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2 **Flowing down the river: influence of hydrology on scale and accuracy of**
3 **elemental composition classification in a large fluvial ecosystem**

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12 **Highlights**

- 13 • Limitations of fish otolith chemistry analyses depend on regional water chemistry
- 14 • Hydrodynamics and geology alter the spatiotemporal composition of water
- 15 • Discharge–concentration relationships in the St. Lawrence River tributaries are weak
- 16 • Regional variations suggest that otolith chemistry analyses at the tributary scale are
- 17 possible

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22

23 **Abstract**

24 Trace metals found in the calcified structures of fish (i.e. otolith, scales and vertebrae) serve as
25 proxies for the ambient water composition at the time of mineralization, and these trace metals are
26 increasingly used as a tool for assessing population structure and the migratory patterns of fish.
27 However, the appropriate scale (e.g. resolution) for such applications can be uncertain because of
28 a poor understanding of the spatiotemporal variations of metal-to-calcium ratios (Me:Ca) in the
29 studied watersheds. This study aims to assess Me:Ca spatiotemporal variability within the St.
30 Lawrence River and nine major tributaries and evaluate the ability of random-forest models to
31 correctly identify rivers on the basis of their elemental composition. We tested the influence of
32 daily discharge on four measured ratios (Sr:Ca, Ba:Ca, Mg:Ca and Mn:Ca) to document local and
33 regional trace element sources and dynamics. The four element ratios displayed a low
34 spatiotemporal variation, reflecting a marked stability over time. We observed that most element-
35 and tributary-specific concentration–discharge relationships were either not significant or showed
36 a weak influence, thereby confirming a stable point source dynamic. The classification
37 performance based on a four-element model (Sr:Ca, Ba:Ca, Mg:Ca and Mn:Ca) produced a
38 classification accuracy of 92.5%, which correspond to a small decrease of accuracy compared to
39 the full model (25 elements, 96.6% of correct classification). A classification based on two
40 elements (Sr:Ca and Ba:Ca) produced a lower classification accuracy (72.6%). Classification
41 errors related mainly to tributaries in close proximity, a problem tempered by grouping these
42 geochemically similar watersheds. Our results show that surveys of the elemental fingerprint of
43 regional tributaries within a given region can provide critical information to determine the
44 appropriate scale (tributary or watershed) for trace metal analysis of the hard-calcified parts of
45 fish.

46 **Keywords:** Otolith chemistry, Concentration–discharge, Water chemistry, Random forests

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50 **1. Introduction**

51 In freshwater ecosystems, bedrock chemistry, as a major source of weathering dissolution, can
52 explain much of the water elemental composition (Humston and Harbor, 2006; Liu et al., 2000).
53 Local and regional hydrodynamics can, however, have sufficient influence to preclude prediction
54 based solely on geological information. Covariance among river discharge and weathering-diluted
55 solutes have been widely observed (Moon et al., 2014), providing evidence of a relationship
56 between hydrodynamics and water chemistry. In addition, other factors such as land use,
57 landscape, industrial wastewater and riverine mixing patterns strongly influence the water
58 chemistry of lakes, streams and rivers and could decouple covariance with discharge (Baronas et
59 al., 2017). The monitoring of trace elements and their relationship to water flow—called
60 concentration–discharge (C–Q) relationships (Baronas et al., 2017; O'Connor, 1976)—has
61 successfully contributed to assess water sources and mixing patterns (Neal et al., 1996). Hence,
62 analysis of C–Q relationship in a given freshwater ecosystem could inform on the processes
63 contributing to observed water chemistry variations. This approach could inform the potential
64 factors affecting accuracy of using water chemistry as an ecological tracer of fish origin and
65 movements through the analysis of hard-parts (i.e. scales, otolith, vertebrae) elemental
66 composition (Pracheil et al., 2014).

67 Otoliths (ear stones) are calcified structures of the inner ear of fish. These structures undergo
68 continuous growth through the daily deposition of a new layer, proportional to the fish's somatic
69 growth (Pannella, 1971). Otoliths are composed of calcium carbonate (CaCO_3 , approximately
70 98%) and a smaller proportion (0.1%–2.3%) of organic matter (soluble and insoluble protein
71 matrix), as well as approximately 45 elements in either minor (>100 ppm) or trace (<10 ppm)
72 amounts (Campana, 1999; Sturrock et al., 2012). Some elements are incorporated into the otolith
73 matrix at concentrations representative of the ambient water experienced by fish at that specific
74 time (Doubleday et al., 2013; Izzo et al., 2018). Hence, the trace element composition of otoliths
75 (or otolith chemistry) can provide information related to individual fish life histories (Campana,
76 1999). Otolith chemistry, assessed by bulk otolith analysis or precise point–based sampling, is
77 increasingly used to clarify fish migratory history (Elsdon and Gillanders, 2003a; Morissette et al.,
78 2016; Secor and Piccoli, 2007), define population structure (Lazartigues et al., 2018; Lazartigues
79 et al., 2016; Wright et al., 2018) and assess mixed stocks composition (Tanner et al., 2016). The
80 detection of biologically significant variations in otolith chemistry arises from the movement of

81 virtually any teleost fish within a sufficient “elemental gradient” or from the fish using habitats
82 characterized by divergent “elemental fingerprints”. Those elements gradients, however, are
83 seldom described by measurements of trace elements in the water as a preliminary step of otolith
84 chemistry studies (Pracheil et al., 2014), which could cause uncertainty on the appropriate scale to
85 based inference on.

86 The most-studied chemical gradient experienced by fish is, without a doubt, the drastic freshwater-
87 saltwater interface observed in estuarine habitats. The meeting of these two distinct water masses
88 produces a steep gradient where elements such as strontium (Sr) and barium (Ba) vary by more
89 than an order of magnitude across a short geographical distance (Taddese et al., 2019; Tanner et
90 al., 2013). Elemental gradients in freshwater ecosystems are, however, generally weaker than
91 estuarine gradients (Radigan et al., 2017; Zeigler and Whitley, 2011) and are characterized by
92 much less predictable, sporadic local variations (Chapman et al., 2013). For instance, large fluvial
93 ecosystem such as the St. Lawrence River (Canada) exhibit significant spatial variations in
94 elemental composition (Lum et al., 1991; Yeats and Loring, 1991). Understanding elemental
95 composition dynamics is a prerequisite to develop applications for the management of various
96 stressors on the fish community commonly observed in large fluvial ecosystem such as
97 overexploitation, poaching, habitat fragmentation or invasion of exotic species.

98 Inferences related to fish migration that are derived from otolith chemistry are generally based on
99 a few elements. Studies documenting the relationship between water and otolith trace metal
100 concentrations, summarized as the partition coefficient $(Me: Ca)_{otolith}/(Me: Ca)_{water}$ (Campana,
101 1999), have demonstrated that Sr:Ca and Ba:Ca coefficients are positive for most species and
102 environments (Brown and Severin, 2009). These conclusions leave little doubt that the
103 concentrations of these elements in water influence the composition of the accreted otolith material
104 to some extent. The published coefficients for manganese (Mn:Ca) are somewhat contradictory
105 (Miller, 2009), although they suggest an environmental influence (Elsdon and Gillanders, 2003b);
106 this is not the case for magnesium (Mg:Ca) because the partition coefficients of magnesium are
107 generally low (<0.001) or nonsignificant (Elsdon and Gillanders, 2003a; Wells et al., 2003). The
108 most parsimonious classification models should then be limited to the use of Sr:Ca and Ba:Ca.
109 Nonetheless, adding Mg:Ca and Mn:Ca to the model can improve the accuracy of watershed
110 classification, as observed in multiple studies, including ones in the St. Lawrence River

111 (Lazartigues et al., 2018; Morissette et al., 2016), and the possible contribution of these ratios to
112 classification models should not be dismissed completely.

113 Otolith chemistry has been successfully used in the St. Lawrence River system to delineate the
114 reproductive stock structure of yellow perch (*Perca flavescens*) fisheries (Lazartigues et al., 2018),
115 identify the early reproductive and migratory behavior of the reintroduced striped bass (*Morone*
116 *saxatilis*, Morissette et al., 2016; Vanalderweireldt et al., 2019) and determine the migratory
117 behavior of the American eel (*Anguilla rostrata*, Benchetrit et al., 2015). These studies, however,
118 have generally been conducted at a low resolution (i.e. salinity front). Finer resolution analyses
119 can improve the delineation of the stock structure of numerous fish populations for which the
120 relative importance of the St. Lawrence River and its tributaries during the various phases of their
121 life cycle and consequent population dynamics remains to be quantified. Otolith chemistry also
122 represents a promising tool for studies related to biological invasions (Crook et al., 2013;
123 Morissette et al., submitted; Thibault et al., 2010) and wildlife forensic sciences (Bourret and
124 Clancy, 2018), a field for which it remains underutilized.

125 This study has the objective of summarizing the elemental composition of the fluvial section of
126 the St. Lawrence River and its major tributaries in order to explore regional spatiotemporal
127 variations in elemental composition and the consequent effects on the accuracy of identifying and
128 classifying rivers based on their water chemistry. We analyzed element concentrations in the
129 various rivers to quantify the intra- and interannual variability of elemental compositions in the
130 tributaries and isolate the influencing factors. Additionally, we assessed the C–Q relationships for
131 Sr:Ca, Ba:Ca, Mg:Ca and Mn:Ca in the studied tributaries in relation to estimates of daily
132 discharge to define elemental source dynamics and the origin of temporal variations. Finally, we
133 tested the accuracy of multiple classification models to explore the potential of using fish hard-
134 parts chemistry for identifying rivers at the tributary scale. This study aimed to provide a detailed
135 depiction of the trace metal fingerprints of a large fluvial ecosystem and its tributaries and improve
136 the understanding of trace metal dynamics to promote diverse applications in fisheries
137 management, and conservation for large river ecosystems.

138 **2. Methods**

139 We obtained the trace metal concentration data from the *Banque de données sur la qualité du*
140 *milieu aquatique* (BQMA) database compiled by the province of Québec's *Ministère de*

141 *l'environnement et de la lutte contre les changements climatiques* (MELCC, 2020). This database
142 includes concentrations ($\mu\text{g/L}$) of 25 trace elements (Ag, Al, As, Ba, B, Be, Ca, Cd, Co, Cr, Cu,
143 Fe, K, Mg, Mn, Mo, Na, Ni, Pb, Sb, Se, Sr, U, V, Zn) and the physicochemical parameters for each
144 sample station. Concentrations of trace metal have been quantified by inductively coupled plasma
145 mass spectrometry (ICP-MS). Raw data and detailed analytical protocols are available via open
146 access through an interactive interface (MELCC, 2020). To quantify the spatiotemporal variation
147 of trace metal concentrations, we restricted our analysis to freshwater tributaries for which we had
148 more than four years of monitoring. Sites of water sampling were chose to be easily accessible and
149 exempt of any exogenous source of water (i.e. small tributaries or storm water channel). Our data
150 set, therefore, covers nine tributaries of the St. Lawrence River, all located between Montréal and
151 Québec City (Fig. 1). Between May 2009 and October 2017, water samples were retrieved on a
152 monthly basis from a single station near the mouth of each sample tributary (Fig. 1). In addition to
153 these sites, we created a single site by combining two St. Lawrence River sampling regions (St.
154 Thérèse and Tracy), each including three sites (North Shore, Center and South Shore). We chose
155 to pool these data into a single site, as high intersite similarities in St. Lawrence River produced
156 high confusions in subsequent classifications models (data not shown).

157 We normalized trace metal concentrations ($\mu\text{g/L}$) according to the calcium concentration measured
158 at the same sampling site (mg/L). Metal-to-calcium ratios (Me:Ca) were expressed as molar
159 concentrations ($\text{mmol}\cdot\text{mol}^{-1}$). Since the uptake of trace elements by fish gills decreases as the
160 concentration of dissolved calcium increases (Mayer et al., 1994), the absolute concentration of
161 trace metals is generally a poor predictor of their environmental availability and their eventual
162 accretion in otoliths (Campana, 1999). Ratios of trace element concentrations as a function of
163 calcium offer a better representation of the elements available for integration into the carbonate
164 matrix of the otolith, and these ratios should be privileged when available. We therefore conducted
165 subsequent analyses using the Me:Ca values for every studied element.

166 We tested the differences of Sr:Ca, Ba:Ca, Mn:Ca and Mg:Ca between sites (sites, $N = 10$) using
167 one-way analysis of variance (ANOVA). Intersite differences were identified by post hoc Tukey
168 HSD tests. We tested for intrasite variability using repeated-measure analyses of variance (RM-
169 ANOVA), where the response variables (Me:Ca) were a function of sampling years (fixed factor)
170 and date of sampling nested in years (fixed factor). We treated tributaries (sites) as the repeated

171 factor to account for the non-independence of the monitoring data. We ran both ANOVA models
172 using the function *aov* in the R software (R Development Core Team, 2008) base package.

173 Linear mixed-effects models assessed the intrasite variations of trace metal C–Q relationships as
174 a function of the adjusted discharge at the site of water sampling. This modelling approach have
175 been chose as its offers possibility of considering intrasite variability and producing variable slopes
176 and intercept within a single model. We retrospectively estimated the adjusted discharge using
177 observed discharge at the available gauge stations and their relative position to water sampling
178 stations, following the method of Lachance-Cloutier et al. (2017). No discharge data were
179 available for Des Prairies River and the St. Lawrence River site. The influence of a large dam
180 complex on the St. Maurice River prevented an accurate estimate of daily discharge at the sampling
181 station. We therefore excluded those three sites from the subsequent analysis of C–Q relationships.
182 We modeled Me:Ca ratios (dependent variable) as a function of adjusted daily discharge
183 (dependent variable, natural-logarithm transformation) nested in sites, with adjusted discharge
184 nested in sites treated as a random factor to account for study design and allow varying intercepts
185 and slopes for every site. We considered slopes of site-specific regressions to be significantly
186 different from 0 when $p < 0.05$ (*t*-test). Linear mixed-effects models were run using the ‘nlme’ R
187 package (Pinheiro et al., 2018).

188 To estimate the identification accuracy of the elemental composition of the tributaries, we built
189 three different classification models using the random forest (RF) algorithm. Classification by RF
190 was shown to be outperforming parametric models (ex. linear discriminant analysis) for otolith
191 chemistry data when raw data are not respecting multi-normality assumptions (Jones et al., 2016;
192 Mercier et al., 2011), which is the case for the present study. Three classification models were built
193 to evaluate the accuracy of identification on the basis of (1) the full elemental fingerprints (25
194 elements) of the sites, which represent the higher classification accuracy reachable with this
195 dataset; (2) the most frequently used trace elements in otolith chemistry, as noted in the published
196 literature (Sr:Ca, Ba:Ca, Mn:Ca and Mg:Ca); and (3) elements exhibiting the most robust partition
197 coefficients between water and otolith concentrations (Sr:Ca and Ba:Ca). To assess the
198 performance of each model, we subdivided the data set at random to create training (75% of the
199 data) and test (25% of the data) data sets. The RF model was adjusted a priori, using the sampling
200 sites as a classification factor, and the adjustment was carried out by 1000 classification trees. We
201 reclassified the test data set and estimated the correct reclassification to test the performance of the

202 RF model. Random-forest modeling was run using the ‘randomForest’ R package (Breiman and
203 Cutler, 2018).

204 3. Results

205 Our water chemistry data set consisted of 580 samples distributed over ten sites. Each site was
206 sampled for at least four years (5–6 monthly samples per year) and for a maximum of eight years.
207 ANOVA showed significant between-site differences for the concentrations of the four elements
208 (Fig. 2). Moving upstream to downstream, the Sr:Ca differences between sites (ANOVA, $F_{9,569} =$
209 $229, p < 0.001$) increased, with the St. Charles River having the highest value in the data set
210 ($\text{Sr:Ca}_{\text{avg}} = 4.85 \pm 0.56 \text{ mmol} \cdot \text{mol}^{-1}$). Conversely, Ba:Ca (ANOVA, $F_{9,570} = 859, p < 0.001$) values
211 generally decreased from upstream to downstream, although the St. Maurice River showed the
212 highest values in the studied tributaries ($\text{Ba:Ca}_{\text{avg}} = 1.13 \pm 0.17 \text{ mmol} \cdot \text{mol}^{-1}$). The main stem of
213 the St. Lawrence River and all the tributaries of the floodplain displayed high Mg:Ca values
214 (ANOVA, $F_{9,580} = 265, p < 0.001$), except for the St. François, Chaudière and St. Charles rivers.
215 Finally, Mn:Ca (ANOVA, $F_{9,569} = 271, p < 0.001$) tended to increase heading downstream, with
216 the exception of sites located within the St. Lawrence River, which had low and variable values
217 ($\text{Mn:Ca}_{\text{avg}} = 0.07 \pm 0.06 \text{ mmol} \cdot \text{mol}^{-1}$). The analysis of intrasite variability showed that there were
218 no significant interannual or intra-annual differences for Sr:Ca (RM-ANOVA_{inter}, $F_{4,2} = 4.31, p =$
219 0.20 ; RM-ANOVA_{intra}, $F_{3,2} = 0.51, p = 0.72$) and Mg:Ca (RM-ANOVA_{inter}, $F_{4,2} = 2.95, p = 0.27$;
220 RM-ANOVA_{intra}, $F_{3,2} = 0.24, p = 0.86$). However, Ba:Ca exhibited a significant intra-annual
221 variability (RM-ANOVA_{intra}, $F_{3,2} = 238.89, p = 0.004$), and Mn:Ca was variable both inter- and
222 intra-annually (RM-ANOVA_{inter}, $F_{4,2} = 24.55, p = 0.04$; RM-ANOVA_{intra}, $F_{3,2} = 88.83, p = 0.01$).
223 These patterns are attributable mainly to the large variations observed in the St. Maurice and St.
224 Charles rivers for this element (Fig. 3).

225 Modeling the effect of the estimated daily discharge on Me:Ca values showed an overall marginal
226 influence. Hence, most site/element combinations (57.1%, 16/28 tests) displayed weak or
227 nonsignificant C–Q relationships (Appendix A). We observed significant relationships for the
228 Assomption, Yamaska, St. François, Chaudière and St. Charles rivers (Fig. 4). Specifically, Sr:Ca
229 values correlated negatively with discharge in the Yamaska, Chaudière and St. Charles rivers.
230 Ba:Ca values showed a positive correlation with discharge in the Chaudière and Des Prairies rivers
231 and a negative correlation with discharge in the St. François and St. Charles rivers. Mn:Ca values

232 correlated positively with discharge in the Yamaska, Chaudière and St. François rivers, but this
233 pair of variables were negatively correlated in the St. Charles River. Finally, Mg:Ca and discharge
234 had a positive correlation for the Assomption River. Me:Ca values in the Des Prairies and
235 Richelieu rivers were weakly influenced by discharge through the observed range of discharge
236 levels. Calcium concentrations were negatively correlated with discharge for all rivers except the
237 Des Prairies and Richelieu Rivers, where concentrations were mostly constant for all estimated
238 discharges. The absence of an influence of daily discharge on Me:Ca in most tributaries showed
239 that calcium and trace metal concentrations remain highly coupled under fluctuating hydrologic
240 conditions, thereby explaining the stability of the ratios.

241 The accuracy of the RF model classification varied depending on the number of elements
242 considered in the model. The full model (25 elements) produced a classification accuracy of
243 96.6%. Most classification errors in the model were attributed to the Des Prairies River (91.8%
244 correct) and Richelieu River (93.8% correct). Ba:Ca and Mn:Ca were, respectively, the first and
245 second most important elements for classification within the full model. The four-element (Sr:Ca,
246 Ba:Ca, Mn:Ca and Mg:Ca) model produced a classification accuracy of 92.5%. We observed most
247 classification errors for the Des Milles-Iles (78.1% correct) and Des Prairies rivers (89% correct),
248 two rivers originating upstream of the Ottawa River. Eight of ten sites showed classification
249 accuracy greater than or equal to 90%. The RF classification model based on the concentration of
250 the two major trace elements (Sr:Ca and Ba:Ca) were 72.6% accurate. The classification accuracy
251 of the Richelieu River was the lowest (61.9% correct), most often confused with the two adjacent
252 tributaries (St. François and Yamaska rivers, Table 1). These three rivers are all located along the
253 south shore of the St. Lawrence River and are located relatively close to each other. Six sites had
254 classification accuracy greater than 85%. The St. Charles and St. Maurice rivers were perfectly
255 classified (100% correct), a result potentially attributed to their high Sr:Ca and Ba:Ca values,
256 respectively (Fig. 5). Pooling the data of the Des Prairies and Des Milles-Iles rivers (Ottawa River
257 water mass) and the Richelieu, St. François and Yamaska rivers (south shore tributaries) increased
258 the classification accuracy of the two-element model to 84.1%.

259 **4. Discussion**

260 Assessment of the geographical and seasonal variability of trace elements represents an
261 opportunity for a more objective evaluation of the appropriate resolution for any analyses of otolith

262 chemistry or other fish hard-parts (Walther, 2019). Knowledge of the local, regional or continental
263 elemental gradients can also establish the possibilities and limits of achievable inferences (Pracheil
264 et al., 2014; Whitley, 2009). Such portraits also represent an important step for identifying
265 redundancy and anomalies within the trace element landscape; for example, the identification of
266 the Sr-rich waters of the Sandusky River (western basin of Lake Erie) was vital for using otolith
267 chemistry to assess the recruitment and migration of invasive grass carp in the Great Lakes
268 (Chapman et al., 2013). In our study, the spatiotemporal variability of dissolved Me:Ca values,
269 assessed over a major portion of the ice-free season, was relatively low. Most intrasite variations,
270 if present, were lower than the observed intersite variations over the study period (>5 years). This
271 result highlights the relative stability of the observed Me:Ca values of the major St. Lawrence
272 River tributaries. Our results showed that the limited intrasite variability and the consistent intersite
273 differentiation will most likely allows for accurate identification of rivers elemental fingerprint
274 and the resulting fish hard-parts chemistry.

275 *4.1 Elemental landscape of the St. Lawrence River system*

276 We observed the presence of positive (Sr:Ca and Mn:Ca) and negative (Mg:Ca) upstream–
277 downstream gradients among the study sites. Such gradients are concordant with previous
278 observations from the St. Lawrence main stem and tributaries (Benchetrit et al., 2015; Lum et al.,
279 1991; Yeats and Loring, 1991). Most studies of trace element concentrations in the system,
280 however, targeted water quality assessments and did not apply *per se* to fish hard-parts chemistry.
281 This specificity caused key information to be either missing (i.e. temporal variability, ratios to
282 calcium content and missing elements) or inappropriate. Our study is, to the best of our knowledge,
283 one of the first to evaluate the potential of tributary- or watershed-scale classifications through the
284 analyses of otolith chemistry in the St. Lawrence system. Admittedly, such conclusions have to be
285 confirmed using actual otolith data in a similar study design. The observed large-scale variations
286 reflect the potential influence of regional geology of the St. Lawrence River valley, characterized
287 by a pronounced difference between North shore and South shore bedrock composition and a
288 gradual upstream/downstream succession. Our result suggest the relatively high likelihood of
289 predicting water composition based on geological data (Liu et al., 2000). In general, tributaries in
290 close proximity, originating and flowing over similar rock formations (e.g. same shore), will share
291 a similar chemical composition, which could highly differs from a close tributary located on the
292 opposite shore.

293 In our data set, the St. Maurice River did not follow the upstream–downstream trend, showing
294 notably higher Sr:Ca and Ba:Ca values relative to the other studied tributaries. Such a marked
295 difference in water chemistry for this river is likely due to the particular geography and geological
296 characteristics of this watershed. Located on the north shore of the St. Lawrence River, the St.
297 Maurice River flows through the black spruce–dominated forest lying upon the Grenville Province
298 of the Canadian Shield (Saint-Jacques and Richard, 2002). The river therefore contains a high
299 concentration of humic acid and a low concentration of calcium; this contrasts with the calcium-
300 rich waters of the St. Lawrence River and its south shore tributaries situated on top of the
301 limestones of the St. Lawrence Lowland Platform. The latter originate from within the
302 Appalachian Province, characterized mainly by sedimentary rocks (i.e. limestone, shale,
303 mudstone), with the river flowing over surface deposits of clay and silt. The low calcium content
304 of the St. Maurice River produces higher Me:Ca values, hence trace element availability.
305 Classification of the St. Maurice River was therefore efficacious, attributed mainly to its
306 comparatively unique bedrock as it was the sole boreal shield river in the data set. Other tributaries
307 flowing over the Canadian Shield, albeit smaller streams originating from southern parts of the
308 Shield, likely share similar trace metal compositions; therefore, the probable accuracy of our model
309 would be reduced with the addition of these other streams. Incidentally, the St. Charles River could
310 have also been considered as a northern shore boreal shield stream, although its watershed is more
311 restricted and mostly urbanized, reducing boreal forest contribution to water chemistry, hence
312 acidity, and increasing Ca concentrations, which is consistent with its less extreme composition.
313 Difference in morphology and landform feature between St. Maurice and St. Charles Rivers are
314 likely to exacerbate the discrepancy in water trace elements compositions. These results also
315 emphasize the importance of using metal-to-calcium ratios when considering ambient water
316 chemistry values and its relationship with otolith chemistry (i.e. elemental availability for
317 deposition), especially in systems where water chemistry varies over short distances owing to
318 differences in geology and land use (e.g. forest vs urban vs agriculture).

319 *4.2 Temporal variation and hydrological influence*

320 Analyses of temporal variations provided evidences that annual variations were mostly negligible
321 for Sr:Ca and Mg:Ca. Annual and seasonal Ba:Ca and Mn:Ca variations were important only in
322 tributaries (i.e. St. Charles and Chaudière) exhibiting significant concentration-discharge
323 relationships (C-Q). The modeled C–Q relationships showed that most Me:Ca concentrations were

324 either not influenced by discharge or presented evidence of stable point-source dynamics.
325 Decreases of Me:Ca in relation to discharge suggest that trace metal inputs are likely from deep
326 (i.e. bedrock) or single point sources (i.e. industrial or municipal wastewater or city effluent)
327 contributing at a relatively constant rate (Neal et al., 1996). Increased flow, associated with waters
328 intermittently flooding lowlands and/or increased sediment mobilization and leaching, appears to
329 have only a marginal influence on the observed Me:Ca values in our studied tributaries and the St.
330 Lawrence River. The positive C–Q relationship observed for the Chaudière River, however,
331 provides possible evidence of intermittently accessible external sources of Ba and Mn that are
332 linked to the flooding regime. The Chaudière River is characterized by a high flooding frequency
333 due to its shallow channel and large (approximately 1 km wide) alluvial plain over a 60 km stretch,
334 which ends only 35 km from the river mouth (Hamelin, 1958). This hydrological regime releases
335 important quantities of dissolved trace metals, which could have a high residency time given the
336 large dam and reservoir complex located approximately 3 km upstream of the river mouth. Even
337 if Me:Ca varied with discharge, we did not observe much overlap with other tributaries in periods
338 of either high or low flow. In this model, intermittent variations of water chemistry related to
339 discharge of the Chaudière River were not sufficient to affect the classification performance at a
340 tributary scale.

341 *4.3 Classification and potential use in otolith chemistry*

342 The St. Lawrence River is a large unhindered freshwater system composed of two highly structured
343 parallel-flowing water masses (Great Lakes and Ottawa River water mass). Previous studies have
344 suggested a limited longitudinal mixing of those water masses from their initial confluence at the
345 eastern tip of the island of Montréal to Lake St. Pierre. From this point, mixing increases until
346 arriving at the St. Lawrence estuary, slightly downstream of Québec (Lum et al., 1991). We
347 observed that, at least for the studied fluvial section, limited variability in chemical composition,
348 in agreement with previous studies (Benchetrit et al., 2015; Yang et al., 1996; Yeats and Loring,
349 1991). Contributions of most tributaries were marginal to the trace element composition of the St.
350 Lawrence River, except for stations very near the outflow of the tributaries. In light of these results,
351 a tributary-scale reconstruction of fish displacement or analyses of stock structure is possible;
352 nonetheless, fish movement within the main stem could be missed owing to a lack of a sufficient
353 element gradient, at least in the studied freshwater reach. Our results also demonstrate that
354 tributary-based classification is difficult or unlikely on the basis of a single element classification.

355 The Chaudière, St. Maurice and St. François rivers are, for example, nearly indistinguishable (Fig.
356 3) when only considering Sr:Ca, despite these rivers being located within separate watersheds
357 characterized by different geology. However, adding only one other element (Ba:Ca) overcomes
358 this problem and provides a 100% success in identifying the rivers (Table 1).

359 Admittedly, identification of substantial trace element gradients in the studied system is only a part
360 of the overall assumption supporting otolith chemistry (Pracheil et al., 2014). This assumption
361 states that ambient variability in the water environment will be reflected in the accreted increments
362 of the otoliths (Campana, 1999) to provide a useful proxy of habitat use. As most trace element
363 concentrations in otolith are not representative of surrounding water (partition coefficient null or
364 non-significant), the full model (25 elements) is presented here as the maximum attainable
365 accuracy, but is not a plausible option for use in otolith chemistry analyses. The elemental
366 composition of water is the major influence on Sr and Ba concentrations in otoliths, an aspect that
367 is particularly relevant in freshwater where salinity, the second most influential factor (Izzo et al.,
368 2018), is absent. Manganese has been related to environmental concentrations (Dorval et al., 2007;
369 Elsdon and Gillanders, 2003b), although the accretion mechanism needs to be better defined
370 (Sturrock et al., 2015). Studies assessing the Mg partition coefficients show that $Mg:Ca_{\text{otolith}}$ is not
371 influenced by Mg concentrations in water (Miller, 2011; Wells et al., 2003), although other
372 extrinsic (e.g. water temperature) and intrinsic (i.e. ontogeny and growth) factors appear
373 responsible for the accretion rate (cited in Hüsey et al., 2020). The influence of these intrinsic and
374 extrinsic factors (such as hypoxia), which affect the rates of elemental uptake and otolith
375 deposition in its inorganic and organic components (Limburg et al., 2015), also need to be
376 considered for improved comprehension and predictive power and to avoid too simplistic views
377 of such complex phenomena (Hüsey et al., 2020). Such developments are certainly not trivial but
378 could lead to much more effective applications. Luckily, the decrease of classification accuracy of
379 a three-element RF model (Sr:Ca, Ba:Ca, Mn:Ca; 85.6% correct) was minimal compared to our
380 four-element RF model (92.5% correct) and likely provides sufficient definition to envision a
381 tributary-scale resolution.

382 **5. Conclusion**

383 Based on water regional chemistry data, we suggest that otolith chemistry analyses in the St.
384 Lawrence River system could be possible at a tributary-scale resolution. Our analysis of the

385 interannual variations of water samples collected from the St. Lawrence River and its tributaries
386 highlighted the relative stability of the chemical composition of water flowing in the tributaries.
387 The simple experimental design applied in this study could be replicated in other aquatic
388 ecosystems. Trace elements concentrations and discharge gauging stations are generally part of
389 routine analyses from federal or states environmental agencies and could provide, even if spatio-
390 temporally limited, valuable a priori information before initiating otolith chemistry studies. We
391 acknowledge that our analysis of elemental composition was limited to the main tributaries of the
392 St. Lawrence Lowlands, located between Montréal and Québec City. These analyses should be
393 extended to the more than 40 tributaries that enter into the freshwater portion of the St. Lawrence
394 River. Moreover, it have been assumed that surface water chemistry of St. Lawrence main stem
395 and tributaries is representative of the whole water column, owing to the high water velocity,
396 however, local vertical variations of water chemistry in main stem deep channels of the St.
397 Lawrence River may exist and will need to be characterized. It is likely that the inclusion of a
398 greater number of tributaries and extreme (deep) habitats into the models will require creating
399 groups of proximal streams/rivers of similar elemental composition, which will decrease the study
400 resolution (e.g. the creation of classification groups combining streams of similar elemental
401 composition). Future studies should aim to refine and improve the spatiotemporal (inter- and intra-
402 tributary) elemental data from large fluvial ecosystems to better understand partition coefficients
403 for the most common elemental species of the system and test classification models at the tributary
404 scale on actual otolith chemistry data. Such developments would increase the value and usefulness
405 of otolith chemistry beyond its actual state in the St. Lawrence River and in several other fluvial
406 ecosystems around the world.

407

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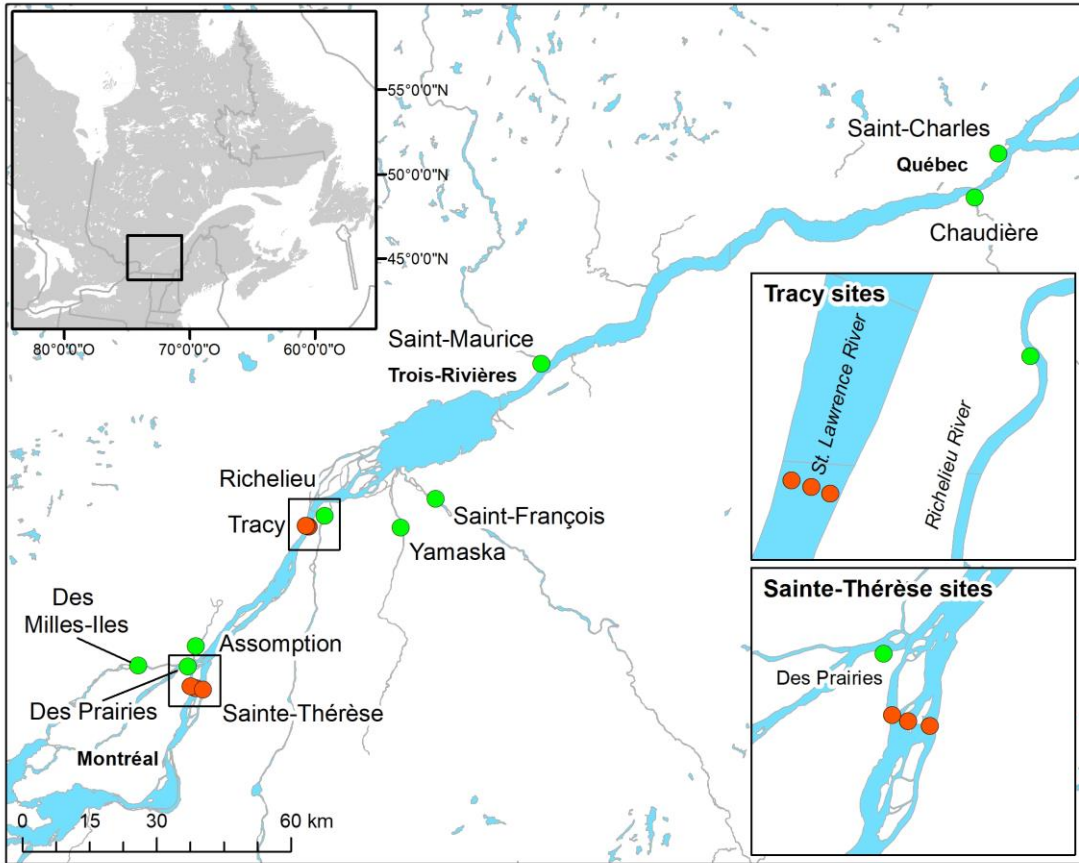
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567

568 **Tables**

569 Table 1. Classification accuracy of the training data of the sample trace metal composition from
 570 the different stations using the two-element RF model (Sr:Ca, Ba:Ca).

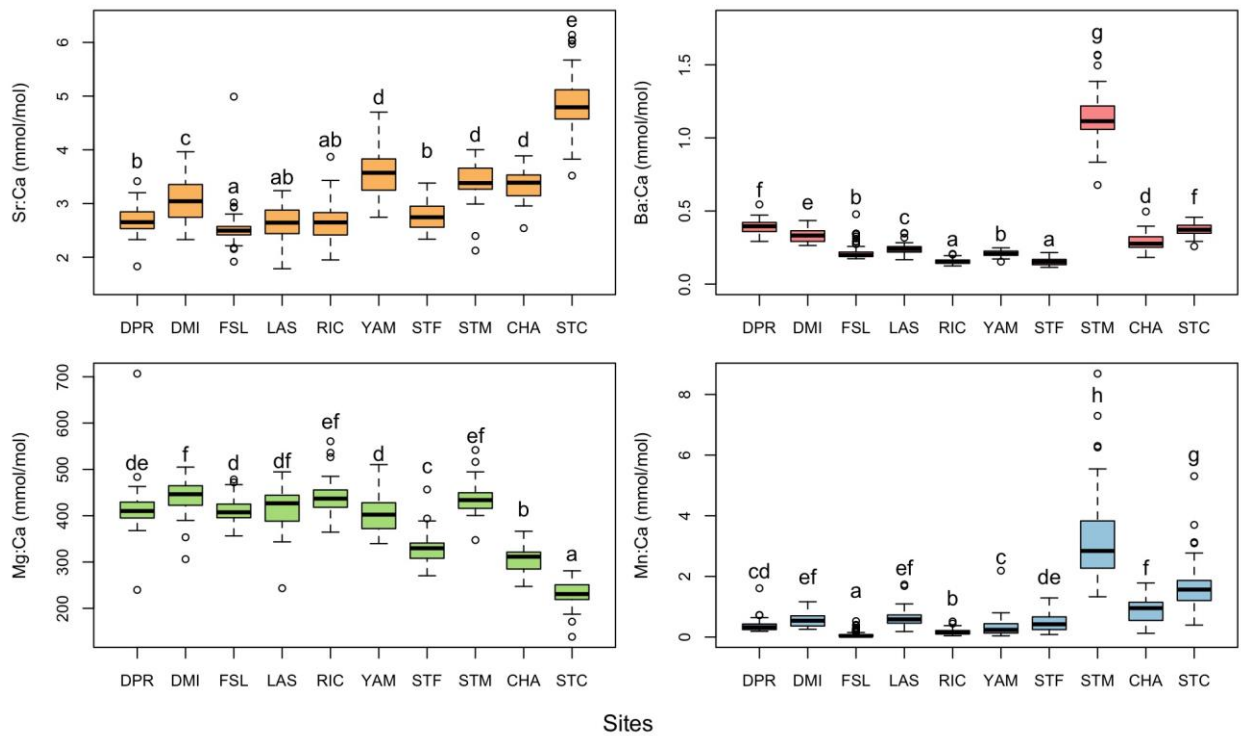
| | DPR | DMI | SLR | LAS | RIC | YAM | STF | STM | CHA | STC | Error |
|-----------------------|-----------|-----------|-----------|----------|-----------|-----------|-----------|-----------|-----------|-----------|-------|
| Des Prairies (DPR) | 18 | 0 | 3 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 0.18 |
| Des Milles-Iles (DMI) | 2 | 11 | 4 | 0 | 0 | 0 | 0 | 0 | 7 | 0 | 0.54 |
| St. Lawrence (SLR) | 4 | 3 | 74 | 10 | 0 | 1 | 0 | 0 | 0 | 1 | 0.20 |
| L'Assomption (LAS) | 0 | 1 | 15 | 8 | 0 | 1 | 0 | 0 | 1 | 0 | 0.69 |
| Richelieu (RIC) | 0 | 0 | 0 | 1 | 16 | 2 | 7 | 0 | 0 | 0 | 0.38 |
| Yamaska (YAM) | 0 | 0 | 0 | 2 | 3 | 22 | 0 | 0 | 2 | 0 | 0.24 |
| St. François (STF) | 0 | 0 | 0 | 0 | 8 | 0 | 12 | 0 | 0 | 0 | 0.40 |
| St. Maurice (STM) | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 25 | 0 | 0 | 0.00 |
| Chaudière (CHA) | 1 | 8 | 0 | 1 | 0 | 3 | 0 | 0 | 12 | 0 | 0.52 |
| St. Charles (STC) | 0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 22 | 0.08 |

571 * The numbers represent the water samples from a site (lines) assigned by the model to the
 572 established elemental fingerprint by model (columns). The numbers in bold are the correct
 573 classifications.



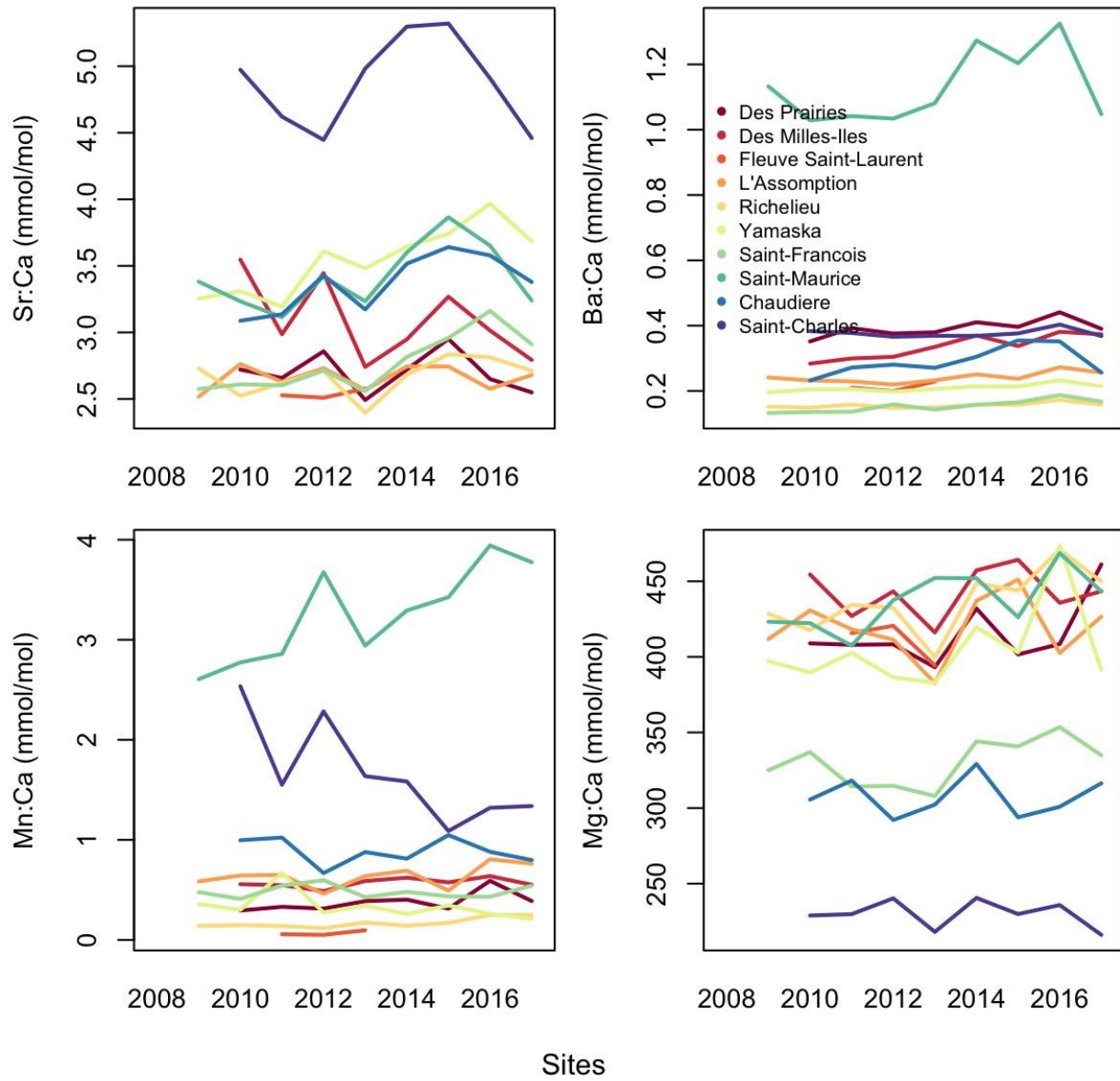
575

576 **Fig. 1.** Sampling sites located in tributaries (green circles) and the St. Lawrence River main stem
577 (orange circles). Inset maps show the locations of the Tracy and St. Thérèse sites (orange
578 circles), which were pooled as St. Lawrence River sites (see text).



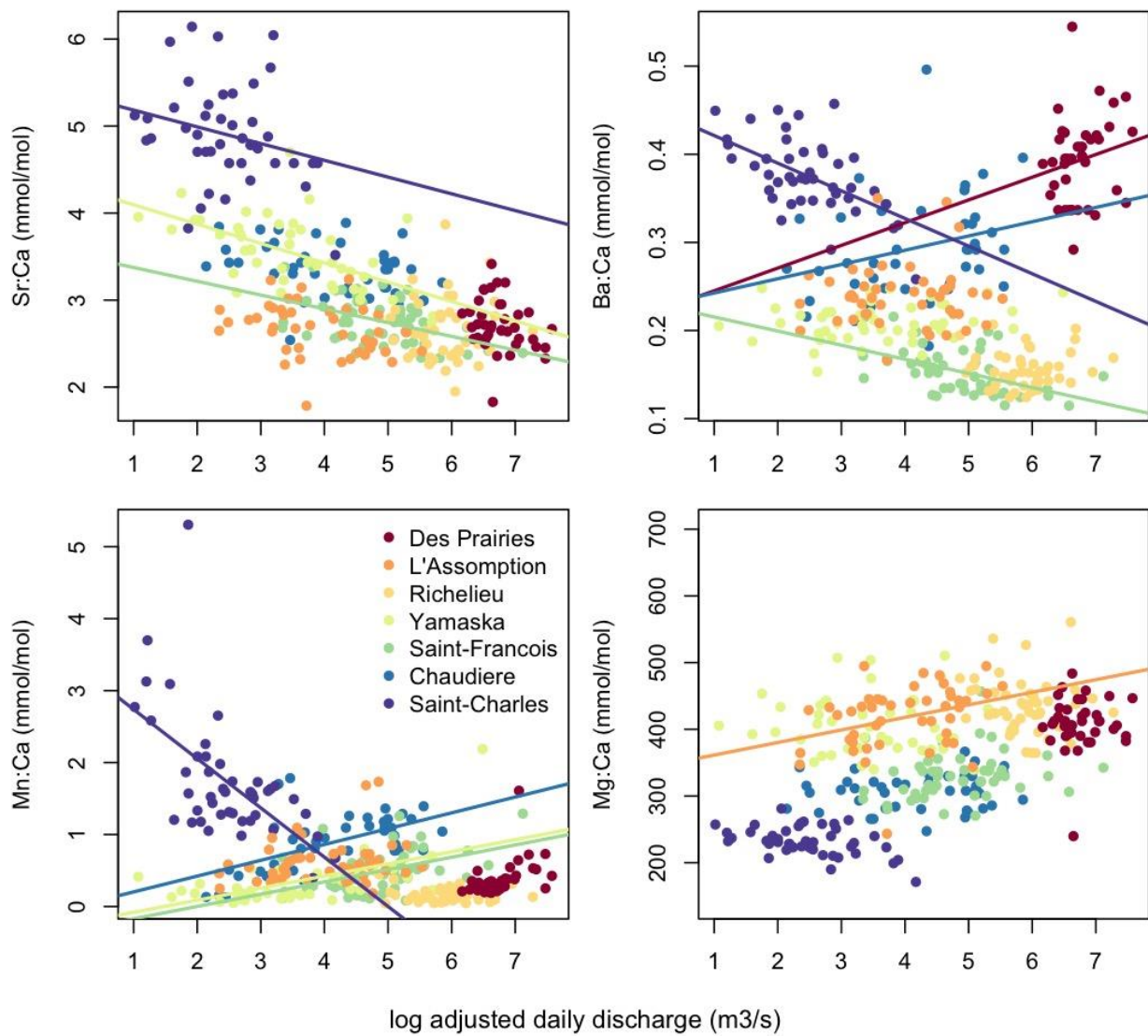
579

580 **Fig. 2.** Average concentrations of Sr:Ca, Ba:Ca, Mn:Ca and Mg:Ca ($\text{mmol}\cdot\text{mol}^{-1}$) from the ten
 581 study sites. In the figure, sites are organized upstream to downstream (from left to right) along the
 582 St. Lawrence River. The Fleuve St. Laurent (FSL) station corresponds to the combined North,
 583 South and Center stations of the St. Thérèse and Tracy Island sites. Letters show significant post
 584 hoc differences between sites.



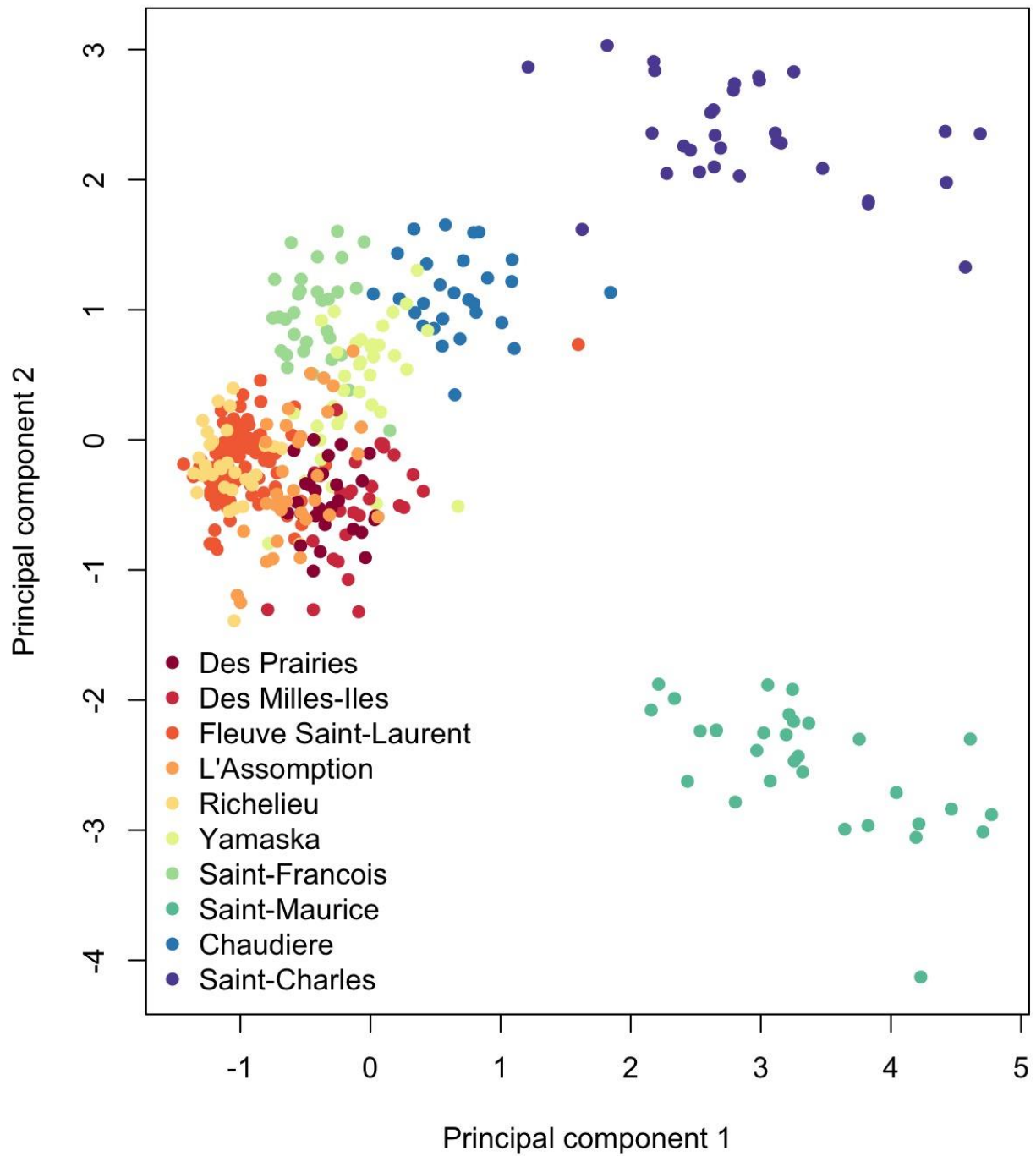
585

586 **Fig. 3.** Annual variations in mean concentrations of Sr:Ca, Ba:Ca, Mn:Ca and Mg:Ca
 587 (mmol·mol⁻¹) for the ten study sites. Each color represents a tributary along the downstream
 588 (warmer colors) to upstream (cooler colors) gradient.



589

590 **Fig. 4.** Regressions of measured concentrations of Sr:Ca, Ba:Ca, Mn:Ca and Mg:Ca in relation to
 591 the adjusted discharge (natural logarithm). Regression lines are displayed for significant
 592 relationships only, according to linear mixed-effect modeling.



593

594 **Fig. 5.** Principal component analysis ordination of the ten sample sites, according to the elemental
 595 fingerprint of Sr:Ca, Ba:Ca, Mn:Ca and Mg:Ca.

596