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Freshwater mussels as a tool for reconstructing climate history

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24 **Abstract**

25

26 Sclerochronology provides an important and widely used tool to examine annual environmental
27 patterns in marine systems, but few similar tools have been developed to establish ecological
28 indicators in freshwaters on a seasonal scale. Previous work using marine mussels as ecological
29 indicators have practiced shell ashing, acetate peels, and thin sectioning, all of which destroy the
30 specimens. We studied the external annual rings of three freshwater mussel species with clear
31 annual bands from the River Medway, UK, in order to non-invasively investigate the relationship
32 between banding patterns, growth parameters, location, and seasonal water temperatures. We
33 tested the accuracy and precision of this method through repetition of measurements and the
34 reproduction of results through separate agers. Overall, mussels living downstream had higher
35 length-at-age curves, asymptotic lengths (L_{∞}), and growth constants (k) than those living
36 upstream. In *Unio pictorum* and *Unio tumidus*, there was a negative relationship between
37 asymptotic lengths and growth constants, indicating that larger asymptotic lengths took more
38 time to reach. *Unio tumidus* had the highest asymptotic lengths and the lowest growth constants
39 across sites while *Unio pictorum* had the lowest asymptotic lengths and the highest growth
40 constants across sites. *Anodonta anatina* showed simultaneously increasing asymptotic lengths
41 with increasing growth constants and had the highest overall lengths of all three species. Summer
42 water temperatures had the largest positive effect on annual growth, followed by spring and
43 winter water temperatures, while autumnal temperatures had a negative impact on growth. The
44 findings of this study suggest that through their annual banding, freshwater mussels can be used
45 as tools to document historical water temperatures, especially in the warmer months of spring
46 and summer, and can therefore serve as powerful indicators of spatial and temporal patterns in
47 water temperature. Such tools can help infer palaeoclimatic conditions from fossil and subfossil

48 shells, establish baselines for understanding future climatic change, and support conservation
49 efforts aimed at protecting temperature-sensitive taxa.

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51 **Keywords**

52 Ecological indicator; Sclerochronology; Unionidae; Climate change; Freshwater ecosystems;
53 Seasonality

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72 **1. Introduction**

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74 Freshwater ecosystems are facing a ‘global crisis’ (Strayer and Dudgeon, 2010) as a result of
75 increasing human demands (i.e. irrigation, potable supplies, aquaculture, industrial uses, and
76 sanitation) combined with greater fragmentation, impoundment, pollution, introduction of non-
77 native species and effects of climate change (Johnson and Evans, 1990; Vörösmarty et al., 2000;
78 Strayer and Dudgeon, 2010). Climate change, in particular, affects freshwaters species more than
79 marine or terrestrial species since freshwater temperatures are highly dependent on air
80 temperatures, and habitats are often fragmented, limiting the ability of species to move as the
81 environment changes (Jenkins, 2003; Woodward et al., 2010). It is therefore essential that we
82 develop effective tools for monitoring freshwater ecosystems, identifying habitat degradation
83 and species decline, and quantifying the effects of rehabilitation programmes. Currently, we are
84 lacking sufficient freshwater temperature records, and there is a need to establish proxies to
85 reconstruct past temperature fluctuations and to set baselines against which future change can be
86 compared. These proxies can be used to assess freshwater quality and then potentially create
87 predictive models (Alizadeh, et al., 2017; Wang, et al., 2014). This will allow us to understand
88 how freshwater communities have responded to climate change in the past and how they might
89 be affected in the future.

90 Common aquatic animals used as environmental proxies include corals, brachiopods, fish,
91 and bivalves (Hudson et al., 1976; Helama et al., 2006; Scourse et al., 2006). Freshwater mussels
92 (*Bivalvia*: *Unionida*) have been identified as effective indicator taxa whose abundance is often
93 positively correlated with associated biodiversity (Aldridge et al., 2007; Chowdhury et al., 2016).
94 Their capacity to filter large volumes of water (up to 40 litres per individual per day (Tankersley

95 and Dimock Jr., 1993)) means that their growth patterns are likely to capture information about
96 their surrounding environment.

97 Annual growth bands of living organisms have been used as ecological indicators in taxa as
98 diverse as trees, fishes and corals. Studies have encompassed terrestrial, marine and freshwater
99 ecosystems and covered geographies including polar, temperate and tropical regions. The
100 greatest attention has focused on tree rings as climatic proxies. By relating spatial variation in
101 annual tree rings to anomalies in annual climate records, dendrochronology can provide a clear
102 picture of recent (Douglass, 1941) and palaeoclimatic (LaMarche, 1974) patterns over long
103 periods of time (Cook et al., 1995). Proxies have also been created for environmental and
104 climatic records of aquatic systems using a similar technique called sclerochronology. The
105 annual increments in fish otoliths and scales, coral skeletons, and bivalve shells can be measured
106 and cross-dated to provide information on factors such as climate, hydrology, pollution,
107 reproduction and migration patterns (Helama et al., 2006; Rypel et al., 2008; Helmle and Dodge,
108 2011).

109 Annual bands on bivalve mollusc shells have become widely used as a tool for ageing and
110 reconstructing population structure. Rings are counted from the shell margin towards the umbo,
111 linking them with growth years (Lefevre and Curtis, 1912; Jones, 1983). Though genetic factors
112 can affect early growth increments of molluscs, external factors mainly drive the visible, annual
113 growth banding on mussel shells (Coker et al., 1921; Marchitto, Jr. et al., 2000; Gooding et al.,
114 2009). Thus, growth bands are good indicators of environmental parameters including
115 temperature (Davenport, 1938; Kennish and Olsson, 1975; Helama et al., 2006), salinity
116 (Navarro, 1988; Marsden and Pilkington, 1996), nutrients (Ansell, 1968; Schöne et al., 2003) and
117 dissolved oxygen (Goodwin et al., 2001). Molluscs develop these dark bands due to a
118 discontinuity in growth and a withdrawing of the mantle during stress or large disturbances (such

119 as colder temperatures each winter). When molluscs start to grow again, they do not start exactly
120 where they left off, but a little distance back, resulting in a slight overlap in the outer layers of
121 the shell which forms a dark band.

122 Annual banding of marine bivalves, specifically *Arctica islandica* and *Spisula solidissima*,
123 has widely been studied for the purpose of sclerochronology and cross-matching (Jones, 1983;
124 Marchitto, Jr. et al., 2000; Schöne et al., 2002; Scourse et al., 2006), whereas very few studies
125 have looked at the potential of freshwater bivalve banding and growth patterns as environmental
126 indicators. Freshwater mussels represent an attractive tool for environmental monitoring due to
127 their clear annual rings (Neves and Moyer, 1988; Haag and Commens-Carson, 2008; Rypel et
128 al., 2008), longevity (Aldridge, 1999), wide geographical distribution (Lopes-Lima et al., 2016),
129 high abundance and accessibility for sampling (Strayer and Smith, 2003), and opportunity for
130 non-destructive ageing of live specimens (Aldridge, 1999; Helama and Nielsen, 2008; Rypel et
131 al., 2008).

132 Freshwater mussels have been effectively used to interpret a variety ecological and
133 hydrological patterns, but to date have not been used to investigate patterns in seasonal climate
134 history. For example, European unionid mussels have been used to study the effects of habitat,
135 hydrology, and sediment composition on shell morphology (Zieritz and Aldridge, 2009), to
136 determine the effect of genetic differentiation on ecophenotypic differences (Zieritz et al., 2010),
137 to compare shell morphology and population dynamics in a sympatric population (Aldridge,
138 1999), and to compare the density of freshwater mussels to overall taxon richness (Aldridge et
139 al., 2007).

140 Studies of growth rates in freshwater mussels of the Illinois River, USA, demonstrate a
141 general increasing trend over the past 150 years which has been attributed to impoundment,
142 eutrophication, water quality improvements, and warming water temperatures (Fritts et al.,

143 2017). However, it is likely that local habitat effects and species-specific responses have
144 additional effects on growth patterns. Furthermore, it can be expected that environmental effects
145 will have a differing influence on growth rates depending on the time of the year, which may
146 render seasonal pattern studies more informative than annual studies. For example, a cold winter
147 or hot summer can suppress bivalve growth, but a warmer spring can facilitate growth (Killam
148 and Clapham, 2018).

149 In this study, we investigated the utility of growth bands as climatic indicators in three
150 widespread European freshwater mussel species, *Unio pictorum* (Linnaeus), *Unio tumidus*
151 (Philipsson) and *Anodonta anatina* (Linnaeus). These mussels were chosen based on their
152 International Union for the Conservation of Nature (IUCN) status of “least concern” (IUCN,
153 2018), their abundance and widespread distribution in the River Medway, and their clear annual
154 growth rings. Other species native to the UK and present in the River Medway were less suited
155 for our study; *Pseudanodonta complanata* (Rossmässler), is of conservation concern (IUCN,
156 2018), while *Anodonta cygnea* (Linnaeus) is rare and localised within the River Medway
157 (Aldridge, 2009).

158 Most previous work on bivalve growth rings have resorted to sectioning shells or sectioning
159 hinge ligaments (Neves and Moyer, 1998; Cerrato, 2000; Goodwin et al., 2001). These
160 approaches are necessary for species that are slow-growing, long-lived or subject to erosion of
161 the external shell. However, the external rings of unionid mussels are typically clear and intact,
162 allowing us to determine age without killing or damaging the specimens. Furthermore, the three
163 species that we used have been successfully aged by their external annuli in previous studies due
164 to their fast growth, short lives, and little umbonal erosion (Aldridge, 1999; Zieritz and Aldridge,
165 2009).

166 Shell annulus patterns along the length of a river were investigated to understand
167 interspecific and habitat-induced effects. It was anticipated that growth rates in these filter
168 feeders would increase towards the lower reaches of the river continuum due to increased
169 temperature and greater abundance of particulate organic matter and algae (Vannote et al., 1980).
170 The three objectives of our study were: (1) to compare mussel growth banding between different
171 species and different locations along the river; (2) to assess the relationship between band
172 spacing and annual water temperatures; and (3) to investigate the effects of seasonal climate on
173 mussel growth, and therefore establish the use of mussel bands to document past climate
174 changes.

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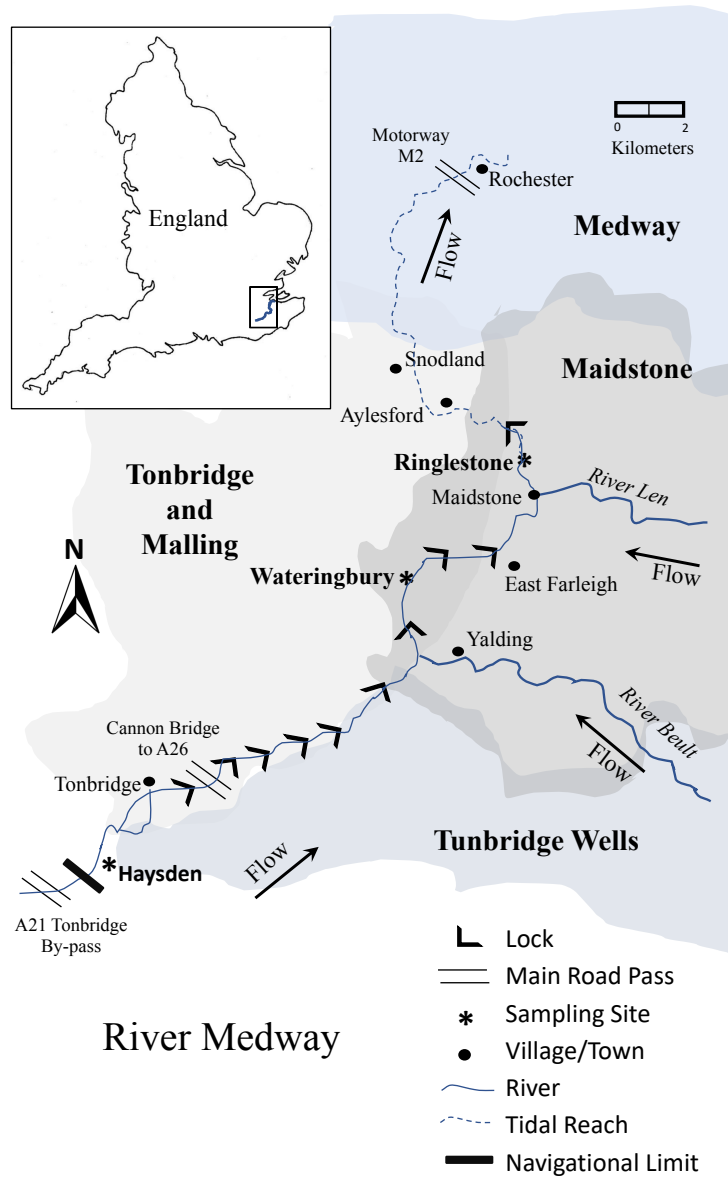
176 **2. Material and Methods**

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178 *2.1 Study site*

179

180 Studies were conducted at three sites within a 27 km navigable reach of the River Medway,
181 Kent, UK (Fig. 1), encompassing the entire range of abundant mussel populations within the
182 river. The most upstream site was located in Haysden (51.19° N, 0.25° E) at the upper
183 navigational limit of the river. The mid site was located 17 km downstream in Wateringbury
184 (51.25° N, 0.42° E) and the lowest site was located in Ringlestone (51.27° N, 0.69° E),
185 approximately 1 km upstream from the river's tidal sluice. The study reach is a highly regulated
186 lowland river, comprising 10 locks and a benthos of gravels and marginal silt. Maximum width
187 ranged from 15 m to 30 m and depth from <2 m to <4 m. Littoral vegetation was dominated by
188 *Nuphar lutea* L. The study reach is unusual in supporting all five UK unionid mussel species: *P.*
189 *complanata*, *A. cygnea*, *A. anatina*, *U. pictorum* and *U. tumidus* (Aldridge, 2009).



190

Fig. 1. Distribution of the three sampling locations on the River Medway, Kent, UK. The four surrounding counties in Kent are listed in bold.

191

192 *2.2 Mussel collection and measurement of shell growth rings*

193

194 In October 2017, approximately 30 individuals from each of three species of mussels were
 195 collected from each site by hand sampling. Studies focused on the three most numerically

196 abundant and widespread species in the river (Aldridge, 2009): *A. anatina*, *U. pictorum* and *U.*

197 *tumidus*. Annual bands, identified as continuous dark bands around the entire shell (Aldridge,
198 1999), were measured from each mussel using a Vernier caliper.

199

200 2.3 Standardisation and data analysis

201

202 Individual growth parameters were determined using the Walford plot model (Walford, 1946;
203 Aldridge, 1999), which uses a linear regression common to de-trending individual growth series
204 (Helama and Nielsen, 2008). Compared to other methods of fitting growth curves, such as the
205 logistic or the Gompertz, this model is convenient for large numbers of individuals (Walford,
206 1946) and assumes von Bertalanffy growth curves from the equation:

207

$$208 \quad L_{t+m} = cL_t + d \quad (1)$$

209

210 where L_t is length of shell at time t , L_{t+m} is length of shell at time $t + m$; m is the interval at which
211 measurements were taken (one year in this study, $m=1$), c is the slope of the Walford plot line,
212 and d is the y-intercept of the Walford plot line. The asymptotic length (L_∞) and the growth
213 constant (k) were then calculated for individuals of all three species at all three sites using
214 equations 2 and 3. The theoretical maximum growth (L_∞) of a mussel if allowed to grow to its
215 full potential is determined by the equation:

216

$$217 \quad L_\infty = d/(1-c) \quad (2)$$

218

219 where L_∞ is the asymptotic length, or the y-intercept of the 45° line. The rate at which the
220 asymptotic length is reached is determined by the equation:

221

$$222 \quad k = -\ln(c)/m \quad (3)$$

223

224 where k is the growth constant.

225 To account for declining growth increments as mussels age and to plot individual mussel

226 growth trends, the natural logarithm of all the raw growth data was taken (Scourse et al., 2006).

227 Without doing so, the growth data for the most recent growth years would have smaller residuals

228 since the majority of our mussels were at an age where growth had already slowed down

229 significantly. This would have displayed a gradual slow of annual mussel growth over time due

230 to ageing, not changing water parameters. Using similar techniques as is used in tree ring width

231 series, each mussel was plotted as the deviation from a standardised index. The ratios of annual

232 actual-to-expected ring width yields a set of dimensionless ring ‘indices’ with a defined mean of

233 1.0 and an approximately homogeneous variance (Fritts, 1976; Cook et al., 1995). Rather than

234 averaging raw growth data, this method allows us to understand good *versus* bad years for

235 growth since it displays each annulus as a positive or negative deviation from the mean index.

236 This method accounts for the fact that mussels, like trees, grow at a faster rate when young. The

237 index of each mussel was calculated as:

238

$$239 \quad I_t = R_t/G_t \quad (4)$$

240

241 where R_t is the ring width measured, G_t is the expected ring width based on the slope and y-

242 intercept of its Walford plot, and I_t is the resultant growth ring standardised index with a mean of

243 1.0. This analysis equalised the growth variations regardless of age or size. By plotting the

244 indices of each mussel’s growth increments, a unique mean standardised growth index was

245 created for each species at all three sites. This ensured differences in growth trends between
246 location and species were documented.

247

248 *2.4 Temperature*

249

250 Since seasonal water temperatures are rarely recorded over long time scales, little research exists
251 on how seasonal temperature changes affect mussel growth. Using logistic modelling based on
252 seasonal air temperatures recorded from the Environment Agency (EA) and Meteorological
253 Office (Met Office), we filled in missing water temperature data to perform seasonal growth
254 analyses. In order to isolate the effects of temperature from the various other factors affecting
255 growth, we also included sex, location, and species in our analyses. Although there are many
256 other driving factors, literature reviews and previous research led us to conclude that temperature
257 is the main driving factor for most other water parameters and is therefore the factor that would
258 supply the most valuable information (Domisch et al., 2013; Mulholland et al., 1997; Woodward
259 et al., 2010).

260 Regular long-term data on water temperature within the River Medway was not available
261 (hence, the need to establish proxies), and since we started our study in 2017 and used annuli
262 dating back to 2009, we did not have the ability to record all the necessary water temperatures
263 directly. However, water temperature for shallow rivers such as the Medway have been shown to
264 closely track air temperature, especially when averaged across seasons (Webb and Nobilis,
265 1999). Daily air temperature records from 2009 to 2016 were acquired from the UK
266 Meteorological Office for the central England district (Met Office, 2018) along with irregularly
267 recorded water temperature data for the same period from the UK Environment Agency

268 (Environment Agency, 2018) at six monitoring stations identified as being closest to the study
269 sites. Water temperature was related to air temperature using a logistic model of the form:

270

$$271 \quad T_w = \frac{\alpha}{1 + e^{\gamma(\beta - T_a)}} \quad (5)$$

272

273 where T_w is the water temperature, T_a the air temperature, α the estimate of the highest water
274 temperature, β the air temperature at the inflection point and γ the steepest slope of the logistic
275 function (see Benyahya et al., 2007). The coefficients α , β and γ were unknown and were
276 calculated using numerical optimization. We followed the methods of Webb et al. (2003),
277 whereby the root mean square error (RMSE) between the observed and calculated water
278 temperatures was minimised. RMSE measures the predicted values minus the observed values in
279 a quadratic sense and can be impacted by outliers (Janssen and Heuberger 1995). Numerical
280 optimization was undertaken using the ‘optim’ function in R (R Core Team, 2017) using the
281 default Nelder and Mead (1965) method. For each site, 80% of the water temperature data was
282 used to train the model. The remaining 20% used to examine model fit, the accuracy of the
283 coefficients and transferability, again using RMSE. All logistic models passed the RMSE
284 validation, each with a difference between validation RMSE and training RMSE within ± 0.4 .
285 This allowed daily predicted water temperatures to be calculated in the Medway, which were
286 summarized for each year and for each season: winter (December – February), spring (March –
287 May), summer (June – August) and autumn (September – November). Furthermore, recent
288 literature suggests that the incorporation of both observed and predicted values into the
289 forecasting model improves performance (Alizadeh et al., 2017). Soft-computing techniques
290 using logistic models are becoming more popular in environmental management research (e.g.
291 Olyaie, et al., 2015).

292 Given that mussels primarily grow between mid-spring and mid-autumn and cease growth
293 during winter (which creates the dark annual band), annual mussel growth measured by the
294 length of each band corresponds to the temperature of the previous year's spring, summer, and
295 autumn and the current year's winter. Therefore, spring, summer, and autumn temperatures were
296 shifted up a year to connect seasonal mean temperatures with their corresponding annual growth
297 indices. For example, a growth ring created in 2012 will demarcate the outer band edge of
298 growth and temperature information for spring 2011, summer 2011, autumn 2011, and winter
299 2011 (December) going into winter 2012 (January and February).

300

301 *2.5 Statistical analyses*

302

303 All statistical analyses were implemented using R programming language (R Core Team, 2017).
304 Analysis of variance (ANOVA) tests were used to determine if length-at-age curves, asymptotic
305 lengths (L_{∞}), or growth constants (k) differed between sites and species. Final models were
306 created after examining Pearson plots of the residuals versus fitted values and Cook's distance.
307 Inspection of the normal Q-Q plots and histograms of the Pearson residuals did not show
308 deviations from homoscedasticity or normality. Where significant differences were identified
309 from the ANOVAs, *post hoc* Tukey's Honestly Significant Difference (HSD) tests were run
310 using the 'multcomp' package (Hothorn et al., 2008) to determine between group differences.

311 A linear mixed-effects model (LME) was employed to analyse the relationship between
312 standardised mussel growth indices, site, species, and antecedent seasonal temperature. Within
313 the model, seasonal temperature, site, and species (without interaction terms) were entered as
314 fixed effects and individual mussel identity as a random effect. Inspection of residual plots did
315 not reveal deviations from homoscedasticity or normality. The likelihood ratio test was used as a

316 means to attain p-values for each variable in the linear mixed-effects model. This was conducted
317 by withdrawing each variable individually and running an ANOVA between the null model and
318 each new model excluding one variable. This allowed us to understand the effect of each variable
319 individually. The likelihood ratio test also helped us to determine which variables to include in
320 the final model through Chi-squared distributions and Akaike information criterion (AIC). The
321 LME model was created using the ‘lme4’ package (Bates et al., 2015) and Tukey HSD tests were
322 again run using the ‘multcomp’ package (Hothorn et al., 2008). All significance was assessed at
323 $P < 0.05$.

324

325 **3. Results**

326

327 *3.1 Length-at-age relationships*

328

329 Length-at-age curves differed significantly between sites and between species (Table 1a).
330 Overall, mussels showed significantly larger annual lengths at Ringlestone than at Haysden or
331 Wateringbury. Lengths increased moving downstream from Haysden to Wateringbury to
332 Ringlestone. *A. anatina* had significantly higher lengths than either *U. tumidus* or *U. pictorum*,
333 and *U. pictorum* had significantly higher lengths than *U. tumidus* (Fig. 2 and Table 1a).

334 Overall, asymptotic length (L_{∞}) differed significantly between sites with largest
335 asymptotic lengths downstream at Ringlestone and smallest lengths at Wateringbury, and
336 between species with significantly larger L_{∞} in *U. tumidus* and *A. anatina* compared to *U.*
337 *pictorum* (Fig. 3a and Table 1b). Growth constant (k) also differed significantly between sites
338 with higher k in Ringlestone and in Wateringbury compared to in Haysden, and between species
339 with higher k in *U. pictorum* and *A. anatina* than in *U. tumidus* (Fig. 3b and Table 1c).

340

341 *3.2 Standardised growth for different years and seasons*

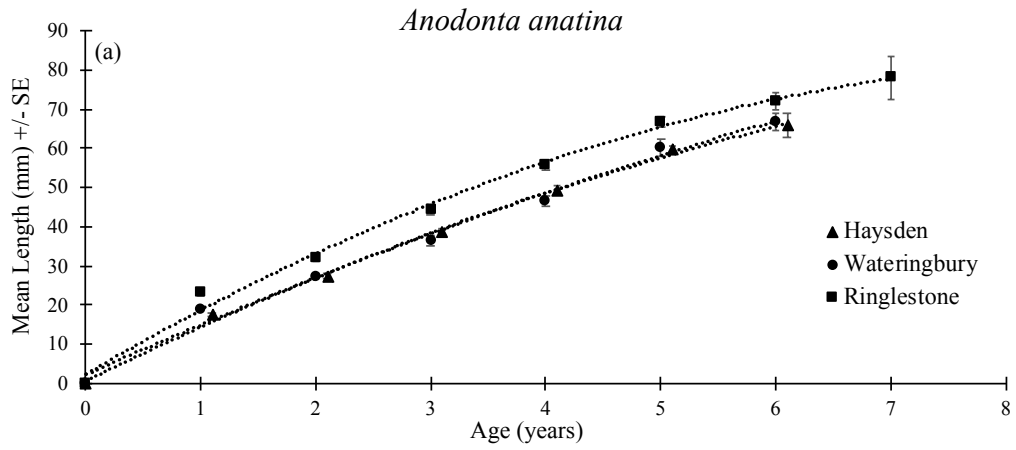
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343 The standardised growth indices showed that growth rates were relatively low across all sites and
344 species in 2012 and 2013, and relatively high in 2015 and 2016 (Fig. 4). There was a strong
345 correlation between annual growth patterns in *A. anatina* and *U. pictorum*, but less similarity
346 between these species and *U. tumidus* (Table 2), suggesting that not all species respond to
347 temperature changes in the same way. At most sites, *A. anatina* and *U. pictorum* grew well in
348 2011 whereas this was a relatively poor growth year for *U. tumidus*. Conversely, 2014 was a
349 relatively good growth year for *U. tumidus* but a below average year for *A. anatina* and *U.*
350 *pictorum*.

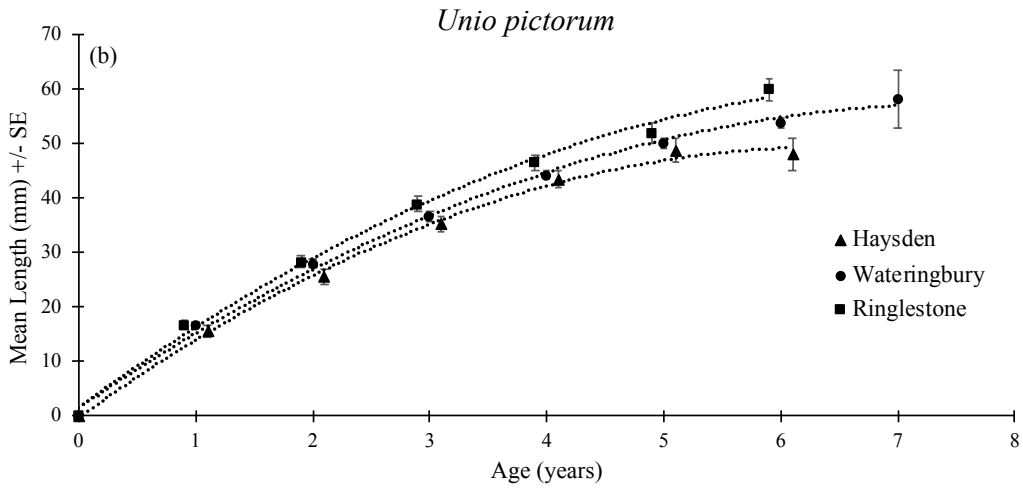
351 The linear mixed-effects model (LME) of mussel growth as a function of site (Table 3)
352 showed significantly lower growth at both Ringlestone and Wateringbury compared to at
353 Haysden. Annual growth indices were significantly higher in *U. tumidus* compared to *U.*
354 *pictorum*. Additional LME models were run separating *U. tumidus* from the other species due to
355 its unique standardised growth trends, but all species showed similar trends in growth as a
356 function of seasonal temperatures and site. Seasonal water temperature was a significant
357 predictor of mussel growth in all species (Table 3 and Fig. 5); warmer water temperatures in the
358 winter, spring, and summer were associated with increased mussel growth, while warmer
359 autumns decreased growth. Summer temperatures had the biggest positive effect on growth,
360 followed by spring, then winter, and finally autumn.

361

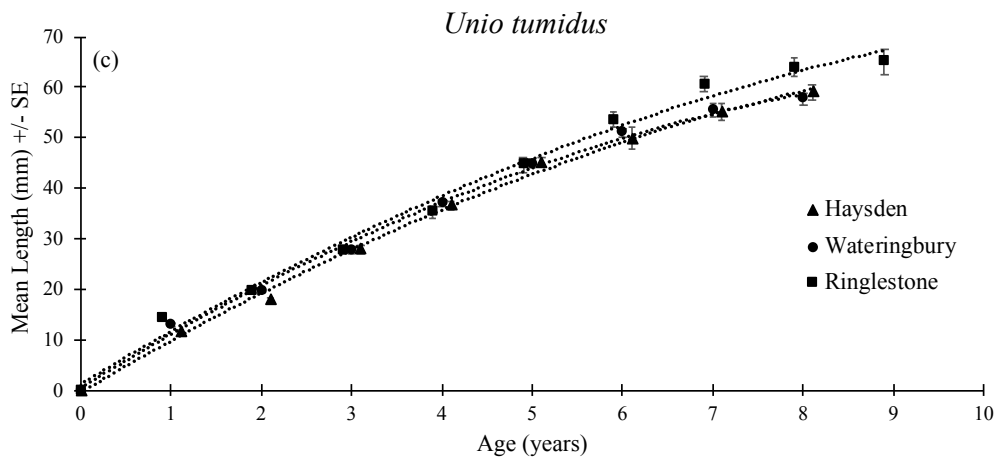
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364



365 **Fig. 2.** Mussel length-at-age curves from three sites on the River Medway separated by species;
 366 (a) *A. anatina*, (b) *U. pictorum*, (c) *U. tumidus*; triangles represent Haysden, circles represent
 367 Watringbury, and squares represent Ringlestone. The points are staggered for clarity. Error bars
 368 represent standard error.

369

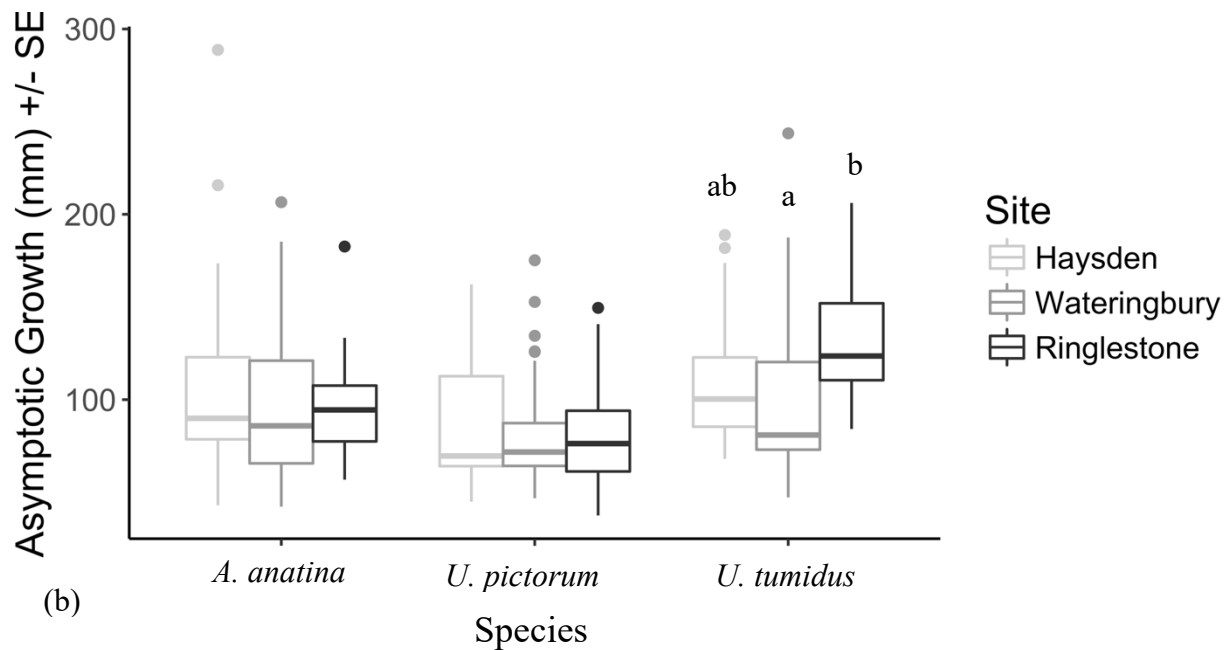
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(a) Length-at-age					
ANOVA					
Variable	d.f.	Sum Sq	Mean Sq	F-value	p-value
Site	2	9168	4584	125.10	<0.001*
Species	2	13321	6660	181.70	<0.001*
Year	1	358598	358598	9782.30	<0.001*
Residuals	1723	63162	37		
(b) L_{∞}					
Site	2	10001	5000	3.59	0.030*
Species	2	33710	16855	12.11	<0.001*
Residuals	253	352287	1392		
(c) k					
Site	2	0.135	0.0676	6.16	0.002*
Species	2	0.864	0.4321	39.34	<0.001*
Residuals	253	2.779	0.011		
(a) Length-at-age					
Tukey HSD					
Variable	Estimate	Std. Error	t-value	p-value	
Ringlestone- Haysden	3.21	0.38	8.46	<0.001*	
Watringbury - Haysden	0.67	0.36	1.87	0.15	
Watringbury- Ringlestone	-2.54	0.35	-7.20	<0.001*	
<i>U. pictorum</i> - <i>A. anatina</i>	-5.27	0.39	-13.47	<0.001*	
<i>U. tumidus</i> - <i>A. anatina</i>	-12.41	0.36	-34.31	<0.001*	
<i>U. tumidus</i> - <i>U. pictorum</i>	-7.15	0.36	-20.09	<0.001*	
(b) L_{∞}					
Ringlestone- Haysden	0.35	6.37	0.06	1	
Watringbury - Haysden	-10.79	5.63	-1.92	0.14	
Watringbury- Ringlestone	-11.14	5.79	-1.93	0.13	
<i>U. pictorum</i> - <i>A. anatina</i>	-16.95	5.90	-2.87	0.01 *	
<i>U. tumidus</i> - <i>A. anatina</i>	10.25	5.92	1.73	0.20	
<i>U. tumidus</i> - <i>U. pictorum</i>	27.20	5.58	4.88	<0.001 *	
(c) k					
Ringlestone- Haysden	0.04	0.02	2.45	0.04*	
Watringbury - Haysden	0.04	0.02	2.42	0.04*	
Watringbury- Ringlestone	-0.01	0.02	-0.34	0.94	
<i>U. pictorum</i> - <i>A. anatina</i>	0.01	0.02	0.78	0.71	
<i>U. tumidus</i> - <i>A. anatina</i>	-0.12	0.02	-6.92	<0.001*	
<i>U. tumidus</i> - <i>U. pictorum</i>	-0.13	0.02	-8.17	<0.001*	

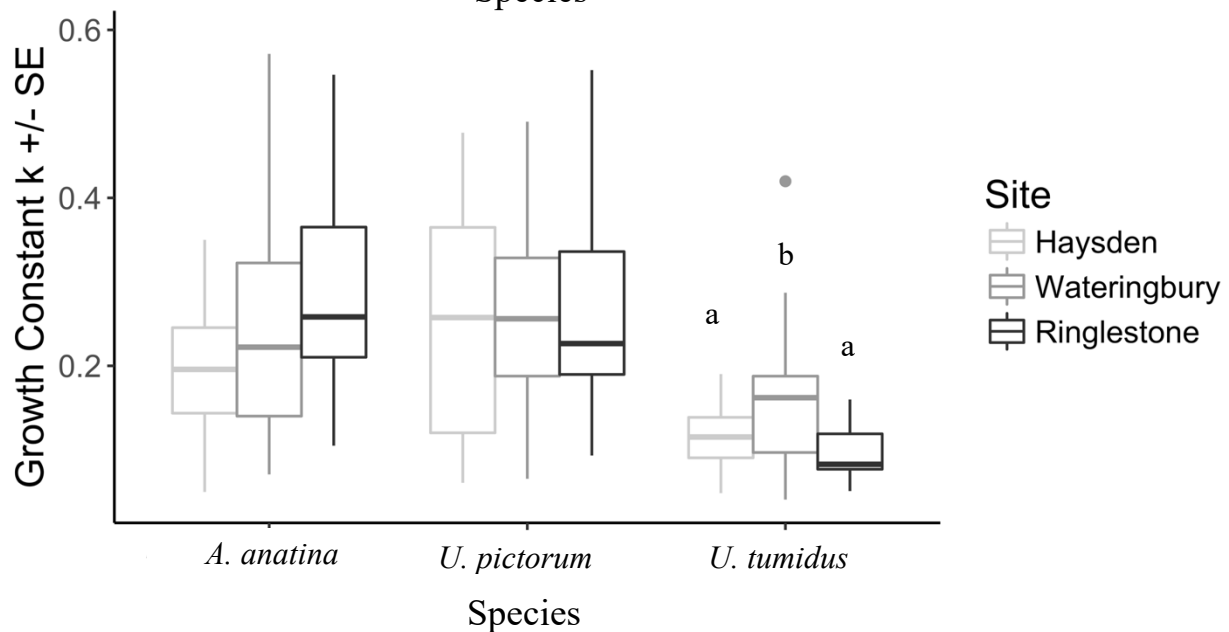
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372 **Table 1** Statistical comparisons of mussel growth parameters between sites and species within
 373 the River Medway; (a) length-at-age curves, (b) asymptotic length L_{∞} , and (c) growth constant k ;
 374 the analysis of variance (ANOVA) tests were run with mean lengths, L_{∞} and k as functions of
 375 site and species; corresponding *post hoc* Tukey HSD estimates represent the difference in
 376 significant response variables between levels; * indicates $p < 0.05$.

377 (a)

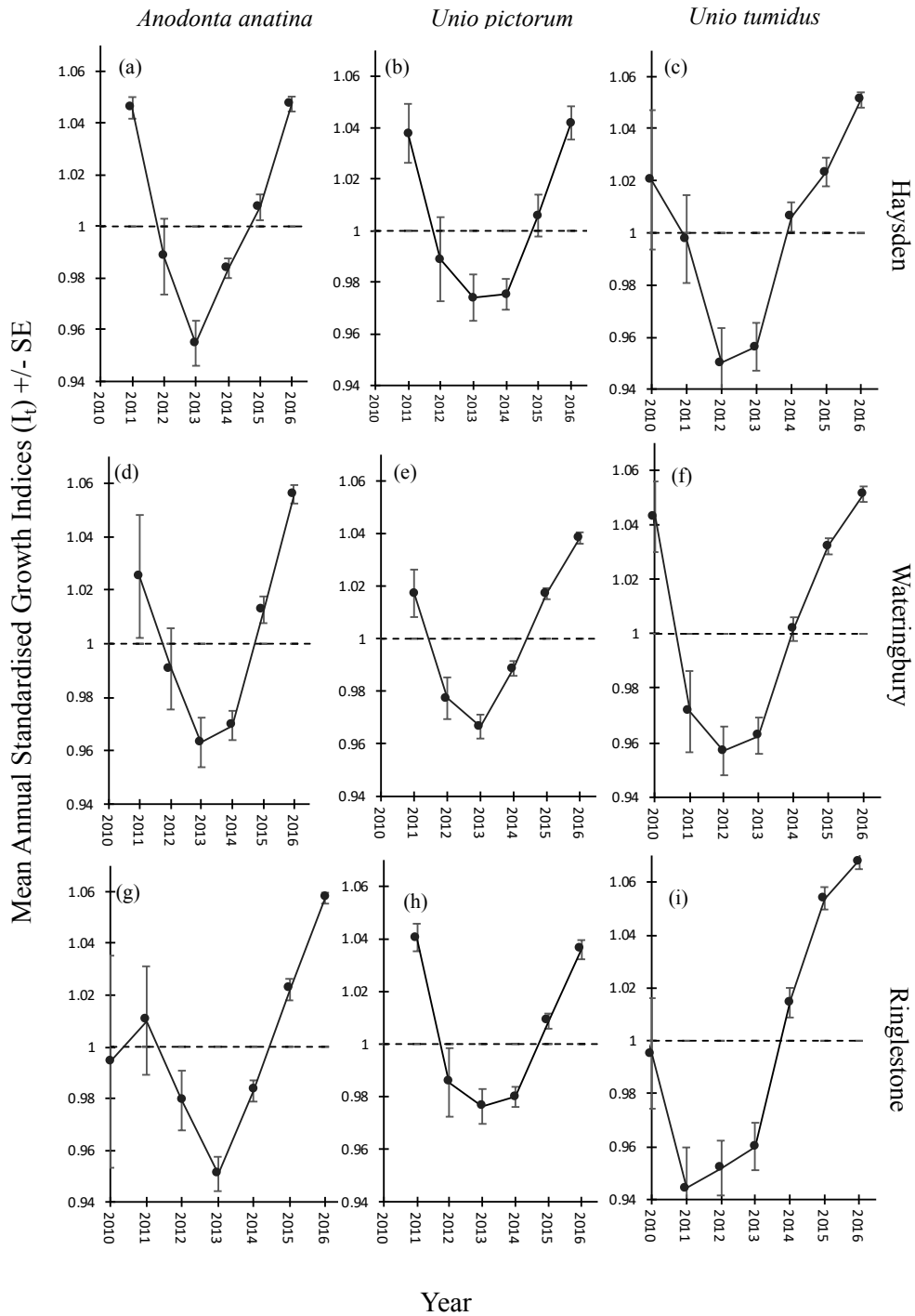


(b)



378

379 **Fig. 3.** Boxplots of the growth parameters for three *Unionidae* species found in three sites along
380 the River Medway; (a) plot of asymptotic lengths L_{∞} and (b) plot of growth constants k ; boxplots
381 annotated with different letters indicate groups that are statistically different ($P < 0.05$); boxes are
382 based on mean, 1st and 3rd quartiles, and error bars represent 95% confidence intervals; dots
383 mark outliers.



384

385 **Fig. 4.** Mean annual standardised growth indices for all records of each species at all three sites;

386 (a) *A. anatina* at Haysden, (b) *U. pictorum* at Haysden, (c) *U. tumidus* at Haysden, (d) *A. anatina*

387 at Waringbury, (e) *U. pictorum* at Waringbury, (f) *U. tumidus* at Waringbury, (g) *A. anatina*

388 at Ringlestone, (h) *U. pictorum* at Ringlestone, (i) *U. tumidus* at Ringlestone; error bars represent
 389 standard error.

390

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392

	UpHays	UpWater	UpRingle	AaHays	AaWater	AaRingle	UtHays	UtWater	UtRingle
UpHays	1.00								
UpWater	0.90	1.00							
UpRingle	0.99*	0.90	1.00						
AaHays	0.97*	0.92	0.97*	1.00					
AaWater	0.96*	0.94	0.93	0.94	1.00				
AaRingle	0.86	0.98*	0.84	0.88	0.95*	1.00			
UtHays	0.65	0.90	0.66	0.70	0.73	0.89	1.00		
UtWater	0.47	0.79	0.46	0.51	0.63	0.83	0.94*	1.00	
UtRingle	0.28	0.65	0.27	0.33	0.48	0.71	0.86	0.98*	1.00

393

394 **Table 2** Correlation matrix of all mean standardised growth indices of mussels from the River
 395 Medway; Up represents species *U. pictorum*, Aa represents *A. anatina*, Ut represents *U. tumidus*,
 396 Hays represents site Haysden, Water represents Wateringbury, and Ringle represents
 397 Ringlestone; * indicates a significant correlation after Bonferroni correction ($P \leq 0.006$).

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Variable	Linear mixed-effects model				Likelihood ratio test			
	Estimate	Std. Error	t-value	p-value	Variable	χ^2	d.f.	p-value
(Intercept)	-0.79	0.09	-8.43	<0.001*				
Summer	0.08	0.005	17.40	<0.001*	Summer	278.76	1	<0.001*
Spring	0.05	0.003	16.02	<0.001*	Spring	239.08	1	<0.001*
Winter	0.05	0.002	20.30	<0.001*	Winter	369.05	1	<0.001*
Autumn	-0.03	0.003	-11.65	<0.001*	Autumn	130.50	1	<0.001*
Ringlestone-Haysden	-0.06	0.004	-13.65	<0.001*	Site	198.05	2	<0.001*
Wateringbury-Haysden	-0.06	0.004	-13.72	<0.001*				
Wateringbury-Ringlestone	-0.0001	0.003	-0.04	1.00				
<i>U. pictorum</i> - <i>A. anatina</i>	-0.0007	0.003	-0.21	0.98	Species	8.51	2	0.01*
<i>U. tumidus</i> - <i>A. anatina</i>	0.007	0.003	2.29	0.06				
<i>U. tumidus</i> - <i>U. pictorum</i>	0.007	0.003	2.57	0.03*				

405

406 **Table 3** Results from linear mixed-effects model (LME) and likelihood ratio test for the effects

407 of seasonal temperature, site, and species on annual mussel growth in the River Medway; the

408 LME was run with standardized annual mussel growth as the response variable versus all

409 analysed variables as fixed effects: seasonal temperatures, site, and species; estimates for

410 continuous variables (i.e. seasonal temperatures) represent the difference in standardised growth

411 indices as temperatures rise; estimates for discrete variables (i.e. site and species) represent the

412 difference in standardised growth indices between two levels; the likelihood ratio test was run

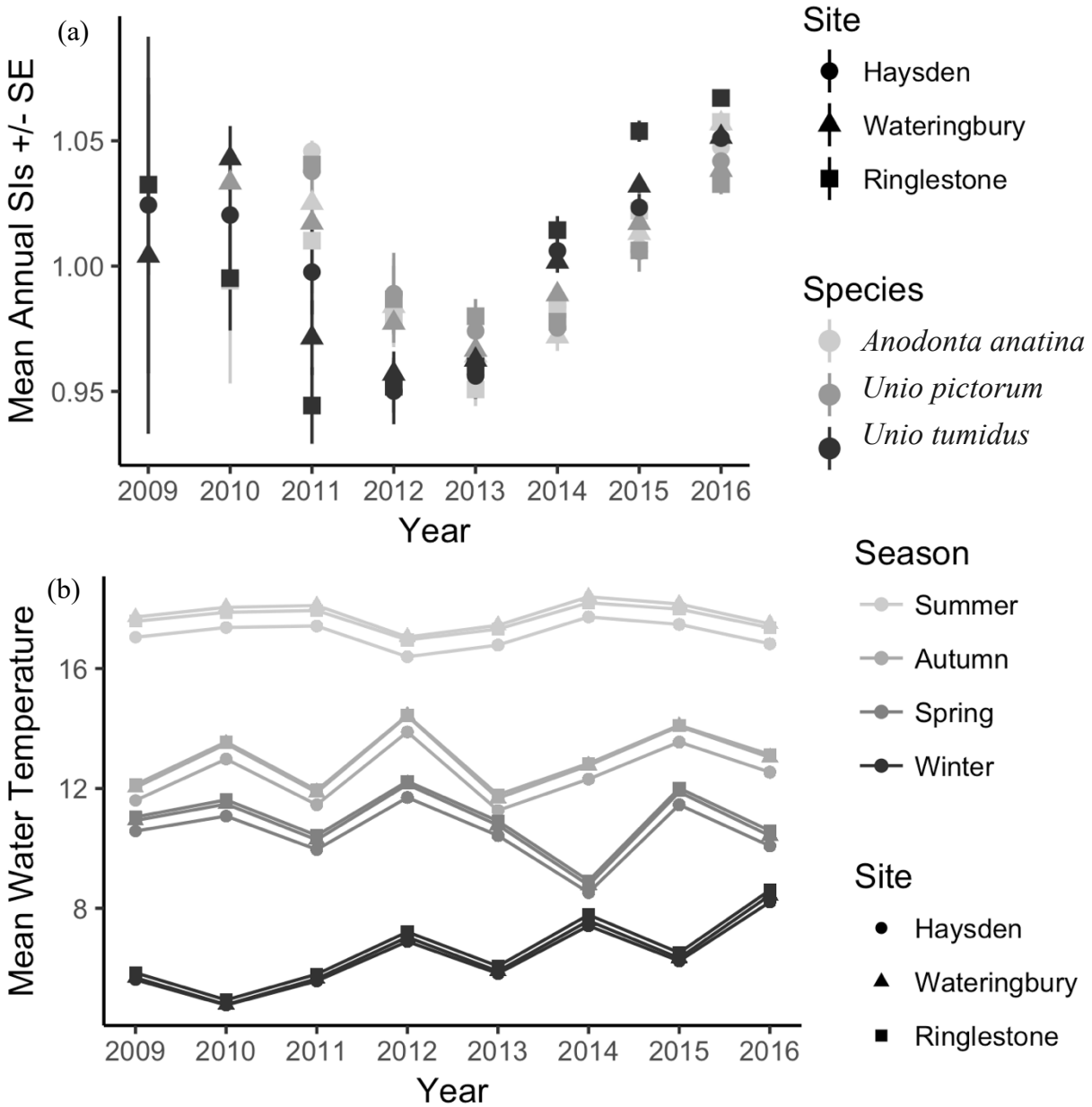
413 with standardised annual mussel growth as a function of seasonal temperatures, site, and species;

414 * indicates $P < 0.05$.

415

416

417



418

419 **Fig. 5.** Mean annual standardised mussel growth indices (SIs) compared to the corresponding
 420 mean seasonal water temperatures in the River Medway; (a) standardised growth indices for all
 421 three *Unionidae* species at all three sites over a period of eight years; (b) seasonal water
 422 temperatures for each site across the same time period; error bars represent standard error.

423

424

425 4. Discussion

426

427 Our data demonstrate that growth banding of freshwater mussels can serve as a powerful
428 indicator of historic water temperatures. In the River Medway, downstream populations were
429 typically larger age-to-age. *U. pictorum* and *A. anatina* shared similar years for above average
430 and below average growth, while *U. tumidus* responded differently. Warmer winters, springs and
431 summers were correlated with wider growth bands and warmer autumns were correlated with
432 narrower growth bands.

433

434 4.1 Length-at-age patterns

435

436 All three species of mussel had lowest length-at-age growth trends at the upstream river site,
437 Haysden, and higher trends downstream. These patterns were further supported by our finding
438 that downstream mussels had higher asymptotic lengths, indicating that these individuals have
439 the potential to reach larger sizes. Mussels also showed larger growth constants downstream, but
440 these statistics were mainly driven by *A. anatina* from Ringlestone. The relationship between
441 growth constant and the asymptotic length is usually negative so that larger growth constants
442 often correlate with lower asymptotic lengths and indicate that the asymptotic length is reached
443 at a faster rate (Bauer, 1992; Aldridge, 1999). Both *U. pictorum* and *U. tumidus* had lowest
444 growth constants in Ringlestone, meaning they would reach their asymptotic lengths at a slower
445 rate. This could be because all mussels in Ringlestone had greater asymptotic lengths and would
446 therefore need more time to reach their maxima. *U. tumidus* had the highest asymptotic length
447 and the lowest growth constant overall, whereas *U. pictorum* had the lowest asymptotic length

448 and the highest growth constant. Thus, within these two species, a higher asymptotic growth
449 length was coupled with having a lower growth constant.

450 Elevated growth rates at downstream sites may be driven by higher algal productivity and
451 greater abundance of fine particulate organic matter typically found downstream (Vannote et al.,
452 1980; Zieritz et al., 2010; Moss, 2011). It is also possible that faster sporadic flow rates in
453 upstream locations of lowland rivers such as the Medway may impose further restrictions on
454 shell morphology that constrain growth rates (Zieritz and Aldridge, 2009). Similar downstream
455 growth patterns have been found from rivers elsewhere (e.g. mussels in the Colorado River
456 Delta, Mexico and fish in the Warta River, Poland) and have been explained by increasing
457 salinity, reduced oxygen uptake, increasing amounts of fine-grained sediments, and increasing
458 food availability downstream (Ansell, 1968; Navarro, 1988; Przybylski, 1996; Schöne et al.,
459 2003a).

460

461 *4.2 Temperature effects on growth*

462

463 Water temperature variation in the River Medway affected mussel growth in all four seasons
464 during our study period. This relationship was positive in summer, spring, and winter, meaning
465 mussel growth increased with rising water temperatures. Conversely, this relationship was
466 negative in autumn, and mussel growth declined with rising water temperatures.

467 The increase in growth with temperature from winter to summer is most likely due to the
468 fact that higher temperatures increase the metabolic activity of mussels and extend the annual
469 growing period, therefore increasing the rate of shell production (Bolotov et al., 2018). Along
470 with the direct physiological effects that water temperature has on mussel growth, there are many
471 indirect effects of temperature which may also impact growth, reproduction, and survival.

472 Previous studies have shown strong correlations between temperature and various environmental
473 conditions which drive mussel growth, many of which increase with rising temperatures; these
474 include phytoplankton abundance (Robarts and Zohary, 1987), particulate organic carbon (POC)
475 (Page and Hubbard, 1987; Kasai et al., 2010), chlorophyll *a* concentration (an indicator of
476 phytoplankton levels) (Kautsky, 1982; Kasai et al., 2010; Kasai and Hirakawa, 2015; Tanabe et
477 al., 2017), excretion of dissolved organic carbon (DOC) from phytoplankton (Zlotnik and
478 Dubinsky, 1989) and nutrient availability (nitrogen, phosphorous, and silicate) (Rhee and
479 Gotham, 1981; Kasai et al., 2010), all reaching peak levels around the same time as the main
480 growing season of mussels.

481 While warmer temperatures generally resulted in positive growth, our study also
482 documented a decline in growth at higher autumnal temperatures. Similar reduction in growth of
483 unionid mussels during elevated temperatures has been reported by Mann and Glomb (1978),
484 Mann (1979), and Bayne and Worrall (1980). The negative relationship between increasing
485 autumnal temperatures and mussel growth may result from attempting to grow (driven by an
486 increased metabolism) at a time of limited food quality and quantity, resulting in an overall net
487 loss of energy and a decrease in the ability to put on weight or increase shell length. In addition
488 to the stress of high temperatures, species *U. tumidus* and *U. pictorum* already have high
489 metabolic demands during the autumn due to the energy requirement of tissue histolysis and
490 histogenesis following their summer spawning period (Mann, 1979; Aldridge, 1999). Therefore,
491 the temperature-induced additional metabolic drive during a season of re-growth and low food
492 availability may hinder overall shell production.

493 It is important to note that an increase in growth rate does not necessarily indicate a
494 healthier population. Although increases in nutrients and phytoplankton (following temperature
495 rises) promote increased growth rates in mussels, these higher growth rates can result in reduced

496 longevity, abundance, and species richness (Haag and Rypel, 2011; Ridgway et al., 2011; Fritts
497 et al., 2017). While this study demonstrates that mussel growth banding is significantly affected
498 by seasonal water temperatures, and therefore documents valuable climate change data within
499 freshwater ecosystems, understanding the full impact of temperature-accelerated growth rates on
500 mussel communities and other freshwater biota requires further investigation. Additionally, other
501 water parameters that often differ per location, such as hydrology, and biotic factors, such as
502 population density and parasite load (Zieritz & Aldridge, 2011), were not measured in this study
503 and may have had further effects on annual growth.

504

505 *4.3 Freshwater mussels as indicators of climate patterns*

506

507 Among our study species, annual summer water temperatures followed by spring water
508 temperatures had the biggest effects on growth indices, supporting the knowledge that mussels
509 grow mainly in the warmer months of spring and summer. The significant effects of seasonal
510 temperatures on growth indices suggest that it is possible to use mussel growth increments to
511 study past freshwater climate trends, particularly those of warmer seasons, and indicate that
512 projected climate change will affect the future growth patterns of freshwater mussels.

513 By using growth rings to represent warmer or colder years, sclerochronology records can
514 be created for the UK river systems to study long-term temperature fluctuations, which may help
515 us to understand environmental changes occurring in these freshwater ecosystems. Long-lived,
516 easily identifiable species at a single site may allow for long-term chronologies to be established,
517 as is developed for marine bivalves. Since *U. tumidus* has the highest asymptotic lengths and
518 takes the longest amount of time to reach these lengths, individuals of this species could contain
519 the most rings and thus hold longer environmental records. Upon looking at individual species

520 trends, *U. tumidus* also shows the greatest separation in asymptotic length and growth constant
521 trends across sites suggesting that it might be the best species for studying water parameters
522 between locations. While annual mussel growth increments had significant relationships with
523 seasonal climate data, the species in this study only provided between three to seven growth
524 rings, limiting the amount of climate data we could match. It would be beneficial, therefore, to
525 create chronologies from longer-lived species, such as *Anodonta cygnea* (Linnaeus) (the swan
526 mussel) (Aldridge, 1999) to create more comprehensive freshwater histories. Additional use of
527 museum collections with known collection dates would allow further cross-matching of *in situ*
528 shell material which can last for decades with intact annuli (Bauer, 1992). Such studies may
529 establish robust baselines against which future changes in growth rates can be compared, and
530 thus may serve as an early warning of impending changes to mussel populations. This idea has
531 already been utilised in dendrochronology; a study performed in Iran used tree-rings and an
532 artificial neural network (ANN) to create a model that simulated groundwater levels from tree-
533 ring diameter and precipitation inputs (Gholami et al., 2015). Given the ecosystem engineering
534 role of freshwater mussels, changes in mussel size or population structure could impact entire
535 freshwater communities. Documenting valuable information, their annual bands might allow us
536 to determine how whole ecosystems are responding to environmental changes.

537 Species-level and habitat-level effects can make a difference on interpretations of growth
538 when using mussels as indicators of temperature change. The annual pattern of mean
539 standardised growth indices of *A. anatina* and *U. pictorum* were highly correlated within all three
540 sites, implying that annual effects dominated site effects. However, while these annual growth
541 trends were relatively consistent down the river, they often opposed the growth trends of *U.*
542 *tumidus*. This further emphasizes the importance of species in the study of environmental
543 parameters since they may differ in habitat preferences and respond differently to environmental

544 change. Between our three species, *A. anatina* prefers sandy and gravel substrate (Bauer and
545 Wächtler, 2001), *U. pictorum* prefers firm, muddy substrate (Aldridge, 1999), and *U. tumidus*
546 prefers finely grained clay substrate (Annie et al., 2013). Zajac et al. (2016) revealed that while
547 *A. anatina* and *U. pictorum* occupied similar niches in a single lake, *U. tumidus* was the only
548 mussel species of the five studied that showed no relation to any of the measured habitat
549 variables. While all three of our study sites support diverse and abundant mussel communities,
550 they differ in physical composition. Haysden consisted mainly of gravel sediment, Wateringbury
551 was dominated by organic pellets and silt, and Ringlestone was characterised by a sticky clay
552 substrate. Ringlestone contained the most favourable substrate for *U. tumidus*, which could
553 explain why the greatest differences in *U. tumidus* growth compared to the other species
554 occurred at this site (years 2010, 2011, 2014, and 2015). Since *U. tumidus* may have greater
555 growth potential in Ringlestone (significantly higher L_{∞} at this site), drastic temperature changes
556 might have a more noticeable effect on growth.

557 Different spawning and fertilisation times between species may also have an impact on
558 their response to changing seasonal temperatures. Aldridge (1999) found mature glochidia in the
559 outer demibranchs of both *Unio* species from May to June (summer), and of *A. anatina* from
560 December to March (winter into spring). This could account for some of the varying trends in the
561 standardised growth, with particularly slowed growth associated with periods when brooding
562 coincides with especially unfavourable temperatures for growth. For example, in 2012, low
563 summer temperatures correlated with a low growth index in *U. tumidus*, and in 2014, low spring
564 temperatures correlated with a low growth index in *A. anatina*.

565 When using annuli of mussels as indicators to past climate trends, understanding the
566 competing effects of seasonality as well as species-level and habitat-level effects can help us to
567 build better chronologies based on the specific mussels being studied. Overall, using longer-lived

568 species from downstream populations could provide the strongest data on historical spring and
569 summer temperatures in river systems. Eventually, by using established relationships between
570 growth and temperature, we could cross-reference banding patterns on older, un-dated *in situ*
571 shells to create longer temperature chronologies and reconstruct historical climates.

572 While seasonal temperatures have been shown to be significant correlates of growth in
573 this study, it is possible that additional parameters may be important and some of these may be
574 directly associated with temperature. Further studies might investigate the effects on growth of
575 phytoplankton levels, POC, DOC, water flow rates, and nutrient levels such as N, P, and silicate.
576 Future work could also include the effects of sex, additional locations, and more species
577 (preferably longer living species). Patterns in additional river systems and over longer
578 timeframes could also be considered. Ultimately, it may be possible to use mussel annuli patterns
579 to reconstruct historical climatic patterns.

580

581 **5. Conclusions**

582

583 Freshwater ecosystems are highly vulnerable to climate change and are experiencing global
584 habitat loss, species declines, and population shifts (Dudgeon et al., 2006; Markovic et al., 2014).
585 Current biomonitoring schemes may be undermined by climate change since they are largely
586 based on responses to organic pollution with little consideration of the increasing influence of
587 climate (Woodward et al., 2010). Presently, we have relatively few tools available for
588 understanding temperature patterns in freshwater systems, and those that exist often require
589 destructive sampling and specialist equipment. Mussels offer an attractive tool for documenting
590 water temperatures due to their clear annuli (capable of documenting large amounts of
591 environmental information), longevity, global dispersal, accessibility, and the ability to measure

592 their growth non-destructively. The established relationship between winter, spring, and summer
593 temperatures and annuli length will aid in the use of mussels to study past freshwater
594 temperatures and to understand the impacts that climate change may have on freshwater
595 ecosystems in the future.

596

597 **Declarations of interest**

598

599 None.

600

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602

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