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Detroit River Phosphorus Loads: Anatomy of a Binational Watershed

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25 **Abstract**

26 As a result of increased harmful algal blooms and hypoxia in Lake Erie, the US and Canada
27 revised its phosphorus loading targets under the 2012 Great Lakes Water Quality Agreement.
28 The focus of this paper is the Detroit River and its watershed, a source of 25% of the total
29 phosphorus (TP) load. Its load declined 37% since 1998, due chiefly to improvements at the
30 regional Great Lakes Water Authority Water Resource Recovery Facility (WRRF) in Detroit
31 (WRRF) and phosphorus sequestered by zebra and quagga mussels in Lake Huron. In addition to
32 the 54% of the load from Lake Huron, nonpoint sources contribute 57% of the TP load and 50%
33 of the dissolved reactive phosphorus load, with the remaining balance from point sources. After
34 Lake Huron, the largest source is the WRRF, which has already reduced its load by over 40%.
35 Currently, loads from Lake Huron and further reductions from the WRRF are not part of the
36 reduction strategy, therefore remaining watershed sources will need to decline by 72% to meet
37 the Water Quality Agreement target- a daunting challenge. Because other urban sources are very
38 small, most of the reduction would have to come from agriculturally-dominated lands. The most
39 effective way to reduce those loads is to apply combinations of practices like cover crops, buffer
40 strips, wetlands, and applying fertilizer low below the soil surface on the lands with the highest
41 phosphorus losses. However, the simulations suggest even extensive conservation on those lands
42 may not be enough.

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47 **Introduction**

48 Among the Laurentian Great Lakes, Lake Erie is the warmest, shallowest, and most productive,
49 contributing to its sensitivity to nutrient inputs. In the 1960s and 70s, increasing phosphorus
50 inputs led to severe algal blooms in its western basin and periods of low oxygen (hypoxia) in the
51 bottom waters of its central basin. Phosphorus abatement programs, initiated in response to the
52 1972 Great Lakes Water Quality Agreement (GLWQA), prompted wastewater treatment
53 facilities to add secondary treatment, removed phosphorus from most soaps and detergents, and
54 enhanced land conservation programs, resulting in substantial water quality improvements
55 (DePinto et al, 1986, Ludsin et al. 2001).

56 However, in the mid-1990s, harmful algal blooms and hypoxia returned to conditions similar to
57 the 1960s and 70s (Scavia et al. 2014). Results from a synthesis of models (Scavia et al. 2016)
58 showed that the increasing spring load of dissolved reactive phosphorus (DRP) from the
59 Maumee River was the primary driver of the western basin blooms, and that the annual load of
60 total phosphorus (TP) to the western and central basins was the primary driver of hypoxia (Zhou
61 et al. 2015, Bridgeman et al. 2013, Michalak et al. 2013, Scavia et al. 2014, 2016; Rucinski et al.
62 2014, 2016; Obenour et al. 2014, Stumpf et al. 2016; Bertani et al. 2016; Bocaniov et al. 2016).

63 In 2012, the US and Canada revised the GLWQA, calling for new Lake Erie phosphorus loading
64 targets and associated action plans. In response to this commitment, they adopted the following
65 targets, each compared to a 2008 baseline (GLWQA 2016).

- 66 ● For central-basin hypoxia, a 40% reduction in the western and central basin TP load.

67 ● For healthy nearshore ecosystems, a 40% reduction of spring (March-July) TP and DRP
68 loads from the Thames River, Leamington tributaries, Maumee River, River Raisin,
69 Portage River, Toussaint Creek, Sandusky River, and Huron River (Ohio).

70 ● For western-basin algae blooms, a 40% reduction in Maumee spring TP and DRP loads.

71 US and Canadian domestic action plans placed substantial attention on loads from Detroit and
72 Maumee rivers because they contribute, respectively, 41% and 48% of the TP load to the western
73 basin, and 25% and 29% of the TP load to the whole lake (Maccoux et al. 2016; Scavia et al.
74 2016). The plans were developed within adaptive management frameworks and the initial phase
75 of review and potential adaptation is underway in 2019.

76 There have been several assessments of the relative contributions and potential controls of
77 phosphorus loads from the Maumee watershed (e.g., Scavia et al. 2017, Muenich et al. 2016,
78 Kalcic et al. 2016). However, the sources of nutrients contributing to the Detroit River load have
79 been somewhat uncertain due to limited data and an historical lack of attention to its watershed,
80 which includes both intensive agriculture and major urban areas. This river system is also
81 complicated by the presence of the large, shallow Lake St. Clair, which processes the nutrient
82 load from its 15,000 km² watershed, as well as from the St. Clair River. Whether the lake is an
83 ultimate source of, or sink for, phosphorus, and whether loads from its different tributaries (e.g.,
84 Clinton, Sydenham, Thames, St. Clair rivers) have equally significant impacts downstream, have
85 been unclear. It has also been difficult to measure accurately the Detroit River load because it is
86 not well mixed in transverse direction to flow, requiring extensive sampling across the river and
87 over time, and because Lake Erie storm surges and seiches occasionally can push lake water into
88 the river (Derecki and Quinn, 1990), introducing large uncertainties and hampering estimates of
89 river discharge and nutrient load.

90 Understanding nutrient sources is critical for developing load reduction plans and for deciding
91 the level of emphasis that should be placed on different tributaries or different source types (e.g.,
92 point sources, agricultural runoff). To help reduce these uncertainties, this project set the
93 following objectives with the help of stakeholders from the public and private sectors: 1)
94 estimate how different sources contribute to the Detroit River phosphorus load to Lake Erie, and
95 2) evaluate options for reducing those loads.

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97 **METHODS**

98 **Study region** - The St. Clair-Detroit River system (Figure 1) receives water and nutrients from
99 Lake Huron and the 19,040 km² watershed that covers parts of southeastern Michigan (40% of
100 watershed area) and southwestern Ontario (60% of watershed area). It delivers nutrients to Lake
101 Erie through the Detroit River. The Detroit River provides approximately 80% of the water flow
102 into Lake Erie and 25% of the lake's annual TP inputs, and its phosphorus concentrations are
103 relatively low compared to the Maumee River. Because of the low concentrations and high flow,
104 it tends to dilute nutrients in the western basin, creating a zone where the Detroit River and the
105 western basin water mix, and algae and total suspended solids concentrations are low (Figure
106 S1). However, the river's annual TP load contributes significantly to central-basin algal
107 production and ultimately to the extent of hypoxia there.

108 The watershed is composed of about 49% cropland, 21% urban area, 13% forest, 7% grassland,
109 and 7% water bodies (Dagnew et al. 2019a). Overall, 79% of the watershed's agricultural land is
110 in Canada and 83% of the urban land is in the US. The Clinton and Rouge sub-watersheds are
111 heavily urbanized (about 56% and 89% urban, respectively), whereas the St. Clair, Sydenham,

112 and Thames sub-watersheds are dominated by agriculture (63%, 89%, and 87% agricultural,
113 respectively).

114 The US portion of the watershed has three watersheds (St. Clair, Clinton, and Rouge) drained
115 primarily by the Black, Clinton, and Rouge rivers, respectively. These sub-watersheds often
116 include multiple drainage areas. For example, in addition to the Black, the St. Clair sub-
117 watershed includes the Pine and Bell river systems (see Figure 9 below), and the Rouge sub-
118 watershed includes the Rouge River system as well as land that drains directly in the Detroit
119 River. The Canadian portion of the watershed has three tertiary watersheds (Upper Thames,
120 Lower Thames, and Sydenham) drained by the Thames and Sydenham rivers. The study region
121 also includes the Essex watershed in Canada and the Lake St. Clair watershed in the US.

122 Five of the six sub-watersheds drain into the 1115 km², 4.25 km³ Lake St. Clair (Figure S2), a
123 shallow, polymictic lake with a mean depth of 3.8 m, a maximum natural depth of 6.5 m, and an
124 8.2 m deep navigation channel (Bocaniov and Scavia, 2018). It processes water and phosphorus
125 from lakes Superior, Michigan, and Huron via the St. Clair River, as well as from its proximate
126 15,000 km² watershed that is roughly 63% in Canada and 37% in the United States. While the
127 lake's theoretical flushing time is roughly 9 days, that flushing time varies seasonally and, more
128 significantly, spatially (Bocaniov and Scavia, 2018) such that during summer, water in the south-
129 eastern part of the lake flushes more slowly than the north-western part. This, in combination
130 with different timing and magnitude of tributary loads, leads to spatial segmentation of primary
131 production resulting in the northwest part of the lake being oligotrophic and southeast part
132 mesotrophic.

133 As part of this assessment, three urban regions received special focus. The National Land Cover
134 Database (NLDC, 2011) and the Annual Crop Inventory (Agriculture and Agri-food Canada

135 2011) were used to select HUC-12 subbasins with more than 80% urban land cover in the US
136 and more than 60% in Canada (Hu et al. in review). This resulted in study areas in southeast
137 Michigan and around London, Ontario and Windsor, Ontario (Figure 2), and more accurately
138 captured urban areas than using political boundaries. The Michigan urban study area covered
139 2,390 km² with over 3.1 million people. It includes the Great Lakes Water Authority's Water
140 Resource Recovery Facility (GLWA WRRF), one of the largest wastewater treatment facilities
141 in the world, treating sewage from 3 million residents across 77 communities, as well as
142 stormwater from the region's combined sewer system. The Windsor and London areas cover 149
143 km² and 138 km², respectively, with populations of 211,000 and 366,000.

144 **Models** – The assessment was built on the construction and use of four models (Figure 2) that
145 collectively simulate the dynamics of this complex watershed.

- 146 • A nutrient mass balance model based on closed water budget and accounting for all
147 phosphorus inputs and outputs on a water-year annual basis between 1998 and 2016
148 (Scavia et al. 2019a), and an accounting of phosphorus sources from within the three
149 major urban areas (Hu et al. in review).
- 150 • A watershed model simulating flow and dynamics of water, nutrients, and sediment on
151 daily-to-annual time scales for 2001-2015, based on the Soil and Water Assessment Tool
152 (SWAT) (Dagnew et al. 2019a).
- 153 • A 3-dimensional (3D) coupled hydrodynamic and ecological model of Lake St.
154 Clair (ELCOM-CAEDYM) simulating thermo- and hydrodynamics, nutrient and algal
155 dynamics for 2009 and 2010 (Bocaniov and Scavia 2018).

156 • An urban model simulating the Great Lakes Water Authority (GLWA) sewer service area
157 based on the Storm Water Management Model (SWMM) (Hu et al. 2018).

158 **Project Guidance** - An advisory group was established at the project inception to help
159 understand policy contexts, and provide feedback on approach and resulting products. The group
160 included US and Canadian representatives from federal, state, and provincial governments;
161 regional conservation authorities; non-profits; universities; and local organizations actively
162 involved in watershed management, policy development, or research (Scavia et al. 2019b).
163 Through more than a dozen in-person meetings, periodic conference calls, and individual
164 consultations, the 30-person advisory group helped ensure the research would be credible
165 scientifically, and the results would be relevant and usable for the Great Lakes policy and
166 management communities. Preliminary interviews and ongoing feedback from the group helped
167 identify key areas of interest, potential concerns, and new data sets and related projects that
168 influenced the team’s approach, baseline assumptions, and specific scenario analyses for
169 modeling runs (Goodspeed et al. 2018). Although all members of the advisory group had
170 opportunities to comment on project results and research summaries, the content of this paper is
171 solely the responsibility of the project team.

172 **Mass balance estimates** - Scavia et al. (2019a) compiled and analyzed data from US and
173 Canadian water quality monitoring programs between 1998 and 2016 (Tables 1, 2, Figure S3),
174 and used the Weighted Regressions on Time, Discharge and Season (WRTDS) method (Hirsch
175 et al. 2010) to calculate tributary phosphorus loads based on concentrations and flow data for
176 gauged tributaries. Area-weighted estimates based on nearby streams were used for unmonitored
177 areas prior to adding upstream point sources (see Figure 9 below). Because WRTDS is not
178 appropriate for the connecting channel (St. Clair and Detroit rivers), their loads were estimated

179 by multiplying flow times concentrations. Atmospheric loads to Lake St. Clair were from
180 Maccoux et al. (2016), and loading from Lake St. Clair shoreline erosion was estimated by
181 multiplying the shoreline length by the annual P loading rate for the Lake St. Clair basin
182 (Monteith and Sonzogni 1976). Monthly industrial and municipal point source data were
183 collected from US EPA, the Great Lakes Water Authority, and the Ontario Ministry of
184 Environment and Climate Change databases (Scavia et al. 2019a). Urban runoff was calculated
185 based on precipitation and impervious area (Arnold et al. 2012).

186 **Lake St. Clair analysis** - Lake St. Clair's annual phosphorus retention estimates were based on
187 the TP and DRP mass balances (Scavia et al. 2019a) for water years 1998-2016. Whole-lake
188 estimates, as well as estimates at smaller spatial and temporal scales for 2009 and 2010 were
189 based on a three-dimensional ecological model (Bocaniov et al. in review). In both cases,
190 percent retention was calculated as the sum of all inputs minus outputs, divided by inputs. The
191 previously calibrated, validated, and applied ecological model (Bocaniov and Scavia 2018) was
192 the Computational Aquatic Ecosystem Dynamic Model (CAEDYM) driven by the 3D
193 hydrodynamic model (Estuary, Lake and Coastal Ocean Model: ELCOM). For this application,
194 the model simulates dynamics of phosphorus, nitrogen, silica, oxygen, carbon and total
195 suspended solids, and five functional groups of phytoplankton (Bocaniov et al. 2016, Bocaniov
196 and Scavia 2018). This model was also used to explore the relationship between major tributary
197 loads to the lake and loads leaving the lake.

198 **Watershed analysis** - The Soil and Water Assessment Tool (SWAT) was applied to the full
199 watershed to explore options for reducing TP and DRP loads (Dagnew et al. 2019a, b). The
200 watershed was divided into 800 subbasins, approximately 24 km², and each sub-basin was
201 further divided into Hydrologic Response Units (HRUs) corresponded to farm fields

202 (approximately 171 acres each), the first time this has been done for a watershed of this size.
203 Given the variability in agricultural management between the US and Canada, the advisory
204 group was engaged extensively over the course of the project to both verify and augment the
205 available data and to provide new data where appropriate (Scavia et al, 2019b). The model was
206 calibrated (2007-2015) and validated (2001-2006) to loads estimated from measurements at the
207 mouths of the six major tributaries (Figure 3) at daily, monthly, and annual time scales, and then
208 used to simulate loads from each of those tributaries. Simulation results were reported for each
209 of these major tributary watersheds, and neighbor watersheds with similar characteristics were
210 assumed to respond similarly (e.g., the Black for the Belle and Pine; the Thames for the Essex).
211 The model was then used to test the watershed’s sensitivity to seven practices. Reduced nutrient
212 application rates (Rate), subsurface placement of nutrients (PL), controlled drainage, and cover
213 crops (CC) practices were applied to all croplands. The wetlands (WT), filter strips (FS), and
214 grassed waterways practices were applied to all lands, including permeable urban areas. Based
215 on analysis of the individual practices and discussions with the advisory group, five bundles of
216 practices were selected, and each bundle was evaluated under three adoption strategies: (1)
217 applied to all appropriate land, (2) applied randomly to 55% of the appropriate land¹, and (3)
218 focused on the 55% of the land with high TP or DRP yields (Dagnew et al. 2019b). When
219 applied in bundles: WT assumed that 1% of every subbasin’s land area was converted to a
220 wetland and those wetlands were positioned such that 50% of the flow in a sub-basin passed
221 through them; PL placed 80% of nutrients sub-surface and 20% on the surface; FS assumed 1.7%
222 of a farm field was converted from crops to a filter strip/buffer strip; CC assumed cereal rye was

¹ Here, and throughout this paper, “appropriate lands” are lands where a practice can be implemented. For example, cover crops, subsurface placement, and fertilizer reduction can only be implemented in croplands while wetlands can be implemented for any land use type.

223 planted in the fall on fields growing corn and soybeans; and Rate assumed a 25% reduction in N
224 and P inputs to a farm field, including both inorganic fertilizers and manure.

225 **Urban analysis** – To examine the effects of green infrastructure across broad urban/suburban
226 areas, Dagneu et al. (2019b) used SWAT to test the effects of increasing pervious area with and
227 without additional vegetation in urban areas in the Clinton and Rouge watersheds (Figure 1) To
228 explore the potential for reducing combined sewer overflows (CSOs) in the GLWA WRRF
229 sewer service area (Figure 2), the calibrated Storm Water Management Model (SWMM), which
230 included 402 subcatchments with unique land cover, soil, gray infrastructure, and connectivity
231 (Hu et al. 2018) was used. The model was calibrated for volume at outfalls of 12 retention basis,
232 two wet weather outfalls at the WRRF, and inflows to the WRRF. To identify subcatchments
233 that contribute most to wet weather discharge at the WRRF as well as to the total system CSO
234 volume, rainfall was eliminated for one subcatchment at a time, and the resulting percent
235 reductions were calculated. This analysis is analogous to converting that catchment to a separate
236 stormwater system. The model was also used to simulate implementing two forms of green
237 infrastructure under average and extreme storms (Hu et al. in review).

238

239 **RESULTS**

240 **New estimates for the Detroit River load**

241 **Phosphorus from Lake Huron dominates the Detroit River load** - Burniston et al. (2018)
242 noted that the TP concentrations entering Lake St. Clair were considerably higher than those
243 leaving Lake Huron, especially for particulate P. Scavia et al. (2019a) found similar results, and
244 showed that the difference was not caused by additional phosphorus from the St. Clair River

245 watershed. They estimated that 54% of the Detroit River load originates in Lake Huron.
246 Satellite imagery revealed frequent large sediment resuspension events along Lake Huron's
247 southeastern shore that can persist for days and evade detection at the two monitoring stations.
248 While sampling at the Point Edward station could detect such events, it was shown to be not
249 frequent enough to catch many of them (Scavia et al. 2019a). This unmeasured load increased
250 over the study period from 2001-2016, in concert with climate-driven declines in ice cover and
251 increased frequency of large storms, approaching the sum of the measured loads from Lake
252 Huron and the St. Clair River watershed (Figure 4).

253 This updated estimate of the Lake Huron contribution does not impact the Scavia et al. (2019a)
254 or Burniston et al. (2018) estimates of the Detroit River load because they are based on
255 measurements at the outlet of Lake St. Clair and measurements in the Detroit River, respectively,
256 effectively capturing the full Lake Huron contribution. However, as discussed below, this
257 unmeasured load does impact our understanding of the relative importance of different nutrients
258 sources and therefore the potential allocation of load reduction targets.

259 After Lake Huron, the largest phosphorus contributors are nonpoint sources, followed by the
260 WRRF in Detroit and other point sources (Figure 5a). Average annual TP loads from the US
261 (798 MTA) are higher than those from Canada (601 MTA).

262 **Lake St. Clair is a TP sink** – On average between 2001 and 2015, Lake St. Clair retained 20%
263 of its TP inputs annually (Scavia et al. 2019a), *albeit* with substantial inter-annual variability
264 (Figure 5b). While measurements of DRP are less reliable, it appears that its annual retention is
265 much lower, perhaps approaching zero. Results from the ecological model (Bocaniov et al. in
266 review), indicated that, for the simulation period March through October, 17.3% of the TP was
267 retained and 34.8% of the DRP was retained. This seasonal TP retention rate is slightly lower

268 than the annual rate, likely because the model could only run for the ice-free season, and ice
269 cover would increase retention via reduced mixing and elevated settling rates during times when
270 ice-cover shields the lake surface from the wind stress. The model's high seasonal DRP
271 retention is driven by rapid uptake by algae during the growing season. To the extent that the
272 annual DRP retention rate is accurate, it suggests that much of the DRP retained during the
273 growing season is recycled back into the water and exported during the colder months.

274 Scavia et al. (2019a) suggested zebra and quagga mussels could have contributed to the
275 sequestration of phosphorus into the bottom sediment of Lake St. Clair. Nalepa et al. (1991)
276 estimated that the mussel-related TP retention between May and October represented about 8.6%
277 of the external TP load during the same period, but because the study was done prior to the zebra
278 and quagga invasion, they suggest that value is likely an underestimate. Lang et al. (1988)
279 estimated macrophyte growth to be roughly 7% of TP loads. So, together these could account
280 for much of the retention. However, Bocaniov et al. (in review) showed that wave-induced
281 bottom shear stress (the driver of sediment resuspension in shallow lakes) is not strong enough to
282 resuspend sediments in the 30% of the lake with depths greater than 5 m. So, deposition of
283 sediment in those areas is also a likely contributor to phosphorus retention. They also showed
284 that both TP and DRP retention rates are correlated negatively with average wind speeds,
285 suggesting that wind-dependent resuspension in the other 70% of the lake could explain the year-
286 to-year variability in the annual retention estimates (Figure 5b).

287 **Revised Detroit River loads** – As described above the new Lake Huron load estimate and Lake
288 St. Clair retention estimates are important, but they do not affect the updated Detroit River TP
289 load estimates because those are based on the load leaving Lake St. Clair. The new estimates
290 (Scavia et al. 2019a) (Figure 5c) are higher than those estimated by Maccoux et al. (2016) and

291 lower for two of the three years estimated by Burniston et al. (2018). The variations among these
292 estimates are likely because the Maccoux et al. used the earlier low estimate for the Lake Huron
293 load, and Burniston et al. used LOADEST (Runkel 2013), which may not be appropriate for
294 connecting channels. The Detroit River load declined roughly 37% from 1998 to 2016 due to
295 declines in Lake Huron phosphorus concentrations after the 2000-2005 invasion of zebra and
296 quagga mussels, and significant improvements in WRRF operations around 2010. There was no
297 statistically significant trend in other sources over this time period.

298 **Options for reducing loads**

299 **Meeting a 40% reduction for the Detroit River** – A 40% reduction from the updated 2008
300 Detroit River load estimate (3,096 MTA, Scavia et al. 2019a) results in a 1,858 MTA target. Our
301 estimates indicate the Detroit River TP load has already declined to 2,425 MTA (based on an
302 average for 2013-2016), so 567 MTA remains to be reduced (Figure 6). This is equivalent to
303 23% of the phosphorus load coming from all sources, including Lake Huron. .

304 After Lake Huron, the largest sources of phosphorus are the WRRF, followed by the Thames
305 River watershed, unmonitored loads to Lake St. Clair, and the Sydenham and Clinton river
306 watersheds (Figure 7). The remaining 10% comes from unmonitored load to the Detroit and St.
307 Clair rivers, and the Black, Rouge, Belle, and Pine river watersheds.

308 ***Contributions and potential reduction of point sources*** – Point sources contribute 43% of the
309 TP watershed load (that is, the load excluding the Lake Huron contribution) and 50% of the
310 watershed DRP load. When considering point source contributions, roughly 83% of the TP load
311 and 85% of the DRP point source loads come from the US (Figure 8), representing 15% and 25%
312 of the Detroit River's TP and DRP loads to Lake Erie.

313 Detroit's WRRF's TP load declined by 44.5% since 2009 (MDEQ 2016; Hu et al. in review), but
314 still currently contributes 54% of the total point source TP and DRP load. However, while
315 beyond the scope of this study, treatment processes and technologies will likely continue to
316 improve, and it could be possible for some of these advances to be implemented in the future.
317 While non-trivial in technological, human resource, and financial costs, improving treatment
318 operations could potentially have one of the biggest impacts on reducing the watershed's
319 phosphorus load. Treatment improvements at some of the other point source facilities could also
320 be possible. Beyond that, the focus in urban areas turns to CSOs and runoff, and they each
321 constitute only about 2% of the Detroit River's load to Lake Erie.

322 Because both CSOs and runoff are primarily driven by rainfall and the amount of impervious
323 surface (Dagnew et al. 2019b, Hu et al. in review), reducing phosphorus load from these sources
324 would likely require increasing pervious areas. SWAT analyses (Dagnew et al. 2019b) for the
325 Rouge and Clinton watersheds demonstrated that both TP and DRP loads are reduced as pervious
326 surfaces increase, and that because of increased evapotranspiration, the reductions were roughly
327 doubled if a transition from impervious to pervious cover included added vegetation. The
328 SWMM analyses (Hu et al. in review) suggested that within the WRRF sewer service area, green
329 infrastructure such as bioretention cells and increasing pervious areas could work well for some
330 upper reaches of the system, but more complex interventions are likely needed downstream.

331 ***Contribution and potential reduction of nonpoint sources*** – Nonpoint sources contribute 57%
332 and 50% of the TP and DRP loads from the watershed, respectively. Dagnew et al. (2019b)
333 estimated that 59% of the watershed's nonpoint source TP and 68% of the nonpoint source DRP
334 come from Canadian agricultural lands, compared to 12% and 6% from US agricultural lands.

335 Runoff from urban and suburban lands make up about 10% of the watershed's nonpoint source
336 TP and DRP loads. (Figure 9).

337 Estimated loss of nonpoint source DRP and TP per hectare (loss yields) from agricultural lands
338 showed that losses were generally higher in Canada than in the US, especially for DRP (Figure
339 10). While this difference may be due to higher fertilizer application rates and more intense
340 drain tile spacing in Ontario, running the SWAT model with the same fertilizer application rates
341 and tile systems in both the US and Canada produced essentially the same patterns in loss yields.
342 Thus, those differences are more likely driven by differences in precipitation and soil
343 characteristics (Figure S1). Those characteristics in Canada are more similar to the Maumee
344 River watershed, which delivers almost half of the phosphorus to the western basin. While the
345 slopes in both the US and Canadian agricultural areas are similar to the Maumee, average annual
346 precipitation in in the upper Sydenham and Thames is similar to that in the Maumee watershed
347 and greater than that in the St. Clair and Detroit River watersheds. Similarly, the Canadian soils
348 are largely poorly drained like those in the Maumee, whereas the US soils are well drained
349 (Figure S4). Edge of field analysis, by the Watershed Evaluation Group at the University of
350 Guelph, for a very small (19.5 km²) subbasin (Upper Medway watershed) within the Upper
351 Thames watershed indicated that average TP yield (2002-2016) at field level ranges from 0.25 -
352 5 kg/ha, averaging at 0.62kg/ha at the subbasin outlet (WEG 2018), which is similar to the
353 results of this study.

354 The highest single-practice TP and DRP load reductions were achieved with wetlands (WT),
355 followed by filter strips (FS), subsurface placement of nutrients (PL), cover crops (CC), and
356 reduced fertilizer application rates (Rate) (Dagnew et al. 2019b). The edge of field study for the
357 Medway watershed (WEG 2018) indicated that TP and DRP reduction by using wetlands, buffer

358 strips, and grassed waterways vary among fields. As a result, even with the extreme case of
359 100% adoption, none of the practices implemented alone achieved a 40% load reduction at their
360 sub-watersheds' outlets. Hence, the need for implementation of multiple practices seems
361 inevitable. In our analysis, the bundle of practices that included filter strips, wetlands, and cover
362 crops on 100% of the appropriate lands performed best, followed by one that included fertilizer
363 subsurface placement, wetlands, and cover crops (Figure 11). These bundles each reduced TP
364 and DRP loads from the agriculturally-dominated Sydenham, Thames, and Black river
365 watersheds by as much as 60-80%. Other combinations could potentially achieve at least a 40%
366 reduction from those watersheds (Dagnew et al. 2019b).

367 The CC-PL bundle performed almost as well as CC-PL-Rate bundle, suggesting that it may not
368 be necessary to reduce fertilizer application rates if cover crops and subsurface placement of
369 fertilizer are implemented. Adding filter strips to the CC-PL bundle further decreased the TP
370 and DRP loads from the Sydenham and Thames rivers, and it was particularly effective for
371 reducing the TP load from the Black watershed.

372 Dagnew et al. (2019b) also showed that placing the practices on just the 55% of the land with the
373 highest TP and DRP yields also surpassed target-level reductions. For example, a 55% focused
374 implementation of CC-FL-WT could achieve a 50% load reduction in the Sydenham sub-
375 watersheds for both TP and DRP (Figure 11, upper right). The Thames River may require
376 slightly more than 55% to reach the same reduction levels. It is important to note, however, that
377 while the model demonstrates the benefits of focusing practices on high phosphorus loss lands,
378 in practice those areas will have to be identified on the ground using farm- or field-level
379 management information (e.g., Muenich et al. 2017).

380 ***The Thames River*** - The binational agreement also calls for a 40% reduction in spring (March-
381 July) TP and DRP loads for, among other watersheds, the Thames River. So, we tested the
382 impacts of key bundled scenarios on the Thames River spring load and the Sydenham and Black
383 rivers for comparison. In testing the bundle most effective for annual TP reductions (CC-FS-
384 WT), one that replaced cover crops with subsurface placement (PL-FS-WT), and one that tested
385 fertilizer application rates and subsurface placement (Rate-PL), Scavia et al (2019b) showed that
386 in all cases, the spring load reductions equal or surpass the annual load reductions for those sub-
387 watersheds (Figure 12). Thus, practices selected to address annual TP loads would also be
388 effective for spring TP and DRP loads.

389 The Thames River is also of particular importance because changes in its load lead to more
390 substantial changes in the load leaving Lake St. Clair (Bocaniov et al. in review). That load,
391 along with re-suspended material, is transported along the shallower east and southeast shore
392 toward the lake's outflow. In addition, its load is largest in late winter, early spring, and late fall
393 when algal uptake is low and circulation favors shorter river water residence times (~11 days).
394 In contrast, the Sydenham is located further from the lake outlet and separated from it by a basin
395 deep enough (≥ 5 m) to support sediment accumulation. However, as Bocaniov et al. (in review)
396 pointed out, because the load to Lake St. Clair is dominated by the St. Clair River, even a 50%
397 decrease in any of its other tributaries would result in a less than 5% decrease in the load leaving
398 the lake.

399 **Climate change will likely make reaching targets more difficult** - Using the delta change
400 method based on six downscaled climate model results for the Maumee River Watershed, Scavia
401 et al. (2019b) used monthly average precipitation and temperature changes between the present
402 (1996-2015) and mid-century (2046-2065) to assess the potential impacts of climate change. All

403 but one climate model projected increases in annual precipitation, and all models projected an
404 increase in temperature. The 6-model average changes in annual precipitation and temperature
405 were +6.2% and +2.7°C, respectively.

406 Similar to other analyses for this region (Daloğlu, et al. 2012, Bosch et al. 2014, Verma et al
407 2015, Jarvie et al 2017) and most of the US (Sinha et al. 2017), increases in the timing and
408 intensity of spring precipitation led to increased runoff and loads. Also similar to recent analysis
409 for the Maumee watershed (Kalcic et al. 2019), increased temperature appears to mitigate some
410 of the spring runoff because reduced snowpack reduces the intensity of spring runoff and
411 increased evapotranspiration reduces the amount of water available to run off. Based on the
412 output from the six climate models, SWAT projected that, on average, higher precipitation alone
413 increased TP loads by 25% and DRP loads by 20%. Combining higher precipitation and
414 temperature increased TP loads by 9.3% and DRP loads by 7.2%.

415 **Discussion**

416 In February 2016, the US and Canada called for a 40% reduction from 2008 levels in annual TP
417 inputs to Lake Erie's western and central basins and spring TP and DRP from the Thames River
418 watershed. The fact that 54% of the TP load to Lake Erie originates in Lake Huron, even though
419 20% of the load is retained by Lake St. Clair, is a reminder that the Great Lakes are an
420 interconnected system, and that upstream nutrient sources are important to consider.

421 The current contribution to the Detroit River load from Lake Huron appears to be more than
422 twice the load estimated from measurements, and that unmeasured contribution has been
423 increasing due to climate change. This unmeasured contribution appears to come from sediment
424 resuspended along Lake Huron's southeast shore, and future efforts to reduce that load will
425 require additional analyses of its sources, phosphorus content, event frequency, and movement

426 toward the outflow to the St. Clair River. It should be possible, however, to at least improve load
427 estimates by including continuous measurement of phosphorus surrogates, such as turbidity, that
428 can be correlated with phosphorus concentrations (e.g., Robertson et al. 2018).

429 Taking into consideration the potential difficulty in controlling the Lake Huron load illustrates
430 the challenge of meeting a 40% load reduction from the Detroit River, even though that load
431 already declined by almost 22% since 2008. A modest 23% reduction of all loads would be
432 needed to achieve the remaining 567 MTA reduction required to meet the target; however, if
433 reductions from Lake Huron are not included, then a 51% reduction would be required from
434 watershed sources. If further reductions from the GLWA WRRF are also not included because it
435 has already been reduced by over 40%, then a 72% load reduction would need to be achieved
436 from the remaining sources - a daunting challenge. However, reducing the Lake Huron and
437 GLWA WRRF loads each by 10-15%, leaves 40-50% to be reduced from watershed sources,
438 which simulations show are possible.

439 Because point sources contribute 43% of the watershed's TP and 50% of the DRP (not including
440 the Lake Huron contribution), they are logical targets. The WRRF in Detroit contributes 54% of
441 the TP and DRP point source load in this watershed; however, substantial load reductions have
442 already been made from this facility, and the high costs of further technological improvement
443 may therefore be difficult to justify at this time. There are about 150 other point sources in the
444 watershed that together contribute 46% of TP and DRP point source load, so additional
445 reductions at those facilities should help. Because CSOs and urban runoff contribute little to the
446 overall load, reductions from them would contribute little. However, to address other public
447 health and environmental concerns, CSO reduction is generally a good practice and could be
448 achieved through a portfolio of complementary green and gray infrastructure strategies.

449 Nonpoint sources contribute the remaining 57% and 50% of the TP and DRP loads and, similar
450 to results from Maumee River watershed assessments (Muenich et al. 2016, Kalcic et al. 2016;
451 Scavia et al. 2017), bundling agricultural management practices appears to work better than
452 implementing single practices. Combining practices, such as cover crops, filter strips, wetlands,
453 and subsurface placement of fertilizer, resulted in TP reductions greater than 50%. Bundled
454 scenarios designed to address the annual TP load reductions for the Detroit River were even
455 more effective for reducing the spring TP and DRP loads for the Thames, Sydenham, and Black
456 rivers. As in the Maumee analyses, focusing practices on land with the highest phosphorus
457 losses resulted in reductions that approach levels achieved from applying them on all agricultural
458 lands. This focused approach, coupled with the relative effectiveness of different combinations
459 of practices, suggests flexibility, where practices can be combined and applied to match the
460 needs and preferences of producers. However, the simulations suggest that even extensive
461 conservation on those lands may not be enough if the strategy is to get a 72% reduction from
462 those lands alone, especially because the future climate is projected to increase loads.

463 It is also important to recognize that increased air temperature favors longer periods of lake
464 stratification leading to an earlier and longer algae growing season, as well as increased organic
465 matter that promotes more hypoxic waters. For example, Rucinski et al. (2016) showed that
466 variation in meteorology (driving lake thermal stratification) explained almost nine times as
467 much interannual variability in hypoxic area compared to variation in phosphorus loading, and
468 that deeper stratification caused by warmer, longer summers led to larger hypoxic areas.

469 Bocaniov and Scavia (2016) also showed that inter-annual differences in weather significantly
470 influenced the spatial extent, duration and severity of anoxia and hypoxia. To advance scientific
471 progress and better inform management, the interactions between climate and land management,

472 as well as between climate and the lake, must be better evaluated to assess future changes in both
473 the watershed and Lake Erie.

474 **Domestic Action Plans Adaptive Management** - To understand and assess the relative sources
475 of and potential actions to reduce loads to Lake Erie from the Detroit River required assembling
476 large data sets from both the US and Canada; developing, calibrating, and validating diverse
477 models at different time and space scales; and using both data and models to explore potential
478 management options. This effort, coupled with similar ones developed for the Huron River (e.g.,
479 Xu et al. 2017), the River Raisin (e.g., Muenich et al. 2017), and the Maumee River (Muenich et
480 al. 2016, Kalcic et al. 2016; Scavia et al. 2017), provide tools that can be used to guide policies
481 and practices as the countries work within the GLWQA adaptive management framework. As
482 new information becomes available, that framework enables both adjustments to action plans and
483 improvements in models and other assessment tools.

484 Each Domestic Action Plan emphasizes that the targets and approaches are not static. For
485 systems this complex and dynamic, it is critical to set targets, take action, monitor the results,
486 and make adjustments as necessary. Much of what has been compiled, analyzed, and assessed
487 herein is new since the targets were set and the action plans developed. Therefore, we anticipate
488 our results will be helpful in evaluating both the overall load reduction targets and their
489 allocation.

490 Potential plan adaptations could include 1) enhancing conservation to reach a 72% reduction
491 from nonpoint sources, 2) designing programs to reduce the Lake Huron and WRRF loads each
492 by 10-15% so that the nonpoint source load reduction is more within reach, 3) relax the
493 expectation of a 40% load reduction from the Detroit River and make up the difference from

494 other watersheds, or 4) relax the overall 40% load reduction target for the western and central
495 basins and accept more hypoxia. Of course, combinations of the above could also be effective.

496

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660

661 Table 1. Total phosphorus load estimates (MTA) from monitored stations. Note minor
662 differences between this table and the one in Scavia et al. 2019a are due to updates in
663 original sources.

TP (MTA)	From Lake Huron	Into the St. Clair River				Into Lake St. Clair						Into the Detroit River				Lake Erie Inflow
		Black	Belle	Pine	Other	St. Clair River	Clinton	Sydenham	Thames	Other	Atmos + Erosion	Lake St. Clair Outlet	Rouge	GLWAP	Other	
1998	2261	114	35	24	59	2493	202	230	541	129	85	3062	47	727	120	2871
1999	2096	22	9	3	38	2168	144	103	218	60	72	2722	36	727	98	2522
2000	1944	38	17	9	62	2071	192	285	649	111	91	2531	56	545	110	2340
2001	2599	87	34	24	50	2548	167	155	507	126	75	2447	37	545	109	2294
2002	1959	120	35	24	57	2502	156	206	502	145	85	2590	51	636	120	2495
2003	1386	22	9	2	38	2115	90	104	247	59	92	2380	26	588	94	2369
2004	1333	137	53	39	73	1954	174	306	603	170	73	2476	47	632	118	2562
2005	1405	91	32	22	58	1785	126	218	434	106	64	2467	35	618	123	2645
2006	1331	110	37	26	61	1622	145	198	415	103	81	2388	46	634	123	2652
2007	1162	87	33	23	63	1529	146	246	463	104	68	2304	53	630	129	2659
2008	1083	82	31	21	58	1547	145	210	394	104	72	2246	51	672	128	2703
2009	1137	230	89	68	85	1767	254	312	550	161	82	2403	56	599	128	2989
2010	1076	40	19	10	37	1900	115	98	149	45	67	2272	42	600	98	2838
2011	954	143	54	40	61	1914	189	231	390	136	77	2099	68	472	122	2557
2012	980	42	17	9	43	2034	127	168	263	74	61	2054	43	368	106	2365
2013	914	160	47	34	52	1936	142	165	350	104	76	1890	39	323	96	2031
2014	890	58	22	14	49	1965	144	175	425	115	76	1911	52	313	113	1907
2015	966	46	17	9	41	2072	117	115	241	73	76	1991	39	336	92	1823
2016	1152	79	27	17	40	2307	111	89	265	78	76	2035	36	331	100	2479

664

665

666 Table 2. Dissolved reactive phosphorus load estimates (MTA) from monitored stations.

Water Year	DRP (MTA)	From Lake Huron	Into the St. Clair River				Into Lake St. Clair						Into the Detroit River				Lake Erie Inflow
			Black	Belle	Pine	Other	St. Clair River	Clinton	Sydenham	Thames	Other	Atmos + Erosion	Lake St. Clair Outlet	Rouge	GLWAP	Other	
1998			42	13	8	24		30	78	91	24	38		9		46	
1999			12	5	2	18		24	37	31	7	38		7		42	
2000			12	5	2	22		32	74	280	36	38		13		50	
2001			32	13	8	22		29	61	111	35	38		8		46	
2002			32	10	6	27		29	108	126	41	38		12		49	
2003			12	5	1	15		24	18	73	19	38		7		44	
2004			39	15	10	25		30	81	186	53	38		29		56	
2005			29	11	6	25		25	87	122	29	38		9	337	54	
2006			38	13	8	27		27	84	119	27	38		13	360	55	
2007			24	9	5	34		30	152	153	28	38		16	403	58	
2008	488		25	10	5	26	454	31	91	130	28	36	765	17	451	59	837
2009	529		68	26	19	41	513	50	174	221	45	41	836	18	369	61	1071
2010	535		13	6	2	16	502	29	30	48	10	34	810	17	262	47	1127
2011	497		46	17	12	26	494	42	91	154	44	38	769	28	242	58	1067
2012	485		12	5	1	24	505	37	113	114	25	30	777	18	228	53	996
2013	423		52	16	10	24	449	37	76	140	35	38	742	19	165	48	858
2014	401		19	7	4	22	455	40	74	180	46	38	780	29	157	58	832
2015	400		15	6	2	19	523	38	46	103	28	38	847	23	158	47	845
2016			30	11	6	19		38	41	121	32	38		22	178	52	

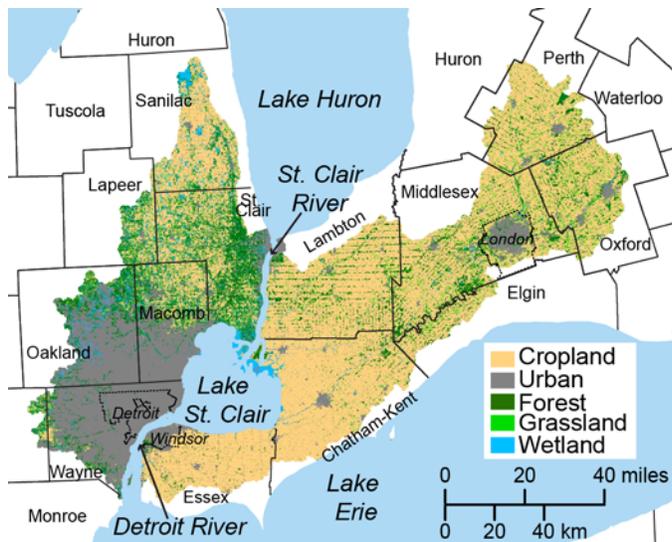
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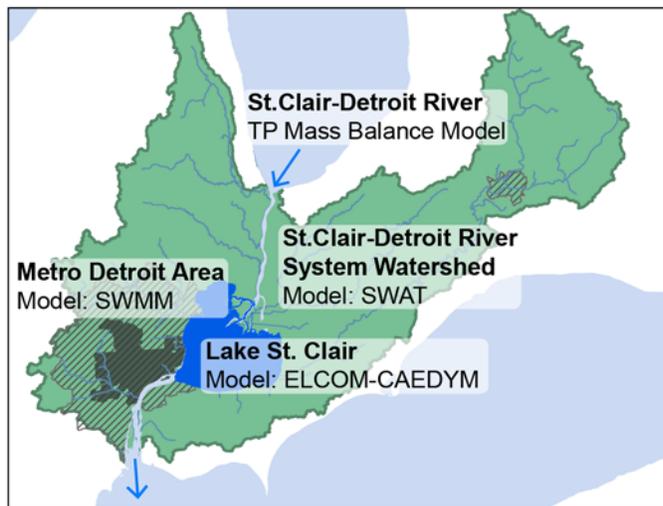
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671 Figure Legends



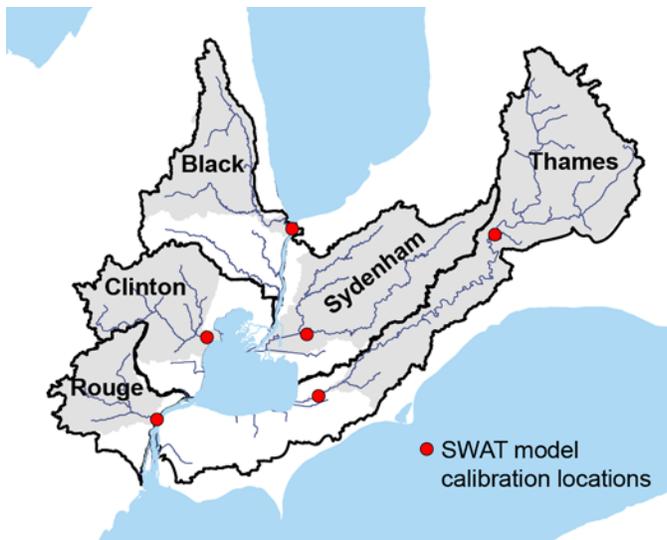
672

673 Figure 1. Land use in the St. Clair-Detroit River System watershed. The watershed is composed
674 of about 49% cropland, 21% urban land, 13% forest, 7% grassland, 7% surface water (including
675 Lake St. Clair), and 3% wetlands.



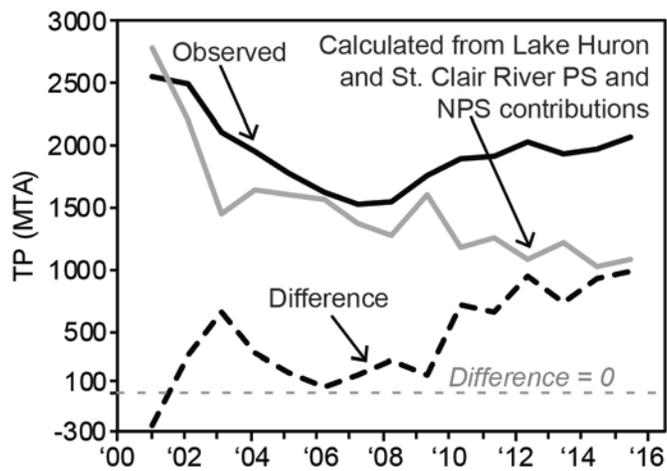
676

677 Figure 2. The four models used in this study. Areas with diagonal lines are the study areas for
678 the analysis of urban sources.



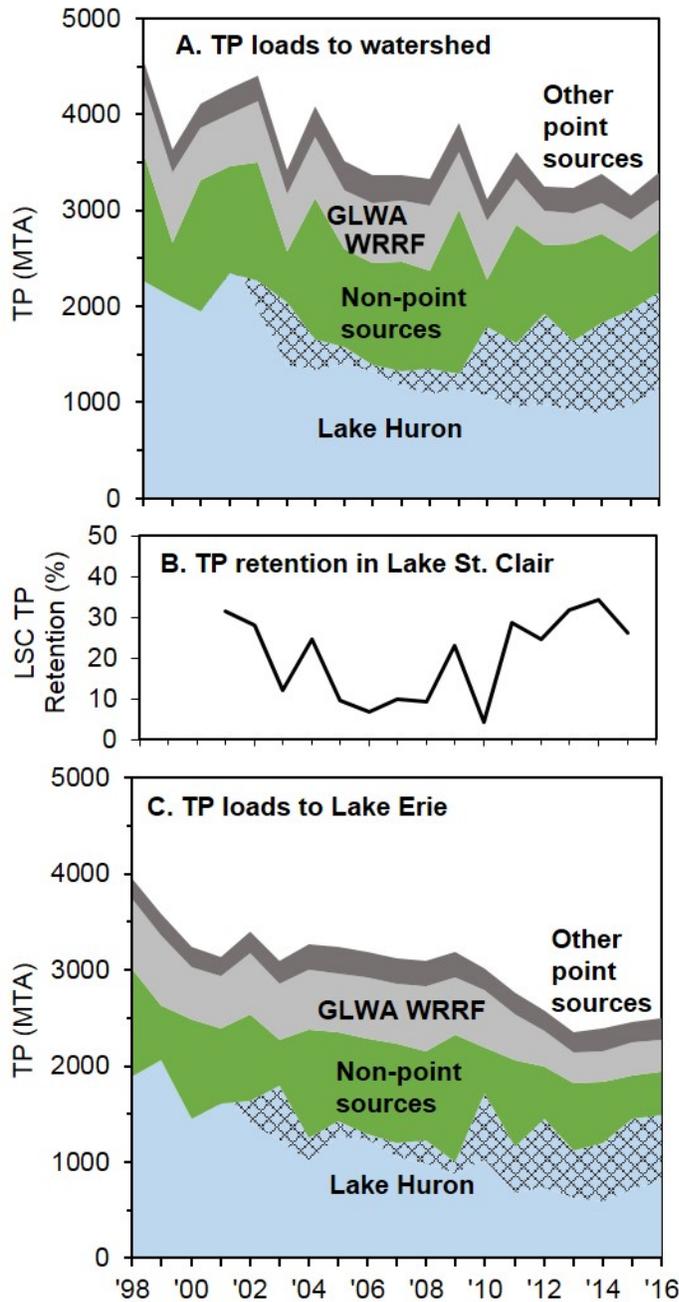
679

680 Figure 3. SWAT model calibration locations. Areas shaded gray and labeled with bold text
 681 represent the calibrated river watersheds. Calibration and scenario results for those watersheds
 682 are assumed to be representative of adjacent areas (not shaded) within the bold black lines.



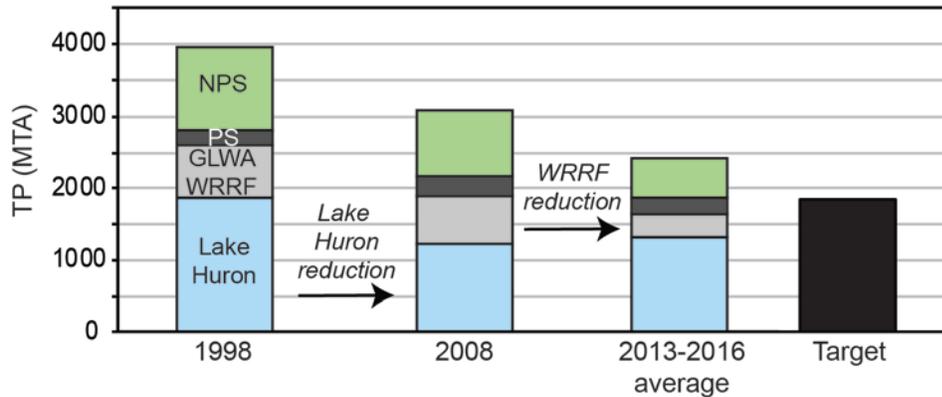
683

684 Figure 4. TP inputs to Lake St. Clair measured at Algonac and Port Lambton (black line), and
 685 calculated from Lake Huron and the St. Clair River point and nonpoint source contributions
 686 (gray line). The difference (dashed line) represents the portion of the load that is entering Lake
 687 St. Clair but not accounted for in monitoring data.



688

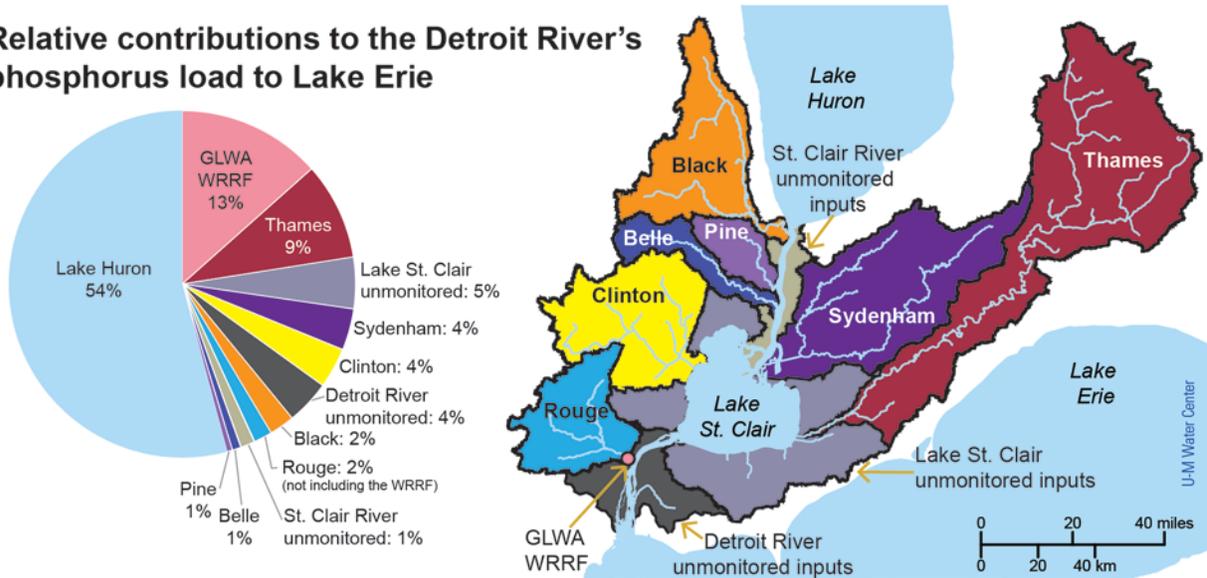
689 Figure 5. A: Time series of the TP load components from the watershed (not accounting for Lake
 690 St. Clair retention). Hatched lines represent the unmeasured load from Lake Huron. B: Percent
 691 Lake St. Clair TP retention. C: TP loads to Lake Erie derived from the sum of the load from
 692 Lake St. Clair and other loads to the Detroit River.



693

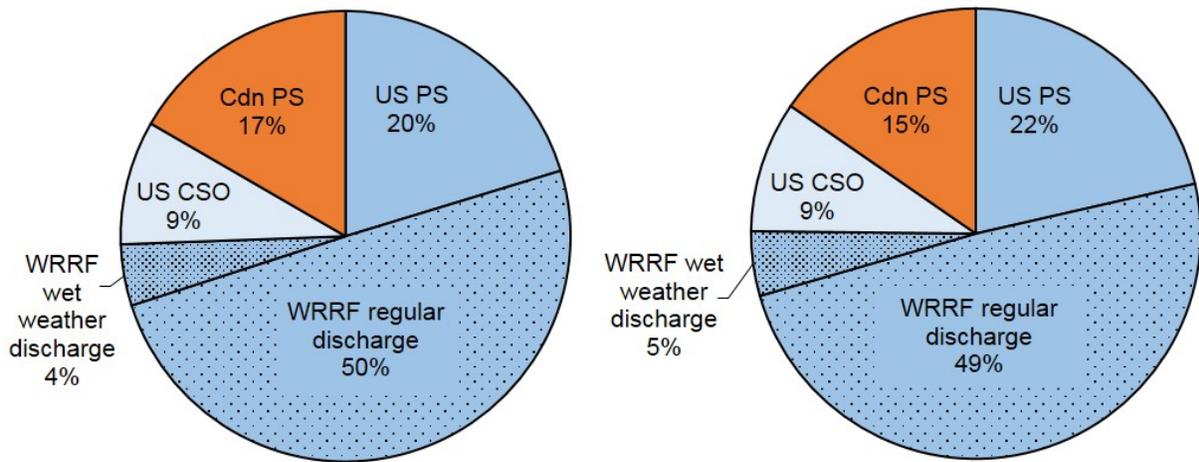
694 Figure 6. Contributions from non-point sources (NPS), point sources (PS), the Great
 695 Lakes Water Authority WRRF, and Lake Huron to the Detroit River TP load to Lake Erie at several
 696 time periods, accounting for Lake St. Clair retention. The target represents a 40% reduction from
 697 the 2008 load.

Relative contributions to the Detroit River's phosphorus load to Lake Erie



698

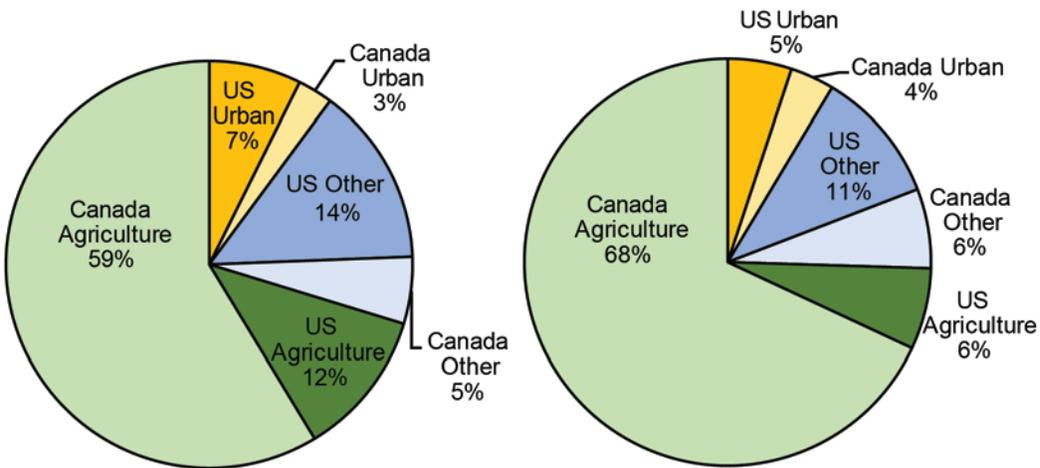
699 Figure 7. Proportions of the Detroit River's TP load to Lake Erie from all sources. The Great
 700 Lakes Water Authority Water Resources Recovery Facility (GLWA WRRF) in Detroit is shown
 701 separately from the Rouge watershed in this case.



702

703 Figure 8. Proportions of the watershed TP (left) and DRP (right) loads from US and Canadian

704 point sources. The load from Lake Huron is not included here.



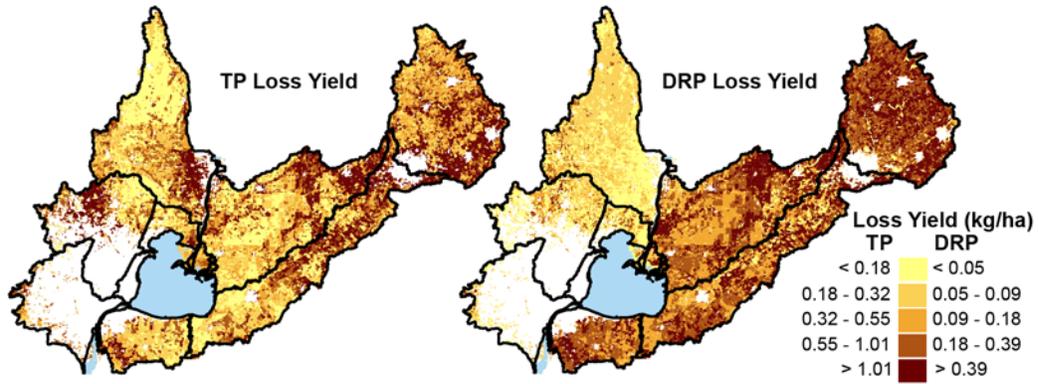
705

706 Figure 9. Proportions of the watershed TP (left) and DRP (right) loads from US and Canadian

707 non-point sources (NPS) coming from agricultural land (i.e., cropland and pastureland), urban

708 land, and other land (i.e., forests and wetlands) derived from SWAT. The load from Lake Huron

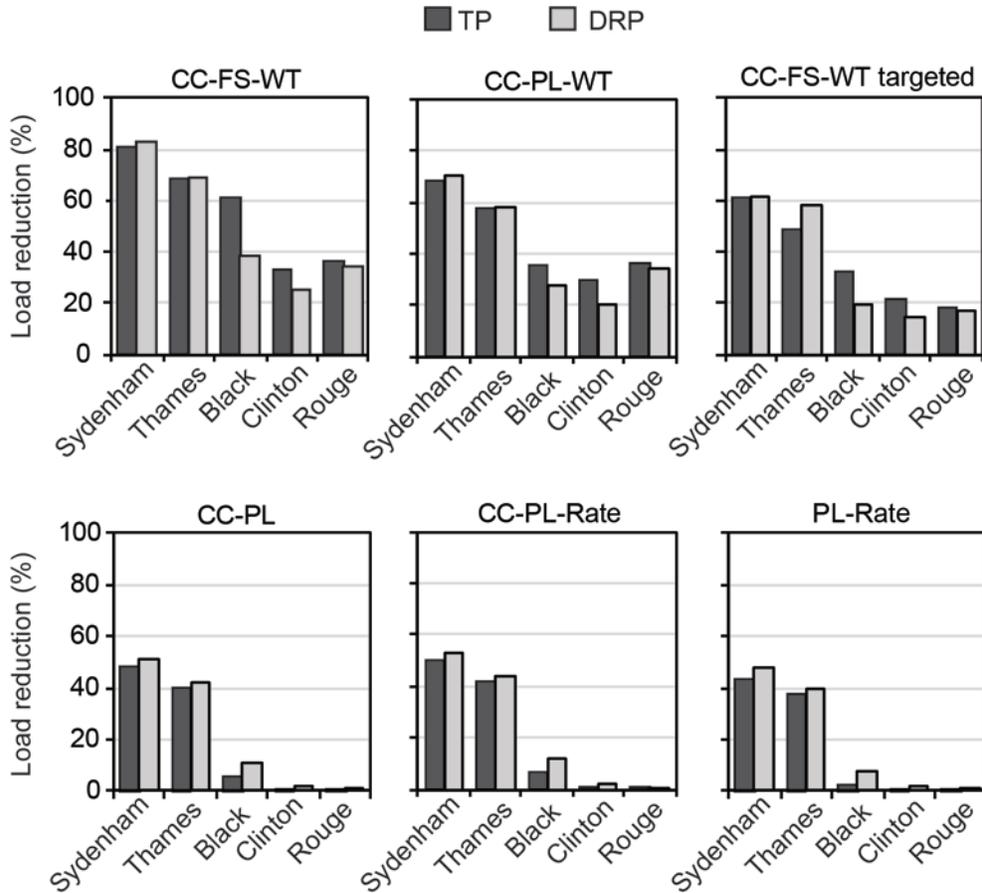
709 is not included here



710

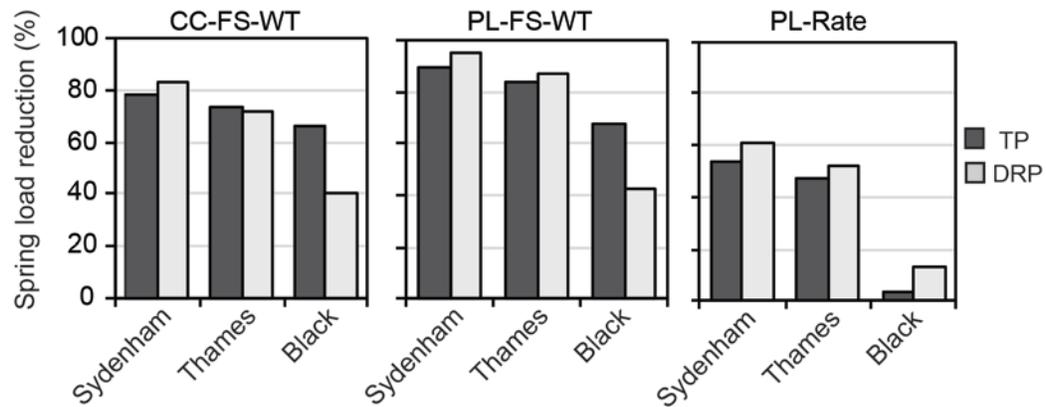
711 Figure 10. Modeled TP and DRP loss yields (kg/ha) for each SWAT model unit (HRU). Data
 712 from urban areas (shown in white) are not included so comparisons can be made across
 713 agricultural lands only.

714



715

716 Figure 11. Percent reductions of TP and DRP for bundled scenarios. Each bundle assumes 100%
 717 implementation, except the “targeted” scenario, which places practices on the 55% of land with
 718 the highest DRP and TP yields. For bundles that altered fertilizer rates, we assumed a 25%
 719 reduction in fertilizer application rates.



720

721 Figure 12. Percent spring (March-July) TP (black) and DRP (gray) load reductions for three

722 bundled scenarios. Each bundle assumes 100% implementation.