Four decades of water quality change in the upper San Francisco Estuary

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1 Abstract

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Recent methods for trend analysis have been developed that leverage the descriptive potential of multi-decadal monitoring data. We apply an estuarine adaptation of the Weighted Regressions on Time, Discharge, and Season (WRTDS) model to describe water quality trends over four decades in the Delta region of the San Francisco Estuary (SFE). Results from multiple stations in the Delta provided novel descriptions of historical trends and relationships between key species of dissolved inorganic nitrogen (ammonium, nitrate/nitrite, total). Trend analysis with WRTDS flow-normalized data demonstrated the potential to misinterpret changes using observed data that include flow effects, such that several trends with flow-normalized data had changes in magnitude and even reversal of trends relative to the observed. We further demonstrate use of WRTDS to provide insight into mechanisms of change with two case studies that 1) evaluate downstream changes in nitrogen following upgrades at a wastewater treatment

plant, and 2) interactions between biological invaders, chlorophyll, and flow in Suisun Bay. Overall, this analysis provides an ecological and management-based understanding of historical trends in the Delta as a means to interpret potential impacts of recent changes and expected trends.

18 Introduction

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Trend analysis is a broad discipline that has been applied to time series for the interpretation of environmentally-relevant changes. Direct evaluation of an observed time series is often insufficient, given that a long-term change can be masked by variation at shorter time scales 21 or the observed variation represents the combined effects of many variables. 1,2 Climate, local, 22 and regional environmental effects may act individually or together over time to impose 23 a change on time series, such that methods that account for variation at different scales have been used for trend analysis. 3-6 As a practical approach for water quality evaluation, 25 trend analysis of ecosystem response indicators often focuses on tracking the change in concentrations or loads of nutrients over many years. Response indicators can vary naturally with changing flow conditions and may also reflect long-term effects of management or policy 28 changes. For example, chlorophyll a (chl-a) concentration as a measure of phytoplankton 29 response to nutrient inputs can follow seasonal patterns with cyclical variation in temperature 30 and light changes throughout each year, whereas annual trends can follow long-term variation 31 in nutrient inputs to the system.^{7,8} Similarly, nutrient trends that vary with hydrologic 32 loading also vary as a function of utilization rates by primary producers or decomposition 33 processes. 9-11 Time series analysis of ecosystem response indicators must simultaneously 34 consider effects of processes at multiple scales and interactions between variables of interest 35 to develop a more comprehensive description of system change. 36 Appropriate methods for the analysis of change depend largely on the question of in-37

Appropriate methods for the analysis of change depend largely on the question of interest and characteristics of the environmental dataset. Trend analyses for aquatic systems have traditionally focused on comparisons between discrete periods of time to estimate di-

conventional approaches addressed limitations in historical monitoring datasets related to 41 infrequent sampling and relatively few years of continuous data. Increased availability of multi-decadal datasets, particularly for high profile environments, has accelerated develop-43 ment of trend analysis methods that leverage the descriptive potential of long-term time series from continuous monitoring programs. 6,14 These methods are often data-driven where 45 the parameterization of a simple functional model can change smoothly over time given that relationships between water quality variables and potential drivers are dynamic. The 47 Weighted Regressions on Time, Discharge, and Season (WRTDS) approach was developed in this context and has been used to characterize decadal trends in running-water systems. 15-20 This method has the potential to provide a spatially and temporally robust description of 50 trends by fitting a dynamic model with parameters that change relative to the domain of 51 interest. More recently, the WRTDS method was adapted for trend analysis in tidal waters, with a focus on chl-a trends in Tampa Bay²¹ and the Patuxent River Estuary,²² and tidally-influenced time series of dissolved oxygen from continuous sonde measurements.²³ These studies have demonstrated the potential of WRTDS for trend analysis in tidal waters. Application to additional datasets could provide further insight into environmental dynamics of aquatic systems. 57 The Sacramento - San Joaquin River Delta (hereafter 'Delta') is a mosaic of inflows 58 upstream of the San Francisco Estuary (SFE) that receives and processes inputs from the larger watershed. 24-26 Sediment export downstream of the Delta and wastewater treatment plant (WWTP) inputs are primary sources of nutrients for the larger Bay. Background 61 nutrient concentrations in SFE often exceed those associated with excessive primary pro-62 duction, although ecosystem responses symptomatic of eutrophication have historically been 63

rection and magnitude of a trend using non-parametric tests. 12,13 Development of these

infrequent. Changes in response to additional stressors (e.g., variation in freshwater in-

puts/withdrawals, invasive species, climate change) suggests that recent conditions have not

followed past trajectories and more subtle spatial and temporal variation could provide clues

that describe underlying properties of this system.²⁷ A comprehensive monitoring dataset has been collected at several fixed locations in the upper estuary and Delta for the last four decades.²⁸ Moreover, nutrient dynamics in the Delta are inherently linked to flow variation from inputs, withdrawal, impoundments, and downstream transport,²⁹ suggesting that an approach that explicitly considers flow effects is critical for trend analysis. To date, the regional monitoring dataset for the northern SFE, including the Delta, is an under-utitilized data source and a comprehensive analysis with WRTDS could facilitate an understanding of historical and recent changes in water quality.

The goal of this study was to provide a comprehensive description of nutrient trends in 75 the northern SFE and Delta region to inform understanding of ecosystem response dynamics and potential causes of water quality change. We applied the newly-adapted method 77 of weighted regression for tidal waters to describe nitrogen trends in different spatial and temporal contexts. The specific objectives were to 1) quantify and interpret trends over four decades at ten stations in the Delta, including annual, seasonal, and spatial changes in nitrogen analytes and response to flow variation, and 2) provide detailed descriptions 81 of two case studies in the context of conceptual relationships modeled with WRTDS. The second objective evaluated two specific water quality stations as additional case studies to demonstrate complexities with nutrient response to flow, effects of nutrient-related source controls on ambient conditions, and effects of biological invasion by benthic filter feeders on primary production. Although quantitative descriptions of change can be ends in themselves, the results provide means to a more detailed understanding of ecosystem properties. Products derived from WRTDS can be used to inform additional analyses, such as water quality response after removing annual, seasonal, or flow effects.

90 Materials and Methods

91 Study system

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estuary on the Pacific coast of North America. The watershed provides water to over 25 million people and irrigation for 18 thousand km² of agricultural land. Water enters the SFE through the Sacramento and San Joaquin rivers that have a combined inflow of approximately 28 km³ per year, with the Sacramento accounting for 84% of inflow to the Delta. The SFE system includes the Delta and subembayments of San Francisco Bay, with Suisun Bay immediately downstream of the Delta, San Pablo Bay to the north, South Bay, and the Central Bay that drains to the Pacific Ocean through the Golden Gate (Figure 1). 99 Water dynamics in the SFE and Delta are governed by inflows from the watershed, tidal ex-100 change with the Pacific Ocean, and water withdrawals for municipal and agricultural use.²⁴ 101 Seasonally, inflows from the watershed peak in the spring and early summer from snowmelt, 102 whereas consumption, withdrawals, and export have steadily increased from 1960 to present, 103 but vary depending on inter-annual climate effects. 27 Notable drought periods have occurred 104 from 1976-1977, 1987-1992, and recently from 2013-2015. ³⁰ 105 Orthophosphate (PO_4^{3-}) and dissolved inorganic nitrogen (DIN) enter the Delta primarily 106 through the Sacramento and San Joaquin rivers and from municipal WWTP inputs. Annual 107 nutrient export from the Delta region has been estimated as approximately 30 thousand 108 kg d^{-1} of total nitrogen (varying with flow²⁹), with 90% of ammonium (NH₄⁺) originating 109 solely from the Sacramento Regional WWTP.²⁶ Although nitrogen and phosphorus inputs 110 are considerable, primary production is relatively low and not nutrient-limited. 25,31 The re-111 sistance of SFE to the negative effects of eutrophication has historically been attributed to its unique physical and biological characteristics, including strong tidal mixing that limits stratification in the larger estuary 7,32 and limits on phytoplankton growth from high turbidity and filter-feeding by bivalve mollusks in the northern portion. 32,33 However, recent

The Delta region drains a 200 thousand km² watershed into the SFE, which is the largest

water quality trends have suggested that resistence to nutrient inputs is decreasing given documented changes in chlorophyll biomass, ³⁴ increased occurrence of hypoxic conditions, ³⁵ and increased abundance of phytoplankton species associated with harmful algal blooms. ^{36,37} These recent changes have been attributed to biological invasions ³⁸ and departures from the historical flow record, ^{27,39} among other factors acting at global scales (e.g., variation in sea surface temperatures). ³⁴ The role of nutrients in stimulating primary production in SFE has been the focus of several recent investigations. ^{40–42}

Quantitative descriptions of nutrient dynamics in the Delta are challenging given multiple 123 sources and the volume of water that is exchanged with natural and anthropogenic processes. 124 A comprehensive evaluation using mass-balance models to describe nutrient dynamics in the 125 Delta demonstrated that nitrogen enters the system in different forms and is processed at 126 different rates before export or removal.²⁹ For example, a majority of ammonium entering 127 the system during the summer is nitrified or assimilated, whereas a considerable percentage 128 of total nitrogen load to the Delta is exported. Although, the focus of our analysis is not to 129 quantify sources or sinks of nitrogen species, a quantitative evaluation of long-term trends 130 will provide a more comprehensive historical interpretation to hypothesize the effects of 131 future changes in the context of known dynamics. Nutrients in the Delta also vary with 132 seasonal and annual changes in the delivery of water inflows and water exports directly from the system. ^{24,26} Our analysis explicitly accounts for the effects of flow changes on nutrient 134 response to better understand variation both within the Delta and potential mechanisms of 135 downstream tranport. 136

137 Data sources

Multi-decadal time series of nutrients and flow records were used to develop a quantitative description of nitrogen trends in the northern SFE region. The Interagency Ecological Program (IEP) is a consortium of state and federal agencies that have maintained the Environmental Monitoring Program (EMP) in the Delta region since 1975.⁴³ The EMP

collects monthly water quality samples at 19 stations in the Delta, Suisun Bay, and northeastern San Pablo Bay. Water samples were collected using a Van Dorn sample, a sub-143 mersible pump, or a flow through system depending on site. All samples were processed 144 with standard QA/QC at the California Department of Water Resources Bryte Laboratory 145 in Sacramento. 43 Nutrient time series from 1976 to 2013 were obtained from the IEP website 146 (http://water.ca.gov/bdma/meta/Discrete/data.cfm) for ten discrete sampling stations 147 (Figure 1). Stations were grouped by location in the study area for comparison: peripheral 148 Delta stations C3 (Sacramento inflow), C10 (San Joaquin inflow), MD10, P8; interior Delta 149 stations D19, D26, D28; and Suisun stations D4, D6, and D7. These stations were chosen 150 based on continuity of the water quality time series and significance of their geographic lo-151 cation for understanding regional trends. Time series were complete for all stations except 152 for an approximate ten year gap from 1996-2014 for D19. Data were minimally processed, 153 with the exception of averaging replicates that occurred on the same day. The three nitrogen 154 analytes that were evaluated were ammonium, nitrite/nitrate, and DIN (as the sum of the 155 former two). Less than 3\% of all observations were left-censored, although variation was 156 observed between analytes and location. The ammonium time series had the most censored 157 observations at sites C10 (25.4% of all observations), MD10 (18.1%), D28 (17.8%), D19 (12%), and D7 (7.9%).

Daily flow estimates for the Delta region were obtained from the Dayflow software pro-160 gram that provides estimates of average Delta outflow. 44 Because of the complexity of water 161 inflow, exports, and outflows from the Delta, the Dayflow program combines observations 162 with estimates based on mass balance to reconstruct historical and daily flow estimates. 163 The WRTDS models described below require a matched flow record with the appropriate 164 station to evaluate nutrient trends. Given the complexity of inflows and connectivity of the 165 system, only the inflow estimates from the Sacramento and San Joaquin rivers were used 166 as measures of freshwater influence at each station. Initial analyses indicated that model 167 fit was not significantly improved with flow estimates from locations closer to each station, 168

nor was model fit improved using lagged times series. As such, the Sacramento daily flow time series was used to account for flow effects at C3, D19, D26, D28, and MD10, and the San Joaquin time series was used for C10 and P8 based on station proximity to each inflow. Salinity observations at D4, D6, and D7 in Suisun Bay were used as more appropriate measures of freshwater variation, given the stronger tidal influence at these stations. Salinity has been used as a tracer of freshwater influence for the application of WRTDS models in tidal waters.²¹

176 Analysis method and application

A total of thirty WRTDS models were created, one for each nitrogen analyte at each station.

The functional form of WRTDS is a simple regression ¹⁵ that models the log-transformed response variable as a function of time, flow, and season:

$$\ln(N) = \beta_0 + \beta_1 t + \beta_2 \ln(Q) + \beta_3 \sin(2\pi t) + \beta_4 \cos(2\pi t) \tag{1}$$

where N is one of three nitrogen analytes, time t is a continuous variable as decimal time 180 to capture the annual or seasonal trend, and Q is the flow variable (either flow or salinity 181 depending on station). The seasonal trend is modeled as a sinusoidal component to capture 182 periodicity between years. The WRTDS model is a moving window regression that fits unique 183 parameters at each observation point in the time series. A unique set of weights is used for 184 each regression to control the importance of observations used to fit the model relative to 185 the observation at the center of the window. The weights are based on a scaled Euclidean 186 distance to estimate the differences of all points from the center in relation to annual time, 187 season, and flow. The complete model for the time series contains a parameter set for every 188 time step that considers the unique context of the data. As such, predictions from WRTDS 189 are more precise than those from more conventional models that fit a single parameter set 190 to the entire time series. 21,45 The WRTDS models applied herein were based on a tidal adaptation of the original method²¹ and were fit to describe the conditional mean response using a weighted Tobit model for left-censored data.⁴⁶ All analyses used the WRTDStidal package written by the authors for the R statistical programming language.⁴⁷?

A hallmark of the WRTDS approach is the description of flow-normalized trends that 195 are independent of variation from freshwater inflows. Although variation in nutrients can be 196 caused by the combined effects of several variables acting at different temporal and spatial 197 scales, flow-normalization provides a basis for further exploration by removing a critical 198 confounding variable that could affect the interpretation of trends. A flow-normalized value 199 is the average of predictions at a given observation using all flow values that are expected 200 to occur for the relevant month across years in the record. Flow-normalized trends for each 201 analyte at each station were used to describe long-term changes in different annual and 202 seasonal periods. Specifically, flow-normalized trends in each analyte were summarized as 203 both medians and percent changes from the beginning to end of annual groupings from 1976-204 1995 and 1996-2013, and seasonal groupings of March-April-May (spring), June-July-August 205 (summer), September-October-November (fall), and December-January-February (winter) 206 within each annual grouping. These annual and seasonal groupings were chosen for continuity 207 with similar comparisons reported in Ref. 28 and as approximate twenty year midpoints in 208 the time series.

Trends in each annual and seasonal grouping were based on seasonal Kendall tests of the 210 flow-normalized predictions. This test is a modification of the non-parametric Kendall test 211 that accounts for variation across seasons in the response variable. 48 Results from the test 212 can be used to evaluate the direction, magnitude, and significance of a monotonic change 213 within the period of observation. The estimated rate of change per year is also returned 214 as the Theil-Sen slope and was interpreted as the percent change per year when divided by 215 the median value of the response variable in the period of observation.²⁶ Trends in annual 216 groupings were based on all monthly observations within relevant years, whereas seasonal 217 groupings were based only on the relevant months across years. Seasonal Kendall tests were 218

also used to describe trends in the observed data. These trends were compared with those
based on the flow-normalized trends to evaluate the improved ability of WRTDS to describe
trends that are independent of flow. Functions in the EnvStats package in R were used for
the seasonal Kendall tests. 49

Results and Discussion

224 Observed Data

The observed time series for the ten Delta - Suisun Bay stations had substantial variation 225 in scale among the nitrogen analytes and differences in apparent seasonal trends (Figure 2). 226 In general, long-term (inter-annual) trends were not easily observed from the raw data. 227 DIN for most stations was dominated by nitrite/nitrate, whereas ammonium was a smaller 228 percentage of the total. However, C3 had a majority of DIN composed of ammonium and 229 other stations (e.g., P8, D26) had higher concentrations of ammonium during winter months 230 when phytoplankton assimilation is lower.²⁹ By location, observed concentrations of DIN for 231 the entire time series were higher on average for the peripheral stations (C3, C10, MD10, P8; 232 maximum likelihood estimation of mean \pm standard error: 1.04 ± 0.03 mg L⁻¹) and similar for 233 the interior (D19, D26, D28, 0.43 ± 0.01) and Suisun Bay stations (D4, D6, D7, 0.44 ± 0.01). Average concentrations were highest at P8 $(1.63\pm0.05 \text{ mg L}^{-1})$ and lowest at C3 (0.4 ± 0.01) for DIN, highest at P8 (0.28 ± 0.02) and lowest at D28 (0.05 ± 0.003) for ammonium, and highest at C10 (1.4 \pm 0.04) and lowest at C3 (0.15 \pm 0.004) for nitrite/nitrate. Mean observed 237 concentrations were also higher later in the time series for all analytes. For example, average 238 DIN across all stations was $0.61\pm0.01 \text{ mg L}^{-1}$ for 1976-1995, compared to 0.7 ± 0.01 for 1996-239 2013. Seasonal changes across all years showed that nitrogen concentrations were generally 240 lower in the summer and higher in the winter, although observed patterns were inconsistent 241 between sites. For example, site MD10 had distinct seasonal spikes for elevated DIN in the 242 winter, whereas other stations had less prominent seasonal maxima (e.g., C3, D7, Figure 2). 243

244 Trends

Application of seasonal Kendall tests to evaluate trends in observed data provided explicit in-245 formation on the direction, magnitude, and statistical significance of changes between years. 246 Trends estimated from the observed data for 1976-1995 and 1996-2013 varied considerably 247 between sites and analytes (Figure 3). Significant trends were observed from 1976-1995 for 248 eight of ten sites for DIN (seven increasing, one decreasing), eight sites for ammonium (six 249 increasing, two decreasing), and six sites for nitrite/nitrate (five increasing, one decreasing). 250 Decreasing trends were more common for the observed data from 1996-2013. Eight sites 251 had significant trends for DIN (four increasing, four decreasing), seven sites for ammonium 252 (five increasing, two decreasing), and eight sites for nitrite/nitrate (four increasing, four de-253 creasing). P8 had a relatively large decrease in ammonium (-8.3%) change per year) for the 254 second annual period compared to all other sites, which is described in detail in the next 255 section. Trends by season were similar such that increases were generally observed in all 256 seasons from 1976-1995 (Figure S1) and decreases were observed for 1996-2013 (Figure S2). 257 Trends for the seasonal comparisons were noisier and significant changes were less common 258 compared to the annual comparisons. 259

Relationships between flow and observed water quality are complex and can change signif-260 icantly through space and time. 15,20 These principles have been demonstrated for monitoring 261 data in the Delta region, ^{26,28,29} suggesting that trend analyses using the observed time se-262 ries are confounded by flow effects. As such, a comparison of flow-normalized results from 263 WRTDS relative to observed data identified changes in the magnitude, significance, and di-264 rection of trends. For all sixty trend comparisons in Figure 3 (flow-normalized values in Table 265 S1) regardless of site, nitrogen analyte, and time period, thirteen comparisons had trends 266 that were insignificant with the observed data but significant with flow-normalized results, 267 whereas only one trend changed to insignificant. This suggests that time series that include 268 flow effects had sufficient noise to obscure or prevent identification of an actual trend of a 269 water quality parameter. Further, changes in the magnitude of the estimated percent change 270

per year were also apparent for the flow-normalized trends, such that fourteen comparisons showed an increase in magnitude (more negative or more positive) and twenty five had a decrease (less positive or less negative) compared to observed trends. Eleven comparisons showed a trend reversal from positive to negative estimated change, nine sites went from no 274 change to negative estimated change, and one site went from no change to a positive trend for 275 the flow-normalized results. Differences by season in the observed relative to flow-normalized 276 trends from WRTDS were also apparent (Figures S1 and S2 and Tables S2 and S3). The 277 most notable changes were an overall decrease in the estimated trend for most sites in the 278 summer and fall seasons for 1996-2013, including an increase in the number of statistically 279 significant trends. 280

Differences in apparent trends underscore the importance of considering flow effects in 281 the interpretation of environmental changes, particularly if trend evaluation is used to assess 282 the effects of nutrients on ecosystem health or the effectiveness of past nutrient management 283 actions. Our results demonstrated the potential to misinterpret trends if flow effects are not 284 considered, where the misinterpretation could vary from a simple change in the magnitude 285 and significance of a trends, to more problematic changes where the flow-normalized trend 286 could demonstrate a complete reversal relative to the observed (e.g., DIN trends for all Suisun 287 stations from 1996-2013, Figure 3). A more comprehensive evaluation of flow in the Delta 288 demonstrated that flow contributions of different end members vary considerably over time 289 at each station.²⁹ For example, flow at MD10 represents a changing percentage by season 290 of inputs from the Sacramento, San Joaquin, Cosumnes, Mokelumne rivers, and agricultural 291 returns. For simplicity, water quality observations in our analyses were matched with large-292 scale drivers of flow into the Delta where most sites were matched to Sacramento or San 293 Joaquin daily flow estimates. Given that substantial differences with flow-normalized results 294 were apparent from relatively coarse estimates of flow contributions, more precise differences 295 could be obtained by considering the influence of multiple flow components at each location. 296 Output from the Dayflow software program 44 provides a complete mass balance of flow in 297

the Delta that could be used to develop a more comprehensive description of flow-normalized trends that considers changing contributions over time.

300 Selected examples

Two stations were chosen to demonstrate use of WRTDS to develop a more comprehensive description of decadal trends in the Delta. The selected case studies focused on 1) effects of wastewater treatment upgrades upstream of P8, and 2) effects of biological invasion on nutrient dynamics in Suisun Bay using observations from D7. Each case study is built around hypotheses that results from WRTDS models were expected to support, both as a general description and for additional testing with alternative methods.

307 Effects of wastewater treatment

Significant efforts have been made in recent years to reduce nitrogen loading from regional 308 WWTPs given the disproportionate contribution of nutrients relative to other sources (e.g., 309 watershed agricultural load, sediment flux, etc.)^{29,50} Several WWTPs in the Delta have 310 recently been or are planned to be upgraded to include tertiary filtration and nitrification 311 to convert biologically available ammonium to nitrate. The City of Stockton WWTP was 312 upgraded in 2006 and is immediately upstream of station P8 (Figure 1),²⁸ which provides 313 a valuable opportunity to assess how nutrient or nutrient-related source controls and water 314 management actions have changed ambient concentrations downstream. A modal response 315 of nutrient concentrations at P8 centered around 2006 is expected as a result of upstream 316 WWTP upgrades, and water quality should exhibit 1) a shift in the ratio of the components 317 of DIN from the WWTP before/after upgrade, and 2) a flow-normalized annual trend at P8 318 to show a change concurrent with WWTP upgrades. 319

Effluent concentrations measured from 2003 to 2009 from the Stockton WWTP showed a gradual reduction in ammonium concentration relative to total DIN (Figure S3). Ammonium and nitrate concentrations were comparable prior to 2006, whereas nitrate was a majority

of total nitrogen after the upgrade, with much smaller percentages from ammonium and nitrite. As expected, flow-normalized nitrogen trends at P8 shifted in response to upstream 324 WWTP upgrades (Figure 4a), with ammonium showing an increase from 1976 followed by 325 a large reduction in the 2000s. Interestingly, nitrite/nitrate concentrations also showed a 326 similar but less dramatic decrease despite an increase in the WWTP effluent concentrations 327 following the upgrade. Percent changes from seasonal Kendall tests on flow-normalized 328 results showed that both nitrogen species increased prior to WWTP upgrades (2% per year 329 for nitrite/nitrate, 2.8% for ammonium), followed by decreases after upgrades (-1.9% for 330 nitrite/nitrate, -16.6% for ammonium, Table 1). Seasonally, increases prior to upgrades 331 were highest in the summer for nitrite/nitrate (2.4%) and in the fall for ammonium (4.9%). 332 Similarly, seasonal reductions post-upgrade were largest in the summer for nitrite/nitrate 333 (-4.3%) and largest for ammonium in the winter (-26.7%). 334

Relationships of nitrite/nitrate with flow described by WRTDS showed an inverse flow 335 and concentration dynamic with flushing or dilution at higher flow (Figure 4b). Seasonal vari-336 ation was even more apparent for ammonium, although both nitrite/nitrate and ammonium 337 typically had the highest concentrations at low flow in the winter (January). Additionally, 338 strength of the flow/nutrient relationship changed between years. Nitrite/nitrate typically had the strongest relationship with flow later in the time series (i.e., larger negative slope), whereas ammonium had the strongest relationship with flow around 2000 in January. Using WRTDS, an empirical link is created between upstream changes and observed effects 342 downstream that is characterized by differences in analytes between years and season. A 343 general conclusion is that ammonium reductions were concurrent with WWTP upgrades, 344 but the reduction was most apparent at low-flow in January. These dynamics are difficult 345 to characterize from the observed time series, and further, results from WRTDS can be used 346 to develop additional hypotheses of factors that influence nutrient concentrations at P8. For 347 example, estimated ammonium concentrations in July were low for all flow levels which sug-348 gests either nitrogen inputs were low in the summer or nitrogen was available and uptake 349

by primary consumers was high. Seasonal patterns in the relationship between flow and nitrite/nitrate were not as dramatic as compared to ammonium, and in particular, low-flow events in July were generally associated with higher concentrations. This could suggest that ammonium concentrations at P8 are driving phytoplankton production at low flow during warmer months, and not nitrite/nitrate given the higher estimated concentrations in July at low flow. As such, these simple observations from WRTDS provide quantitative support of cause and effect mechanisms of nutrient impacts on potentially adverse environmental conditions as they relate to nutrient-related source controls upstream.

358 Effects of biological invasions

Invasion of the upper SFE by the Asian clam Potamocorbula amurensis in 1986 caused severe 359 changes in phytoplankton abundance and species composition. Reduction in phytoplankton 360 biomass has altered trophic networks in the upper SFE and is considered an important 361 mechanism in the decline of the protected delta smelt (Hypomesus transpacificus) and other 362 important fisheries. 51,52 Changes in the physical environment have also occurred with the most notable invasion effect being increased water clarity following a reduction of phyto-364 plankton by biofiltration.⁵² The clams are halophilic such that drought years are generally 365 correlated with an increase in biomass and further upstream invasion of the species. ^{27,53} We 366 hypothesized that WRTDS models applied to water quality observations in the upper estuary 367 would show 1) a decline in annual, flow-normalized chlorophyll concentrations over time co-368 incident with an increase in abundance of invaders, and 2) variation in the chlorophyll/clam 369 relationship through indirect or direct controls of flow. The application of WRTDS to water 370 quality observations at station D7 in Suisun Bay and comparison with clam abundance and 371 biomass data (see Ref 33) provides an approach to assess the competing effects of climate 372 variability, hydrology, and ecology on ambient conditions. 373

Results from WRTDS demonstrated complex relationships between clam abundance and chlorophyll concentrations, which were further affected by flow changes over time (Figure 5).

Invasion in the 1980s showed a clear displacement of the native Corbicula fluminea with establishment of P. amurensis (Figure 5a), where biomass of the latter was negatively as-377 sociated with flow from the Sacramento river (Figure 5b). The increase in clam abundance 378 was associated with a notable decrease in annually-averaged chl-a from WRTDS results (Fig-379 ure 5c). A seasonal shift in the flow-normalized results was also observed such that chl-a 380 concentrations were generally highest in July/August prior to invasion, whereas a spring 381 maximum in April was more common in recent years (Figure 5f). An increase in annually-382 averaged silicon dioxide (Figure 5e) was coincident with the chl-a decrease, with the largest 383 increases occurring in August (Figure 5g). These relationships suggest that diatoms were the 384 dominant genera early in the time series, particularly in late summer, whereas the spring 385 peak observed in later years represents a shift to an earlier seasonal maxima. This sup-386 ports past research that showed a decrease in silica uptake by diatoms following invasion.^{7,54} 387 Further, DIN trends were similar to silicon-dioxide in both annual and seasonal changes 388 (compare Figures 5e and 5e with 5d and 5g) such that an increase in both nutrients earlier 389 in the time series corresponded with the decrease in chl-a. Overall, these results suggest the 390 increase in nutrient concentrations are from decreased uptake by phytoplankton in Suisun 391 Bay. 392

The relationship of chl-a with clam biomass was significant (Figure 5g) with lower chl-a 393 associated with higher biomass, confirming results from earlier studies. 32,55 However, the effect of flow on both clams and phytoplankton as a top-down or bottom-up control changed 395 throughout the time series. The chl-a/flow relationship showed that increasing flow (de-396 creasing salinity) was associated with a slight increase in chl-a followed by a decrease early 397 in the time series (Figure 5j), whereas overall chl-a was lower but a positive association with 398 flow (negative with salinity) was observed later in the time series. In the absence of benthic 399 grazing prior to invasion, this dynamic suggests that chl-a production may be limited at low 400 flow as less nutrients are exported from the Delta, stimulated as flow increases, and reduced 401 at high flow as either nutrients or phytoplankton biomass are exported to the larger bay. 402

Following clam invasion, chl-a concentrations were reduced by grazing but showed a positive and monotonic relationship with increasing flow. The increase in clam abundance was con-404 current with decline in chl-a concentration, although variation in abundance between years 405 was also observed. Clam abundance was reduced during high flow years in the late 1990s, 406 2006, and 2011 (5a). In the same years, WRTDS predictions for chl-a were higher than the 407 flow-normalized component (Figure 5c), which further suggests a link between increased flow 408 and phytoplankton production. As such, chl-a production in early years is directly related 409 to flow, whereas the relationship with flow in later years is indirect as increased flow reduces 410 clam abundance and releases phytoplankton from benthic grazing pressure. These relation-411 ships have been suggested by others, ^{26,53,55} although the precise mechanism demonstrated by 412 WRTDS provides a quantitative description of factors that drive water quality in the Delta. 413 As demonstrated by both case studies and the overall trends across all stations, wa-414 ter quality dynamics in the Delta are complex and driven by multiple factors that change 415 through space and time. At a minimum, WRTDS provides a description of change by focusing on high-level forcing factors that explicitly account for annual, seasonal, and flow 417 effects on trend interpretations. We have demonstrated the potential for imprecise or in-418 accurate conclusions of trend tests that focus solely on observed data and emphasize that flow-normalized trends have more power to quantify change. Combined with additional data, WRTDS results can support hypotheses that lead to a more comprehensive understanding 421 of ecosystem dynamics. Additional factors to consider include the effects of large-scale cli-422 matic patterns, more detailed hydrologic descriptions, and additional ecological components 423 that affect trophic interactions. For example, a more rigorous matching of flow time series 424 with water quality observations at each station that considers varying source contributions 425 over time could provide a more robust description of flow-normalized results. Alternative 426 methods for time series analysis could also be used to address a wider range of questions, 427 particularly those with more generic structural forms that can explicitly include additional 428 variables (e.g., generalized additive models). 22 Overall, statistical interpretations of multiple 429

factors can provide a basis for quantitative links between nutrient loads and adverse effects
on ecosystem conditions, including the identification of thresholds for the protection and
restoration of water quality.

433 Acknowledgments

We thank the staff of the San Francisco Estuary Institute and the Delta Regional Monitoring
Program. We thank Larry Harding for providing comments on an earlier draft. This study
was funded, reviewed, and approved for publication by the US EPA, National Health and
Environmental Efects Research Laboratory. The views expressed in this paper are those of
the authors and do not necessarily reflect the views or policies of the US EPA.

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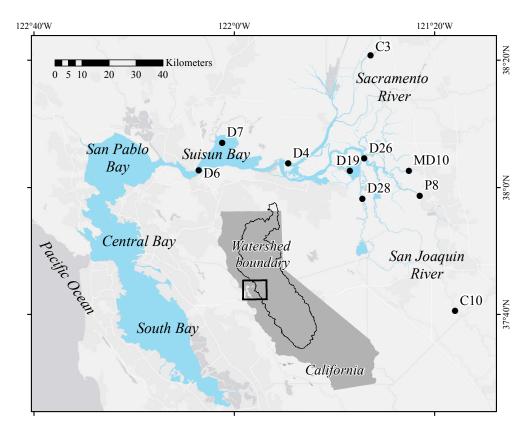


Figure 1: The San Francisco Estuary and Delta region with monitoring stations used for analysis. The Delta drains the combined watersheds of the Sacramento and San Joaquin rivers (inset). All data were obtained from the Interagency Ecological Program website (http://water.ca.gov/bdma/meta/Discrete/data.cfm). 43

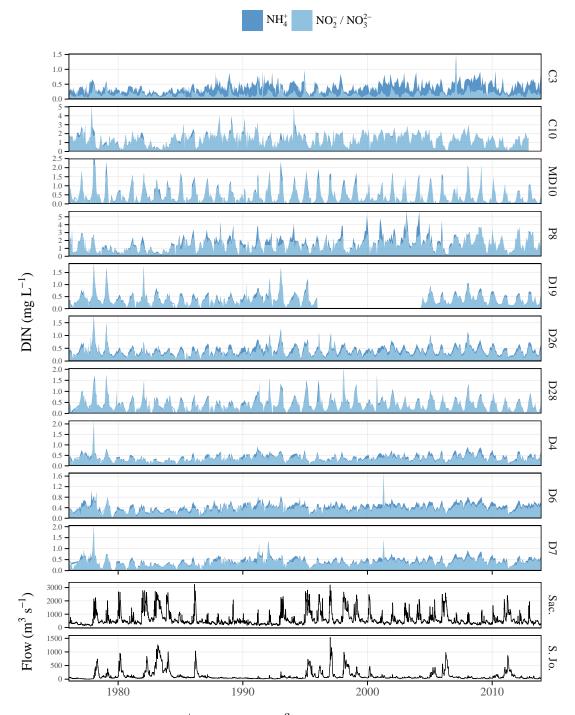


Figure 2: Observed DIN ($\mathrm{NH_4^+} + \mathrm{NO_2^-/NO_3^{2-}}$) from ten stations in the upper SFE Delta and flow from the Sacramento and San Joaquin rivers. Data were collected monthly and evaluated with WRTDS models using daily flow estimates from 1976 to 2013. Note different y-axis scales. See Figure 1 for station locations.

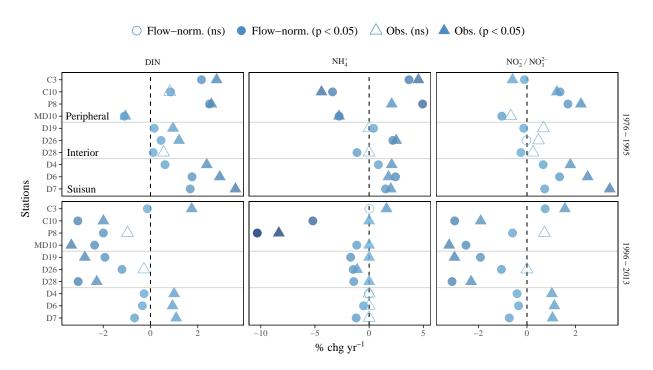


Figure 3: Results from seasonal Kendall tests on observed data (triangles) and flow-normalized predictions (circles) from WRTDS for nitrogen analytes. Results are shown as the percent change per year as the estimated Theil-Sen slope divided by the median for a given aggregation period (significance evaluated at $\alpha = 0.05$, based on τ). Trends are shown separately for different annual groupings. See Figures S1 and S2 for seasonal groupings.

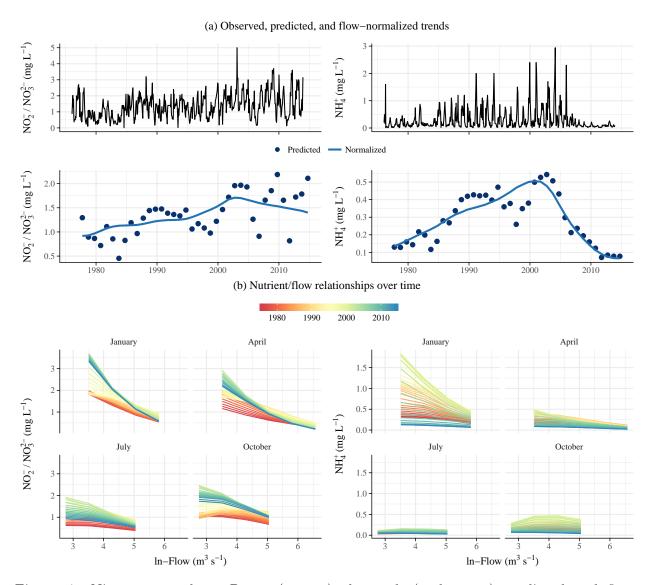


Figure 4: Nitrogen trends at P8 as (a, top) observed, (a, bottom) predicted and flow-normalized estimates from WRTDS, and (b) relationships with flow over time from WRTDS. Nitrite/nitrate trends are on the left and ammonium trends are on the right. Wastewater treatment plant upgrades at the City of Stockton (San Joaquin County) were completed in 2006 (Figure S3).

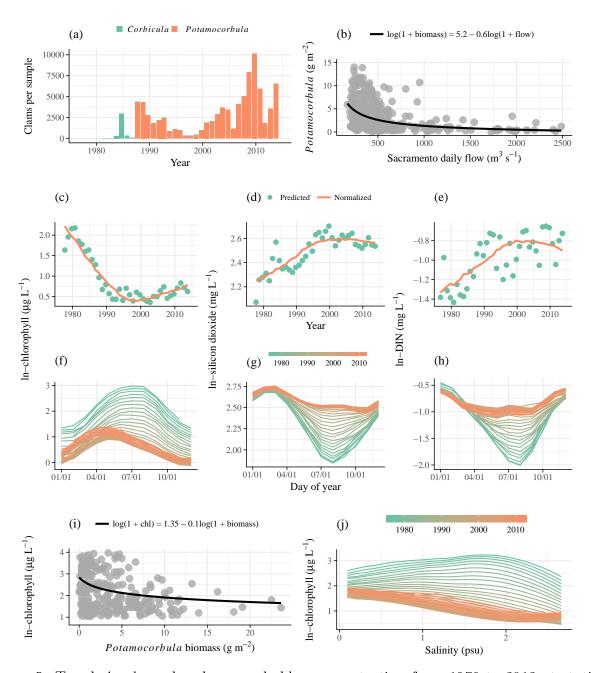


Figure 5: Trends in clam abundance and chl-a concentration from 1976 to 2013 at station D7 in Suisun Bay. Invasion by Potamocorbula amurensis clams in the late 1980s and displacement of Corbicula fluminea was shown by changes in clam density (a, annual means), with biomass linked to salinity (b). A decrease in chl-a concentration was also observed by changes in annual (c) and seasonal trends (f) based on WRTDS results. Reductions in chl-a concentration were coincident with an increase in SiO₂ and DIN concentrations (d, e), with the greatest increases in August (g, h). A significant (p < 0.001) relationship between clam biomass and chl-a concentration is shown in subfigure (i). Flow relationships with chl-a concentration shown by WRTDS have also changed over time (j, observations from June).

Table 1: Summaries of flow-normalized trends in nitrite/nitrate and ammonium (mg L⁻¹) concentrations before and after WWTP upgrades upstream of station P8. Upgrades were completed in 2006 at the City of Stockton WWTP (San Joaquin County, Figure S3). Summaries are medians and percent change per year in parentheses (increasing in bold-italic). Changes and significance estimates are based on seasonal Kendall tests of flow-normalized results within each time period. Increasing values are in bold-italics. Months for each season are Spring: MAM, Summer: JJA, Fall: SON, Winter: DJF.

| Period | $\mathrm{NO}_2^-/\mathrm{NO}_3^{2-}$ | | \mathbf{NH}_4^+ | |
|-------------------------|--------------------------------------|---------------|-------------------|---------------|
| | Median | % change | Median | % change |
| Annual | | | | |
| 1976-2006 | 1.3 | 2** | 0.2 | 2.8** |
| 2007-2013 | 1.4 | -1.9** | 0.1 | -16.6** |
| Seasonal, pre | | | | |
| Spring | 1.2 | <i>1.6</i> ** | 0.2 | <i>1.4</i> ** |
| Summer | 1 | 2.4** | 0.1 | 3.3^{**} |
| Fall | 1.3 | 2.2** | 0.2 | <i>4.9</i> ** |
| Winter | 1.5 | <i>2.1</i> ** | 0.7 | <i>4.8</i> ** |
| Seasonal, post | | | | |
| Spring | 1.3 | -1.6** | 0.1 | -16.2** |
| Summer | 0.9 | -4.3** | 0.1 | -15.7** |
| Fall | 1.5 | -1.7** | 0.1 | -19.3** |
| Winter | 2.2 | -0.8** | 0.2 | -26.7** |

p < 0.05; *p < 0.005