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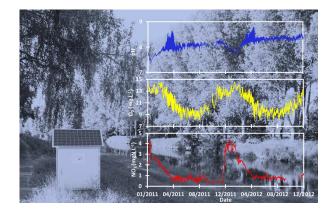
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1	Sensors in the stream: the high-frequency wave of the present
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33 Abstract

34 New scientific understanding is catalysed by novel technologies that enhance measurement 35 precision, resolution or type, and that provide new tools to test and develop theory. Over the last 50 36 years, technology has transformed the hydrologic sciences by enabling direct measurements of 37 watershed fluxes (evapotranspiration, streamflow) at time scales and spatial extents aligned with 38 variation in physical drivers. High frequency water quality measurements, increasingly obtained by in-situ water quality sensors, are extending that transformation. Widely available sensors for some 39 40 physical (temperature) and chemical (conductivity, dissolved oxygen) attributes have become 41 integral to aquatic science, and emerging sensors for nutrients, dissolved CO₂, turbidity, algal 42 pigments, and dissolved organic matter are now enabling observations of watersheds and streams at 43 timescales commensurate with their fundamental hydrological, energetic, elemental, and biological 44 drivers. Here we synthesize insights from emerging technologies across a suite of applications, and 45 envision future advances, enabled by sensors, in our ability to understand, predict, and restore 46 watershed and stream systems.

47



48 TOC Art

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51	Recent progress in in-situ sensor monitoring
52	Just over a decade ago, Kirchner et al. ¹ envisioned the hydrologic sciences being
53	transformed by the increased availability of stream chemistry measurements at time scales
54	commensurate with hydrologic forcing, a theme echoed in the U.S. National Research Council's
55	"Challenges and Opportunities in the Hydrologic Sciences" in 2012
56	(http://www.nap.edu/read/13293/chapter/1). At the same time, ecologists were recognizing the
57	transformative potential of sensors that allow ecosystem processes to be measured at time and
58	space scales that match relevant physical, chemical and biological drivers ²⁻⁴ . The vision of Kirchner
59	et al. ¹ has been realised, in part, with significant progress in estimating solute residence times in
60	watersheds ⁵ , but it is the converging vision across hydrological, biogeochemical, biological and
61	ecological disciplines that highlights the significant intellectual payoff from new sensor technologies
62	in watershed and stream science. Now stream water chemistry data are available every hour, or
63	even every minute across a broad range of analytes, and commensurate biological data are available
64	at fortnightly to daily intervals for sustained periods greater than one year. These advances allow the
65	study of multiple solutes at sub-daily intervals, not just single solute time series, and enable
66	interpretations and hypothesis testing of ideas around river biogeochemistry, biology and ecology, in
67	addition to catchment signals. These novel measurements have revealed complex temporal
68	dynamics that were obscured by traditional sampling frequencies ⁶⁻⁸ and have enabled new insights
69	into the inner-workings of watersheds and streams.
70	While automated collection and traditional laboratory processing of discrete samples have
71	yielded enormously informative sub-daily data ⁸⁻⁹ , the transformation of stream and watershed
72	science will occur primarily in response to increasing availability of automated in situ sensors.
73	Indeed, electrode-based measurements of pH, conductivity, temperature, and dissolved oxygen (DO)
74	have been available for over half a century 10 and are now essential tools for stream and watershed

75 studies; however, sensor technology has been extended through the development of other methods

76	such as optical, wet analytical chemical or flow cytometry techniques (a laser- or impedance-based,
77	biophysical technology employed in cell counting), recent advances in field deployment engineering
78	(anti-fouling, batteries, micropumps), and electronics (detectors, emitters) that have reduced costs.
79	This, in turn, has increased the number of sites at which in situ measurements are now made.
80	Among the solutes for which sensors are most widely available is nitrate. Early colorimetric based
81	sensors ¹¹ for nitrate were constrained by performance and reagent wastes, and have largely given
82	way to spectrophotometers ¹² enabling very high frequency (0.5 Hz, samples per second) sampling
83	that has proven enormously informative for understanding riverine dynamics ¹³⁻¹⁵ . For other solutes,
84	wet analytical chemistry remains the most viable approach, with "lab-on-a-chip" sensors lowering
85	power requirements and reducing the interferences that are intrinsic in optical absorbance
86	measurements ^{16,17} . For example, measurements of orthophosphate using standard reagent-based
87	colorimetry has emerged as a robust field-deployable technology, permitting automated hourly
88	sampling and a host of attendant informative inferences enabled by this increase in temporal
89	resolution ^{6,18} . Other deployable optical sensors include fluorimeters that can measure chlorophyll-a
90	and other photosynthetic pigments, as well as fluorescent dissolved organic matter ¹⁹ ; while these
91	sensors have a long history in marine and estuarine settings, their use in streams and small
92	watersheds has revealed a variety of novel insights ²⁰ . Indeed, Fast repetition rate Fluorimetry (FrrF),
93	a technique which measures the variability of light emission from chlorophyll a, can be used to
94	measure photosynthetic rates in situ which reduce when algae are stressed due to the prevailing
95	environmental conditions (e.g. drought), and these measurements are supported by weekly
96	(imaging) flow cytometry (that can discriminate and assess abundance among phytoplankton and
97	phytobacterial functional groups) and environmental DNA techniques (that can characterise
98	microbial communities and detect invasive species) ^{21,22} . In short, the suite of widely used
99	parameters that hydrologists, geochemists, and stream ecologists consider relevant is almost
100	uniformly possible in real time and at high spatial or temporal resolution.

Environmental Science & Technology

101	These new data sets have the ability to transform our understanding of a diverse range of
102	fundamental aquatic processes, from watershed dynamics to nutrient spiralling to ecosystem
103	response to disturbance. The potential for sensors to unravel ecosystem functioning and realize
104	improved environmental management was illustrated by the recent commissioning of a "national
105	nutrient sensor challenge" by the White House Office of Science and Technology Policy. This effort
106	seeks to enable the next generation of long-term deployable, high accuracy, high precision in situ
107	sensors, and to drive down costs to ensure broad adoption by academic, private and government
108	scientists ²³ .

109 Here we present four examples (inferring nutrient sources and transport, measuring in situ 110 nutrient processing, detecting ecological effects, and temporal scaling of solute export), spanning 111 continents and time scales, in which recent utilization of sensor technologies have advanced our 112 understanding of stream and watershed systems. While rivers and their watersheds are our focus, 113 the use of novel measurement technologies in other aquatic ecosystems such as lakes, estuaries and oceans has been equally transformative ²⁻⁴. In addition to the insights that have already been made, 114 115 we highlight ongoing trends in sensor development and suggest areas in which sensors will enable 116 new insights and allow tests of watershed and ecological theory.

117

118 Identifying nutrient sources and transport pathways in watersheds

Sub-daily monitoring of nutrient hydrochemistry has traditionally utilized automatic watersamplers, but these are expensive to run in terms of regular sample collection and subsequent
laboratory analysis, and can have chemical and biological stability issues during sample storage ²⁴,
which usually is in the range of days. Through the deployment of *in situ* sensor and colorimetric
based auto-analyser technology, Bowes et al. ²⁵ measured hourly total reactive phosphorus (TRP)
and nitrate concentrations and used these to characterise, on a storm-by-storm basis, the nutrient

- source changes to a rural river in southern England over a two year period by analysing the
- 126 hysteresis in the relationship between concentration and flow during storm events when the stream
- 127 or river flow increases and then recedes. Differences in the hysteresis behavior between storms
- 128 provide information on nutrient sources and pathways and the findings are summarized in Figure 1.
- 129 In this case study of the River Enbourne in the UK, the results highlighted the importance of the
- acute mobilisation of sewage-derived phosphorus in bed sediment and the large diffuse phosphorus
- 131 inputs entering the stream from manure applications during May storms, thereby helping to target

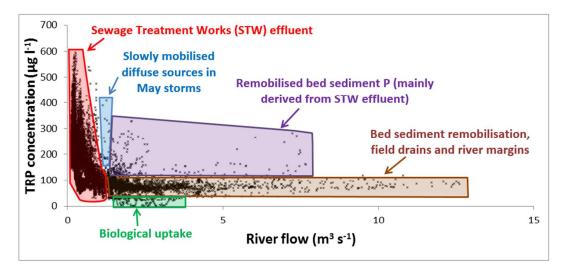
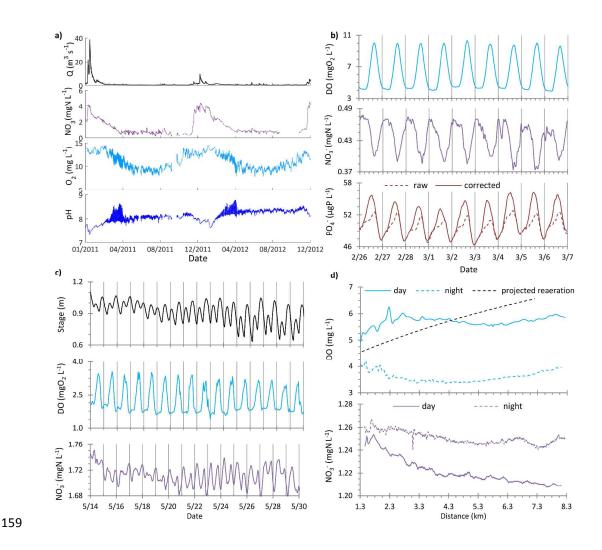


Figure 1: Phosphorus sources to the River Enborne (southern England) identified using two years of hourly
 total reactive phosphorus (TRP) and flow data. The clusters were derived from storm hysteresis analysis ²⁴.

135 future remediation measures. Additional analysis showed clear double-peaked diel phosphorus and 136 nitrate cycles during low flows, which pointed to chronic pollution related to the daily pattern of effluent discharges from sewage treatment works and septic tank systems ²⁶. Recently Mellander et 137 138 al. ²⁷ also used 20 minute phosphorus concentration monitoring to identify that subsurface flows in 139 bedrock cracks were the dominant phosphorus transport pathways in a karst landscape in County 140 Mayo, Ireland, and Mellander et al.²⁸ demonstrated that phosphorus could predominantly be 141 transferred to streams via groundwater during winter in the south of Ireland using sub daily 142 colorimetric based auto-analyser measurements. Both these studies allowed a deeper

143	understanding of phosphorus transfer pathways and retention in the aquifer to be developed with
144	advice on location of critical source areas for phosphorus loss in a Karst landscape resultant.
145	Furthermore, regular sub-daily measurements of dissolved organic matter (DOM)
146	composition (which plays a central role in carbon dynamics and participates in the complexation of
147	trace metals and the mobilization of pollutants) by in situ Chromophoric Dissolved Organic Matter
148	(CDOM) fluorescence have revealed a complex short-term variability in DOM composition. This
149	variability is a function of source, flow pathway and instream photochemical and biologically
150	mediated processes ^{20,29} .
151	Together, these results highlight that sub-daily observations have high potential to
152	accurately make source assignments and that watershed management can greatly benefit from high
153	frequency measurements to identify site specific loss mechanisms and pathways and potential
154	legacy issues ^{25,28} , such as groundwater nitrogen and stream bed phosphorus retention. The
155	development of sensors for an increasing range of water quality constituents, with more widespread
156	deployment, will lead to a greater ability to fingerprint chemical sources through seasons and
157	individual storm events in the future.

158 Quantifying coupled nutrient processing and metabolism



160 Figure 2. Continuous high frequency oxygen, nitrate and pH data reflect the seasonal pattern of primary production and assimilatory N uptake due to flow and light variability (panel a; adapted from Rode et al. ³²). 161 162 The relative magnitude of diel variation in DO, NO_3^{-} and PO_4^{-} (panel b; adapted from Cohen et al. ¹⁸) is strongly 163 correlated with autotroph stoichiometry. The assimilatory P uptake (PO4 signal must be corrected for 164 precipitation) notably appears temporally de-coupled from primary production and assimilatory N uptake by 165 several hours. Comparison of day-night profiles can be used to partition assimilatory versus dissimilatory pathways (panel c^{33,35}), while high resolution longitudinal profiling (panel d; adapted from Hensley et al.³³⁾ 166 167 has been used to identify spatial heterogeneity with N processing in a tidal river shown to be strongly 168 influenced by residence time variation.

Drainage networks are not passive conduits, but are important for chemical retention and
 transformation. High frequency data have proven especially useful in quantifying in-stream nutrient

Environmental Science & Technology

171	processing and understanding the stoichiometric coupling of autotrophic uptake across the periodic
172	table. In addition to seasonal patterns (Fig. 2a), finely resolved time-series have revealed strong diel
173	nitrate variability (Fig. 2b) similar to that observed for DO and interpreted as autotrophic N
174	assimilation ^{14,15} . Actively measuring nutrient uptake rates (e.g., via isotope or nutrient dosing) is
175	complex and expensive, limiting measurements to short (hours to days) periods ³⁰ , typically under
176	steady baseflow conditions, and with a significant bias towards small streams ³¹ . In Florida's spring
177	fed rivers, autotrophic nitrate uptake amounted to less than 20% of total net N retention ¹⁸ . In two
178	central European streams, percentage daily autotrophic N uptake peaked at 47% (agricultural
179	stream) and 75% (forest stream) of the daily N loading input to the stream network of the whole
180	watershed ³² . There were different ranges of autotrophic areal rate of nutrient uptake (U, analogues
181	to the mass of nutrient removed from water per unit area of streambed (m^{-2}) per unit time (d)) with
182	30-160 mg N in the Florida rivers ¹⁴ and 0-270 mg N and 0-97 mg N in the central European
183	agricultural and forest stream ³² , respectively. Dissimilatory pathways such as denitrification, which
184	account for the balance of net retention, were also coupled with primary productivity through
185	secondary relationships such as the availability of labile carbon ¹⁴ . In a separate study, a more
186	complex retention signal (Fig. 2c) arose in a tidal river, representing the convolution of diel
187	assimilatory uptake and tidally varying denitrification based on residence time and benthic surface
188	area ³³ .
400	

189 While early insights into coupled nutrient processing have focused on nitrate dynamics, sensors for other solutes have proven equally valuable. Cohen et al. ¹⁸ used an ortho-phosphate 190 191 sensor, along with optical nitrate, and electrode based DO and specific conductance sensors. The 192 high frequency signals enabled identification and deconvolution of geochemical P-retention 193 pathways that created overlapping diel P signals. The data also revealed clear coupling of N and C 194 assimilation, and that while also coupled, P uptake was not synchronous with the timing of N and C 195 assimilation. This asynchronous N and P assimilation may represent timing differences in protein and 196 ribosome production in aquatic plants. It has been suggested that temporal nutrient coupling occurs

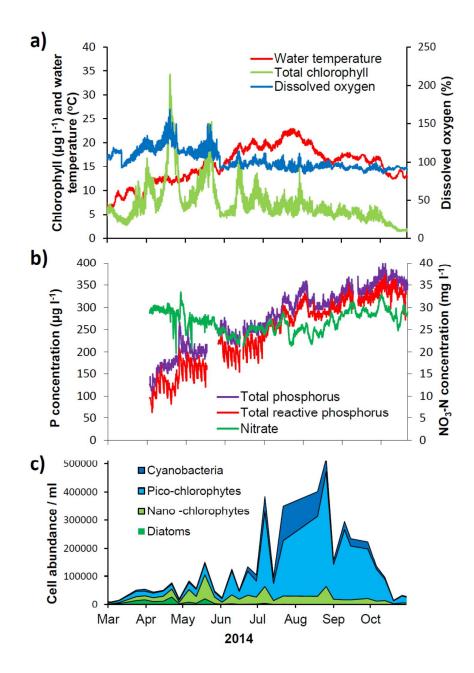
197	only when an input nutrient is limiting and therefore the identification of such temporal coupling,
198	through <i>in situ</i> high frequency monitoring, is a useful indicator of ecosystem limitation status ³⁴ . Diel
199	concentration variation for biologically active trace metals (e.g., Ba, Fe, Mn, and U) has similarly
200	been observed in spring systems in Florida, USA, and suggests that aquatic plant metabolism
201	controls diel and seasonal cycling of metals ³⁵ . As sensors emerge for measuring other solutes such
202	as other nutrients (e.g., Si, Fe, Mg), the organic nutrient forms (e.g. DON, DOP), measurements of
203	total concentrations which include the particulate fraction and therefore all the total nutrient
204	potentially available, and organic pollutants, their dynamics can be compared to metabolic, thermal,
205	flow and photolytic forcing, which will enable a rich new arena for understanding aquatic systems.
206	Sensors can also be applied using an alternative Lagrangian approach ³⁶ . Reach-scale
207	nutrient processing rates such as areal rate of nutrient uptake (U) and uptake length (average
208	distance a nutrient molecule, typically nitrate, moves downstream in dissolved form before being
209	assimilated by the biota) have been estimated from longitudinal changes in stream solute
210	concentrations (Fig. 2d), revealing a large degree of spatial heterogeneity in nutrient uptake which
211	appears related to changes in river morphology ³³ . Furthermore, estimates of reach-scale metabolism
212	and nutrient processing show diel variations along a continuous gradient from headwaters to mouth,
213	as envisioned in the River Continuum Concept ^{37,38} , and highlight how river regulation disrupts the
214	continuum, for example, in terms of stream metabolism through increased total dissolved N uptake
215	below dams ^{38,39} .

These early insights suggest that the emergence of sensor-derived high-frequency time series for multiple solutes will better allow us to observe the stoichiometric coupling of metabolic and geochemical processes and thereby test stoichiometric theory. It will also enable a deeper understanding of variation in retention rates and pathways with flow, temperature and other abiotic drivers across watersheds spanning a gradient of size and geochemical and physical features.

221

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222 Separating the effects of multiple processes on aquatic ecology



223

Figure 3: Combined physical (a), chemical (b) and biological monitoring (a, c) of the River Thames at Goring, UK
(Unpublished Data, supplied by Centre for Ecology & Hydrology, Wallingford, UK, and the UK Environment
Agency). The time axis is the same in the three panels.

The effects of multiple pressures on freshwater ecosystems are difficult to separate because
 of the multiple, interrelated abiotic-biotic interactions. New biological monitoring techniques have

229	been developed that allow the high-frequency characterisation of river plankton composition and
230	function for the first time. Fluorimeters now reliably measure total chlorophyll and other
231	photosynthetic pigments at sub-hourly frequencies to estimate phytoplankton concentrations
232	(Figure 3a). FrrFs are able to monitor changes in the photosynthetic stress of both chlorophyte and
233	cyanobacterial communities at sub-hourly intervals ⁴⁰ . These <i>in situ</i> techniques can be further
234	supplemented by in situ flow cytometry (Figure 3c), which provides a rapid and simple methodology
235	to characterise, at high-frequency, the river phytoplankton community by quantifying the cell
236	concentrations of diatoms, chlorophytes, cryptophytes and different classes of cyanobacteria ⁴¹ . By
237	combining these new biological data with physico-chemical data of the same high temporal
238	resolution, it has been demonstrated that it is water temperature, flow and light conditions that are
239	controlling the onset and magnitude of phytoplankton blooms in the River Thames, rather than
240	increases in nutrient concentration ⁴² . However, phosphorus and silicon may ultimately terminate
241	large phytoplankton blooms due to nutrient depletion and limitation ⁴² . Such insights are only
242	possible through long-term, sub-daily biogeochemical observations that are able to capture the
243	conditions at the precise time points where chlorophyll concentrations begin to increase or
244	decrease. In the coming years, high frequency next generation DNA sequencing will provide an even
245	greater understanding of river microbiological dynamics and their biotic-abiotic interactions ²² .

247 Quantifying water quality across multiple time scales

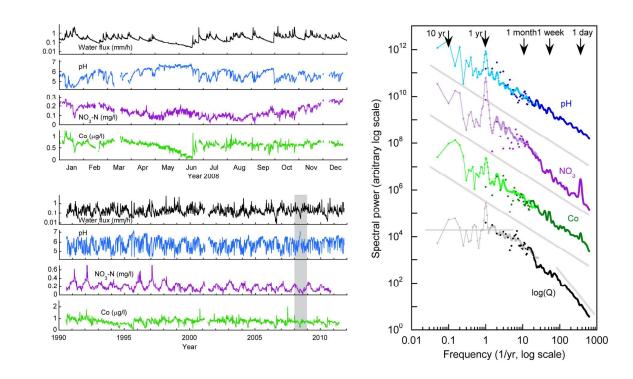


Figure 4. Water quality time series in Upper Hafren streamwater, Plynlimon, Wales, at 7-hour intervals for one year (left upper panel) and weekly intervals for 21 years (left lower panel) (Unpublished Data). Three water quality parameters are shown: pH (an indicator of acid-base status), nitrate (a nutrient that exhibits both diurnal and seasonal cycles) and cobalt (Co). The shaded band in the lower panels shows the time interval covered by the upper panels. The right panel shows power spectra of weekly and 7-hourly time series (light and dark points, respectively), calculated using the methods of Kirchner and Neal⁸.

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255 Detecting water quality trends requires an understanding of water quality fluctuations over
256 many time scales, including those that are invisible in typical weekly or monthly sampling <sup>43</sup>. At
257 several small research watersheds, broad suites of chemical parameters have been measured at
258 daily or higher frequency <sup>8,9,44</sup>, facilitating chemical dynamic characterization on shorter timescales
259 as well as those captured by typical monitoring programs (weekly or monthly) (Figure 4 – left
260 panels).
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261 Spectral analysis decomposes a time-series into a spectrum of cycles of different 262 wavelengths with the power spectrum defined by the contribution of each frequency, f, to the time 263 series. The combination of high-frequency and long-term chemical analysis has demonstrated that,

264	on time scales from hours to decades, the power spectrum of mulitple solute time-series, in this case
265	pH (or H ⁺), NO ⁻ ₃ and Co, can be characterised as $1/f^{\alpha}$ noise, where α , the scaling exponent, is
266	approximately equal to 1 (known as "pink noise") ⁸ . This is shown in Figure 4 (right panel), where the
267	water quality parameters have power-law slopes of -1 (parallel to the long-grey reference lines),
268	indicating that spectral power is inversely proportional to frequency (the 1/f pink noise – which is
269	sometimes referred to as "fractal" noise to emphasise that the scaling exponent, α , can be a non-
270	integer - it does not mean there is self-similarity in the power spectrum). The result implies that
271	three ions do not simply flush through the catchment with the water but that the catchment has a
272	long chemical memory. The stream discharge spectrum, by contrast, has a power-law slope near 0 at
273	low frequencies and -2 at high frequencies (as indicated by the shorter grey reference lines).
274	Whilst the idea of catchment solute storage is not new, analysis of rainfall and stream water
275	power spectra, for the same solute, allows a transfer function to be derived which can be used to
276	quantify the travel time distribution which is very useful in understanding solute retention in
277	different geographical settings. In addition, the 1/f behaviour of the chemical time series has
278	important implications ⁴⁵ . Such time series are "non-self-averaging"; they do not converge to stable
279	averages when sampled for longer periods, because their fluctuations do not average out over time.
280	This non-self-averaging behaviour implies that even purely random time series can exhibit spurious
281	trends, on all time scales, which appear to be statistically significant when evaluated by conventional
282	statistics ^{9,46} . Even more disconcertingly, collecting more data makes this problem worse; non-self-
283	averaging time series exhibit more spuriously "significant" trends (not fewer, as one would expect)
284	when sampled for longer periods, or at higher frequency. Thus, environmental trends should be
285	analysed with more sophisticated statistical methods that are not confounded by the multi-scale
286	correlations that characterize these time series. A recent example of such analysis includes the
287	application of Dynamic Harmonic Regression to use this non-stationary technique to explore stream
288	water nitrate dynamics across decadal to sub-daily timescales and to derive the main cause and
289	effect links at long-term, seasonal and diel time-scales ^{47,48} .

291 Additional advances

292	Where a particular water constituent cannot yet be measured directly with an in situ sensor,
293	it may still be possible to construct a high frequency time-series for that constituent if there is a
294	strong relationship with a water quality parameter that can be readily measured, such as turbidity,
295	dissolved oxygen, temperature and pH. In this way, such sensor measurements have recently been
296	used as proxies for a range of water quality parameters, including total suspended solids ^{49,50} ,
297	alkalinity ⁴⁹ , total nitrogen ⁴⁹ , total phosphorus ^{49,50} , sodium ⁴⁹ , chloride ⁴⁹ , fluoride ⁴⁹ , sulfate ⁴⁹ , fecal
298	coliform bacteria ⁴⁹ , fluoranthene and mercury ⁵¹ , polycyclic aromatic hydrocarbons ⁵² , , and when
299	coupled with discharge measurements allow flux estimation. For example, high resolution in situ
300	measurements of turbidity and fluorescence were used to estimate total mercury transport between
301	the San Francisco estuary and an adjacent tidal wetland ⁵³ . High frequency water quality
302	measurements can also be used to reconstruct concentration patterns in combination with other
303	commonly available continuous data, such as precipitation or discharge. In this way, Rozemeijer et
304	al. ⁵⁴ reduced the bias of total phosphorus load calculations by up to 63% using 20 events sampled at
305	15 minute intervals.

306 Measurements of lake, reservoir, wetland and estuarine diel dynamics help identify internal 307 processing of nutrients and metals. High frequency monitoring in lakes and reservoirs using 308 autonomous vertical profiling systems is increasingly exploited for safeguarding high water quality 309 (e.g., for drinking water abstraction ⁵⁵). Such systems detect river intrusions that may quickly reach water abstraction infrastructure ^{56,57}. High frequency measurements also allow new insights into lake 310 311 metabolism and help constrain biogeochemical budgets or to differentiate the importance of internal versus external factors⁵⁸. It has been recently shown that monitoring external watershed 312 313 loading, as well as within lake chemistry, at high frequency, enables separation of carbon 314 accumulation due to internal phytoplankton dynamics versus external inputs of organic carbon from

runoff events⁵⁹. High frequency oxygen measurements revealed that external seasonal forcing plays
a key role in determining the extent to which a lake ecosystem is a seasonal carbon sink or source to
the atmosphere ⁶⁰. Further experimental uses of high-frequency sensors extends to monitoring
tracers ⁶, changes in artificial environments (e.g., benthic chambers) and in process-control systems,
for example, to control iron dosing to co-precipitate phosphorus at Wessex Water's Keynsham
Sewage Treatment Works in the UK (http://www.worldpumps.com/view/316/control-of-chemicaldosing-in-wastewater-treatment/).

322 For parameters with a strong diel variation, such as DO, the value obtained, and thus the 323 classification of the European Union Water Framework Directive (WFD), which commits European 324 Union member states to achieve good qualitative and quantitative status of all water bodies, can 325 depend markedly on the frequency of sampling. In lowland UK river-systems, monthly sampling for a 326 year can result in the same water body being assigned to three or four of the WFD classes with 95% 327 confidence, due to random sampling effects, although the specific effect on WFD classification 328 depends on the closeness of the range of measured concentrations to the class boundaries. Where 329 water body status is estimated using parameters, such as water temperature, that are assessed using extreme percentiles in a distribution of measurements, such as the 98th percentile for water 330 331 temperature as done in the UK, then monthly sampling does not capture the full variance observed and causes an inaccurate estimate of the true value ⁶¹. 332

333

334 Implications for environmental modelling

From a modelling perspective, the emerging evidence for excessive nutrient contribution of short-term events puts into question the ubiquitous applications of the data-driven models, such as the water quality balance model SPARROW in North America ^{62,63}. Underestimating nutrient export by a factor of two or three and missing the timing of greatest nutrient delivery into a waterbody

Environmental Science & Technology

339 impedes efforts to delineate watershed "hot-spots" or time periods with increased likelihood of violations of water quality targets ⁶⁴. The advent of high resolution data offers a new perspective on 340 341 process-based model parameterization and our capacity to accommodate threshold-type of behaviours when locating critical source areas of non-point source pollution ⁶⁵. In this regard, Wellen 342 et al. ⁶⁶ presented a Bayesian hierarchical framework which postulated that the watershed response 343 344 to precipitation occurs in distinct states, depending, for example, on precipitation and catchment 345 storage. The proposed calibration framework enabled the identification of extreme states and the 346 characterization of different watershed behaviours and improved model performance by allowing 347 parameter values to vary between low and high flow conditions. In addition, estimates of instream 348 assimilation and denitrification help to constrain catchment nitrogen delivery and transport 349 models³², and sub-daily chemistry data coupled with weekly biological monitoring are providing the 350 basis on which to develop a process-based description of aquatic biotic-abiotic interactions, thus 351 enabling an enhanced understanding compared to using ecological indicators alone. Furthermore, 352 development of intelligent water body-specific, cost efficient monitoring schemes combining 353 modelling tools with high frequency monitoring would also help to optimize monitoring schemes 354 and make these technologies accessible for large scale water management.

355

356 Limitations of current in situ technologies

In case of conventional *in situ* chemical and biological measurements there are major issues related to calibration (requiring stable reagents and standards) and supporting infrastructure (e.g., of pumped flow systems) and frequency of servicing intervals which in turn affect the scalability of *in situ* deployment. *In situ* optical sensors, such as those for nitrate, require cleaning to remove biofilm. They can also suffer from interferences due to turbidity and from co-absorbing species like humic acids. However, performing multi-parameter sensing, such as monitoring turbidity and nitrate simultaneously, enables the robustness of the nitrate measurements to be assessed. Given the costs,

364	service requirements, the risks of theft and vandalism, and instrument power requirements, which
365	has decreased recently, there is a need for a cost-benefit analysis to assess the utility of in situ
366	sensors for widespread operational and regulatory monitoring. Furthermore, "big" data streams
367	from in situ measurements pose a challenge to environmental scientists because traditional
368	approaches to data quality assurance and quality control are no longer practical when confronted
369	with the demands of real-time processing. Despite routine maintenance and calibration of sensors,
370	there is a pressing need for the development of automated tools and standards for quality assurance
371	and quality control of sensor data ⁶⁷ .

373 Future directions

The use of high frequency sensors has moved beyond the realm of purely academic research⁶⁸ and these sensors are now employed by numerous national, state, and municipal level environmental authorities. In the United States there are over 500 stations with continuous DO sensors, and over 100 stations with continuous nitrate sensors (http://waterdata.usgs.gov/nwis). There are similar levels of deployment in other developed nations.

379 The knowledge gained from new sensor technology has and will continue to stimulate 380 further advancement. Already microfluidic sensors for measuring nutrients based on colormetric techniques have advantages of small size and limited reagent and power requirements ^{16,17}, though 381 382 further improvements of these devices are necessary to increase robustness and reduce 383 maintenance during permanent deployment. There is still a clear need for further development of 384 new types of sensors, particularly for chemical and organism-based measurements of freshwater 385 ecosystems. Increasing the number of analytes to include redox sensitive elements, micro-nutrients 386 and pesticides would be highly beneficial for more complete environmental assessment. The exciting

prospect of micro-scale inductively coupled plasma spectrometers would allow the measurement of
 a wide-array of elements in water⁶⁹.

389 The inferences drawn from the examples we present above are broadly applicable, as 390 suggested by their geographic range and variation in temporal scales, and move beyond findings 391 that can be obtained from single experiments. Real time sensor deployment for measuring water 392 quality properties continuously from multi-parameter probes offers new prospects to develop 393 sensor networks for whole river networks, watersheds, and lakes. High frequency measurements will 394 expand from the water column to hot spots of biogeochemical transformation and ecological 395 significance, such as the interfaces between aquatic and terrestrial sites (e.g., hyporheic and riparian 396 zones, wetlands, and river-estuarine transition zones). This would significantly increase our 397 understanding of the interaction between sources, uptake (e.g., primary production) and retention 398 (e.g., denitrification) in whole river networks. The co-location of isotope and dissolved anions and 399 cations measurements will also enable enhanced understanding of pollutant storage and transfer 400 and integration of hydrological and water quality models through better characterisation of water 401 and ion transit times⁷⁰.

402 Current use of real time sensors is still restricted to fundamental aquatic attributes such as DO, pH value, SRP and NO₃ 71 , but the field of sensor development is rapidly advancing and we see 403 404 great potential for developing observational data sets that can substantially improve our ability to 405 understand and predict the causes and consequences of environmental changes of aquatic 406 ecosystems. Furthermore, such high temporal resolution data streams can be complemented by 407 additional data types like satellite products for a synoptic survey of water quality of wetlands, large 408 rivers, and lakes to create new scope for validating ecosystem models across multiple scales ⁷². The 409 utility of in situ sensor measurements has already begun to transform routine monitoring in the U.S., 410 with federal (e.g., U.S. Geological Survey), and state agencies (e.g., St Johns River Water 411 Management District in Florida) investing heavily in the structural and personnel capacity to deploy

412 and interpret high-resolution solute time series. In Germany, routine high resolution sensor 413 deployments are not restricted to highly sensitive water bodies like drinking water reservoirs and the gain of scientific transport process understanding⁶⁵; German state water authorities (e.g., state 414 415 environmental agency of Hesse and Baden Wurttemberg) increasingly use high frequency 416 monitoring to quantify matter fluxes (especially at the outlet of large rivers) and for early warning 417 systems for drinking water river bank infiltration facilities (e.g. at the Rhein river). There is also 418 potential to use high frequency monitoring more widely to measure intermittent discharges from 419 Combined Sewer Overflows.

420 Automated sensors that collect novel data, or even traditional data at novel time scales, can enable analyses that inspire new paradigms in aquatic ecology⁴. The susceptibility of an ecosystem 421 422 to changing drivers or random events depends on the characteristics of critical thresholds, such as in 423 ecosystem metabolism⁷³ or in the physical drivers of change, such as flow, light and temperature. 424 Fundamental progress in ecology requires better understanding of thresholds and the rate of 425 anthropogenic induced change in aquatic ecosystems. Emerging technology such as FrrF, "lab-on-a-426 chip", and DNA technology for observing time series data at high temporal resolution will make a growing contribution to this field ⁷⁴. High frequency measurements gained by automated sensors 427 428 will increase our opportunities to better determine the severity of extreme events in terms of water 429 quality and freshwater ecological impacts, and identify the most important variables for assessing the links to environmental change ⁷⁵ at different spatial scales and for different aquatic ecosystem 430 431 types⁴¹. Based on this, we will be in a stronger position to spot early warning signals of critical 432 transitions of watershed biogeochemistry and aquatic ecosystems, and identify and evaluate 433 management options to help mitigate adverse water quality and ecological impacts⁷⁶.

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656 Biography

657 Dr. Michael Rode is Deputy Head of the Department of Aquatic Ecosystem Analysis at Helmholtz 658 Centre for Environmental Research-UFZ in Magdeburg; he leads a research team on new water 659 quality monitoring techniques and modelling of matter fluxes at watershed scale. Andrew J Wade is 660 a Professor of Hydrology in the Department of Geography and Environmental Science at the 661 University of Reading. His research focuses on catchment hydrology, water quality and the links to 662 ecology. Dr. Cohen is a professor in the School of Forest Resources and Conservation at the 663 University of Florida in Gainesville, Florida. He works on a variety of watershed and aquatic systems 664 research. Dr. Hensley is a post-doctoral associate in the Ecohydrology Lab at the University of 665 Florida. He is part of a research team using in-situ sensors to better understand riverine metabolism 666 and nutrient spiraling. Dr. Bowes leads the Water Quality Processes Group at the Centre for Ecology 667 & Hydrology. His research investigates the sources and fates of phosphorus and nitrogen in river 668 catchments, and how they impact on river ecology. James Kirchner is the Professor of the Physics of 669 Environmental Systems at ETH Zurich and the former Director of the Swiss Federal Research Institute 670 WSL; his group conducts research at the interfaces between hydrology, geomorphology, and 671 watershed geochemistry. Professor Arhonditsis is the Chair of the Department of Physical and 672 Environmental Sciences at the University of Toronto. He leads the Ecological Modelling Laboratory 673 that has developed integrated biogeochemical models to guide the policy making process in several 674 impaired systems in North America over the past decade. Phil Jordan is Professor of Catchment 675 Science, Ulster University, Northern Ireland. His research focuses on the dynamics and fate of 676 nutrients and sediment in catchments with an emphasis on the capture and analysis of high 677 resolution water quality data. Professor Brian Kronvang is Head of Section of Catchment Science and 678 Environmental Management at Department of Bioscience, Aarhus University. Dr. Sarah Halliday 679 works as a researcher at the Department of Geography and Environmental Science, University of 680 Reading, UK; her research focuses on the new insights into catchment biogeochemical processing 681 attainable from high-frequency hydrochemical data. Dr. Richard Skeffington is Professor of 682 Geography at the University of Reading, with research interests in water quality monitoring and 683 modelling and understanding biogeochemical processes. Dr. Joachim Rozemeijer coordinates the 684 Environmental Monitoring and Modeling research program at Deltares, The Netherlands; he is 685 scientific researcher and advisor in the Groundwater and Soil Quality department and holds a PhD in 686 Dynamics in Groundwater and Surface water Quality. Dr. Aubert was a post-doctoral fellow at Justus 687 Liebig University in Giessen working on high-frequency nitrate time series analysis ; now she is a 688 fellow at Eawag (the Swiss federal institute of aquatic science and technology) in Dübendorf studying 689 how water guality can be considered as part of structured decision making. Karsten Rinke is heading 690 the Department of Lake Research at the Helmholtz Centre for Environmental Research-UFZ in 691 Magdeburg, Germany; he is a limnologist working on lake modelling and the application of online 692 water quality monitoring in lake and reservoir management. Dr. Seifeddine Jomaa is a research 693 scientist at the Helmholtz Centre for Environmental Research-UFZ in Magdeburg. He leads a 694 research effort on hydrological water quality monitoring and modelling at the watershed scale.