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3
4 **Quantitative biomonitoring in the Detroit River using *Elliptio complanata*: Verification of steady state**
5 **correction factors and temporal trends of PCBs in water between 1998-2015.**

6
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13
14 **Abstract**

15 Quantitative biomonitoring methods were applied to determine PCB concentrations in water from the
16 Detroit River over a 17 year period. During 2014, mussels were deployed for and extended duration (21-
17 364 d) and time dependent PCB concentrations were fit to a bioaccumulation model to estimate
18 elimination coefficients (k_{tot}) and provide site specific calibration of mussel toxicokinetics. The site
19 specific calibration and different k_{tot} versus K_{OW} relationships from the literature were used to correct for
20 steady state. \sum PCB concentrations in water were not significantly dependent on the k_{tot} values used
21 indicating that individual variation exceeds error contributed by steady state correction factors. The
22 model was then applied to estimate \sum PCB concentrations in water using the long term (1998-2015)
23 data. \sum PCBs concentrations in water exhibited a significant decreasing trend with a half life of 9.12 years
24 resulting in a drop in yearly geometric mean residues from 198.1 pg/L to 43.6 pg/L.

25 **Keywords:** Biomonitoring, toxicokinetics, polychlorinated biphenyls, persistent organic pollutants, water
26 concentrations

27
28 **Introduction**

29 Mussels are considered ideal biomonitors of bioaccumulative pollutants in water as a
30 result of their sessile nature, slow growth, low metabolic biotransformation capabilities and low
31 trophic position (O'Connor, 2002). Mussel biomonitors have been widely used to monitor
32 polychlorinated biphenyl (PCB) and other organic contaminants in the Huron-Erie Corridor of
33 the Laurentian Great Lakes since the 1980's (Kauss and Hamdy, 1985). A biomonitoring
34 program that continues to operate is the long running City of Windsor Mussel Biomonitoring
35 Program which commenced in 1996 (Drouillard et al., 2013). The program was developed to
36 perform upstream and downstream sampling across the city's two wastewater treatment plant
37 plumes in order to determine if the city's wastewater utilities and storm water overflows were
38 contributing to elevated loads of persistent organic pollutants (POPs) to the Detroit River. The
39 program uses a well calibrated freshwater mussel species, (*Elliptio complanata*), which are
40 collected from a reference location and caged at each of the program's fixed monitoring sites
41 over a set of defined deployment periods in each year. The study design makes the data
42 amenable to quantitative biomonitoring approaches and as a result has made the program one of
43 the most valuable data sets available for monitoring temporal trends of PCBs in water of the
44 Detroit River.

45 Quantitative biomonitoring requires the interpretation of biomonitoring data using
46 bioaccumulation models with the goal of converting accumulated chemical residues in the
47 biomonitor to a time integrated estimate of water concentrations for the target contaminants
48 (Gewurtz et al., 2003; Drouillard et al., 2013). This conversion requires three operational steps:
49 1) perform control correction of reference site accumulated residues; 2) convert time-dependent
50 chemical concentrations in the biomonitor to a steady state residue and 3) translate steady state
51 biomonitor residues into a water concentration estimate. Steps 1 and 2 rely on accurate
52 information about chemical and animal specific toxicokinetics, most notably the whole body
53 elimination rate coefficient (k_{tot}) for the biomonitor species being used. Step 3 requires
54 knowledge of the steady state animal/water bioconcentration factor (BCF). Past studies have
55 been performed using *E. complanata* to characterize organic contaminant k_{tot} values under
56 laboratory (Russell and Gobas, 1989; Gewurtz et al., 2002; O'Rourke et al., 2004; Drouillard et
57 al., 2007) and field conditions (O'Rourke et al., 2004; Raeside et al., 2009). Raeside et al. (2009)
58 performed *in situ* calibration of mussels deployed throughout the Huron-Erie corridor by pre-
59 dosing caged animals with non-environmental PCBs and tracking the rate at which dosed PCB
60 were lost from caged animals at each deployment site. The latter study demonstrated that field
61 deployed mussels eliminated PCBs faster compared to laboratory held mussels as well as
62 considerable variation in chemical toxicokinetics across different deployment locations
63 indicating that mussel filtration rates were dependent on site specific conditions. Error in the
64 magnitude of k_{tot} used in steps (1) and (2) of the quantitative biomonitoring model process leads
65 to incorrect water concentration estimates. It is therefore critical to understand what level of error
66 is introduced as a result of steady state correction factors and how to minimize this error when
67 interpreting biomonitoring extrapolated water concentrations.

68 During the 2014 implementation of the City of Windsor's Mussel Biomonitoring
69 Program, the mussel deployment was extended from the normal 182 deployment period to a 364
70 d deployment period. This provided a unique opportunity to apply non-linear regression methods
71 to fit the time-dependent mussel data to a generalized bioaccumulation model allowing an
72 alternative estimate of *in situ* k_{tot} values for individual PCB congeners and to contrast this against
73 previous calibration studies. Different k_{tot} values from across the various calibration studies were
74 then applied to estimate PCB concentrations in water (C_w) to determine the level of error (or
75 bias) that may occur when using non-site specific k_{tot} estimates derived from the literature within
76 the model. Finally, the fitted *in-situ* k_{tot} values were extended to the full 17 year biomonitoring
77 data base of mussel data from this field location to provide temporal trends of PCB
78 concentrations in water of the Detroit River over the 1998 to 2015 study period.

79 **Methods**

80 The Detroit River is designated as an International Joint Commission Great Lakes Area
81 of Concern owing to assessed degraded status of a series of beneficial uses, many of which are
82 tied to organic and metal pollutants present in sediments and water (Drouillard et al., 2006). The
83 City of Windsor Mussel Biomonitor Program routinely cages mussels at 9 locations that include
84 effluent holding tanks of one of its wastewater treatment plants, two smaller tributaries of the
85 Detroit River (Little River and Turkey Creek) and four locations within the Canadian waters of
86 the Detroit River. For purposes of this study, the focus was on the Riverside Marina location
87 (42°20'27.93"N; 82°55'56.04"W) given this was the site where a 364 d deployment was

88 collected. This station is located in the upstream portion of the Detroit River just downstream of
89 the mixing zone between Little River and the Detroit River. The site is likely jointly influenced
90 by both the Little River and Detroit River inflow from Lake St. Clair. Mussels are collected each
91 year from Balsam Lake, Lindsey, Ontario (typically in late April to early May) and retained at a
92 local aquaculture farm until deployment. Mussels are placed in wire minnow traps, with the
93 ends crushed and suspended in the water column 1 m below the surface. Triplicate mussels are
94 typically collected from each location after 21, 63, 126 and 182 days of deployment and stored
95 frozen until chemical analysis. The exception was for 2014 where the 182 d time point became a
96 364 d time point.

97 Shucked mussels were analyzed for PCBs, a suite of 18 organochlorine pesticides and
98 PAHs according to the procedures described by Lazar et al. (1992). PCBs, the focus of this
99 study, were determined by gas chromatography electron capture detector following liquid-liquid
100 extraction, florisil clean-up and instrumental conditions as specified in Lazar et al. (1992). The
101 commonly detected PCB congeners were IUPAC #'s (28/31, 52, 49, 44, 74, 66/95, 101, 99, 87,
102 110, 151, 149, 118, 153, 105, 138, 158, 182/187, 183, 171, 180, 170/190, 201, 195, 194 and
103 206). Lipid content was determined gravimetrically as was moisture content following drying at
104 110°C for 12 h. Instrument detection limits for individual PCBs were determined using a signal
105 to noise ratio of 10 for the equivalent peak width of the baseline signal determined for 7 blank
106 samples. Method detection limits were verified by spiking a set of PCB congeners into synthetic
107 triolein at the Instrument Detection limit for 5 samples and verifying that the spiked
108 concentration was not significantly different from the spiked value. Detection limits ranged from
109 0.01 to 0.06 ng/g wet weight for individual PCBs. Quality control measures include running a
110 method blank and in-house reference homogenate (Detroit River Carp) along with each batch of
111 six samples extracted. All reference homogenates PCB concentrations must be in compliance
112 within the control chart values used by the quality control officer before release of the data,
113 otherwise corrective actions are performed and the batch is re-run for analysis until QA
114 compliance is achieved. Each sample was also spiked with a recovery standard to determine
115 chemical recovery. The recovery standard changed over the years of the program, originally it
116 was 1,3,5-tribromobenzene and was later switched to PCB 34. The mean±standard error of
117 recoveries for the data set at Riverside Marina was 88.9±0.07%. Data were not recovery
118 corrected.

119 For the 2014 data (which included the 364 d deployment), non-linear regression was
120 performed to fit control corrected PCB concentrations in mussels according to the following
121 equation:

$$122 \quad C_{mcc(t)} = BCF \cdot C_W \cdot (1 - e^{-k_{tot} \cdot t}) \quad (1)$$

123 where $C_{mcc(t)}$ is the control corrected time dependent mussel concentration (pg/kg lipid
124 equivalent), BCF is the bioconcentration factor associated with animal lipids (kg/L) for a given
125 PCB congener, C_W is water concentration (pg/L), k_{tot} is the whole body elimination rate
126 coefficient (d^{-1}) for a given PCB congener and t is the deployment time (d). Lipid equivalent
127 concentrations are generated by establishing lean dry weight as having a partition capacity of
128 0.05 of lipids and adding this to the fraction of lipid in the sample (Drouillard et al. 2013).
129 Control correction was performed according to:

$$130 \quad C_{mcc(t)} = C_{m(t)} - C_{m(o)} \cdot e^{-k_{tot} \cdot t} \quad (2)$$

131 Where $C_{m(t)}$ is the time dependent mussel concentration (pg/kg lipid equivalent) and $C_{m(o)}$ is the
 132 concentration in the non-deployed control mussels (pg/kg lipid equivalent). BCF is set to the n -
 133 octanol/water partition coefficient value of each PCB congener as reported by Hawker and
 134 Connell (1988). Non-linear regression solutions (and 95% confidence intervals) were generated
 135 for C_w and k_{tot} by least squares procedures fit to Eq. 2 using Systat 13 Statistical Software after
 136 providing estimates of initial starting C_w and k_{tot} values used in the fitting procedure. Several
 137 starting estimates were generated to ensure the best model solution was achieved. Initially, k_{tot}
 138 values were assigned based on the Raeside et al. (2009) expression (see below) to establish
 139 control correction and solve Eq. 1. Once site specific k_{tot} values were generated from Eq. 1, the
 140 process was repeated using the newly solved k_{tot} in Eq. 2 and then the data were re-fit to Eq. 1.
 141 Only congeners where the model fit yielded an R^2 greater than 0.25 were accepted as an
 142 adequate model fit. Linear regression analysis was subsequently performed on $\log k_{tot}$ values
 143 against $\log K_{OW}$ to generate a predictive relationship for site specific in situ k_{tot} values. Non-
 144 linear regression model estimates of C_w were not interpreted but rather estimated by the
 145 procedure below.

146 For the 2014 and the larger temporal data base consisting of 17 years of data generated at
 147 Riverside Marina (inclusive of 2014), the below model was used to extrapolate water
 148 concentration estimates for each mussel analyzed. The model is given by:

$$149 \quad C_w = \frac{C_{m(t)} - C_{m(o)} \cdot e^{-k_{tot} \cdot t}}{(1 - e^{-k_{tot} \cdot t})} \cdot \frac{1}{BCF} \quad (3)$$

150 The k_{tot} values used with Eq. 3 were derived from 3 different predictive relationships. The first
 151 was based on the site-specific in situ k_{tot} relationship generated for k_{tot} values fitted against
 152 chemical hydrophobicity as generated using the approach described in Eq. 1. The second was
 153 based on the k_{tot} relationship for PCBs measured in laboratory held *E. complanata* (O'Rourke et
 154 al., 2004):

$$155 \quad k_{tot} = -0.59 \cdot K_{OW} + 2.05 \quad (4)$$

156 The third was based on the combined in-situ relationship for *E. complanata* deployed in the
 157 Huron-Erie corridor reported by Raeside et al. (2009):

$$158 \quad k_{tot} = -0.34 \cdot K_{OW} + 1.13 \quad (5)$$

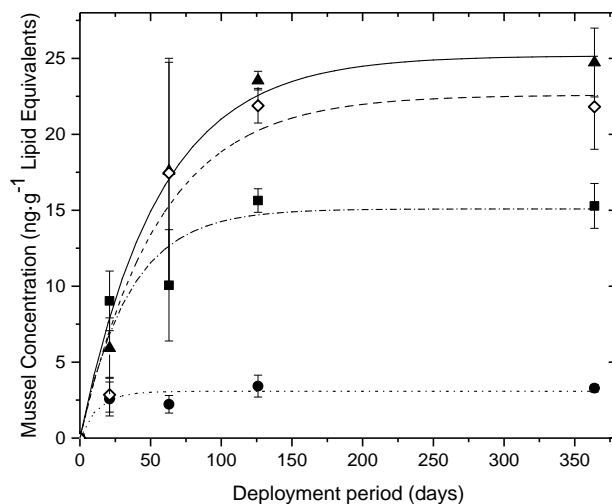
159 C_w was estimated for each individual PCB congener detected in the database. Sum PCBs
 160 ($\sum PCB$) are designated as the sum of C_w concentrations across the detected congeners identified
 161 in methods. Non-detected values were censored and not included in the $\sum PCB$ estimate. Linear
 162 regression analysis was performed on $\ln C_w$ ($\sum PCB$ s) versus Julian date to examine for temporal
 163 trends at the biomonitoring station. Analysis of variance (ANVOA) was used to determine if the
 164 above slope was significantly different from zero. The water concentration half life was
 165 estimated from the slope of the above relationship, where $t_{1/2} = \ln(2)/\text{slope}$. Analysis of
 166 covariance (ANCOVA) was performed to determine if the slope of sum PCB concentration with
 167 time differed when temporal trends were determined using the full data set or when the data were
 168 truncated to include only the longest deployment time in each year. Two way ANOVA was
 169 performed on log-transformed $\sum PCB$ C_w data to test for difference in C_w estimates with
 170 deployment time for the 2014 data set and also to determine if significant differences in C_w
 171 occurred depending on which k_{tot} versus K_{OW} relationship was used within the bioaccumulation
 172 model.

173 **Results and Discussion**

174 Satisfactory fit to Eq. 1 was obtained for 21 of 26 congeners measured using the 2014
 175 dataset where the R^2 of model fit to the empirical mussel data exceeded the criteria ($R^2 > 0.25$)
 176 used to specify adequate model prediction. The R^2 values of successfully fitted congeners
 177 ranged from 0.27 to 0.76. Congeners excluded due to lack of satisfactory model fit ($R^2 < 0.25$)
 178 included PCB 31/28, 52, 66/95, 158 and 183. The bioaccumulation pattern and model fit for four
 179 selected PCB congeners (PCBs 49, 110, 138 and 180) to the 2014 data are presented in Figure 1.
 180 Goodness of fit tests (observed vs predicted linear regression relationship) were significant for
 181 PCB 110 ($p < 0.05$) and highly significant for PCBs 138 and 180 ($p < 0.01$). The goodness of fit
 182 test was not significant for PCB 49 given that range of concentrations was low across time points
 183 owing to rapid achievement of steady state for low K_{OW} compounds. Most congeners exhibited
 184 linear uptake over the first 30 days and approached steady state by 122 days. Model estimated
 185 k_{tot} values for individual PCBs demonstrated a highly significant decreasing trend with
 186 increasing hydrophobicity described by:

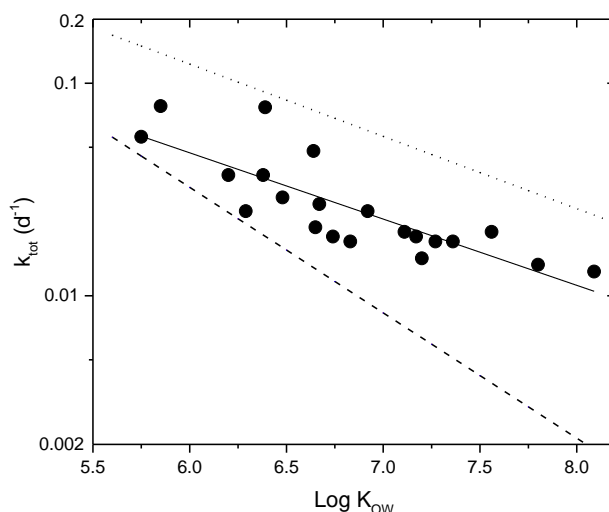
187
$$\log k_{tot} = -0.31 \pm 0.05 \cdot \log K_{OW} + 0.54 \pm 0.034; R^2 = 0.66; p < 0.001 \quad (6)$$

188 Figure 2 presents the model fitted congener specific k_{tot} values as well as k_{tot} versus K_{OW}
 189 relationships described by Eqs. 4, 5 and 6. Model fitted k_{tot} values were generally bracketed by
 190 k_{tot} predictions generated by Eqs. 4 and 5. The slope from Eq. 6 was most consistent (± 1
 191 standard error) for the slope reported by Eq. 5, although the intercept was lower. This is likely
 192 due to the fact that both Raeside's k_{tot} relationship (Eq. 5) and those generated from the present
 193 research reflect in-situ calibrated measures whereas O'Rourke's study (Eq. 4) was conducted
 194 under laboratory conditions. Mussel filtration rates are known to be affected by site specific
 195 characteristics including temperature, dissolved oxygen, food quality and quantity characteristics
 196 (Björk and Gilek, 1997; Heinonen et al., 1996). Across the various biomonitor calibration
 197 studies, field deployed mussels consistently appeared to exhibit higher filtration rates and
 198 chemical toxicokinetics compared to laboratory held animals (O'Rourke et al., 2001; Raeside et
 199 al., 2009). The lower intercept associated with Eq. 6 relative Eq. 5 is probably related to
 200 overwintering temperatures which would have slowed mussel filtration rates during this period.



202 **Figure 1.** Bioaccumulation rates of PCBs 49 (●), 110 (■),
203 138 (▲) and 180 (◇) in mussels biomonitors during 2014.
204 Lines represent model fit to equation 1.
205

206 PCB concentrations in water were subsequently estimated for the 2014 data using Eq. 2
207 and predictive k_{tot} relationships generated from either Eqs. 4, 5 or 6. Results are presented in
208 Figure 3 as mean±standard error concentrations at each deployment period and as yearly average
209 water concentration estimates denoted by horizontal lines. All three methods generated the same
210 water concentration for the 364 d deployment because each of the k_{tot} relationships predict that



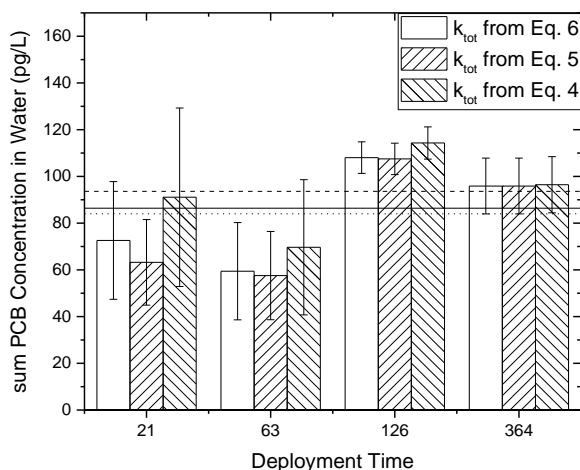
211 **Figure 2.** Whole body elimination coefficients of PCBs from
212 in mussels biomonitors as a function of chemical hydrophobicity.
213 Symbols (●) represent fitted model estimates of k_{tot} . Solid line is the
214 linear regression fit to the data (Eq. 6; $p < 0.001$), dashed lines are
215 predicted k_{tot} values based on Eqs. 4 and 5.
216

217
218 mussels fully achieve steady state over this time frame. Water concentrations for the 21 d
219 deployment were the most variable across k_{tot} algorithms because shorter deployments
220 necessitate the largest steady state correction factors. For example, the Eq. 4 k_{tot} expression
221 generated a mean 21 d \sum PCB water concentration that was 20.3% higher than that that derived
222 from the k_{tot} relationship using Eq. 6. Use of Eq. 5. yielded a mean concentration estimate 13%
223 lower compared to that using Eq. 6. By day 63, Eq. 5 produced a mean \sum PCB water
224 concentration that was within 3.3% of that generated by the site specific expression (Eq. 6),
225 while Eq. 4 yielded a water concentration within 17.2% of the Eq. 6 generated value. The yearly
226 mean \sum PCB concentrations from Eq. 4 and 5 were within 2.8 and 8.4% of the yearly mean
227 generated from Eq. 6, respectively.

228 An ANOVA was performed to determine the effect of deployment time and estimation
229 procedure (k_{tot} relationship utilized) on \sum PCB water concentrations generated for the 2014 data.
230 PCB concentrations in water were significantly different across the deployment periods

231 ($F_{1,41}=5.11$; $p<0.05$) but not significantly dependent ($F_{2,41}=0.16$; $p>0.8$) on the k_{tot} relationship
 232 utilized. This implies that between replicate variability of mussel accumulated residues was
 233 greater than the error introduced by using different correction factors associated with different
 234 k_{tot} relationships.

235 The k_{tot} relationship from Eq 6. was subsequently applied to estimate congener specific
 236 and \sum PCB water concentrations for full temporal data set generated at the Riverview Marina
 237 (1998-2015; $n = 209$ observations). The geometric mean \sum PCB concentration in water across
 238 years was 108.4 pg/L and ranged from 6.4 to 826.2 pg/L across individual measurements. Figure
 239 4 presents mean sum PCB concentrations as a function of time. There was a highly significant



240 **Figure 3.** Biomonitor estimated sum PCB water concentrations at
 241 the Riverside Marina site during 2014 over different deployment periods
 242

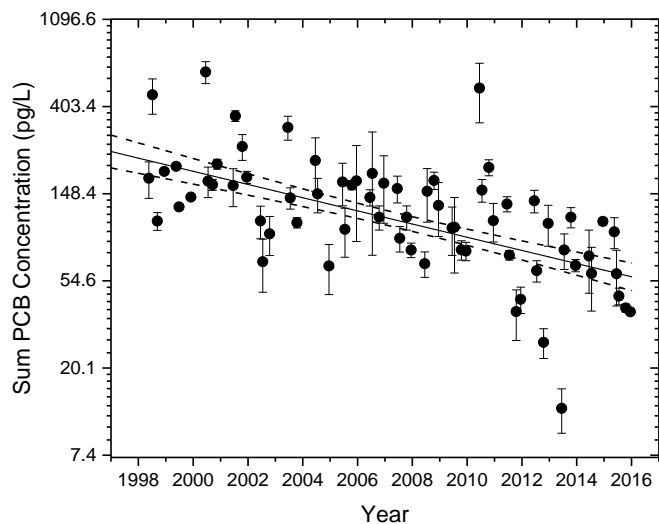
243
 244 ($F_{1,207} = 86.78$; $p<0.001$) decreasing trend in \ln PCB concentrations in water across years
 245 described by:

$$246 \ln C_{wat} = -0.076 \pm 0.008 \cdot Year + 156.6 \pm 16.3; R^2 = 0.29 \quad (7).$$

247
 248 Based on Eq. 7, the half life of \sum PCBs in water at this location in the Detroit River is 9.12 years.
 249 As a cross check against the temporal trend outlined in Eq. 7, a second regression was performed
 250 omitting the early time points and limiting the data to only the longest deployment time in each
 251 year. This was performed in order to reduce potential error associated with steady state
 252 correction factors. The second regression yielded a slope of -0.057 ± 0.012 that was not
 253 significantly different (ANCOVA; $F_{1,262}$; $p>0.2$) from the regression generated using the full data
 254 set. On the truncated data set, the estimated half life was somewhat slower at 12.1 years.

255 This research demonstrates that chemical toxicokinetics in deployed biomonitors will
 256 vary spatially and are also likely to vary at the same location in different years due to changing in
 257 situ conditions. However, the impact of this variation on PCB water concentration estimates will
 258 likely be small except in cases where mussels are deployed for only short periods of time (e.g. 21
 259 days or less). When mussels are deployed for 60 d or longer, the potential error in PCB water
 260 concentration estimates using an assumed k_{tot} relationship will not likely to exceed 20%, even

261 when laboratory as opposed to in-situ calibration of the biomonitor is performed. Additional
 262 sources of error not considered in this research may be generated as a result of error in the BCF
 263 expression used within the model. For examples, setting the $BCF = K_{ow}$ in Eq. 1 and 3 assumes
 264 that mussel lipids achieve equilibrium with dissolved water concentrations. However, mussels
 265 may also accumulate and potentially biomagnify PCBs from ingested seston along with
 266 exchange of water contaminants across their gills. *E. complanata* is a size selective filter feeder
 267 which consumes only small algal particles (<10 μm in size; Mueller et al., 2004). Small
 268 phytoplankton are typically assumed to be in equilibrium with the water, but can also achieve
 269 much lower chemical fugacity relative to water during high growth (Swackhamer and Scoglund,
 270 1993). Thus the depression of ingested phytoplankton fugacity during growth would offset
 271 biomagnification taking place by the mussels owing to their exposure to this diluted food source.
 272 However, further research is necessary to verify this is the case for the study species.



273
 274 **Figure 4.** Biomonitor estimated sum PCB water concentrations at
 275 Riverside Marina as a function of time. Solid line represents linear
 276 regression fit (Eq. 6). Dashed lines are the upper and lower
 277 confidence intervals of the regression fit.

278
 279 The present research also demonstrated that PCB concentrations in water at one location
 280 in the Detroit River are undergoing significant declines with time and the estimated half life of
 281 PCBs was approximately 9 years. Over the period of 1998 to 2015, the seasonal geometric mean
 282 PCB concentration declined from 198.1 pg/L to 43.6 pg/L representing a 454% decrease in
 283 measured residues. Based on Eq. 7, geometric mean residues at the biomonitoring station were
 284 predicted to have dropped 362% over this period of time. Thus, the Canadian waters of the
 285 Detroit River are exhibiting long term improvements with respect to PCB contamination in
 286 water.

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292

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