-State Recommendations are that no
204 P fertilizer is required, yet they provide an estimate of P to apply as a means to “safeguard
205 against sampling or analytical variation”.³⁹ This safety factor built in to the recommendations
206 may be one factor responsible for over-application of P in the Midwest.⁴¹

207 [FIGURE 4]

208 By modifying the SWAT code (see SI), we were able to report simulated values corresponding to
209 the soil test P. In doing so, we found reducing SWAT’s inorganic soil P initialization⁴² to 1/5th
210 of the default value resulted in a more reasonable range of soil test P (range: 10-145; median: 35)
211 that corresponded well with existing data for soils tested in the Ohio Lake Erie Basin (range: 2.8-
212 291; median: 35.8).³⁸ To further emphasize this point, we tested the effect of initializing SWAT
213 at soil P levels higher than the calibrated value, and found that there is a “critical” SWAT soil
214 test value of ~10 ppm below which corn and soybean production decline in the MRW model (see
215 SI). In these tests, the soil could maintain critical levels for the duration of the entire simulation
216 (>80 years) and thus maintain crop growth, which is unlikely even in this intensively managed
217 system.

218 **Response to Agricultural Conservation Practices** - While the previous section focused on
219 system responses to a theoretical elimination of fertilizer under average, wet, and dry weather,

220 here we evaluate the impact of conservation practices driven by historical climate. The following
221 conservation practice scenarios were simulated under observed 1981-2010 climate conditions
222 and compared to the baseline model under those same conditions (Figure 5).

223 [FIGURE 5]

224 *In-Field Practices* - When inorganic P application rates were reduced to 75%, 50%, 25%, and
225 0% of the rates used in the baseline model, both DRP and TP loads decreased (Figure 5) with
226 minimal impacts on crop yields (SI Figure S3). Even the 0% inorganic P scenario still exceeded
227 the proposed TP and DRP target loads in some years with little impact on yields, indicating that
228 manure applications, combined with legacy P, may be sufficient for plant growth for this 30 year
229 period. This is consistent with the previous section's results that suggested at least 25 years may
230 be required for decreased fertilization rates to impact corn and soybean yields under average
231 weather conditions. Including a winter cover crop in rotations across all cropland reduced TP
232 loads, but did not effectively decrease DRP (Figure 5). Winter cover crops reduce erosion and
233 loss of sediment-bound P, while DRP concentrations in tile drainage flow remain unaffected. It
234 should be noted that cereal rye modeled in this study is only one of many cover crops used by
235 farmers, and while it is one of the best performing cover crops in poorly drained soils and
236 establishes late in the fall, it may not be representative of some of the other common crops,
237 particularly those that winter kill.

238
239 *Edge-of-Field Practice* - Simulated vegetated buffer strips varied in their ability to intercept TP
240 and DRP (Figure 5). With higher quality filter strips placed along every agricultural waterway,
241 the loading target could often be met for TP. However, this average masks weather-related inter-
242 annual variability, and it is likely filter strips would not prevent wetter-than-average years from

243 exceeding the targets (SI Figure S4). DRP load was reduced with filter strips because some
244 portion of DRP moves with surface runoff, but it did not meet targets on average, even at full
245 watershed implementation. One of the reasons is that approximately 40% of DRP is delivered
246 through tile drains in our model (a value in the range reported in King et al.¹⁴) and those drains
247 empty directly to ditches and streams, bypassing filter strips. While this percent of DRP through
248 tiles was calibrated to match the reported range of field values, further improvements in the
249 modeling of P transport through tiles could influence the results of this scenario, especially with
250 respect to DRP loads. Additionally, filter strips are given no spatial area of converted land in the
251 model, and while surface runoff from farm fields passes through before reaching the stream, they
252 do not treat streamflow when waters rise above stream banks. Future improvements to the
253 mechanics of filter strips in SWAT could also influence prediction of P intercepted by filter
254 strips.

255
256 *Alternative Land-Use Strategies* – The impacts of changing all agricultural land to alternative
257 row crop rotations (continuous sunflower, continuous lentil, or sunflower-lentil rotations) varied
258 as a function of how much P was applied (Figure 5). In the high P application scenarios (roughly
259 equivalent to current corn/soybean application rates), the alternative row crops generally had
260 higher DRP loads and very similar TP loads to the baseline model. In general, the alternative row
261 crops performed similarly to the baseline, suggesting that the amount of P applied to row crops
262 was more important than the type of crop planted, especially as modeled by SWAT. However,
263 when comparing the recommended P fertilizer rates from the Tri-State Standard for corn and
264 soybean to growth handbooks for sunflowers and lentils,⁴³⁻⁴⁴ under similar soil P levels, these

265 alternative crops require less P. Alternative crops could have a greater impact on P losses to the
266 extent that they have require lower P applications.

267

268 Changing the entire MRW from conventional row crop agriculture to perennial biofuel grasses
269 would have a significant impact on P loads (Figure 5). These crops were fertilized with nitrogen
270 at rates required for their growth,⁴⁵⁻⁴⁸ and with no inorganic P. Thus, each biofuel scenario was
271 run with and without manure application for better comparison to the baseline model. Compared
272 to the baseline, if all agricultural lands produced either switchgrass or miscanthus, DRP and TP
273 loads would diminish greatly. Continuing manure applications prevented achieving targets for
274 DRP in some years, whereas these grasses grown without manure additions met targets every
275 year. Their impact on TP loads is much greater because these crops are grown in stands (not
276 rows), don't winter kill, and develop deeper and larger root systems. This allows these crops to
277 trap sediment and runoff more easily, tap into legacy nutrients within the soil profile, and
278 provide cover for soil in the winter to prevent erosion. For these reasons the TP targets were met
279 under all scenarios and variations. Biofuel crop yields remained relatively constant throughout
280 the time period (SI Figure S5), and were similar to previously reported values for a similar
281 climate.⁴⁷

282

283 **Policy Implications**

284 Our results highlight information which should help those responsible for developing action
285 plans to reduce P loads to Lake Erie from agricultural sources. First, our results support
286 conclusions from Bosch et al.¹² that it will take more than one agricultural conservation practice
287 with widespread implementation to see consistent P load reductions because no single practice

288 could achieve the target loads every year. Second, the alternative biofuel crops did show
289 significant reductions in both TP and DRP. Although these were also implemented across the
290 entire MRW, partial and perhaps targeted conversion of marginal croplands back to grassland
291 has the potential to decrease P loads, and using biofuel crops could allow farms to remain
292 economically viable given the right policy incentives. Third, our results demonstrate the
293 importance of P stored within the soil and its long-term impact on restoring water quality. For
294 example, in our simulations, crop growth continued for ~25 years and it took as much as 30-40
295 years to reach P load targets even after fertilization was stopped, depending on climatic
296 conditions. This suggests a large amount of legacy P exists in the MRW soil and this legacy will
297 need to be addressed in agricultural conservation. These same runs also demonstrated that DRP
298 loads could drop relatively quickly under average weather conditions, highlighting the
299 importance of current P fertilizer applications in driving a given year's loads. From this it is clear
300 that conservation strategies will need target losses due to current fertilizer applications, as well as
301 losses due to legacy P in the system. Lastly, this work highlights the importance of climatic
302 impacts on P given the stark differences between TP and DRP loads under wet, dry, and average
303 weather conditions. Therefore, it will be crucial for policymakers to understand how climate
304 change, or even inter-annual variability, may impact proposed management solutions. For
305 example, previous work in the Maumee has shown that precipitation may decrease in May-to-
306 October, but increase in the rest of the year⁴⁵ so targeting conservation strategies temporally will
307 be important. Overall, this work demonstrates that significant changes in agricultural land
308 management in the Western Lake Erie Basin are needed to reach target P loads in order to
309 prevent or mitigate future harmful algal blooms.

310 AUTHOR INFORMATION

311 **Corresponding Author**

312 *rlogsdon@umich.edu; phone: (734) 647-2495

313 **Author Contributions**

314 The manuscript was written through contributions of all authors. All authors have given approval
315 to the final version of the manuscript.

316 ACKNOWLEDGMENT

317 This work was funded in part by the University of Michigan Graham Sustainability Institute, the
318 National Science Foundation (NSF) under grant 1313897, and the Joyce Foundation under grant
319 15-36415.

320 ABBREVIATIONS

321 BAU, business as usual; DRP, dissolved reactive phosphorus; GLWQA, Great Lakes Water
322 Quality Agreement; HABs, harmful algal blooms; M3P, Mehlich-3 Phosphorus; MRW, Maumee
323 River watershed; NCWQR, National Center for Water Quality Research ; No Fert, no fertilizers;
324 No inorganic P Fert, no inorganic phosphorus fertilizers; No P Fert, no phosphorus fertilizers; P,
325 phosphorus; SI, supporting information; SWAT, Soil and Water Assessment Tool; TP, total
326 phosphorus.

327 SUPPORTING INFORMATION. Includes additional information on the SWAT model set-up,
328 details for year selection in legacy phosphorus runs, details for management scenario runs,
329 additional results, and description of methods used to evaluate soil test phosphorus in SWAT.

330 REFERENCES

- 331 1. Conroy, J.D.; Kane, D.D.; Briland, R.D.; Culver, D.A. Systematic, early-season *Microcystis*
332 blooms in western Lake Erie and two of its major agricultural tributaries (Maumee and
333 Sandusky rivers). *J. Great Lakes Res.* **2014**, *40* (3), 518-523.
- 334 2. Scavia, D.; Allan, J.D.; Arend, K.K.; Bartell, S.; Beletsky, D.; Bosch, N.S.; Brandt, S.B.;
335 Briland, R.D.; Daloğlu, I.; DePinto, J.V.; Dolan, D.M.; Evans, M.A.; Farmer, T.M.; Goto,
336 D.; Han, H.; Höök, T.O.; Knight, R.; Ludsın, S.A.; Mason, D.; Michalak, A.M.; Richards,
337 R.P.; Roberts, J.J.; Rucinski, D.K.; Rutherford, E.; Schwab, D.J.; Sesterhenn, R.; Zhang,
338 H.; Zhou, Y. Assessing and addressing the re-eutrophication of Lake Erie: Central Basin
339 Hypoxia. *J. Great Lakes Res.* **2014**, *40*, 226–246;
340 DOI: <http://dx.doi.org/10.1016/j.jglr.2014.02.004>.
- 341 3. Michalak, A.M.; Anderson, E.J.; Beletsky, D.; Boland, S.; Bosch, N.S.; Bridgeman, T.B.;
342 Chaffin, J.D.; Cho, K.; Confesor, R.; Daloğlu, I.; DePinto, J.V.; Evans, M.A.; Fahenstiel,
343 G.L.; He, L.; Ho, J.C.; Jenkins, L.; Johengen, T.H.; Kuo, K.C.; LaPorte, E.; Liu, X.;
344 McWilliams, M.R.; Moore, M.R.; Posselt, D.J.; Richards, R.P.; Scavia, D.; Steiner, A.L.;
345 Verhamme, E.; Wright, D.M.; Zagorski, M.A. Record-setting algal bloom in Lake Erie
346 caused by agricultural and meteorological trends consistent with expectant future
347 conditions. *Pro. Natl. Acad. Sci.* **2013**, *110* (16), 6448-6452.
- 348 4. Obenour, D.R.; Gronewold, A.D.; Stow, C.A.; Scavia, D. Using a Bayesian hierarchical model
349 with a gamma error distribution to improve Lake Erie cyanobacteria bloom forecasts.
350 *Wat. Resour. Res.* **2014**, *50*, 7847-7860; DOI 10.1002/2014WR015616.

- 351 5. Stumpf, R.P.; Wynne, T.T.; Baker, D.B.; Fahenstiel, G.L. Interannual variability of
352 cyanobacterial blooms in Lake Erie. *PLoS ONE* **2012**, 7 (8), e41444; DOI:
353 10.1371/journal.pone.0042444.
- 354 6. *Great Lakes Water Quality Agreement-2012, Between the United States of America and*
355 *Canada*; International Joint Commission: Windsor, Ontario, 2012;
356 http://www.ijc.org/en_/Great_Lakes_Water_Quality.
- 357 7. *Great Lakes Water Quality Agreement of 1978, with Annexes and Terms of Reference Between*
358 *the United States and Canada; International Joint Commission*; Windsor, Ontario,
359 Canada, Nov 22, 1978.
- 360 8. *Recommended phosphorus loading targets for Lake Erie; Annex 4 Objectives and Targets*
361 *Task Team Final Report to the Nutrients Annex Subcommittee May 11, 2015*;
362 [http://www2.epa.gov/sites/production/files/2015-06/documents/report-recommended-](http://www2.epa.gov/sites/production/files/2015-06/documents/report-recommended-phosphorus-loading-targets-lake-erie-201505.pdf)
363 [phosphorus-loading-targets-lake-erie-201505.pdf](http://www2.epa.gov/sites/production/files/2015-06/documents/report-recommended-phosphorus-loading-targets-lake-erie-201505.pdf).
- 364 9. Rucinski, D.; Scavia, D.; DePinto, J.; Beletsky, D. Modeling Lake Erie's hypoxia response to
365 nutrient loads and physical variability. *J. Great Lakes Res.* **2014**, 40 (3), 151-161; DOI
366 10.1016/j.jglr.2014.02.003.
- 367 10. *A Balanced Diet for Lake Erie: Reducing Phosphorus Loadings and Harmful Algal Blooms.*
368 *Report of the Lake Erie Ecosystem Priority*; International Joint Commission (IJC), 2014.
- 369 11. Scavia D, DePinto J, Bertani I A multi-model approach to evaluating target phosphorus loads
370 for Lake Erie. *in review*.

- 371 12. Bosch, N.S.; Allan, J.D.; Selegean, J.P.; Scavia, D. Scenario-testing of agricultural best
372 management practices in Lake Erie watersheds. *J. Great Lakes Res.* **2013**, *39* (3), 429-
373 436.
- 374 13. Daloglu, I.; Cho, K.H.; Scavia, D. Evaluating causes of trends in long-term dissolved reactive
375 phosphorus loads to Lake Erie. *Environ. Sci. Technol.* **2012**, *46* (19), 10660-10666.
- 376 14. King, K.W.; Williams, M.R.; Macrae, M.L.; Fausey, N.R.; Frankenberger, J.; Smith, D.R.;
377 Kleinman, P.J.A.; Brown, L.C. Phosphorus transport in agricultural subsurface drainage:
378 A review. *J. Environ. Qual.* **2015**, *44* (2), 467-485.
- 379 15. Smith, D.R.; King, K.W.; Johnson, L.; Francesconi, W.; Richards, P.; Baker, D.; Sharpley,
380 A.N. Surface runoff and tile drainage transport of phosphorus in the Midwestern United
381 States. *J. Environ. Qual.* **2015**, *44* (2), 495-502.
- 382 16. Tributary Data Download; <http://www.heidelberg.edu/academiclife/distinctive/newqr/data>.
- 383 17. Arnold, J.G.; Srinivasan, R.; Muttiah, R.S.; Williams, J.R. Large area hydrologic modeling
384 and assessment part 1: Model development. *J. Am. Wat. Resour. Assoc.* **1998**, *34* (1), 73-
385 89; DOI 10.1111/j.1752-1688.1998.tb05961.x
- 386 18. Douglas-Mankin, K.R.; Srinivasan, R.; Arnold, J.G. Soil and water assessment tool (SWAT)
387 model: Current developments and applications. *Trans. Am. Soc. Agric. Biol. Eng.* **2010**,
388 *53* (5), 1423-1431.
- 389 19. Van Liew, M.W.; Veith, T.L.; Bosch, D.D.; Arnold, J.G. Suitability of SWAT for the
390 conservation effects assessment project: Comparison on USDA agricultural research

- 391 service watersheds. *J. Hydrol. Eng.* **2007**, 12 (2), 173-189; DOI 10.1061/(ASCE)1084-
392 0699(2007)12:2(173).
- 393 20. Kalcic, M.M.; Kirchoff, C.; Bosch, N.; Muenich, R.L.; Murray, M.; Scavia, D. Engaging
394 Stakeholders to Define Feasible and Desirable Agricultural Conservation in Western
395 Lake Erie Watersheds. *Environ. Sci. Technol.* **In Review**.
- 396 21. Moriasi, D.N.; Arnold, J.G.; Van Liew, M.W.; Bingner, R.L.; Harmel, R.D.; Veith, T.L.
397 Model evaluation guidelines for systematic quantification of accuracy in watershed
398 simulations. *Trans. Am. Soc. Agric. Biol. Eng.* **2007**, 50 (3), 885-900.
- 399 22. *Effects of conservation practice adoption on cultivated cropland acres in Western Lake Erie*
400 *Basin*; United States Department of Agriculture Natural Resource Conservation Service:
401 Washington, DC, 2016;
402 <http://www.nrcs.usda.gov/wps/portal/nrcs/detail/national/technical/nra/ceap/pub/?cid=nrc>
403 [seprd949606](http://www.nrcs.usda.gov/wps/portal/nrcs/detail/national/technical/nra/ceap/pub/?cid=nrcseprd949606).
- 404 23. Schipanski, M.E.; Barbercheck, M.; Douglas, M.R.; Finney, D.M.; Haider, K.; Kaye, J.P.;
405 Kemanian, A.R.; Mortensen, D.A.; Ryan, M.R.; Tooker, J.; White, C. A framework for
406 evaluating ecosystem services provided by cover crops in agroecosystems. *Agric. Syst.*
407 **2014**, 125, 12-22; DOI 10.1016/j.agsy.2013.11.004.
- 408 24. Tonitto, C.; David, M.B.; Drinkwater, L.E. Replacing bare fallows with cover crops in
409 fertilizer-intensive cropping systems: A meta-analysis of crop yield and N dynamics.
410 *Agric. Ecosyst. Environ.* **2006**, 112 (1), 58-72; DOI10.1016/j.agee.2005.07.003.

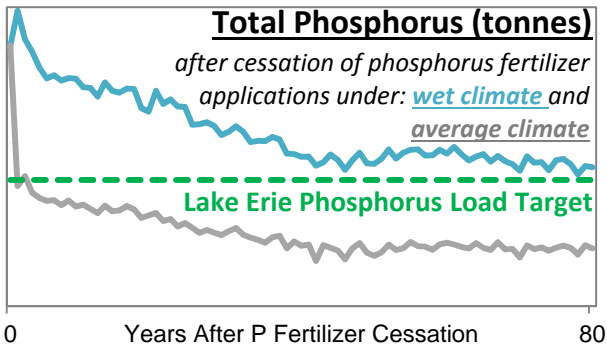
- 411 25. Stutter, M.I.; Chardon, W.J.; Kronvang, B. Riparian buffer strips as a multifunctional
412 management tool in agricultural landscapes: Introduction. *J. Environ. Qual.* **2012**, *41*,
413 297-303; DOI 10.2134/jeq2011.0439.
- 414 26. Zhou, X.; Helmers, M.J.; Asbjornsen, H.; Kolka, R.; Tomer, M.D.; Cruse, R.M. Nutrient
415 removal by prairie filter strips in agricultural landscapes. *J. Soil Wat. Conserv.* **2014**, *69*
416 (1), 54-64; DOI 10.2489/jswc.69.1.54.
- 417 27. *Soil and Water Assessment Tool Theoretical Documentation Version 2009*; Texas Water
418 Resources Institute: College Station, TX; [http://swat.tamu.edu/media/99192/swat2009-](http://swat.tamu.edu/media/99192/swat2009-theory.pdf)
419 [theory.pdf](http://swat.tamu.edu/media/99192/swat2009-theory.pdf).
- 420 28. Oplinger, E.S.; Hardman, L.L.; Kaminski, A.R.; Kelling, K.A.; Doll, J.D. Lentil. In
421 *Alternative Field Crops Manual*; University of Wisconsin-Extension, Cooperative
422 Extension 1990; <https://www.hort.purdue.edu/newcrop/afcm/lentil.html>.
- 423 29. Putnam, D.H.; Oplinger, E.S.; Hicks, D.R.; Durgan, B.R.; Noetzel, D.M.; Meronuck, R.A.;
424 Doll, J.D.; Schulte, E.E. Sunflower. In *Alternative Field Crops Manual*; University of
425 Wisconsin-Extension, Cooperative Extension 1990;
426 <https://www.hort.purdue.edu/newcrop/afcm/sunflower.html>.
- 427 30. Cibin, R.; Trybula, E.; Chaubey, I.; Brouder, S.; Volenec, J.J. Watershed scale impacts of
428 bioenergy crops on hydrology and water quality using improved SWAT model. *Glob.*
429 *Change Biol. Bioenerg.* **2015**; DOI 10.1111/gcbb.12307

- 430 31. Heaton, E.; Dohleman, F.G.; Long, S.P. Meeting US biofuel goals with less land: the
431 potential of Miscanthus. *Global Change Biol.* **2008**, *14*, 2000-2014; DOI 10.1111/j.1365-
432 2486.2008.01662.x.
- 433 32. Sarkar, S.; Miller, S.A. Water quality impacts of converting intensively-managed agricultural
434 lands to switchgrass. *Biomass Bioenerg.* **2014**, *68*, 32-43; DOI
435 10.1016/j.biombioe.2014.05.026.
- 436 33. Dodd, J.R.; Mallarino, A.P. Soil-test phosphorus and crop grain yield response to long-term
437 phosphorus fertilization for corn-soybean rotations. *Soil Sci. Soc. Am. J.* **2005**, *69*, 1118-
438 1128.
- 439 34. Dodd, J.R.; McDowell, R.W.; Condon, L.M. Predicting the changes in environmentally and
440 agronomically significant phosphorus forms following the cessation of fertilizer
441 applications to grassland. *Soil Use Mgt.* **2012**, *28* (2), 135-147; DOI 10.1111/j.1475-
442 2743.2012.00390.x.
- 443 35. McCollum, R.E. Buildup and decline in soil phosphorus: 30-year trends on a Typic
444 Umprabuult. *Agron J.* **1991**, *83*, 77-85.
- 445 36. Sims, J.T.; Edwards, A.C.; Schoumans, O.F.; Simrad, R.R. Integrating soil phosphorus
446 testing into environmentally based agricultural management practices. *J. Environ. Qual.*
447 **2000**, *29*(1), 60-71.
- 448 37. Vadas, P.A.; Kleinman, P.J.A.; Sharpley, A.N.; Turner, B.L. Relating soil phosphorus to
449 dissolved phosphorus in runoff. *J. Environ. Qual.* **2005**, *34* (2), 572-580; DOI
450 10.2134/jeq2005.0572.

- 451 38. Summary of Supplementary Data;
452 http://www.heidelberg.edu/sites/default/files/jfuller/files/Team_Mtg._5.4_Summary_of_s
453 [supplementar_data.pptx](http://www.heidelberg.edu/sites/default/files/jfuller/files/Team_Mtg._5.4_Summary_of_s).
- 454 39. Vitosh, M.L.; Johnson, J.W.; Mengel, D.B. *Tri-state Fertilizer Recommendations for Corn,*
455 *Soybeans, Wheat and Alfalfa*; Michigan State Extension Bulletin E-2567 1995;
456 <https://www.extension.purdue.edu/extmedia/AY/AY-9-32.pdf>.
- 457 40. Watson, M.; Mullen, R. *Understanding soil tests for plant-available phosphorus*. The Ohio
458 State University Extension Report 3373 June 2007.
- 459 41. Smith, D.R.; King, K.W.; Williams, M.R. What is causing the harmful algal blooms in Lake
460 Erie? *J. Soil Wat. Conserv.* **2015**, *70* (2), 27A-29A; DOI 10.2489/jswc.70.2.27A.
- 461 42. Vadas, P.A.; White, M.J. Validating soil phosphorus routines in the SWAT model. *Trans.*
462 *Am. Soc. Agric. Biol. Eng.* **2010**, *53* (5), 1469-1476.
- 463 43. Meyer, R.; Belshe, D.; O'Brien, D.; Darling, R., Eds. *High Plains Sunflower Production*
464 *Handbook*; North Dakota State University, Cooperative Extension Service;
465 http://www.agmrc.org/media/cms/Sunflowers_C84E1143C31B9.pdf.
- 466 44. Kandel, H.; Ashley, R., Eds. *Growing Lentil in North Dakota*; North Dakota State Extension
467 Service Report A-1636 2013; <http://www.ag.ndsu.edu/pubs/plantsci/rowcrops/a1636.pdf>.
- 468 45. Khanna, M.; Dhungana, B.; Clifton-Brown, J. Costs of producing miscanthus and
469 switchgrass for bioenergy in Illinois. *Biomass Bioenergy* **2008**, *32* (6), 482-493. DOI
470 10.1016/j.biombioe.2007.11.003.

- 471 46. Muir, J.P.; Sanderson, M.A.; Ocumpaugh, W.R.; Jones, R.M.; Reed, R.L. Biomass
472 production of ‘Alamo’ switchgrass in response to nitrogen, phosphorus, and row spacing.
473 *Agron. J.* **2001**, *93* (4), 896-901.
- 474 47. Trybula, E.M.; Cibir, R.; Burks, J.L.; Chaubey, I.; Brouder, S.M.; Volenec, J.J. Perennial
475 rhizomatous grasses as bioenergy feedstock in SWAT: parameter development and model
476 improvement. *Glob. Change Biol. Bioenerg.* **2014**, *7* (6), 1185-1202; DOI
477 10.1111/gcbb.12210.
- 478 48. Vogel, K.P.; Brejda, J.J.; Walters, D.T.; Buxton, D.R. Switchgrass biomass production in the
479 Midwest USA: Harvest and Nitrogen Management. *Agron. J.* **2002**, *94* (3), 413-420; DOI
480 10.2134/agronj2002.0413.
- 481 49. Verma, S.; Bhattarai, R.; Bosch, N.S.; Cooke, R.C.; Kalita, P.K.; Markus, M. Climate change
482 impacts on flow, sediment and nutrient export in a Great Lakes watershed using SWAT.
483 *CLEAN--Soil, Air, Wat.* **2015**, *43* (11), 1464-1474; DOI: 10.1002/clen.201400724.
- 484

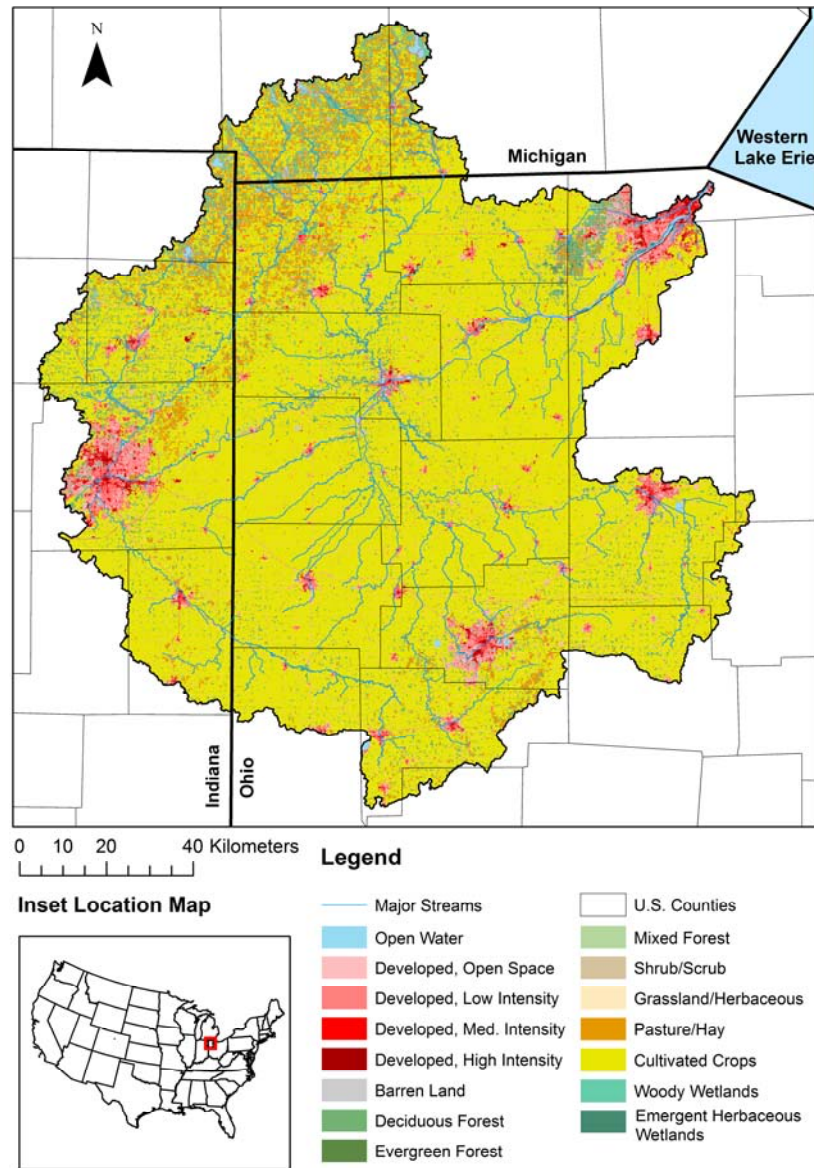
485 TABLE OF CONTENTS GRAPHIC AND SYNOPSIS



486

487

488



490

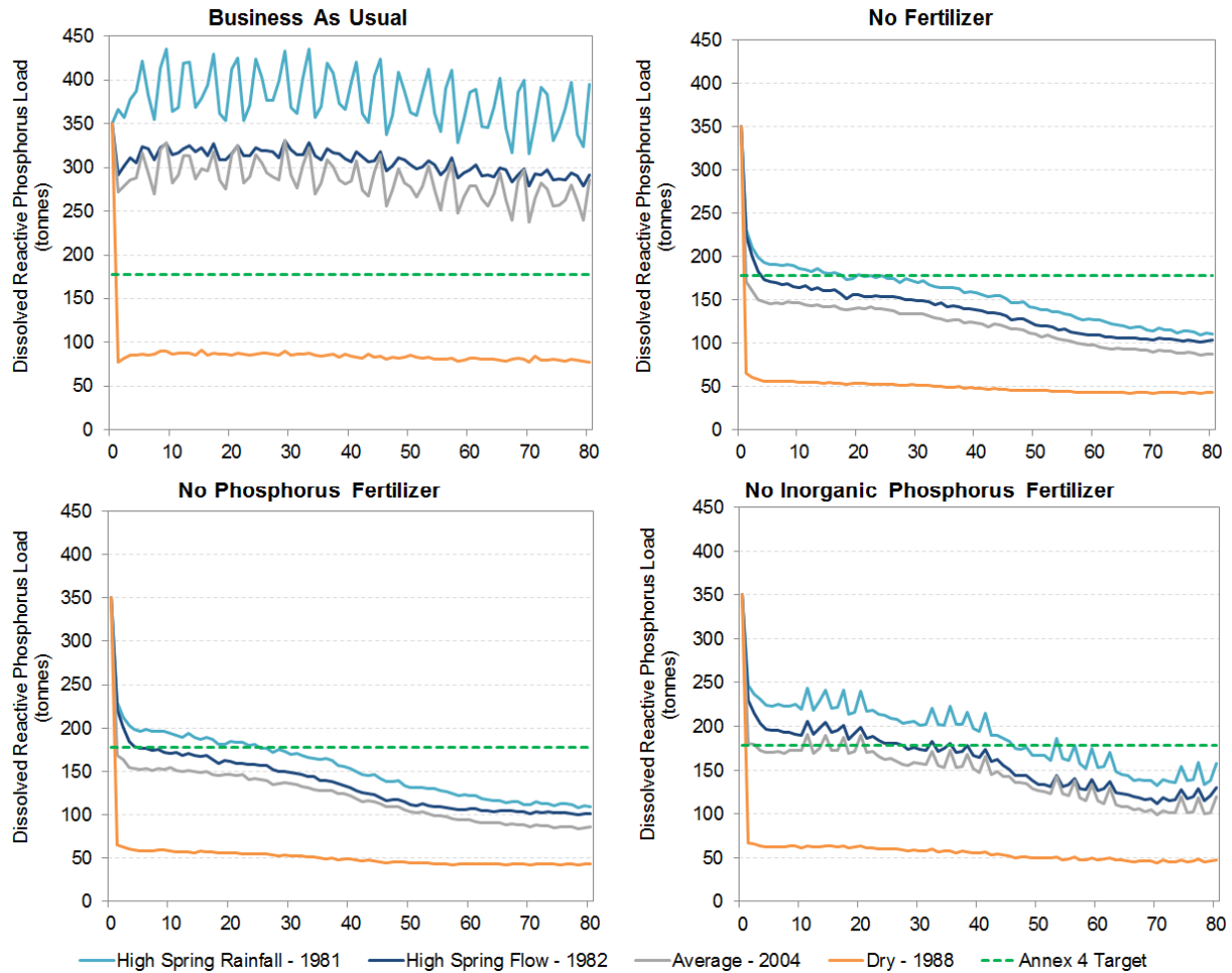
491 **Figure 1.** Location and land-use (from the USGS 2006 National Land Cover Dataset) of the
 492 Maumee River Watershed. The Maumee River Watershed Soil and Water Assessment Tool
 493 model was developed using a combination of the National Land Cover Dataset and the National
 494 Agricultural Statistics Service Cropland Data Layer, extracted for multiple years²⁰.

495 **Table 1.** Alternative P reduction strategy scenarios run under historical climatic conditions.

496 Further details for scenario implementations in SWAT are provided in the SI.

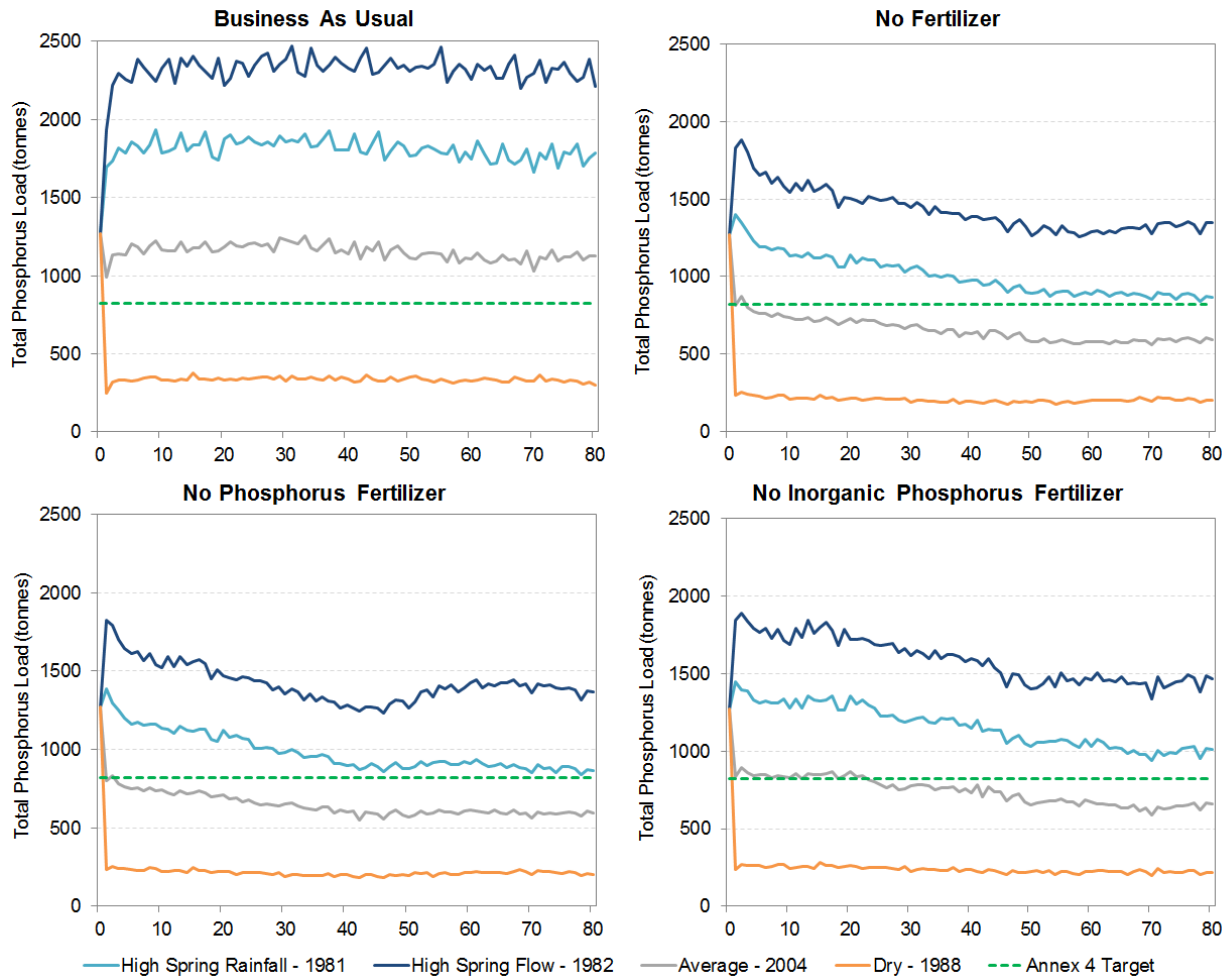
Type	Scenarios	Further variations
Rate of P applications	1. Baseline with 75% inorganic P fertilizer 2. Baseline with 50% inorganic P fertilizer 3. Baseline with 25% inorganic P fertilizer 4. Baseline with 0% inorganic P fertilizer	None
Filter Strips	5. 25% of agricultural lands with filter 6. 50% of agricultural lands with filter 7. 75% of agricultural lands with filter 8. 100% of agricultural lands with filter	Filter strip condition: poor to good
Cover crops	9. Cover crops over 25% of ag. lands 10. Cover crops over 50% of ag. lands 11. Cover crops over 75% of ag. lands 12. Cover crops over 100% of ag. lands	None
Alt. Row Crops	13. High P application 14. Medium P application 15. Zero P application	Three rotations: continuous sunflower, continuous lentil, and sunflower-lentil rotation
Biofuels	16. Manure applied (not incorporated) 17. No manure applied	Two varieties: Shawnee switchgrass (<i>Panicum vigratum</i>) and <i>Miscanthus x giganteus</i>

497



498

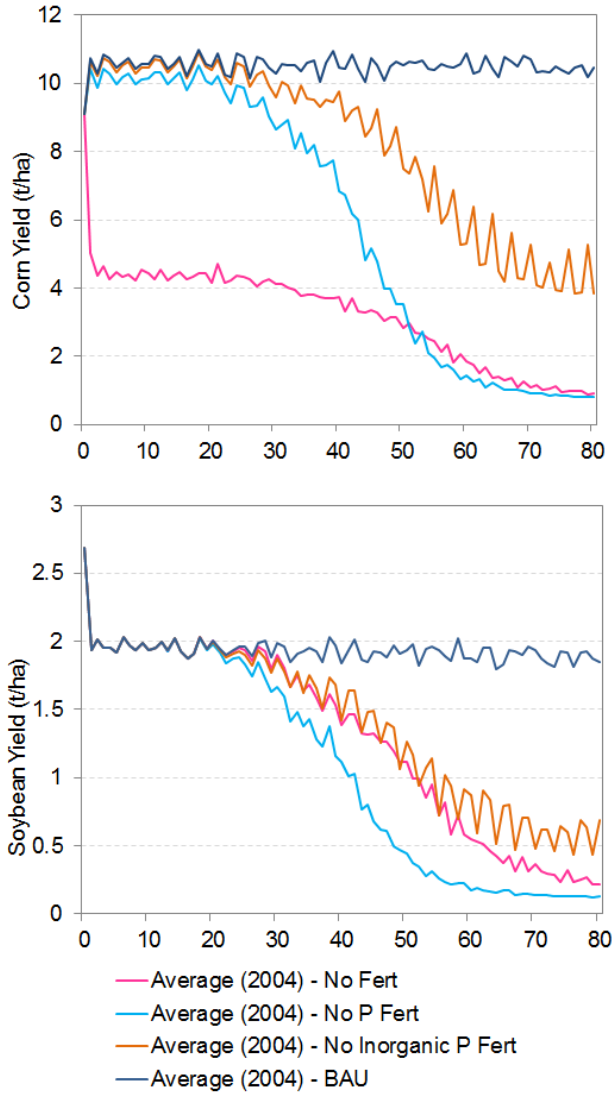
499 **Figure 2.** March-to-July dissolved reactive phosphorus loads under each fertilizer application
500 cessation scenario for each type of climate condition across 80 years of simulation. The **light**
501 **blue line** represents high spring rainfall conditions (1981), the **dark blue line** represents high
502 spring streamflow conditions (1982), the **gray line** represents average climate conditions (2004),
503 the **orange line** represents dry climate conditions (1988) and the **dashed green line** represents the
504 GLWQA targets area-weighted to the Heidelberg water quality station near Waterville, Ohio.



505

506 **Figure 3.** March-to-July total phosphorus loads under each fertilizer application cessation
 507 scenario for each type of climate condition across 80 years of simulation. The light blue line
 508 represents high spring rainfall conditions (1981), the dark blue line represents high spring
 509 streamflow conditions (1982), the gray line represents average climate conditions (2004), the
 510 orange line represents dry climate conditions (1988) and the dashed green line represents the
 511 GLWQA targets area-weighted to the Heidelberg water quality station near Waterville, Ohio.

512

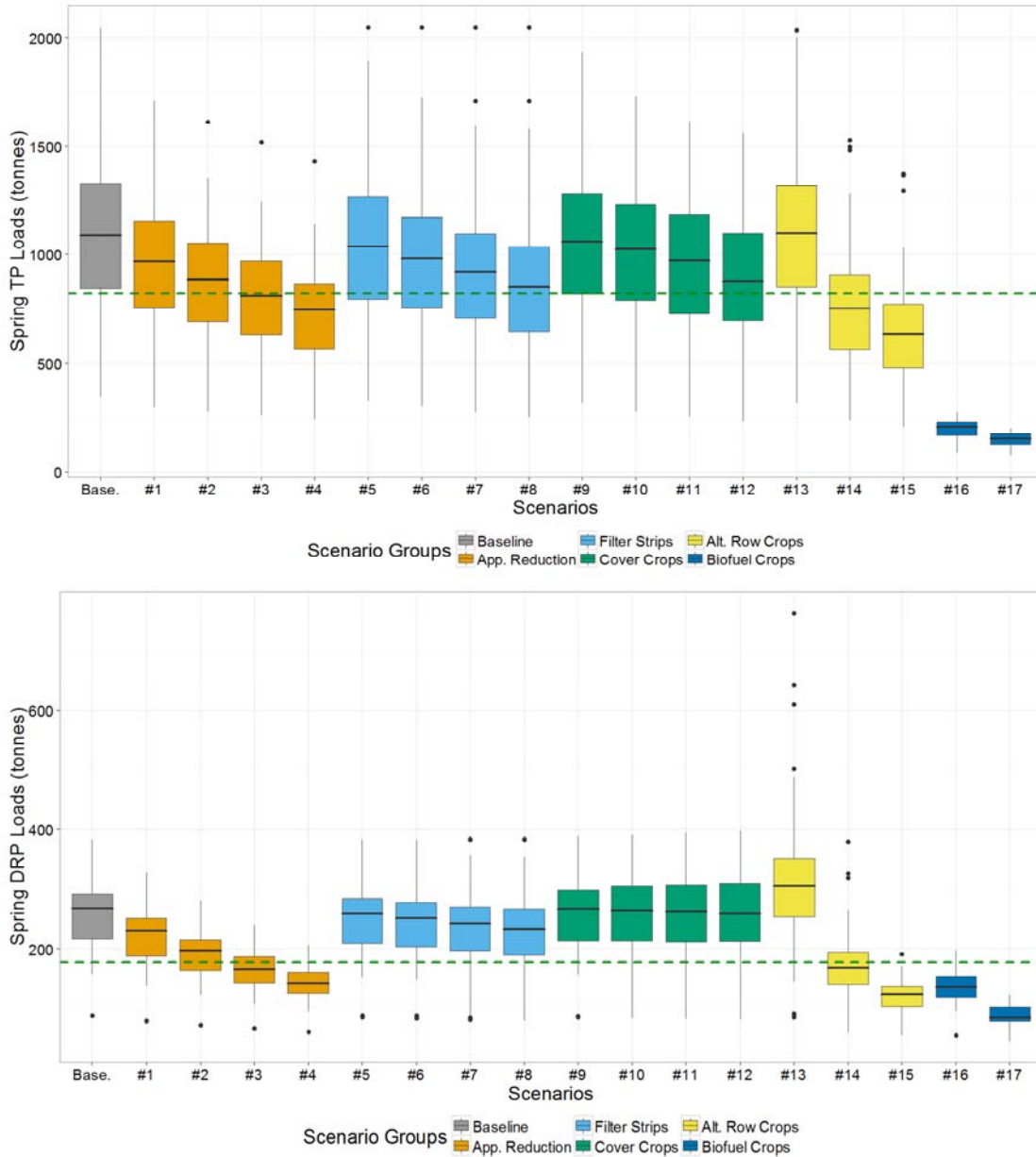


513

514 **Figure 4.** Average MRW yields (t/ha) of corn (top) and soybean (bottom) under average (2004)

515 weather conditions across 80 years of simulation.

516



517

518 **Figure 5.** Scenario results under 1981-2010 climate for springtime (March-to-July) TP (top) and

519 DRP (bottom) loads grouped by scenario types. Colored boxes contain results from 75% of the

520 years in the multi-year simulations. Solid black line represents the median value. The thin

521 vertical lines represent the range of the results, and the dots show extreme outliers. The GLWQA

522 targets are shown with the dashed line. Observed DRP loading differs more from the baseline

523 because the model was calibrated for a period with higher DRP concentrations.