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Human-induced biotic invasions and changes in plankton interaction networks

1   **Human-induced biotic invasions and changes in plankton interaction networks**

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SUMMARY

1. Pervasive and accelerating changes to ecosystems due to human activities remain major sources of uncertainty in predicting the structure and dynamics of ecological communities. Understanding which biotic interactions within natural multitrophic communities are threatened or augmented by invasions of non-native species in the context of other environmental pressures is needed for effective management.

2. We used multivariate autoregressive models with detailed time-series data from largely freshwater and brackish regions of the upper San Francisco Estuary to assess the connectance, topology, direction and strength of trophic interactions following major invasions and establishment of non-native zooplankton in the early 1990s. We simultaneously compared the effects of fish and clam predation, environmental temperature, and salinity intrusion using time-series data from > 60 monitoring locations and spanning more than three decades.

3. We found changes in the networks of biotic interactions in both regions after the major zooplankton invasions. Our results imply an increased pressure on native herbivores; intensified negative interactions between herbivores and omnivores; and stronger bottom-up influence of juvenile copepods but weaker influence of phytoplankton as a resource for higher trophic levels following the invasions. We identified salinity intrusion as a primary pressure but showed relatively stronger importance of biotic interactions for understanding the dynamics of entire communities.

4. *Synthesis and applications.* Our findings highlight the dynamic nature of biotic interactions and provide evidence of how simultaneous invasions of exotic species may alter interaction networks in diverse natural ecosystems over large spatial and temporal

45   scales. Efforts to restore declining fish stocks may be in vain without fully considering

46   the trophic dynamics that limit the flow of energy to target populations. Focusing on

47   multitrophic interactions that may be threatened by invasions rather than a limited focus

48   on responses of individual species or diversity is likely to yield more effective

49   management strategies.

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51   **Key-words:** Bayesian estimation, conservation of interactions, environmental pressures,

52   estuaries, multivariate autoregressive model, pelagic food webs, time-series.

53

54   **INTRODUCTION**

55   Human perturbations of ecosystems range from modifications of the atmosphere and climate to

56   the degradation of habitats through exploitation or introductions of non-native species. These

57   changes directly affect individual species (Walther *et al.* 2002) or taxonomic composition

58   (Winder, Jassby & Mac Nally 2011; Winder & Jassby 2011), but they may also alter the relative

59   importance of biotic interactions within multitrophic networks (Hampton, Scheuerell &

60   Schindler 2006; Francis *et al.* 2012). Networks of biotic interactions are more than the collection

61   of their component species (Montoya, Pimm & Sole 2006) and human perturbations may alter

62   network structure without necessarily affecting species diversity (Tylianakis, Tscharntke &

63   Lewis 2007). Given that biotic interactions affect ecosystem dynamics (McCann, Rasmussen &

64   Umbanhowar 2005), the conservation of network structure recently has been emphasized

65   (McCann 2007; Tylianakis *et al.* 2010), alongside the more traditional foci on individual taxa,

66   community composition or diversity. Changes to interaction networks can transcend biotic

67   boundaries and disrupt ecosystem function, such as biogeochemical cycling (Atwood *et al.* 2013)

68   or distribution of toxic compounds (Stewart *et al.* 2004). Quantifying the strength of biotic

69   interactions modulated by multiple perturbations in diverse natural communities and over large

70   spatial and temporal scales has become a key challenge for community ecology and conservation

71   biology.

72   Estuaries are complex ecosystems that lie at the interface of ocean, terrestrial and

73   freshwater systems and their biotic communities are adapted to great variation in environmental

74   conditions such as oxygen, temperature and salinity (Elliott & Whitfield 2011). Global threats to

75   estuarine community structure, function and services result from the large fraction of the world's

76   human population living near estuaries (Lotze *et al.* 2006). Releases of ballast water from cargo

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77   ships may be the world's largest invasion vector that can restructure entire ecosystems (Ruiz *et*

78   *al.* 1997; Seebens, Gastner & Blasius 2013). Although estuarine communities are adapted to a

79   highly variable environment and therefore potentially resistant to certain anthropogenic pressures

80   (Elliott & Quintino 2007; Elliott & Whitfield 2011), our understanding of their resistance to

81   species invasions is rudimentary.

82   The San Francisco Estuary is a coastal ecosystem that has been much modified by habitat

83   degradation, water harvesting, increased temperature, and a high ratio of non-native to native

84   species (Cloern & Jassby 2012). This estuary, the largest on the US Pacific coast, provides many

85   ecosystem services and is one of the most biotically invaded estuaries in the world (Cohen &

86   Carlton 1998). A series of invasions over the last four decades caused sharp declines of native

87   zooplankton, shifting species composition to a community resembling that of east-Asian

88   estuaries (Orsi & Ohtsuka 1999; Winder, Jassby & Mac Nally 2011). These taxonomic shifts

89   potentially altered negative (competition, predation) and positive (facilitation) interactions within

90   and among trophic levels, with flow-on effects on food-web structure and dynamics. Previous

91   univariate analyses indicated the negative effects of the invasive bivalve *Potamocorbula*

92   *amurensis* (hereafter *Potamocorbula*) on phytoplankton, zooplankton, and several fish species

93   (Alpine & Cloern 1992; Cloern 1982; Kimmerer, Gartside & Orsi 1994; Kimmerer 2006). In

94   addition to individual zooplankton invasions, there was a major simultaneous invasion of three

95   non-native copepod species that became established in the early 1990s and that dominated the

96   zooplankton assemblage thereafter (Orsi & Ohtsuka 1999; Winder & Jassby 2011). The impact

97   of such simultaneous invasions on native assemblages is expected to be more severe than if each

98   of the species had invaded and established asynchronously (Simberloff & Von Holle 1999).

99   Despite the importance and complexity of plankton communities (Hays, Richardson &

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100   Robinson 2005), the strength of interactions in plankton networks and their changes in human-

101   altered ecosystems remain poorly understood. In the upper San Francisco Estuary, in situ

102   interactions among plankton taxa have not been explored systematically in a multivariate

103   framework. A single study quantified strengths of trophic interactions among pelagic fishes and

104   two groups of calanoid copepods (Mac Nally *et al.* 2010). However, that study did not explore

105   interaction strengths among all zooplankton and phytoplankton functional groups and included

106   only chlorophyll-*a* as a proxy for total phytoplankton biomass.

107   Here, we focus on interactions among major phytoplankton and zooplankton functional

108   groups and relate these to pelagic fish biomass, seawater temperature, salinity intrusions and

109   biotic invasions. We employed detailed monitoring data sets for pelagic fish, zooplankton and

110   phytoplankton to compare the topology, directions, and strength of trophic interactions before

111   and after the establishment of non-native zooplankton (hereafter 'pre-invasion' and 'post-

112   invasion' period respectively). We analyzed the shift in the interaction network from the pre-

113   invasion period (1974–1992) to the post-invasion period (1993–2008) and the relative influence

114   of key covariates using multivariate autoregressive (MAR) models (Ives *et al.* 2003). Our

115   specific goal was to explore whether, and how, the shifts in relative abundances of plankton taxa

116   (Winder & Jassby 2011) after the establishment of non-native copepods around 1993 altered the

117   importance of trophic interactions among the major plankton functional groups. We predicted

118   shifts in biotic interactions after the establishment of invasive zooplankton species. In particular,

119   we expected to detect stronger interactive effects of invasive predators and weaker effects of

120   native herbivores because of major abundance shifts within these functional groups (Winder &

121   Jassby 2011). We also predicted that salinity intrusions and the bivalves will influence plankton

122   interaction network as previous analyses showed these variables can strongly alter dynamics of

123   individual plankton groups (Kimmerer, Gartside & Orsi 1994, Jassby *et al.* 1995). We show that

124   species invasions could potentially modify the connectance and topology of interaction networks

125   and highlight the importance of changes to biotic interactions when designing effective

126   conservation and management strategies.

127 **MATERIALS AND METHODS**

128 We focused our analysis on the upper San Francisco Estuary consisting of the brackish, low-

129 salinity Suisun Bay and Marsh (hereafter 'Suisun region') and largely freshwater inverted delta

130 formed by Sacramento and San Joaquin rivers (hereafter 'delta region'). These two regions have

131 been a great focus of natural resource management where major functional groups of the

132 community and environmental covariates have been monitored for more than three decades. We

133 analyzed both regions separately, as they differ in their abiotic conditions and the community

134 composition (Winder & Jassby 2011), and we wished to compare the responses of interaction

135 networks to species invasions in each of them. All data included in this study

136 (http://www.water.ca.gov/iep/) were collected at discrete sampling sites in the upper San

137 Francisco Estuary by the California Department of Fish and Wildlife (DFW) as part of the

138 Interagency Ecological Program (IEP) between 1974 and 2008.

139

140 **Response variables and covariates**

141 *Zooplankton biomass*

142 Zooplankton samples were collected using oblique tow nets (154 μm mesh size) or pumped

143 through a 64 μm mesh, bi-weekly or monthly (for details of the methods see Orsi & Mecum

144 1986). We converted zooplankton densities to carbon biomass and calculated annual means

145 (Winder & Jassby 2011) to capture shifts in the long-term biomass of key functional groups (Fig.

146 1). We used time-series data from six consistently sampled locations in Suisun and from seven

147 consistently sampled locations in the delta.

148   Zooplankton taxa and their life stages were classified by their feeding function into five

149 major groups: nauplii, herbivores, omnivores, small predators and predators. Nauplii are larval

150   stages of copepods and are also an important food for omnivorous and predatory zooplankton,

151   clams, and larval fish. Herbivores are suspension feeders consuming phytoplankton. They can

152   compete with, or serve as food for, other zooplankton groups. Herbivores include native

153   cladocerans, such as *Daphnia* spp., *Bosmina* spp., *Diaphanosoma* spp., and rotifers such as

154   *Asplanchna* spp., *Keratella* spp., *Polyarthra* spp., *Synchaeta* spp. and *Trichocerca* spp.

155   Omnivores feed on phytoplankton, suspended particulate matter, and smaller zooplankton

156   including their juvenile stages; omnivores include three probably native species of *Acartia* spp.,

157   *Eurytemora affinis*, and introduced *Pseudodiaptomus* spp., and *Sinocalanus doerrii*. We

158   characterize the group 'small predators' as zooplankton of smaller sizes (< 500 μm) that likely

159   derive carbon from the microbial pool, feeding on ciliates, flagellates, and some taxa also on

160   detritus and bacteria. This group is composed largely of introduced cyclopoids *Limnoithona* spp.,

161   and *Oithona davisae*, and harpacticoids. The 'predators' include native *Acanthocyclops* spp. and

162   introduced *Acartiella sinensis* and *Tortanus dextrilobatus* which can reach ten times the

163   individual body mass of the 'small predators' (Gould & Kimmerer 2010). This group is at the top

164   of the planktonic food web, largely feeding on other zooplankton species (Hooff & Bollens

165   2004; York *et al.* 2013).

166

167   *Phytoplankton biovolume*

168   The phytoplankton samples were collected with a submersible pump from 1 m depth. We used

169   phytoplankton abundance data from three locations in Suisun and three locations in the delta

170   region that have been sampled consistently. To obtain an estimate of phytoplankton biovolumes,

171   we used median values from direct measurements (IEP) or from published literature, and applied

172   median biovolumes for individual phytoplankton taxa to the entire time-series (Fig. 2). Despite

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173   the simplification of using median values, biovolume data provide information on taxonomic

174   composition and food quality, which greatly affect resource availability for zooplankton, and

175   consequently is a better proxy of resource availability than chlorophyll-*a*. We aggregated the

176   data into high-quality (cryptomonads and diatoms) and low-quality (other taxa including

177   cyanobacteria, dinoflagellates, green algae, chrysophytes, euglenoids, synurophytes and

178   haptophytes) groups (Brett, Müller-Navarra & Park 2000).

179

180   *Biotic and abiotic covariates*

181   We quantified the relative influence of four covariates that are predicted to have substantial

182   effects on plankton groups and potentially on their interactions. Two biotic covariates were:

183   summed biomass of the five dominant planktivorous fish species (northern anchovy *Engraulis*

184   *mordax*, American shad *Alosa sapidissima*, threadfin shad *Dorosoma petenense*, age-0 striped

185   bass *Morone saxatilis,* and longfin smelt *Spirinchus thaleichthys*); and the abundance of an

186   invasive suspension-feeding bivalve, *Potamocorbula amurensis*. Fish samples were collected

187   each month from September to December as a part of the Fall Midwater Trawl Survey conducted

188   by the California Department of Fish and Wildlife (DFW). Fish data were derived from 42

189   locations that were sampled every year over the entire time-series and converted to biomass (mg

190   C m^{-3}) using length-weight regression functions (Kimmerer *et al.* 2005). Most of the fish

191   collected in this program are young-of-the-year planktivores (median lengths from 70 to 81 mm

192   from September to December) and their biomass can therefore be used as an index of

193   planktivorous fish biomass throughout the year. *Potamocorbula* spread rapidly across the low-

194   salinity Suisun region after its introduction in 1986 (Alpine & Cloern 1992), but is not abundant

195   in largely freshwater parts of the delta. The bivalve data were derived from two locations (D4-L

196   and D7-C) that were sampled the same sampling period. Two abiotic covariates were annual

197   means of surface temperature, recorded at the same locations and time as the zooplankton

198   samples were collected, and the distance from the mouth of the estuary of the daily-averaged,

199   near-bottom salinity of 2 (practical salinity scale), locally termed 'X$_2$' (Jassby *et al.* 1995). X$_2$ is

200   a synthetic measure of the physical responses of the estuary to freshwater flow (Monismith *et al.*

201   2002).

202

203   **Statistical methods**

204   We used a variant of a multivariate, autoregressive model (MAR; Ives *et al.* 2003; Supporting

205   Information) to analyze the dynamics of plankton functional groups. The MAR method was

206   developed to estimate ecological interactions from time-series data and has been applied to

207   plankton interaction networks in freshwater (e.g., Ives *et al.* 2003; Hampton, Scheuerell &

208   Schindler 2006) and marine (e.g., Francis *et al.* 2012) ecosystems. We built separate MAR

209   models for each region, which included time-dependent parameters that tested for differences in

210   intra-specific interactions and covariate effects between the pre- (1974-1992) and post- (1993-

211   2008) invasions periods. The Bayesian framework allowed us to incorporate our prior

212   expectations for the direction of biotic interactions and the effects of main covariates. We used

213   normal prior distributions with unit variance and means that reflected prior expectations for the

214   direction of influence, i.e. $N(\mu\ 1)$, with $\mu = -0.675$, $+0.675$ and 0 for parameters with negative,

215   positive, or unknown prior expectations, respectively (Mac Nally *et al.* 2010). Parameters

216   quantifying changes in intra-specific interactions or covariate effects were assigned

217   uninformative, standard normal prior distributions, $N(0,1)$. The informative normal priors

218   corresponded to prior odds of 3:1 in favour of the expected direction of influence (Mac Nally *et*

219   *al.* 2010); uninformative ($\mu=0$) priors correspond to odds of 1:1. We inferred the importance of

220   model parameters from odds ratios (OR, the ratios of posterior odds to prior odds) that

221   parameters were positive or negative. Odds ratios are calculated as $OR = p_1/(1-p_1) : p_0/(1-p_0)$,

222   where p_1 and p_0 are the posterior and prior probabilities, respectively, that the parameter is

223   positive (or negative). If the OR $\geq$ 3.2, we concluded that there was substantial support in the

224   data for the expected relationship. If $1 \leq OR < 3.2$, the data did not invalidate the expectation but

225   there was relatively little support (Jeffreys 1961). We used hierarchical partitioning to determine

226   the relative importance of autoregressive terms, among- plankton interactions, and covariate

227   effects in explaining the variation in data in each region (see Supporting Information for details).

228

229   **RESULTS**

230   Our MAR models for the two regions fitted the time-series data well, as gauged by $R^2 = 0.894$

231   for Suisun and $R^2 = 0.892$ for the delta. The among-functional group components of the models

232   were most important in explaining variation in the data for both regions, followed by the

233   covariates and then the autoregressive components. Ratios for the hierarchical partitioning of the

234   explained variance were 1 : 0.77 : 0.35 in Suisun and 1 : 0.43 : 0.04 in delta, indicating that

235   trophic interactions among the functional groups are likely to be the key determinant of the

236   plankton community dynamics.

237

238   *Relationships in the Suisun region*

239   There appeared to be differences in the importance of biotic interactions and environmental

240   covariates in the pre-invasion and post-invasion time periods in Suisun (Fig. 3, Table 1). Prior to

241   the 1993 invasions, the biomass of omnivores was positively correlated with phytoplankton,

242   especially with high-quality food taxa, and negatively correlated with predators (Fig. 3a). The

243   association between omnivores and phytoplankton was not evident after 1993. Instead, we found

244   negative reciprocal correlations between omnivore and herbivore biomasses (Fig. 3b). Both

245   omnivores and herbivores were negatively associated with X_2 and predator biomass in the post-

246   invasion Suisun (Fig. 3b). Predator biomass was positively associated with omnivore biomass

247   after 1993, but not before (Fig 3). There was some evidence of reciprocal positive associations

248   between omnivores and nauplii after 1993, but these relationships had weak support (OR = 2.9).

249   There was no evidence of omnivore-nauplii associations before 1993. Biomasses of copepod

250   nauplii were negatively related to *Potamocorbula* abundance, and low-quality phytoplankton

251   was negatively related to X_2 before, but not after, the invasions in 1993 (Fig. 3).

13

252   We analyzed the temporal changes in the direction and strength of the relationships

253   between 1974–1992 and 1993–2008 (see strongly supported deviations in Table 1). In the latter

254   period there was a shift toward negative correlation between omnivores and herbivores, an

255   increased negative effect of predators on herbivores and an increased positive effect of nauplii on

256   omnivores. We found a weakening of the negative effect of *Potamocorbula* on nauplii and a

257   strengthening of the negative effects of X_2 on herbivores and omnivores after 1993 (Table 1).

258

259   *Relationships in the delta region*

260   There were fewer correlations among plankton functional groups, and fewer relationships

261   between plankton and environmental covariates in the delta region than in Suisun (Fig. 4, Table

262   1). We did not find any associations between phytoplankton or nauplii and other zooplankton

263   groups in either time period, and all correlations among plankton groups in the delta were

264   negative. In the pre-invasion delta, there was strong support for negative effects of predators on

265   herbivores, small predators on nauplii, and omnivores on small predators (Fig. 4a). Nauplii were

266   positively correlated with X_2 before 1993 (Fig. 4a). None of these relationships was evident in

267   the delta after 1993.

268   In the post-invasion delta, predators were positively associated with X_2, while herbivores

269   were negatively associated with omnivore biomass (Fig. 4b). This last coefficient had marginal

270   support (OR = 3.1). Direct comparison of model parameters for the pre- and post- invasion

271   periods (deviations in Table 1) suggest reduced influence of small predators and X_2 on nauplii,

272   and an increase in the positive effect of X_2 on predators.

273   There was little evidence to suggest that pelagic fish or water temperature affected the

274   plankton communities in either region or time period (Figs 3, 4).

14

275   **DISCUSSION**

276   Our study indicates substantial changes in the strength and direction of interactions among the

277   plankton functional groups following the establishment of invasive copepods in the upper San

278   Francisco Estuary in the early 1990s. Among the strongest shifts were negative reciprocal

279   correlations between omnivores and herbivores in the post-invasion Suisun that were not evident

280   before the invasions. These findings are consistent with intensified competition for the reduced

281   phytoplankton resource in this region. Introduced omnivorous *Pseudodiaptomus* spp. (Orsi &

282   Walter 1991; Winder & Jassby 2011) probably contributed to the negative effects of omnivores

283   on herbivores. A shift toward a stronger negative effect of predators on herbivores in the post-

284   invasion Suisun probably arose from predation by invasive *A. sinensis* and *T. dextrilobatus* after

285   their establishment around 1993 (Orsi & Ohtsuka 1999), while salinity intrusions likely mediated

286   these relationships. Herbivores, mainly consisting of rotifers in Suisun and cladocerans and

287   rotifers in the delta, declined sharply and our analysis suggests that competition and predation

288   from other zooplankton groups may have contributed to this decline, although ammonium

289   discharge (Dugdale *et al.* 2007) and other introductions (e.g., Kimmerer & Orsi 1996) were

290   likely contributing factors.

291   The biomass of omnivorous zooplankton was positively associated with phytoplankton in

292   the pre-invasion Suisun, but with nauplii in the recent post-invasion period. The association

293   between nauplii and omnivores indicates copepod recruitment since the omnivores' principal

294   food is microzooplankton such as ciliates (York *et al.* 2013), which are not sampled by the

295   monitoring programs. A decline in phytoplankton and an increase in nauplii biomass after

296   invasions (Winder & Jassby 2011) reduced the importance of phytoplankton as a food resource.

297   Such changes in the interactions between primary producers and zooplankton consumers that

298   form the food sources for pelagic fish can greatly alter energy transfer to higher trophic levels

299   (Ravet, Brett & Arhonditsis 2010). The absence of correlations between herbivores and

300   phytoplankton in both regions indicates that other processes than mesozooplankton grazing may

301   control phytoplankton dynamics (Kimmerer 2006; Kimmerer & Thompson 2014). A meta-

302   analysis of observational and experimental studies from marine pelagic ecosystems showed

303   similar loose coupling between zooplankton and phytoplankton biomasses (Micheli 1999).

304   We expected to detect competitive effects of *L. tetraspina* on other groups because after

305   1993 this small predator became the numerically dominant in Suisun. However, 'small predators'

306   appeared to have little association with other plankton groups. *L. tetraspina* feeds preferentially

307   on protists, but rarely consumes diatoms (Bouley & Kimmerer 2006) and it is too small to

308   consume calanoid copepods. Our results are consistent with laboratory experiments (Bouley &

309   Kimmerer 2006) and suggest that *L. tetraspina* may not compete with or prey upon native

310   zooplankton groups. The negative effect of small predators on nauplii in the delta declined after

311   the establishment of *L. tetraspina* in 1993 in an agreement with this species being a source rather

312   than a consumer of nauplii. Nevertheless, *Limnoithona* spp. may be a resource for predatory

313   zooplankton (York *et al.* 2013). The weakening of the negative correlation between predators

314   and herbivores after 1993 in the delta region is consistent with a shift from the native predators

315   *Acanthocyclops* spp. to invasive *A. sinensis.*

316   In general, there were fewer statistically important associations and the shift in the

317   plankton network structure was less apparent in the delta region than in Suisun. This may be due

318   to the different taxonomic composition of individual functional groups and the lower densities of

319   predatory copepods established in the delta after 1993. Furthermore, plankton dynamics can be

320   more strongly driven by the changes in the freshwater flow, water clarity, and ammonium

321   discharges in the delta than in the Suisun region. Greater strengths and numbers of potential

322   negative interactions in Suisun than in the delta suggests that changes in a particular functional

323   group, such as nauplii, could potentially cascade to other components of food webs.

324   Our results suggest a strong effect of the isohaline index X_2 on community structure in

325   both regions, which has been reported elsewhere (Jassby *et al.* 1995; Mac Nally *et al.* 2010;

326   Winder, Jassby & Mac Nally 2011). Given that X_2 represents principally the response of the

327   salinity field to freshwater flow into the estuary, and species are distributed spatially along a

328   salinity gradient, association with X_2 can be understood as a consequence of interannual

329   movement of the plankton across the delta – Suisun boundary. X_2 in the post-invasion Suisun

330   was negatively correlated with herbivores and omnivores that may have narrow salinity ranges

331   (e.g., Kimmerer 2002). When freshwater flow increases the delta regions goes from being mostly

332   fresh to entirely fresh, while Suisun region can go from being entirely brackish to largely fresh

333   for much of the year. This potentially underlies some, though not all, of the observed interactions

334   among functional groups and results in differential patterns in both regions.

335   Prior studies have documented a negative effect of *Potamocorbula* on phytoplankton,

336   copepod nauplii, and individual species of zooplankton and fish (Alpine & Cloern 1992;

337   Kimmerer, Gartside & Orsi 1994; Kimmerer 2006). Our study suggests a negative effect of

338   *Potamocorbula* on nauplii in Suisun before 1993, where relatively high nauplius biomass sharply

339   declined soon after the introduction of *Potamocorbula* in 1986 (see also Kimmerer, Gartside &

340   Orsi 1994). Nauplii did not appear to be negatively affected by the clam after 1993, but increased

341   due to the recruitment of new invasive zooplankton, mainly *Limnoithona* spp. (Winder & Jassby

342   2011). The lack of response of plankton to *Potamocorbula* abundance in our study may relate to

343   the highly variable size of clams, and therefore a weak relationship between their numerical

344   abundance and grazing rate.

345   Other covariates appeared to have little effect on plankton community structure in both

346   regions. Abundance of fish was unrelated to zooplankton community dynamics, which is obvious

347   also from the independent dynamics of these groups after a steep drop in production of pelagic

348   fish in the early 2000s (Thomson *et al.* 2010). We found no evidence that water temperature

349   affected the plankton community, but other temperature-related variables, such as variability or

350   the frequency of extreme temperatures, may be also relevant and warrant further investigations.

351   The MAR framework is well suited to explore links among functional groups and to

352   investigate the responses of ecological communities to perturbations in a multivariate framework

353   (Ives *et al.* 2003; Mac Nally *et al.* 2010; Francis *et al.* 2012). Although the positive or negative

354   associations among species may indicate biotic interactions, the MAR analyses rely on

355   correlations and cannot inform the causality of these relationships. We assigned the prior

356   expectations to the relationships based on the assumed trophic interactions before the analyses.

357   Despite the weak influence of the priors, few significant relationships showed opposite

358   directions, so that our findings are largely consistent with the previous knowledge about the

359   system.

360   We used time-series data aggregated annually and across large geographical areas (about

361   1000 km^2) to remove the effect of seasonal events (e.g., Micheli 1999) and to focus our analysis

362   on long-term trends of plankton dynamics across the regions. Therefore, some short-term or

363   locally important interactions may not have been detected. The MAR analysis assumes linear

364   relationships (on the log-scale), so any non-linear relationships may not have been revealed (Ives

365   *et al.* 2003). Despite these limitations, our results suggested several relationships, some

366   previously identified but some new, among plankton functional groups and environmental

367   covariates.

368   The establishment of three dominant zooplankton species around 1993 was not an

369   isolated event and individual invasions have occurred throughout the time-series. Nevertheless,

370   this simultaneous establishment of invasive copepods was followed by major shifts in the

371   relative species abundance within the functional groups and in the relative importance of the

372   functional groups within the entire plankton community (Winder, Jassby & Mac Nally 2011;

373   Winder & Jassby 2011). Our results imply that these changes, in combination with other human-

374   induced perturbations, have the potential to modify the network of biotic interactions. We

375   identified that herbivores are the functional group most affected by negative interactions and

376   their loss may have cascading effects on biodiversity, in particular in the Suisun region. This

377   system likely shifted towards a nutritionally inferior phytoplankton and detrital-based food web,

378   with potential nutritional consequences for secondary consumers. The importance of some

379   environmental covariates shifted between the two time periods, although there appeared to be a

380   relatively stronger influence of biotic interactions than any of the covariates.

381   There are few studies that investigate changes in the biotic network structure following

382   invasions of non-native species. Recent exceptions that focused on the complex plant–pollinator

383   networks in terrestrial systems found no effect of invasive species on overall connectance (the

384   number of realized links relative to the number of possible links; Aizen, Morales & Morales

385   2008; Heleno et al. 2009), but the connectance among native species declined (Aizen, Morales &

386   Morales 2008). Our analysis of a simplified network among plankton functional groups suggests

387   that connectance can either increase (Suisun) or decline (delta), following the invasions (Figs 3,

388   4). Such different responses highlight the importance of understanding the environmental context

389   when assessing the impacts of invasive species on community structure (Tylianakis *et al.* 2008).

390   There is increasing recognition of the importance of altered trophic interactions and

391   energy flows in ecosystems degraded by human activities (e.g., Naiman *et al.* 2012). Attempts to

392   restore individual taxa or diversity may be in vain if altered trophic dynamics limit the flow of

393   energy to target populations. Our results suggest that the planktonic food web in Suisun, in

394   particular, may have been fundamentally altered following zooplankton invasions. Such changes

395   probably alter the flows of energy and resource flows to higher trophic levels, and, therefore,

396   have implications for understanding or reversing recent declines in fish populations in this

397   system.

398

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551

552

553 **SUPPORTING INFORMATION**

554 Additional Supporting Information may be found in the online version of this article:

555 **Appendix S1.** Supporting methods.

556

557   **Table 1.** Statistical comparison of major biotic interactions and covariates in pre-invasion and

558   post-invasion period in Suisun and delta regions of the upper San Francisco Estuary. Parameter

559   details are listed for Bayesian estimation of the multivariate autoregressive (MAR) model.

560   'Expected' directions of the interactions (+ positive, - negative, ? unknown, P0 probably zero, 0

561   zero with high certainty) were specified in the prior probabilities and prior odds were set to 3.

562   'Deviations' represent $\Delta\beta_{ij}$ from the eq. 4 and indicate substantial changes in relationships from

563   pre-invasion to post-invasion period. Odds ratio (posterior odds / prior odds) is expressed as 1/G

564   for results that were contrary to expectation (G represents odds ratio for the unanticipated result).

565

Biotic Interactions & Covariates	Mean ± SD	2.50 – 97.50%	Post-odds	Expected	Odds ratio
Suisun region pre-invasion					
high-quality phytoplankton on omnivores	0.659 ± 0.400	-0.106 – 1.445	20.1	+	6.7
low-quality phytoplankton on omnivores	0.359 ± 0.254	-0.130 – 0.877	13.3	P0	4.4
predators on omnivores	-0.308 ± 0.184	-0.664 – 0.045	23.5	-	7.8
Potamocorbula on nauplii	-0.221 ± 0.165	-0.545 – 0.116	10.5	-	3.5
X_2* on low quality phytoplankton	-0.519 ± 0.329	-1.159 – 0.150	14.5	?	4.8
Suisun region post-invasion					
omnivores on herbivores	-0.646 ± 0.393	-1.408 – 0.173	17.5	-	5.8
herbivores on omnivores	-0.581 ± 0.324	-1.214 – 0.051	27.9	-	9.3
predators on herbivores	-0.376 ± 0.223	-0.829 – 0.072	20.4	-	6.8
omnivores on nauplii	0.658 ± 0.534	-0.416 – 1.698	8.6	-	1/2.9

nauplii on omnivores	0.958 ± 0.744	-0.481 – 2.364	8.6	+	2.9
predators on omnivores	-0.270 ± 0.204	-0.664 – 0.145	9.9	-	3.3
omnivores on predators	0.873 ± 0.583	-0.263 – 1.968	13.2	+	4.4
X_2 on herbivores	-0.790 ± 0.223	-1.234 – -0.379	Inf.	?	Inf.
X_2 on omnivores	-0.737 ± 0.216	-1.158 – -0.296	799.0	?	266.3
Suisun region deviations					
omnivores on herbivores	-0.856 ± 0.458	-1.720 – 0.074	26.6	-	8.9
herbivores on omnivores	-0.873 ± 0.495	-1.811 – 0.129	21.2	-	7.1
predators on herbivores	-0.417 ± 0.264	-0.947 – 0.104	17.3	-	5.8
nauplii on omnivores	0.947 ± 0.720	-0.492 – 2.343	9.8	+	3.3
Potamocorbula on nauplii	0.430 ± 0.305	-0.176 – 1.033	11.6	-	1/3.9
X_2 on herbivores	-0.510 ± 0.314	-1.160 – 0.079	22.3	?	7.4
X_2 on omnivores	-0.588 ± 0.317	-1.211 – 0.046	27.6	?	9.2
Delta region pre-invasion					
predators on herbivores	-0.477 ± 0.331	-1.156 – 0.171	12.5	-	4.2
small predators on nauplii	-0.810 ± 0.347	-1.477 – -0.110	67.6	-	22.5
omnivores on small predators	-0.506 ± 0.379	-1.276 – 0.255	10.5	P0	3.5
X_2 on nauplii	0.503 ± 0.239	0.026 – 0.979	53.6	?	17.9
Delta region post-invasion					
omnivores on herbivores	-0.512 ± 0.396	-1.295 – 0.297	9.2	-	3.1
X_2 on predators	0.402 ± 0.203	-0.009 – 0.804	34.8	?	11.6
Delta region deviations					
small predators on nauplii	0.819 ± 0.441	-0.043 – 1.690	29.0	-	1/9.7
X_2 on nauplii	-0.442 ± 0.303	-1.052 – 0.136	13.3	?	4.4
X_2 on predators	0.401 ± 0.306	-0.215 – 1.009	10.1	?	3.4

566

567

568 **FIGURES**
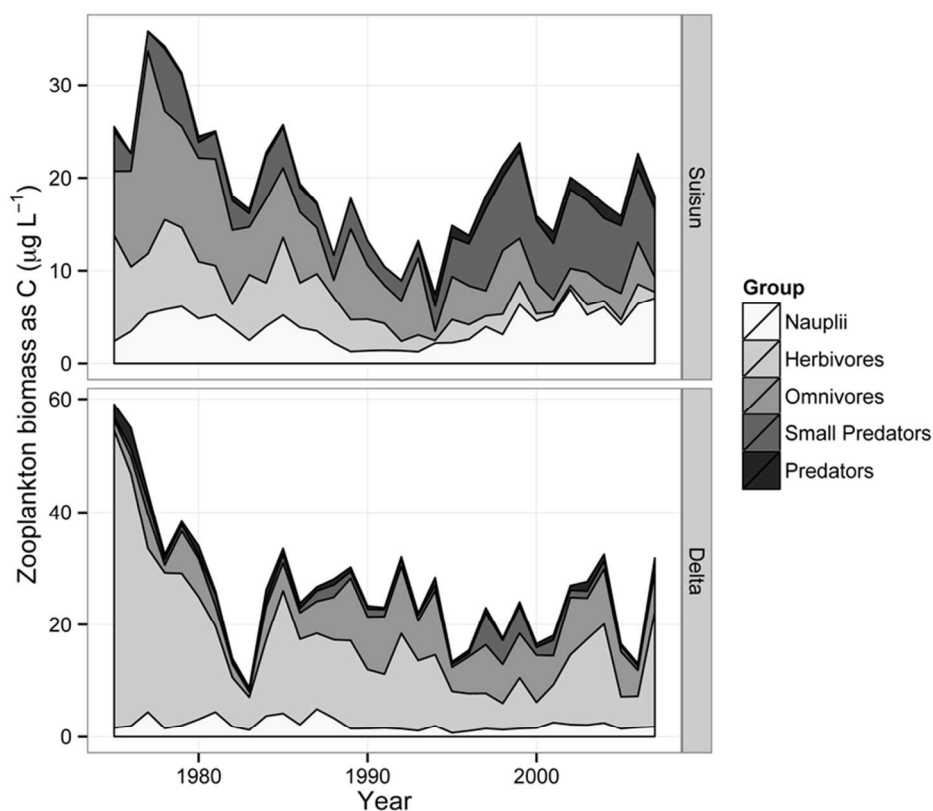
569

570 **Figure 1.** Temporal dynamics of zooplankton annual mean biomass, expressed as μg carbon per

571 L, in two regions of the upper San Francisco Estuary. Zooplankton community in Suisun

572 experienced continuous declines in herbivore (slope of the long-term trend estimated by Mann-
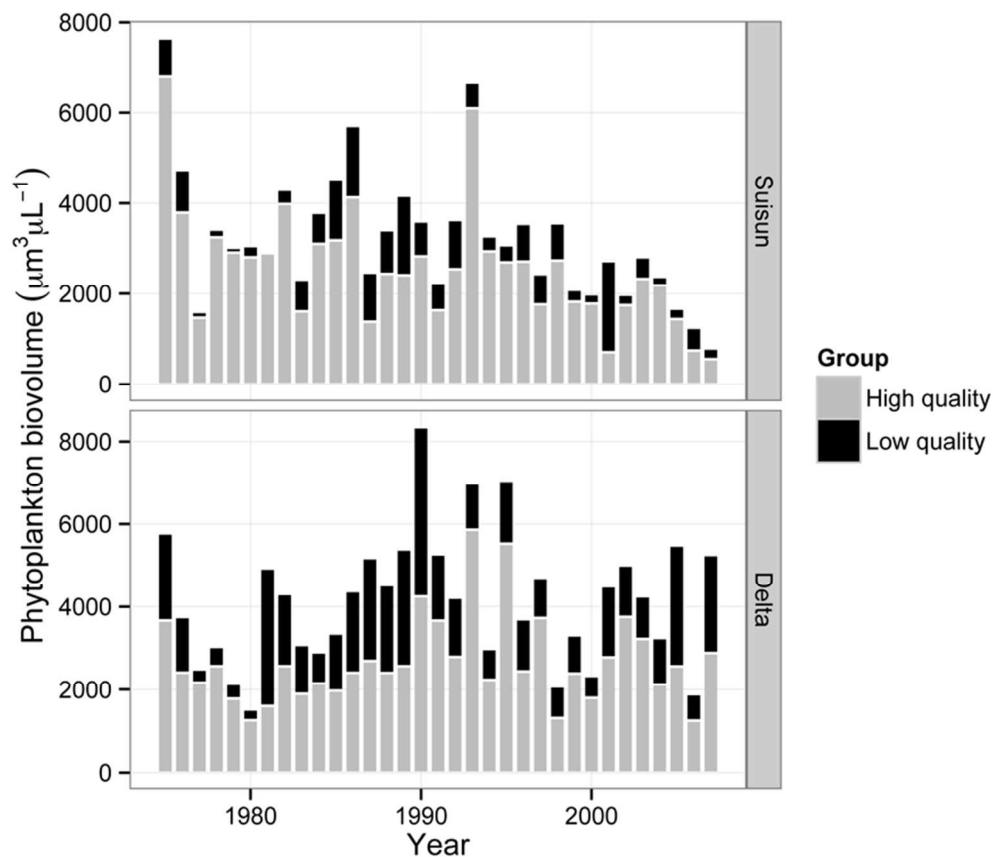
573 Kendall test $b = -6.83\%$ year^{-1}, n = 33, $P < 0.001$) and omnivore ($b = -4.68\%$ year^{-1}, n = 33, $P <$

574 0.001) biomass. Biomass of predators ($b = 6.13\%$ year^{-1}, n = 33, $P < 0.001$) and small predators

575 ($b = 3.71\%$ year^{-1}, n = 33, $P = 0.001$) in Suisun increased. Zooplankton community in delta

576 experienced continuous declines in herbivore ($b = -3.82\%$ year^{-1}, n = 33, $P < 0.001$) and

577 increased in omnivore ($b = 2.30\%$ year^{-1}, n = 33, $P = 0.008$) biomass.

578

Figure 2. Temporal dynamics of phytoplankton expressed as mean annual biovolume in two regions of the upper San Francisco Estuary. High quality phytoplankton are cryptomonads and diatoms (black bars), low quality phytoplankton (grey bars) encompass all other taxa including cyanobacteria, dinoflagellates, green algae, chrysophytes, euglenoids, synurophytes and haptophytes. High quality phytoplankton ($b = -2.50\%$ year^{-1}, n = 33, $P < 0.001$, Mann-Kendall test) and total phytoplankton biovolume ($b = -2.37\%$ year^{-1}, n = 33, $P < 0.001$) in Suisun showed declining trends over the study period.

586
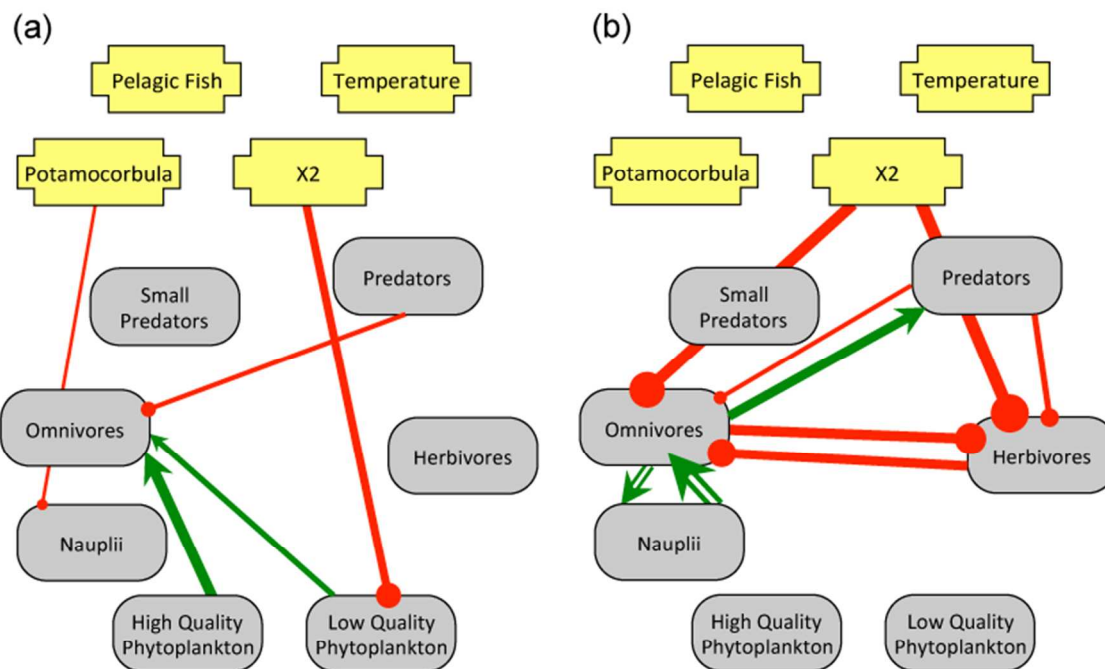
Figure 3. Biotic interactions among plankton functional groups (grey ovals) and influence of

major covariates (yellow boxes) in low-salinity Suisun region before (a) and after (b) invasion of

several zooplankton species in 1993. Green arrows pointing toward a plankton group indicate

positive interactions; red lines indicate negative interactions. The strength of the interactions is

represented by the widths of the lines. Only the relationships with substantial support from the

data are depicted (OR $\geq$ 3.2). Two open lines represent interactions that have marginal support

from the data (OR = 2.9) but do not meet the criterion for substantial evidence.
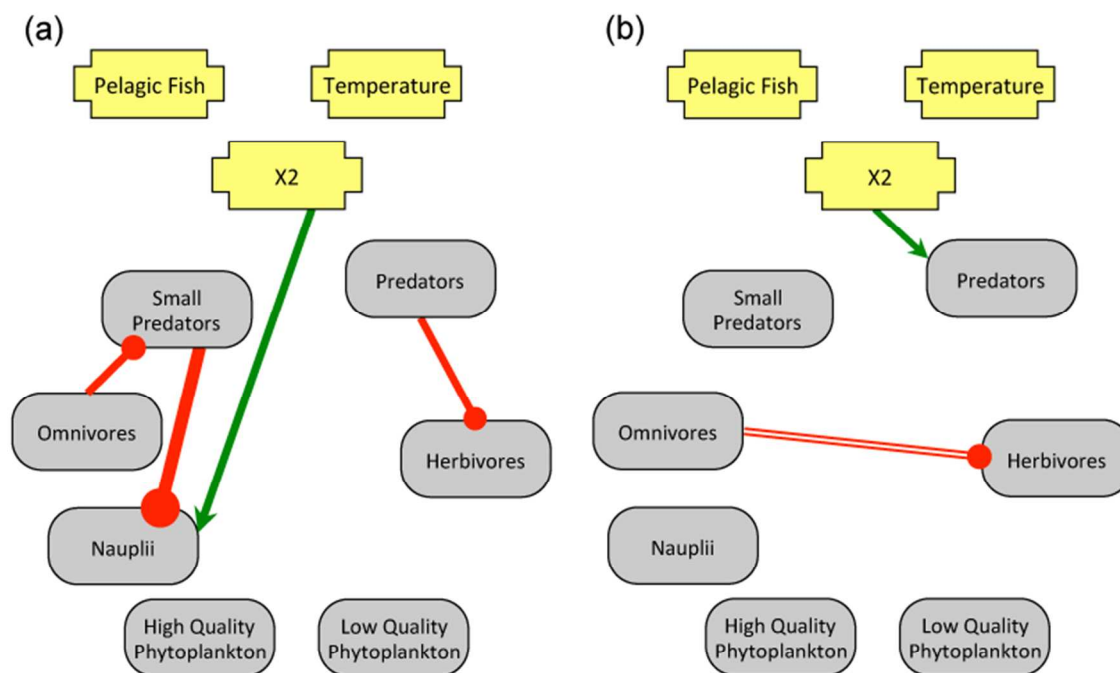
594

595   **Figure 4.** Biotic interactions among plankton functional groups (grey ovals) and influence of

596   major covariates (yellow boxes) in mostly freshwater delta region before (a) and after (b)

597   invasion of several zooplankton species in 1993. Green arrows pointing toward a plankton group

598   indicate positive interactions; red lines indicate negative interactions. The strength of the

599   interactions is represented by the widths of the lines. Only the relationships with substantial

600   support from the data are depicted (OR ≥ 3.2). The open line represents an interaction that has

601   marginal support from the data (OR = 3.1) but does not meet the criterion for substantial

602   evidence.

SUPPORTING INFORMATION

Human-induced biotic invasions and changes in plankton interaction networks

Pavel Kratina, Ralph Mac Nally, Wim J. Kimmerer, James R. Thomson, Monika Winder

Appendix S1. Supporting methods.

Formulation of the multivariate autoregressive model (MAR)

We represented dynamics of the seven plankton groups ('focal groups') with the Gompertz model (Dennis *et al.* 2006), beginning with a deterministic version (Reddingius 1971):

$$n_{i,t} = n_{i,t-1}\exp\left(\gamma_i + \delta_i \ln\left(n_{i,t-1}\right)\right). \tag{1}$$

Here, $n_{i,t}$ is the biomass of functional group i at time t, $n_{i,t-1}$ is the biomass of group i at the previous time step, γ_i is the intrinsic rate of population growth for group i, and δ_i, which has been interpreted as the degree of density dependence. By taking natural logarithms and by setting $x_{i,t} = \ln(n_{i,t})$, (1) becomes:

$$x_{i,t} = x_{i,t-1} + \gamma_i + \delta_i x_{i,t-1}. \tag{2}$$

Interspecific interactions among the J plankton groups were incorporated by appending terms relating to the previous year $\beta_{ij}x_{j,t-1}$, excluding intra-group terms:

$$x_{i,t} = x_{i,t-1} + \gamma_i + \delta_i x_{i,t-1} + \sum_{j=1,j\neq i}^{J}\beta_{ij}x_{j,t-1}. \tag{3a}$$

The interaction terms among the plankton functional groups were implemented as potentially differing between the pre- and post-invasion periods by using an indicator variable I, which was 0 for the pre-invasion period and unity afterwards:

$$x_{i,t} = x_{i,t-1} + \gamma_i + \delta_i x_{i,t-1} + \sum_{j=1,j\neq i}^{J}\left(\beta_{ij}^0 + I_t\Delta\beta_{ij}\right)x_{j,t-1}. \tag{3b}$$

Here, the pre-invasion values are β_{ij}^0 and the post-invasion values are $\left(\beta_{ij}^0 + I_t\Delta\beta_{ij}\right)$.

Changes in the relationships from the pre-invasion to the post-invasion period (i.e.,

deviations) are $\Delta\beta_{ij}$. We included the effects of the K covariates u_k through α-

coefficients, for the current year t, which were also allowed to differ after the invasions:

$$x_{i,t} = x_{i,t-1} + \gamma_i + \delta_i x_{i,t-1} + \sum_{j=1, j\neq i}^{J}\left(\beta_{ij}^0 + I_t\Delta\beta_{ij}\right)x_{j,t-1} + \sum_{k=1}^{K}(\alpha_{ik}^0 + I_t\Delta\alpha_{ik})u_{k,t}.$$

$$(4)$$

Bayesian estimation of the MAR model

We used Bayesian methods to estimate the model because measurement uncertainties can

be incorporated using hierarchical models and sporadically distributed missing data are

handled easily within the same process by which the parameter estimates are made.

Equation (4) was estimated by using the following model in WinBUGS, v. 1.4

(Spiegelhalter, Thomas & Best 2003):

$$z_{i,t}\sim N\left(x_{i,t}, \omega_{i,t}^2\right); x_{i,t}\sim N\left(\mu_{i,t}, \sigma_i^2\right); \acute{c}_{k,t}\sim N\left(u_{k,t}, \zeta_{k,t}^2\right); \qquad (5a)$$

$$\mu_{i,t} = x_{i,t-1} + \gamma_i + \delta_i x_{i,t-1} + \sum_{j=1, j\neq i}^{J}\left(\beta_{ij}^0 + I_t\Delta\beta_{ij}\right)x_{j,t-1} + \sum_{k=1}^{K}(\alpha_{ik}^0 + I_t\Delta\alpha_{ik})u_{k,t}.$$

$$(5b)$$

Here, N denotes the normal distribution. The model states that the (ln-transformed)

observed values ($z_{i,t}$) represent the true values ($x_{i,t}$), which have observation errors.

These were included by the use of (ln-transformed) unobserved values ($x_{i,t}$) and

observation errors, $\omega_{i,t}^2$. The observation errors were estimated from standard errors (SEs)

of mean values for the response variables for each time period. The $z_{i,t}$ were ln-

transformed, so we used the first term of a Taylor functional expansion to approximate

the ln-transformed SEs $[SE(\ln(\bar{n}))] \approx SE(\bar{n})/\bar{n}]$ (Seber 1973; Stuart & Ord 1987).

Process variances (σ_i^2) were group-specific and were implemented with uninformative

priors on σ_i of U(0.01, 10) (Gelman 2005), where U denotes a uniform distribution. The

true, unobserved values ($\mu_{i,t}$) are functions of the groups-dynamic parameters, biotic

interactions, and covariates (eq. 4).

The observed covariates $c_{k,t}$ were standardized [$\acute{c}_{k,t} = \left(c_{k,t} - \bar{c}_k\right)/SD_k$] in order

to aid model convergence and to equalize numerical ranges among different scales of

measurement. Uncertainties in covariate measurements (within-year SEs) were scaled by

the inter-annual standard deviations (i.e., $SE_{k,t}/SD_k$). The model specifies that the true

(standardized) covariate values ($u_{k,t}$) are related to the observed standardized values

($\acute{c}_{k,t}$) but include the covariate-specific uncertainties [$\zeta_{k,t}^2 = (SE_{k,t}/SD_k)^2$].

Prior probabilities and parameter inference

For the important ecological parameters, α^0 and β^0, we used both informative (offset

normal) and uninformative (standard normal) priors. The offsets were

$N(0.675,1)$ and $N(-0.675,1)$ for expected positive and negative relationships; these

values invest a c. 3:1 prior probability mass in favor of the expected influence. While

there are plausible intraspecific density-dependencies in plankton dynamics over short

periods of time (e.g., phytoplankton spring bloom), here we assumed no interannual

intraspecific density dependence within the plankton groups: $\delta_i \sim N(0, 10^{-3})$. Several of

the possible relationships were specifically excluded from the model (i.e., deemed

unlikely to be important). For such relationships, coefficients were assigned $N(0, 10^{-6})$

priors (i.e., 0 with high certainty). For change parameters, $\Delta\alpha$ and $\Delta\beta$, we used

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standard normal prior distributions $N(0,1)$, except for those involving implausible

relationships, which were fixed at zero.

Model convergence and adequacy

The model was estimated with three MCMC chains of 20 000 iterations after 10 000

iterations of burn-in ('model settling'), results for which were discarded. We checked for

MCMC mixing and convergence using the boa package (Smith 2006) in *R* (R

Development Core Team 2013). We determined the adequacy of model fit by using

posterior predictive assessment (Gelman, Meng & Stern 1996). This involves simulating

samples from the $(\mu_{i,t}, \sigma_i^2)$, $\eta_{i,t}$. An adequate measure of fit for an observation for a

normal variate is $|x_{i,t} - \mu_{i,t}|$ for the data and $|\eta_{i,t} - \mu_{i,t}|$ for the simulated values

respectively (Agresti 2002). These values are summed over all (*i,t*) combinations. A good

model fit has the observed summed value in the middle of the distribution of simulated

summed discrepancies, so a posterior predictive probability (*PPP*) of 0.5 indicates a

perfect fit, whereas *PPP* values close to zero or unity indicate poor fit. Our results

demonstrated that posterior predictive assessment was 0.14 for Suisun and 0.18 for the

delta, suggesting that the data plausibly could be generated from the parameters and

structure of the models.

We determined the relative importance of the autoregressive (A), among-focal

group (F), and covariate (C) components in the model. We calculated the R^2 for eight

models: null (fitting constant-only averages for the focal group variables), A, F, C, A + F,

A + C, F + C, A + F + C (full model). Values of R^2 for these models were estimated by

omitting terms from eq. (5) as appropriate. The γ_i terms were retained for all models. R^2

were the squared Pearson correlation coefficients between the z and $\square$ values from the focal group values for all years. We performed hierarchical partitioning to decompose variance (Chevan & Sutherland 1991; Mac Nally 2000), which identifies independent contributions from individual terms (viz., A, F, and C) and joint variance explained; the R package hier.part (Walsh & Mac Nally 2003) was used for this task.

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