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Environmental sensor networks in ecological research

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Summary

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Environmental sensor networks offer a powerful combination of distributed sensing capacity, real-time data visualization and analysis, and integration with adjacent networks and remote sensing data streams. These advances have become a reality as a combined result of the continuing miniaturization of electronics, the availability of large data storage and computational capacity, and the pervasive connectivity of the Internet. Environmental sensor networks have been established and large new networks are planned for monitoring multiple habitats at many different scales. Projects range in spatial scale from continental systems designed to measure global change and environmental stability to those involved with the monitoring of only a few meters of forest edge in fragmented landscapes. Temporal measurements have ranged from the evaluation of sunfleck dynamics at scales of seconds, to daily CO₂ fluxes, to decadal shifts in temperatures. Above-ground sensor systems are partnered with subsurface soil measurement networks for physical and biological activity, together with aquatic and riparian sensor networks to measure groundwater fluxes and nutrient dynamics. More recently, complex sensors, such as networked digital cameras and microphones, as well as newly emerging sensors, are being integrated into sensor networks for hierarchical methods of sensing that promise a further understanding of our ecological systems by revealing previously unobservable phenomena.

I. Introduction

Ecological research is undergoing a major technological revolution as interfaces develop between environmental science, engineering and informational technology. These advances have been spurred by decreasing cost, size and weight, and improved reliability, of environmental sensing hardware and software. Coupled with the increased connectivity afforded by the Internet to transmit and share data, arrays of intelligent sensor networks are emerging as fundamental tools to address complex questions of myriad ecosystems.

Key to these advances has been the development of appropriate cyberinfrastructure, which comprises the computing systems, advanced instruments, data storage systems and data repositories, visualization environments and technically trained individuals linked together by software and high-performance communication networks to improve research productivity (Estrin *et al.*, 2003; Brunt *et al.*, 2007). Sensor networks coupled with associated cyberinfrastructure thus offer a powerful combination of distributed sensing capacity, internet and satellite communication, and computational tools that lend themselves to countless applications in ecological research. Moreover, new designs of sensor networks allow for the observation of systems in near-real time based on incoming data not only from local sources, but also from nested or adjacent networks, and from remote sensing data streams. These advances are providing a new and better understanding of our ecological systems by revealing previously unobservable phenomena and by allowing a potential for second generation of ecological questions that we have not yet addressed (Porter *et al.*, 2005).

Ecological sensor networks with highly developed cyberinfrastructure lie at the core of major new efforts to address

fundamental issues of global change and environmental stability. The National Ecological Observatory Network (NEON), nearing implementation in the USA, is an integrated network of 20 regional observatories designed to gather long-term data on ecological responses of the biosphere to changes in land use and climate, and on feedbacks with the geosphere, hydrosphere and atmosphere (Keller *et al.*, 2008). Using standardized protocols and an open data policy, NEON will gather essential data for the development of the scientific understanding and theory required to manage the nation's ecological challenges. Similarly, the Global Lake Ecological Observatory Network (GLEON) is a network of limnologists, information technology experts and engineers with the goal of deploying a scalable, persistent network of lake ecology observatories to better understand key processes, such as the effects of climate and land-use change and episodic events on lake function. As with NEON, these observatories will consist of instrumented platforms on lakes around the world capable of sensing key limnological variables and moving the data in near-real time to web-accessible databases.

Many fundamental applications of sensor networks for ecological research involve the challenges of environmental monitoring across a wide range of spatial scales from centimeters to kilometers and temporal scales from fractions of a second to hours (Fig. 1). The ability to characterize the spatial and temporal scales of extreme events is of particular significance, as these have a disproportionate role in shaping the ecology, ecophysiology and evolution of plant species (Levine, 1992; Gaines & Denny, 1993; Gutschick & BassiriRad, 2003; Verstraeten *et al.*, 2008).

The conventional paradigm of increasing the density of fixed sensors in deployments to address issues of scale, however, is

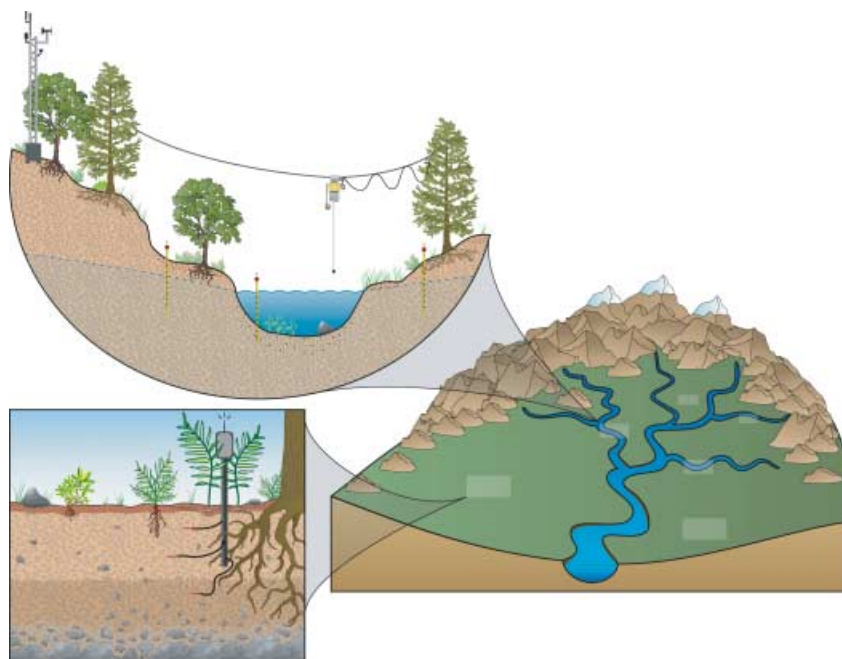


Fig. 1 Diagrammatic representation of a multiscale approach for environmental sensor networks across a landscape. Terrestrial, soil and/or aquatic sensor networks can all be used with multiple sensor modalities and both fixed and mobile sensing platforms. Drawing by Jason Fisher.

Table 1 Examples of major sensor modalities with comments on cost, reliability and power requirements

Sensor category	Example	Comments
Physical	Temperature (e.g. thermocouple, thermistor, IR sensor)	Inexpensive to intermediate cost, reliable, low power requirements
	Relative humidity	Intermediate, reliable, low power
	Leaf wetness	Inexpensive, reliable, low power
	Soil moisture	Inexpensive to moderate, issues with calibration and measurement units, low power; many choices
	PFD, total irradiance	Intermediate, reliable with calibration issues, low power
	Wind speed and direction	
	Cup anemometer	Inexpensive to intermediate, reliable, fails at low wind speed, low power
Chemical	Hot wire anemometer	Intermediate, less reliable, higher power
	2-D/3-D sonic anemometer	Intermediate to expensive, very reliable, moderate power
	Atmospheric carbon dioxide	Expensive, reliable, moderate power, requires careful calibration
	Soil carbon dioxide	Intermediate, reliable, low power, calibration?
	Soil carbon dioxide efflux	Expensive, reliable, moderate power, requires careful calibration
Biological	Nitrate sensor	Expensive, under development for reliable terrestrial deployments
	Phosphorus sensor	Not available for terrestrial deployments
	Digital imagers	Moderately expensive, reliable, moderate power; high bandwidth, software requirements
	Minirhizotron camera	Expensive, variable power requirements
	Sap flow sensors	Commercial probes moderate, control system needed; calibration issues
	Acoustic sensors	Moderate, reliable, moderate power, high bandwidth; software needs

2/3-D, two/three-dimensional; IR, infrared; PFD, photon flux density.

neither economically feasible nor desirable for a variety of reasons. For instance, the sampling and characterization of dynamic phenomena, such as sunflecks on the forest floor, with fixed sensors, whatever their number or position, will invariably be inefficient. Instead, a new paradigm in the design of sensor networks is multiscale sensing based on hierarchical systems that achieve efficient sampling of spatially and temporally dynamic phenomena by optimizing spatial coverage and sensor fidelity. The basic concept in multiscale sampling is that measurements from a low-resolution, wide-area sensor can be used to identify regions of interest, and then higher resolution sensors located in that region are awakened or focused onto that region and tasked for measurement.

As an example, using adaptive sampling, a wide field-of-view camera could sample an area at low resolution and communicate to a fixed sensor in that area to increase the sampling rate or to a mobile sensor to visit the area of interest and better characterize the spatial or temporal extent of the observation. Thus, sensor nodes do not necessarily need to be static, but can also be actively moved, such as on cables, tracks, robotic vehicles and aircraft (Baldocchi *et al.*, 1984; Clements *et al.*, 2003; Gamon *et al.*, 2006b; Laffea *et al.*, 2006). In our work, we have utilized cable-based robotic systems in long-term and rapidly deployable configurations, called Networked Info-Mechanical Systems (NIMS), to complement fixed sensor deployments (Fig. 1; Jordan *et al.*, 2007). The use of mobile sensing platforms allows for cyberinfrastructure with intelligent algorithms to utilize adaptive sampling protocols. Several statistical methods to adaptively sample data have

been proposed in the literature, including stratified methods, in which initial sparse scans extract regions of high variability to be subsequently visited with more precision (Rahimi *et al.*, 2004); Gaussian process models, in which the 'informativeness' of a particular location is derived from the measurements made at already visited locations (Seeger, 2004); and kernel estimators, in which the value of the scalar field at any location is estimated using weighted linear regression, assuming that the closer two locations are, the higher the correlation between the values (Singh *et al.*, 2007).

There is great promise in sensor networks to expand on the traditional sensors for microclimate to involve new sensor modalities (Table 1). These include imagers and acoustic monitoring devices as biological sensors, which are discussed in the context of terrestrial sensor networks, and nutrient sensors, which are discussed in the context of soil and aquatic sensor networks. There are, nevertheless, a number of challenges in designing and deploying successful ecological sensor networks. Some of these issues relate to science-driven questions and requirements that are specific to terrestrial, soil or aquatic domains. Each of these domain needs is discussed in more detail below.

II. Terrestrial sensor networks

Traditional climate monitoring has been transformed by the connectivity afforded by the Internet, combining isolated climate stations into coarse-scale, terrestrial sensor networks that provide data relevant to environmental studies. An example

can be seen in the United States National Weather Service, which first recruited cooperative observers in 1890, and now has more than 11 700 volunteers and 1900 airport-based installations, providing standardized, high-quality, near-real time meteorological data that are freely available through the National Weather Service Forecast Office (<http://www.wrh.noaa.gov>) and through long-lived commercial entities, such as The Weather Underground (<http://www.wunderground.com>). Regional climate monitoring and finer scale meteorological networks have been established to meet the needs of precision agriculture (Ley & Muzzy, 1992; Pierce & Elliott, 2008), and even fine-scale, experiment-driven sensor networks are growing in number. This section discusses both coarser and finer scale terrestrial sensor network deployments, with an emphasis on emerging technologies and newer strategies for data collection.

1. Sensor networks for ecosystem flux measurements

Micrometeorologists have been measuring CO₂ and water vapor exchange between vegetation and the atmosphere since the late 1950s and early 1960s. However, routine application of the eddy covariance methodologies and associated data management to allow continuous flux measurements did not occur until the 1980s, when technological advances were made in sonic anemometry, infrared spectrometry and digital computers. By the early 1990s, further technological developments, including larger data storage capacity and improved stability and precision in instruments, enabled scientists to build on pioneering ecosystem flux studies (Baldocchi *et al.*, 1987; Jarvis, 1989) to make defensible measurements of eddy fluxes for extended periods (Wofsy *et al.*, 1993; Vermetten *et al.*, 1994). The success of these new technologies, coupled with an increasing realization of the critical significance of ecosystem studies of carbon balance, led to the establishment of large multi-investigator experiments, such as the Boreal Ecosystem–Atmosphere Study (Sellers *et al.*, 1997) and the Northern Hemisphere Climate–Processes Land–Surface Experiment (Halldin *et al.*, 1999) that utilized sophisticated sensor networks.

The concept of a global network of long-term flux measurement sites had its genesis as early as 1993 in the science plan of the International Geosphere–Biosphere Program. This interest led to the establishment of the AmeriFlux network in 1997 to quantify spatial and temporal variation in exchanges of carbon, water and energy in major vegetation types across a range of disturbance histories and climatic conditions in the Americas, and to better understand processes regulating carbon assimilation, respiration and storage (Baldocchi *et al.*, 2001).

The AmeriFlux program soon joined with parallel programs in Europe, Japan and Latin America to form FLUXNET (www.fluxnet.ornl.gov), a self-described global network of micrometeorological tower sites that use eddy covariance methods to measure the exchanges of CO₂, water vapor and

energy between terrestrial ecosystems and the atmosphere (Running *et al.*, 1999; Misson *et al.*, 2007). Over 500 tower sites with FLUXNET are now operating on a long-term and continuous basis with core data that include monthly and annual heat, water vapor and CO₂ flux, gap-filled flux products, ecological site data and remote-sensing products, with many sites deploying additional secondary networks of sensors. Data from European and US-American eddy covariance networks have allowed the analysis of seasonal patterns of assimilation and respiration in various ecosystems (Reichstein *et al.*, 2005), and the separation of net ecosystem exchange into gross ecosystem carbon uptake and ecosystem respiration (Falge *et al.*, 2002). The development of these instrumented towers over the past two decades has provided important practical lessons in the deployment and maintenance of complex multimodal sensor networks.

The issue of connecting these ground-based measures of ecosystem fluxes as provided by FLUXNET, and the broader issue of scaling these measurements up to global levels, has pointed to the critical interface between remote sensing and terrestrial sensor networks (Turner *et al.*, 2005). There have been a variety of efforts to bridge this gap, one of which has been through SpecNet (Spectral Network), a network of sites that combine optical sampling with eddy covariance data to address issues of scale (Gamon *et al.*, 2006a). SpecNet optical sampling focuses on spectral reflectance measurements and surface temperature measurements parallel to those generated from satellite sensors (Ustin *et al.*, 2004), but measured at finer spatial scales from towers, mobile trams and low-flying aircraft (Gamon *et al.*, 2006b; Hill *et al.*, 2006).

2. Targeted sensor networks

Both fixed and wireless sensor networks have been deployed successfully in a number of precision agriculture and ecological situations. Agricultural sensor networks provide data from fixed sensors in the field and from those embedded in mobile agricultural machines (Camilli *et al.*, 2007; Pierce & Elliott, 2008). Cyberinformatics and appropriate data modeling become key issues for precision agriculture, where farmers are less interested in masses of data than in decision-making based on acquired data (Beckwith *et al.*, 2004; Burrell *et al.*, 2004).

Ecologists have used wired networks of temperature and light sensors for decades. More recently, however, a number of research groups have successfully developed wireless sensor networks to meet specific research needs for remote sites (Polastre *et al.*, 2004; Porter *et al.*, 2005; Tolle *et al.*, 2005; Collins *et al.*, 2006).

Our Extensible Sensing System (ESS) at the James Reserve in the San Jacinto Mountains of southern California continuously monitors ambient microclimate below and above ground in more than 100 locations with a mix of wired and wireless networks within a 25-ha study area (Hamilton *et al.*, 2007). Individual nodes, each with up to eight sensors, are deployed

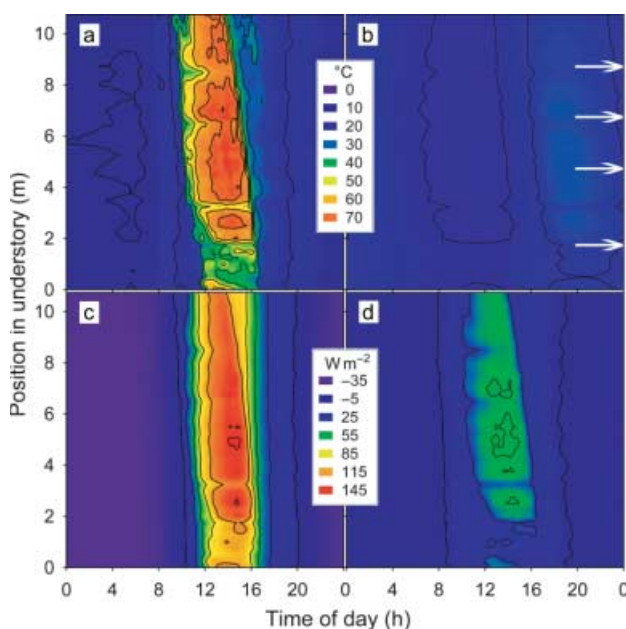


Fig. 2 An example of the density of data collected by a combined fixed and mobile wireless sensor array; data were collected for 24 h along a 10.75-m transect: (a) measured soil surface temperatures every 0.25 m via a mobile sensor platform; (b) measured (indicated by arrows) and calculated soil temperatures at 8 cm depth using soil temperature models; (c) calculated soil heat flux at the surface; and (d) calculated heat storage between the surface and 8 cm depth. Data from E. A. Graham *et al.* (unpublished).

along transects and in dense patches, crossing all major ecosystems and environments on the Reserve. Sensor modalities in the ESS include microclimate sensors, such as for temperature, humidity and photosynthetically active radiation (PAR), as well as a variety of imagers and acoustic sensors. Fixed sensors from the James Reserve networks have been successfully used in conjunction with mobile NIMS units and ecosystem energy flux models to provide a broad temporal and spatial analysis of patterns of soil surface energy balance (Fig. 2).

A notable example of wireless cyberinfrastructure providing core support to ecological research can be seen in the cyber-network of research sites in southern California (Cayan *et al.*, 2003). This network comprises 11 telecommunication sites, seven of which are solar powered, which connect 24 weather stations, three hydrological stations and 13 remote cameras to the Internet via the collaborative infrastructure of the High Performance Wireless Research and Education Network (HPWREN; hpwren.ucsd.edu). The connectivity allows researchers to employ high-bandwidth instruments, such as imaging systems used to measure and monitor ecological and environmental systems, as well as to extend the number and range of conventional remote sensing devices in the terrestrial domain (Hansen *et al.*, 2002).

As mentioned above, there has been increasing interest in the addition of new sensor modalities to sensor networks, such as, for example, with imaging, ecophysiological measurements

such as sap flow, acoustic monitoring and biosensors. Image processing has been used in agricultural studies that combine automatic image capture, analysis and plant physiology (e.g. Slaughter *et al.*, 2008). Ecophysiological studies using cameras range from the detection of CO₂ fluxes in a desiccation-tolerant moss (Graham *et al.*, 2006) to quantitative phenological studies in woody species (Richardson *et al.*, 2007; Graham *et al.*, 2008). The proliferation of Internet-connected cameras that are situated in many natural ecological areas or human-dominated systems provides both challenges and opportunities for image analysis and data reduction. Although many of these systems involve fixed cameras, the addition of pan-tilt-zoom cameras to Internet-connected sensor networks provides a direct means of actuated control over these sensors (Graham *et al.*, 2008). Biosensors are in their early stages of development, but show great promise. Sapflow sensors can be readily added to wireless networks (Burgess & Dawson, 2008), and sensors to measure nutrient concentrations are being developed for soil and aquatic ecosystems, as described below.

We have had notable success in targeted short-term deployments of mobile NIMS to address a variety of ecological research questions. Indeed, although autonomous networks of sensors may seem attractive, early practical experience has indicated the difficulty of specifying field requirements in advance to operate systems remotely. Thus, many of our deployments are now based on dynamic 'human in the loop' scenarios (Wallis *et al.*, 2007), where teams regularly conduct short-term campaigns to collect data. One of these deployments established a replicated set of understory transect measurements of microclimate across the sharp boundary from open clearing to primary tropical rainforest at the La Selva Biological Station in Costa Rica. These measurements have allowed us to examine the diurnal dynamics of microclimate change in a manner that was not possible previously (Fig. 3).

3. Plant–animal interactions

Mobile, networked sensors for environmental monitoring can also be carried by people or animals (Burrell *et al.*, 2004). Although this area of research has a strong zoological and behavioral ecological orientation, systems collecting data on patterns of microclimate and animal–plant interactions, such as herbivory, pollination and seed dispersal, are highly relevant to plant biologists (Cooke *et al.*, 2004; Wikelski *et al.*, 2007). These tracking systems range from highly localized ones, using very high-frequency (VHF) radio-telemetry systems, to satellite-linked systems.

One of the most innovative examples of wireless data collection for large animal tracking is ZebraNet, a system that uses a peer-to-peer network to deliver logged data back to researchers (Juang *et al.*, 2002). The predominant satellite-based system for tracking wildlife is called Argos, a joint venture between the Centre National d'Etudes Spatiales (CNES), the National Aeronautics and Space Administration

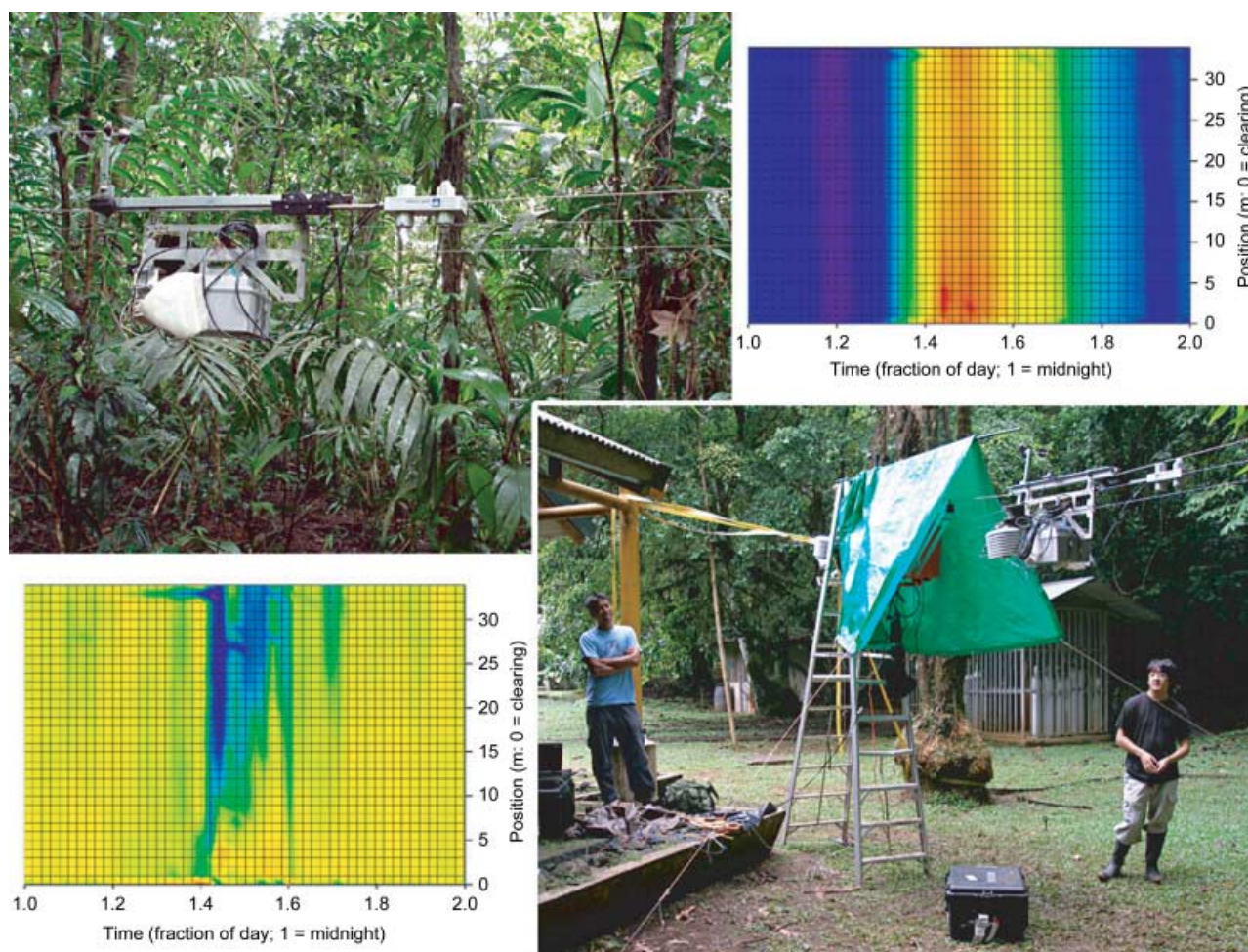


Fig. 3 Use of a mobile sensor platform to measure spatial and temporal patterns of understory microclimate at La Selva Biological Station, Costa Rica. The horizontal node (top left panel) autonomously traveled over an understory transect of 30 m carrying micrometeorological sensors. Stationary sensors and control of the horizontal node were located in a clearing at one end of the transect (bottom right panel). The panels show 24 h of 5-min air temperature data collected at different positions along the transect as absolute temperature (top right: temperature range, 22–30°C) and values relative to the clearing temperature (bottom left: temperature range, –1.0 to 1.0°C).

(NASA) and the National Oceanic and Atmospheric Administration (NOAA).

One of the newest networked systems developed for animal tracking over coarse spatial scales in tropical forests is the Automated Radio-Telemetry System (ARTS) on Barro Colorado Island in Panama. Based on the concept of radio-telemetry systems used to track satellites, ARTS employs a multiple antenna system and software to triangulate transmitter signals and send the resulting information about the location of an animal to a mapping program on a computer. This system has now been used with reasonable success to track ecologically important vertebrates (Cofort *et al.*, 2008).

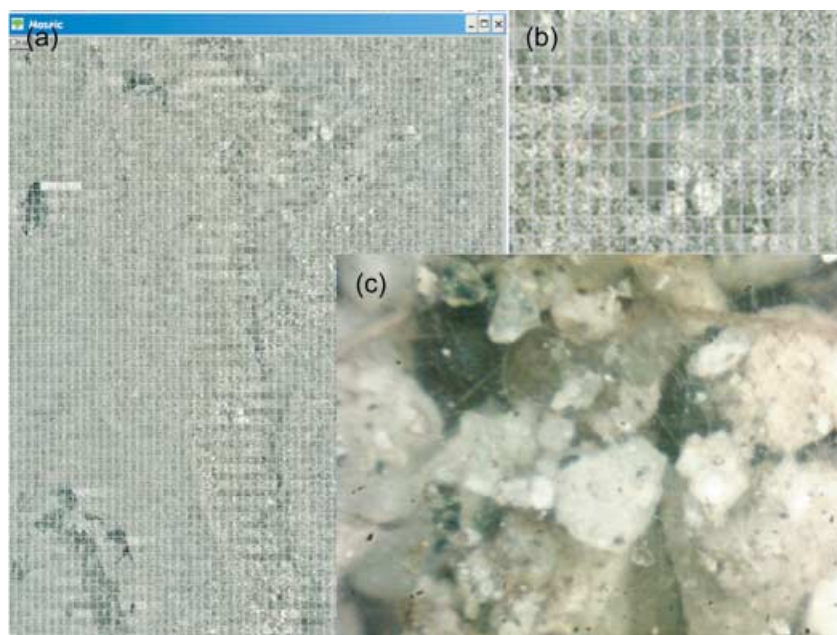
III. Soil sensor networks

Global concerns about the management of carbon and nutrient fluxes rest on an improved understanding of the

exchanges that occur between a myriad of organisms, and coupling these interactions to the exchange between soil and atmosphere. CO_2 is primarily fixed in terrestrial ecosystems by plants, and the major recipients of the fixed carbon are plant roots and mycorrhizal fungi. They, in turn, access nutrients mostly following the decomposition of the plant parts. To date, these exchanges have been largely black-boxed, with exchange rates provided by coarse-scale inputs and outputs. A new approach is to place a network of sensors and imagers into the field to measure naturally occurring dynamics and interactions to evaluate the responses of multiple variables simultaneously (Fig. 1).

Sensor technology has the potential to tell us how the ecosystem partitions dynamics in real time. An example is the demonstration of the temporal dynamics in autotrophic versus heterotrophic respiration in semi-arid central California by Baldocchi and colleagues (e.g. Tang *et al.*, 2005). Pieces of the

Fig. 4 Automated minirhizotron (AMR) test image from the James Reserve, California, USA. The camera takes a 1-mm² high-resolution image completing the scan of each tube. These images can be taken using a defined pattern within the tube, scanning the complete tube surface, or returning to individual points of interest. The individual images are then stitched together to form a mosaic of the tube surface. (a) Mosaic view of c. 30% of a single tube showing an overview of root and soil channels. By focusing on different areas of the mosaic, individual roots can be observed (b) and, by focusing on individual high-resolution images, roots, soil particles, and individual hyphae (c) can be seen and monitored. Using this test mode, we observed that some individual hyphae live for weeks or longer (e.g. coarse runner hyphae of arbuscular mycorrhizal fungi), whereas some of the finer hyphae are produced and disappear with days.



respiration puzzle have been pulled together over the past decade. Högberg *et al.* (2001) girdled c. 360 trees and measured soil respiration over a 2-month study, and reported that much of the soil respiration came from new photosynthate. Although a unique and important study, this method has a number of acknowledged problems, not least of which is the loss of those trees for additional research. Tang *et al.* (2005) undertook continuous measurement of soil respiration (using solid-state Vaisala sensors), coupled with eddy covariance measurements of total stand CO₂ fluxes. Normally, daytime respiration is calculated by subtracting the night-time respiration from the daytime total flux, and correcting for temperature. However, on coupling the soil sensors, they found a pulse of CO₂ coming from the soil that was decoupled from temperature, indicating a 7–12-h lag from photosynthesis until the carbon was received by the roots and mycorrhizal fungi. Without coupling soil sensors, it would have been predicted that there was more daytime soil respiration than actually occurred. This has the potential to have rather dramatic impacts on carbon sequestration models.

1. Soil sensor/imager approach

Taking apart black boxes requires more than simply looking at physical or chemical measurements of ecosystem dynamics. It also requires being able to observe the organisms responsible for those dynamics. *In situ* soil sensing systems for the measurement of CO₂ fluxes with root growth are now available, providing interesting insights into ecosystem functioning (e.g. Tang & Baldocchi, 2005; Tang *et al.*, 2005; Baldocchi *et al.*, 2006). Expanding the range of interactive sensors and replicating deployments in time and space are where sensor networks can be of greatest utility.

Roots have been studied using direct coring, root in-growth bags and minirhizotrons. The problem with focusing only on roots is that they have both metabolic and growth respiration, and their lifespans are too long for production and death to account for short-term or often even seasonal dynamics. Fungi comprise the second largest biomass group in most soils, but their dynamics are rarely studied in the field. Interestingly, individual hyphae grow and die at time scales of days to weeks (e.g. Hobbie & Wallander, 2006; Johnson *et al.*, 2006) that tie very closely to seasonal ecosystem dynamics and even shorter time scale events (Allen, 1993).

Newer automated minirhizotron (AMR) units have the potential to track both root and fungal dynamics *in situ* (Allen *et al.*, 2007), imaging soil volumes multiple times per day. Although these units are still in the testing phase, their use within a soil sensor network is promising (Fig. 4). Part of the problem with studying fine roots and fungal hyphae is simply the timing of production and disappearance. Stewart & Frank (2008) found that monthly measurements used to track root turnover were inadequate, and 3-d intervals were required. Fine roots may grow and die quickly, or can live for years, and rhizomorphs and coarse hyphae often have long life spans (Allen *et al.*, 2003). However, in response to events, rapid changes can occur even between daily observations.

Because the soil must be disturbed to establish a soil sensor network, the use of as much preliminary data as possible for situating sensors is key. Stover *et al.* (2007) used ground-penetrating radar (GPR) to track coarse root turnover. This instrument can provide initial information on the depth to rocks or water table, and the distribution of important features, such as coarse roots or artifacts. At the James Reserve, we were able to determine whether the locations of our sensor nodes were anomalous or representative of a range of characteristics,

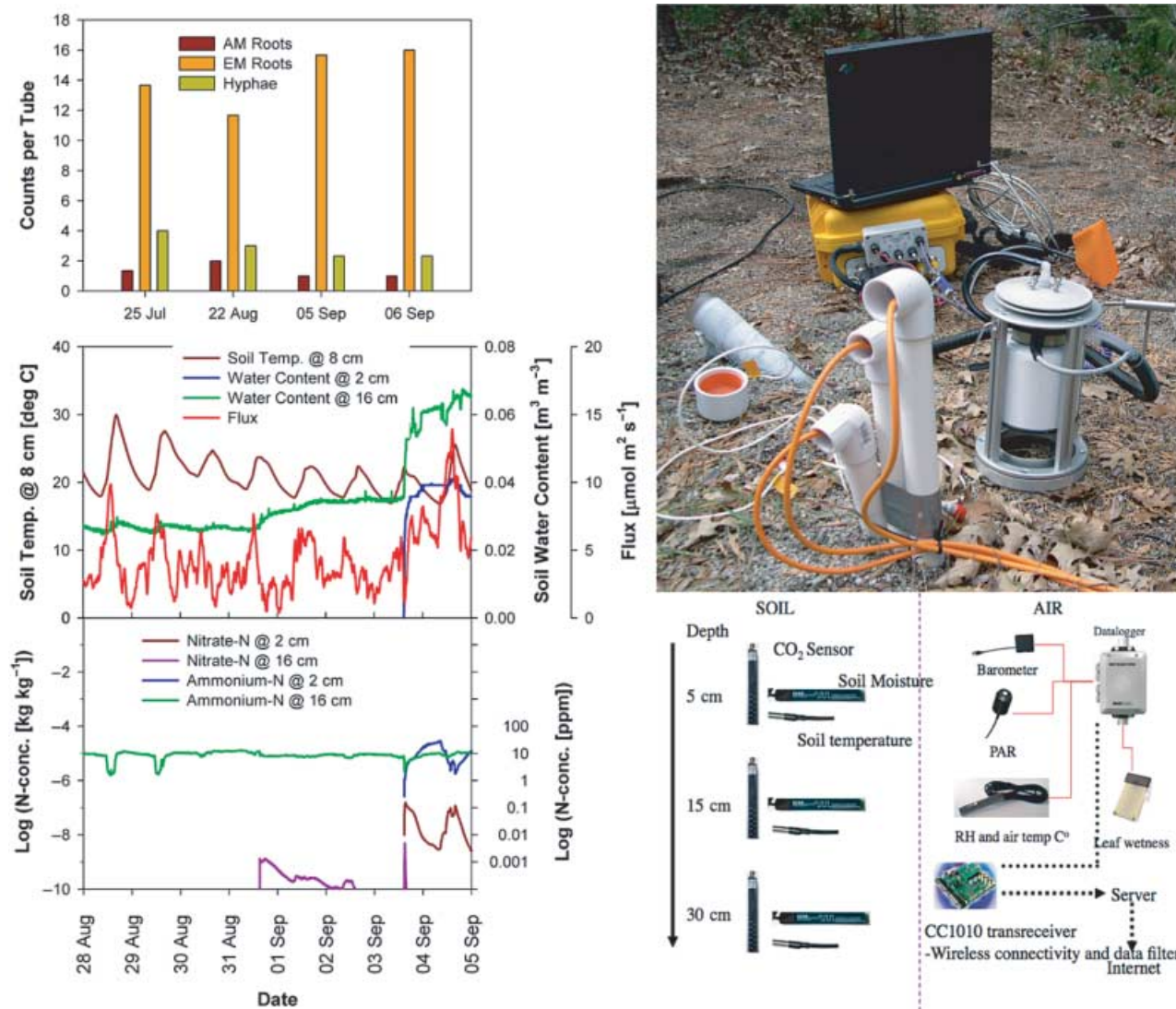


Fig. 5 Sensor output and counts of arbuscular mycorrhizal (AM) and ectomycorrhizal (EM) roots, and rhizomorph hyphae, during a monsoonal event in September 2007 at the James Reserve, California, USA. Shown (left) are the counts per minirhizotron tube, soil temperature, water content, soil CO₂ flux (calculated according to Vargas & Allen, 2008a) and soil nitrate and ammonium concentration. The sensors used include nitrate, ammonium, CO₂, soil water, temperature, photosynthetically active radiation (PAR) and barometric pressure, all needed to provide direct measurements of concentration and to calculate fluxes (see text). The upper right photograph is a location with a minirhizotron tube, entry points for sensors and a LiCor 8100 soil CO₂ flux system used to test against flux models (Vargas & Allen, 2008a).

including depth, rocks, coarse roots or other features. In addition, at the James Reserve, $\delta^{14}\text{C}$ measurements indicated that the coarse roots were too long lived for measurements within the time frames of interest to us (a mean of 17 yr for roots > 1 mm; Vargas & Allen, 2008a).

Because cores must be removed to insert tubes (for minirhizotrons and sensors), these cores should also be used for valuable baseline characterization of soil nutrients and texture. The texture is of special concern, as the calculation of tortuosity (ξ) is essential for modeling the amount of air space, which is tied to calculations of CO₂ production or respiration.

We have integrated a three-dimensional array of sensors, including sensors for CO₂ concentration, T , θ , NO_3^- -N and NH_4^+ -N, using a sensor network. These data are then used to calculate fluxes based on Fick's first law of diffusion. To calculate the fluxes (Fig. 5), it is also necessary to couple these data with the soil texture (which is used to determine tortuosity), soil moisture (coupled with texture to determine air-filled porosity), temperature (with pressure to determine the diffusivity in the soil) and atmospheric boundary conditions (to determine the ratio of diffusivity in the soil to that in the atmosphere, the driving gradient) necessary to model fluxes (Vargas & Allen, 2008a). There are four levels of calibration used to ensure the

integrity of the data. The sensors require periodic calibration. For example, we recalibrate CO₂ sensors every 6 months. The modeled respiration rates are routinely tested against chamber CO₂ measurement systems (e.g. LiCor 8100) or eddy flux towers to evaluate performance. Having multiple sensors at a location allows for the detection of anomalies limited to a single sensor. The outputs of the models are then coupled to both conventional minirhizotron (CMR) or newer AMR imaging systems. This allows us to visualize if 'anomalies' are sensor outliers, or if there is a concentration (or dearth) of biological activity due to fine-scale patch structure common in soils (e.g. Klironomos *et al.*, 1999).

2. Sensing soil heterogeneity

One of the difficulties in soil ecology is defining where and how densely to place sensor nodes. Soil is exceedingly heterogeneous and thus must be sampled at spatial and temporal densities exceeding those for above-ground systems. Sensor networks are thus ideal for use in a physically dense array, such that the spatial structure can be discerned and often placed into a dynamic framework as patches become occupied, depleted and opened again. In addition, using a dense time array, lags and hysteresis can be identified. Finally, both acute and chronic perturbations can be studied over longer time scales with a stable network.

At the James Reserve, networked sensors provide readings at 5-min intervals, and minirhizotron readings are taken weekly, with intermittent daily campaigns. An example output from a single node is shown in Fig. 5. It is critical to note that not all sensor data are appropriate at all times (e.g. the nitrate and ammonium data in the extremely dry soils need to be dropped because of a lack of soil contact). Just as importantly, we have found that coarse hyphae can only occasionally be seen using CMR (Fig. 4) and need to be re-evaluated in the light of the AMR image outputs. Nevertheless, we have been able to begin to put together pictures of dynamics that are not observable using conventional approaches.

Many ecosystems are physically highly patchy. At the James Reserve, a semi-arid mixed conifer forest, there are light gaps and small-to-large meadows scattered across a complex terrain. Using a multiscale sensor network approach, images taken from a tower overlooking the underground study nodes found that shadows covered some nodes earlier than others. These shadows resulted in lower soil temperatures, and subsequently directly changed the diffusion and the measured soil respiration. Snow and rainfall also differentially occur across locations, creating dramatic differences in respiration, moisture extraction and nitrogen mineralization and uptake at quite fine scales. These result in very large differences between sites. We are just beginning to analyze these fine-scale spatial differences, but our preliminary estimates suggest that, when using random or daily measurements of respiration, the cumulative CO₂ released can be incorrect by 80% or more within 20 d.

Because above-soil canopies are not uniform, we also studied a soil transect within our sensor network running from a forest into a meadow (Vargas & Allen, 2008b). In that case, the hysteresis associated with photosynthate pumping in the forest disappeared in the meadow. Respiration was directly associated with diel temperature fluctuations.

3. Sampling remote disturbance events

One benefit of sensors and AMR units is that data can be obtained through significant disturbance events that occur when an investigator is not present. These can include extreme conditions of short-duration or severe events that alter long-term productivity. One such event studied was Hurricane Wilma that entered the Yucatan Peninsula in October 2005. The storm itself was far too severe (200 km h⁻¹ winds, 1500 mm rainfall) for investigators to be at the site, and it took almost 2 months before investigators could reach the site after the storm. However, the sensors had worked well into the storm, before the flooding shorted out the battery system (Allen *et al.*, 2007). Some of the discoveries enabled by the use of this autonomous sensor network included the observation that the drop in barometric pressure probably did not result in a major degassing of the soil CO₂, apparently because the water had already saturated the soil, replacing soil air pockets. Thus, the water probably forced out the CO₂ and, as a result of low O₂ tension, respiration initially was low. Thereafter, surface litter rapidly decomposed because of the high moisture and high temperatures. The higher temperature was caused by intense radiation due to the loss in leaf area resulting from the winds and rapid drop in barometric pressure. The hysteresis pattern observed before the hurricane changed dramatically afterwards, apparently in response to a decoupling of respiration from night-time temperatures (Vargas & Allen, 2008c). This tells us that multiple and often unknown mechanisms account for major changes in ecosystem functioning as a result of perturbation. Having more systems in the field to track more and different types of events could provide a dramatic improvement in our understanding of major disturbances.

IV. Aquatic sensor networks

Aquatic sensors have been employed for decades on moorings and gauging stations to record time series for basic water parameters, such as temperature, stage-based flow and specific conductance (for salinity), and for above-water meteorological sensors. As more aquatic sensors are becoming accessible, conceptual models will be more readily tested and refined (Gawne *et al.*, 2007). Figure 6 illustrates a hypothetical sensor network focused on the observation of distributed environmental properties as they relate to macrophyte community structure (Trempe, 2007). Integrated data streams from such networks can be used to characterize higher order environmental

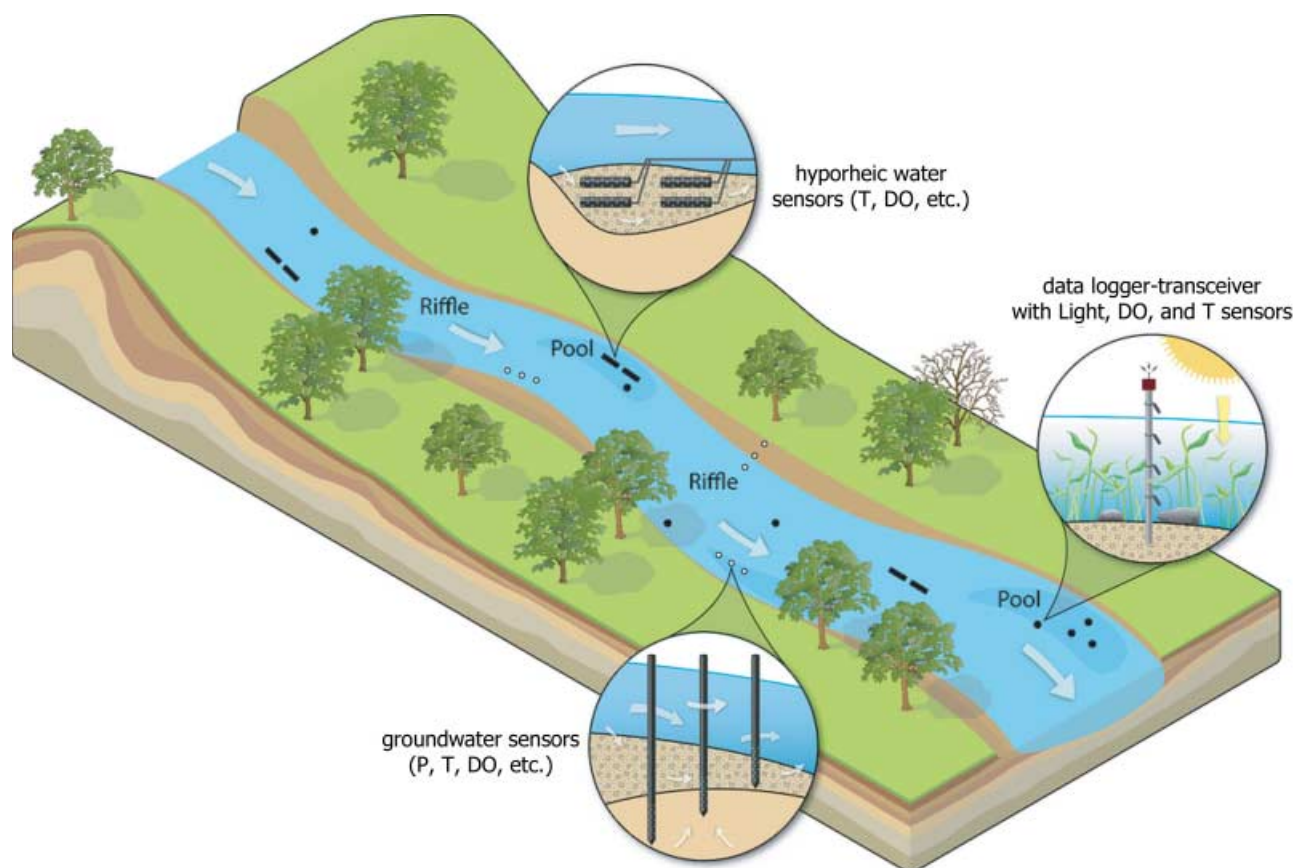


Fig. 6 Hypothetical deployment of an aquatic sensor network sampling pool and riffle patterns along a stream course, with positions of hyporheic water samplers and piezometers. DO, dissolved oxygen. Drawing by Jason Fisher.

properties of aquatic and riparian ecosystems. For example, moorings instrumented with temperature, light, dissolved oxygen (DO) and meteorological sensors have been used to characterize the primary productivity and respiration rates in rivers (e.g. Bott *et al.*, 2006) and lakes (e.g. Coloso *et al.*, 2008). In addition to observing local phenomena, these sensor networks can support scientific inquiry at the continental (Montgomery *et al.*, 2007) and global scale in the context of oceans (Dong *et al.*, 2008; Forget *et al.*, 2008). This section discusses existing and developing sensor network deployment strategies for aquatic and riparian ecosystems. It begins with a brief overview of currently available sensors and emerging sensors for aquatic observations, and extends to a discussion of sensor networks for making higher order observations of streams or rivers, lakes and groundwater.

1. Aquatic sensor modalities

The modes of physical, chemical and biological sensing useful in aquatic networks have been detailed elsewhere (Daly *et al.*, 2004; Goldman *et al.*, 2007; Johnson KS *et al.*, 2007; Prien, 2007). Physical sensors are the most reliable in the aquatic realm and include the previously discussed meteorological

sensors, as well as evaporation, oxygen transfer and other processes at the air–water interface. Physical sensors for water pressure (depth), light penetration and flow velocity are also commonly deployed in aquatic systems. Reliable chemical sensors are available for several water properties, including salinity (in terms of specific conductance) and DO, which have been developing rapidly over the past decade, and long-term deployments are now common for both electrochemical (e.g. Clark cells) and optical luminescent DO sensors (optodes) (Tengberg *et al.*, 2006). The Alliance for Coastal Technologies (ACT) has completed performance demonstrations on several commercial DO sensors, mainly in marine settings (ACT, 2004), and results suggest that 2-wk to 1-month service intervals may be necessary, depending on local biofouling conditions. Other commonly deployed chemical sensors include those for oxidation–reduction potential (ORP), total suspended solids (TSS), nutrients (N, P), dissolved organic matter (e.g. CDOM) and indicators of primary production (e.g. chlorophyll). Among these, electrochemical sensors, such as pH, ORP and various ion selective electrodes (ISEs), remain unsuitable for long-term autonomous deployments within networks as their responses tend to drift excessively over time in the absence of frequent servicing. However, it is

encouraging that several long-term deployment successes have been reported for nitrate ISEs in natural streams and canals (Le Goff *et al.*, 2003; Scholefield *et al.*, 2005). When more accurate assessments are necessary, robust ultraviolet–visible (UV/Vis) absorption (Johnson & Coletti, 2002) and flow cell analyzers (ACT, 2008) are now commercially available for nitrate, and are becoming available for dissolved and filterable phosphorus (ACT, 2008). Both electrochemical (Bobacka *et al.*, 2008) and optical (McDonagh *et al.*, 2008) sensing modes relevant to nitrogen and phosphorus species remain active research areas.

The observation of TSS is important in aquatic systems, particularly in the context of light penetration in the water column. Commercial models can be deployed reliably for as long as 2 wk in systems subject to biofouling (ACT, 2007), and probably significantly longer in nutrient-limited freshwater systems.

Sensors for aquatic flora and fauna are available to a much lesser extent relative to those for physical and chemical properties. Remote sensing platforms, such as hyperspectral imagers, have been used effectively to map vegetation in aquatic ecosystems (e.g. Hestir *et al.*, 2008). Distributed, embedded sensing technologies, including local imaging platforms discussed above, can provide high-resolution spatio-temporal data to complement remote sensing products, which are typically coarser in time and space and broader in spatial coverage. In terms of chemical sensing, fluorometers for indicating CDOM, *in vivo* chlorophyll-*a*, cyanobacteria and other parameters are now commercially available, and several models have been performance tested (ACT, 2006) and have shown excellent promise for observations coordinated with DO, primary production, eutrophication and other issues.

Flow-through cameras equipped with image classification software are available for the *in situ* identification and enumeration of phytoplankton in the water column (Bowen *et al.*, 2006). Similar coarser scale systems are used to identify and enumerate fish in engineered settings, such as fish ladders (Olson & Sosik, 2007; Sosik & Olson, 2007). As noted for terrestrial systems, there are potentially many more applications in the context of aquatic ecosystems for imagers as sensors in multiscale networks.

2. Sensor networks in rivers and lakes

As with terrestrial systems, both static and mobile deployment modes are used in aquatic systems. In general, many of the same types of sensor are carried in both modes, and the key difference is whether the observational objectives are related to time series or synoptic data or both. For observations in rivers and streams, the gauging station mode for static deployments is the most common. Driven mainly by governmental regulatory needs, coarse-scale networks of such stations exist to measure flow and basic water quality, typically temperature and salinity, throughout river basins, such as the California Digital Exchange Center (CDEC) for the

Sacramento–San Joaquin river basins. More complex stations, such as the Columbia River (CORIE) Observation Network, are being developed which are equipped with multiparameter water quality sondes and acoustic Doppler current profilers (ADCPs) for mapping the passing velocity field under different flow regimes (Dang *et al.*, 2007). Such systems are becoming increasingly necessary in human-dominated watersheds, where environmental flows must be managed in terms of quantity and quality to sustain aquatic habitat in the face of other demands on water.

For lentic systems, moorings or tethered buoys have been used to assess basic water quality and key derived metrics, such as primary production (e.g. Gawne *et al.*, 2007; Coloso *et al.*, 2008). Typical buoyed platforms include basic meteorological sensors (including solar radiation) as well as multilevel water temperature thermistor chains, DO, submerged PAR sensors and other water quality parameter sensors. Using primary production/respiration (PP/R) models, the limnologists are able to integrate the network's data stream into PP/R time series. Coloso *et al.* (2008) examined the variation of gross primary production (GPP) and respiration horizontally and vertically in a lake using high-frequency DO observations to reveal that, although GPP declined sharply with depth, respiration was unrelated to depth. In a river restoration project, a similar network was used to assess changes in the introduction of air into the lower water layers, GPP, respiration and net daily metabolism (NDM) before and after canal backfilling, and after the restoration of continuous flow through the river channel (Colangelo, 2007).

Less well tested are sensor systems that support investigation into aquatic plant ecology questions at the terrestrial–aquatic margins, such as in wetlands and in littoral zones associated with lentic and lotic systems (Gratton *et al.*, 2008; Istvánovics *et al.*, 2008). These types of questions could be more efficiently addressed with support from terrestrial and aquatic sensor systems deployed to assess local environmental conditions above the water surface (e.g. vegetation type and cover, air temperature, humidity, PAR, wind), within the water column (e.g. stage, velocity, light transmission, water quality) and within the benthic zone (e.g. groundwater seepage rates, water quality).

Static sensor networks deployed on gauging stations or moored platforms are effective for providing time series data, and are acceptable when stream cross-sections and lakes are reasonably well mixed. There are many applications, however, where understanding stream community structure dictates the need for greater spatio-temporal observational coverage (Johnson RK *et al.*, 2007; Tremp, 2007). In lakes prone to stratification, vertical profiling capabilities may be necessary; moorings with this capability are commercially available and have been used successfully by oceanographers and limnologists (e.g. Doherty *et al.*, 1999; Reynolds-Fleming *et al.*, 2002). More recently, the tethered robotic NIMS discussed in the terrestrial network section was modified for application in lakes and rivers. In one application, NIMS was used to

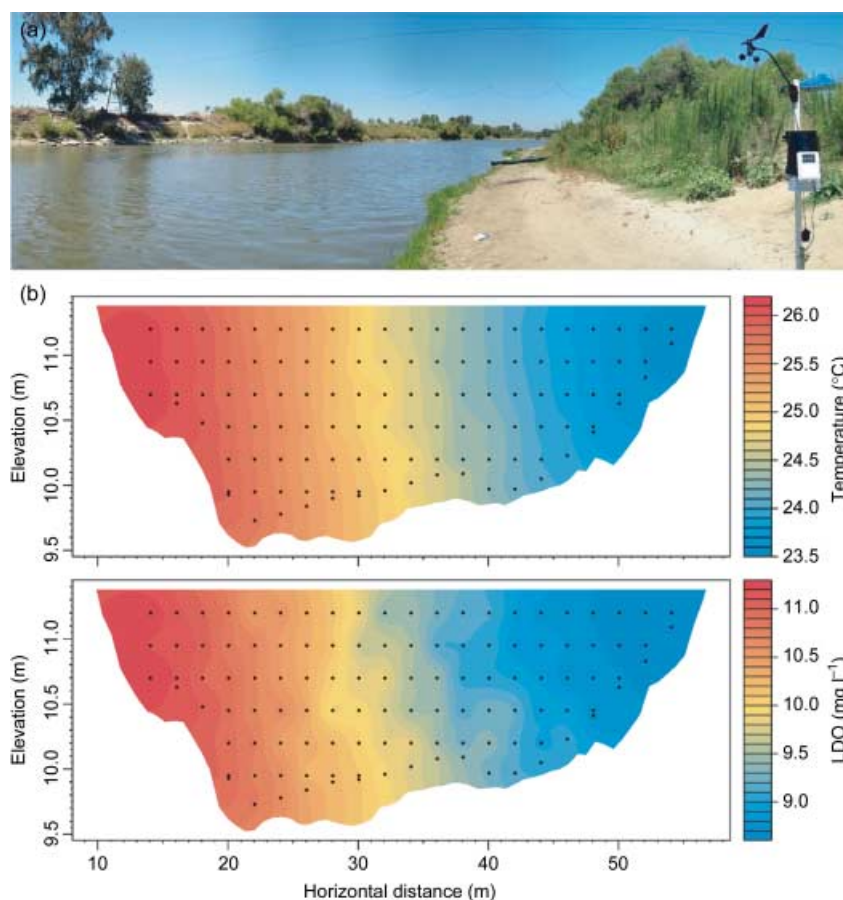


Fig. 7 (a) Networked Info-Mechanical Systems (NIMS) robotic cable system erected on a transect across the San Joaquin River just downstream of its confluence with the Merced River, which enters to the right; (b) significant gradients in temperature and dissolved oxygen (LDO) sampled by sensors conveyed by the NIMS device. The black dots indicate sampling locations.

observe significant horizontal gradients in velocity and water quality (temperature, salinity, DO) within the confluence of two major rivers (Harmon *et al.*, 2007; see Fig. 7). To address three-dimensional space over time, autonomous underwater vehicles (AUVs) have been used extensively by the oceanographic community, and more recently in lakes and river systems (e.g. Laval *et al.*, 2000; Farrell *et al.*, 2005; Sukhatme *et al.*, 2007). These systems can range from powerless drifters, to profiling gliders programmed to regulate their buoyancy to ascend or dive, to fully powered self-propulsion AUVs.

3. Sensor networks in groundwater and the hyporheic zone

Several types of physical and chemical sensor are available in forms suitable for deployment in observation wells or access tubes in the benthic environment. Although they have not yet seen widespread use, such systems would be useful for observing macrophytes, periphyton or other applications in lentic and littoral systems (e.g. Sebestyeni & Schneider, 2004; Westwood *et al.*, 2006; Tremp, 2007). For example, small rugged pressure transducers can be used for mapping the groundwater pressure gradients over broader scales. Basic water quality parameter sensors are also available in compact

form, the most common being conductivity and temperature (CT) sensors for observing trends in salinity and temperature; these are also available with integrated pressure transducers to provide the depth of the water column above the sensor (CTD sensors). Several small-diameter, multiparameter water quality sondes similar to those discussed above are also commercially available.

Groundwater pressure gradients are typically modest and difficult to detect at fine spatial scales of interest, say, in the study of macrophytes in a stream segment. For this reason, methods employing arrays of inexpensive temperature sensors have been developed to map temperature gradients in streambeds, and from which groundwater–surface water exchange rates can be estimated (e.g. Johnson *et al.*, 2005; Essaid *et al.*, 2008). Obviously, as chemical sensors, such as for nitrogen, phosphorus and other nutrients, become more accessible in price and size, they will be extremely useful in the observation of groundwater–surface water exchanges and their relation to aquatic plants.

4. Near-future aquatic sensing systems

Major developments in sensor technology are anticipated in the next 5–10 yr in two general areas: novel sensors and

integrated, miniaturized sensor systems (i.e. 'lab-on-a-chip'), both of which are expected to increase the functionality and decrease the per-sensor cost. Many of the advancements are anticipated to increase the quality and miniaturization potential for existing physical and chemical sensors (Bobacka *et al.*, 2008; McDonagh *et al.*, 2008). Major innovations, however, are anticipated in the area of biosensor development (i.e. sensor-transducer systems which use biological mechanisms to generate responses associated with targeted pollutants or microorganisms), which is growing exponentially in terms of research papers and patents (Marazuela & Moreno-Bondí, 2002; Rodríguez-Mozaz & Lopes de Alda, 2006; Sassolas *et al.*, 2008; Borisov & Wolfbeis, 2008). Although much of the biosensor field remains in the laboratory validation stage, several commercialization efforts are encouraging in the water quality monitoring area, including a toxicity biosensor for wastewater (Farré *et al.*, 2000), and surface plasmon reflectance (SPR) commercial venture. SPR is an extremely promising optical technique which enables real-time observation of molecular interactions, thus enabling a broad spectrum of biosensing opportunities (for a review, see Homola, 2008).

With these exciting new sensor types comes the rapid development of integrated, miniaturized sensor systems (Joo & Brown, 2008). These systems include not only the sensor-transducer components, but microfluidics, to pretreat samples and add reagents, and the associated circuitry to manage the system in terms of sample throughput, energy management and data communications.

V. Challenges for sensor network development

1. Data collection and management

Sensor deployments can generate far more data than can be managed by the traditional methods used in field research, placing data quality and control beyond the capacity for individuals to effectively monitor. Substantial initial effort and attention to quality assurance/quality control (QA/QC) issues from the outset must be expended to capture, maintain and make high-quality data available for use by others. A large variety of faults can impact on data quality, including sensors affected by aging, biofouling or leaking of internal solutions, or simply sensors with bad connections to the data collection device. Criteria for data integrity vary both by context and by individual, and work on tools and services to capture data, metadata and publications is ongoing (Michener *et al.*, 1997; Borgman *et al.*, 2007; Wallis *et al.*, 2007).

Error checking can occur during or after data have been collected, and indeed may be facilitated by cyberinfrastructure and the incorporation of archived data and sources not in the immediate network. An automated system to perform QA/QC either before or after data are inserted into a database is essential, because of the unwieldy amount of data that can be collected by even a modest sensor network, and some work

towards automated fault detection has been made (Sharma *et al.*, 2007). The intercomparison of data from sensors, using Bayesian techniques when multiple sensors of the same type are deployed (Ni & Pottie, 2007), or between sensors and remote sensing data (Grassotti *et al.*, 2003), can be part of data QA. In addition, a model–measurement intercomparison can allow the international scientific community to evaluate the performance of models compared with field observations (Hoffman *et al.*, 2007) to ensure long-term and cross-network comparability.

Systems for managing and making sensor network data accessible include OPeNDAP (www.opendap.org), an open-source project to create a standard Network Data Access Protocol. OPeNDAP also provides software which makes local data accessible to remote locations, regardless of local storage format, facilitating the networking of sensors. In addition, GEON (www.geongrid.org) is a collaboration among a dozen institutions to develop the cyberinfrastructure to support an environment for integrative geoscience research, and EcoGrid (seek.ecoinformatics.org) is a next-generation Internet architecture for data storage, sharing, access and analysis. Work on EcoGrid involves a wide variety of ecological and biodiversity data, and analytical tools for efficiently utilizing data stores to advance ecological and biodiversity science. Although data storage and management in open-source databases, such as MySQL (www.mysql.com), are popular, additional work on publishing and sharing data from MySQL with various users has not been straightforward.

Gaps in the data flow from sensor networks present a widespread and pervasive problem. These gaps may be short-term problems caused by instrument failures, power outages or inclement weather, or more serious long-term gaps (days to weeks) as a result of major instrument problems or maintenance shut-downs. Although many ecologists have the image of perfect fidelity and precision of sensor data, this is far from the truth, even in some of the most sophisticated sensor networks, such as those associated with FLUXNET. Studies have shown that 17–50% or more of individual flux measurements are rejected or missing in such ecosystem studies (Richardson & Hollinger, 2007; Xing *et al.*, 2008). An example of the extent of this problem can be seen in a careful analysis of eddy covariance measurements at the Hesse deciduous forest site in France (Longdoz *et al.*, 2008). Over a full-leaf season, 60% of half-hour values were rejected, with a higher error rate of 69% for night-time measurements (Falge *et al.*, 2001). Even in short-term controlled deployments, sensor networks may fail to record half or even more of the programmed data points (Tolle *et al.*, 2005).

Both short-term and long-term data gaps in ecosystem studies present a serious problem by reducing the quantity and integrity of the sensor outputs and, for long-term deployments, impacting on the accuracy of estimates of ecosystem flux parameters, particularly those related to estimates of net ecosystem productivity, the accuracy of process–climate

linkages and the relative source–sink relationships of carbon balance. As a result, there have been a variety of gap-filling methodologies proposed over the past decade, each with their strengths and weaknesses (e.g. Falge *et al.*, 2001; Ruppert *et al.*, 2006; Moffat *et al.*, 2007; Xing *et al.*, 2008).

With the recent developments in cyberinfrastructure, researchers can have near-real time access to all data streams from their sensor networks. This means that instrument failures, power interruptions and even calibration errors can all be quickly identified and corrected, with major data gaps avoided or minimized. Careful attention to system function can enable researchers to reduce the uncertainties in their data collection. A complementary strategy to continuous vigilance of data quality is to have sensor networks designed with redundancy in sensors, data logging devices, and power supplies – in this way, detected hardware failures may be more quickly resolved to minimize data loss. Nevertheless, planned sensor calibration and maintenance schedules to replace hardware before failure are vital to data integrity.

2. Energy efficiency

An issue that cuts across all research domains is that of energy requirements, which defines the limitations of wireless vs. line-powered systems. The design of larger long-term deployments, such as NEON and GLEON, is dependent on the availability of continuous line power. However, wireless sensor networks offer important solutions to issues related to remote and/or short-term deployments (Puccinelli & Haenggi, 2005; Raghunathan *et al.*, 2006), and power management is of most concern in many of the studies involving battery-operated wireless sensor nodes (Raghunathan *et al.*, 2006; Hart & Martinez, 2006; Moraisa *et al.*, 2008; Ruiz-Garcia *et al.*, 2008). In the early days of design considerations for wireless sensor networks, the focus of energy efficiency was on the energy consumed by wireless communication. Although simple and low-rate sensors, such as those typically used for monitoring temperature, humidity, irradiance and wind, require little energy, this is not the case for other sensor modalities, such as those for imaging and acoustic monitoring, which typically require high-rate and high-resolution analog-to-digital (A/D) converters that can be power hungry.

Without continuous battery changes, sensor networks require sophisticated power management techniques coupled to their communications design. The requirement for on-demand high-performance computing and communication for complex information processing, however, can be addressed by new multiprocessor node hardware and software architectures.

Several energy-harvesting techniques are also now feasible, and solar energy harvesting through photovoltaic conversion currently provides the highest power density, making it the method of choice to power sensor nodes (Raghunathan *et al.*, 2006), although less reliable in some ecosystems (Pierce &

Elliott, 2008). Alternative energy methods, such as wind and water flow, that supply rechargeable batteries have also been explored for sensor networks (Moraisa *et al.*, 2008).

3. Commercialized sensor network systems

Sensor networks are quickly transitioning from being objects of academic research interest to a technology that is being deployed in a wide variety of applications and is rapidly being commercialized. Several commercial ventures for producing consumer-grade hardware for wireless mesh networks exist, including Crossbow (www.xbow.com), ZigBee (www.zigbee.org) and Sentilla (www.sentilla.com), which all offer various forms of off-the-shelf solutions to wirelessly connected sensors. SensorWare Systems (www.sensorwaresystems.com) is a spin-off company from the NASA/JPL Sensor Webs Project, and thus has benefited from research on *in situ* sensor networks and end-to-end solutions for accessible data flow, in real-time, via the Internet (Delin *et al.*, 2005). Common to all these products is relatively inexpensive commercial technology combined from both the computation and telecommunication industries to create practical, field-deployable and embedded systems.

4. Wireless communication

Standardization is missing at many levels of sensor networks (Hart & Martinez, 2006). Hardware platforms and operating software vary and interoperability is difficult. One way to address standardization issues is through the Institute of Electrical and Electronics Engineers (IEEE) 802.11 family of wireless communication standards. The standards (overlapping significantly with Wi-Fi) are able to handle very high data transmission rates, but are simultaneously power-demanding and face interference problems outside of line-of-sight deployments. In response to power issues related to embedded processors, the IEEE has established a 802.15.4, Wireless Personal Area Network (WPAN) standard for communications, enabling the creation of complex *ad hoc* networks to provide ultra-low power consumption (very long battery life of months or even years) and very short wake-up time capabilities at very low power cost. The WPAN standard assumes that the data transmitted are short and that transmissions occur at a low-duty cycle (active/sleep times ratio), reducing the overall power needs and enabling the application of battery-powered embedded systems (Moraisa *et al.*, 2008). The ZigBee Alliance (www.zigbee.org) set of communication protocols is based on the WPAN standard.

Remote locations are particularly challenging, as the lack of power infrastructure and data communication lines hampers coordinated data collection. A joint NASA–Information Sciences Institute (University of Southern California) project called Sensor Processing and Acquisition Network (SPAN) has established a sensor network with satellite communications to

support ecological research (Ye *et al.*, 2008). SPAN relays data from the field from remote locations to the scientist through satellite communication as the wide-area networking (WAN) backhaul.

5. The near-future of sensor networks

One of the benefits of a distributed network is the integration of information obtained from multiple sensors into a larger world view not detectable by any single sensor alone. Going beyond traditional sensor networks and new applications of additional sensor modalities and adaptive sampling, there is now the emergence of a more general notion of model-based active sampling to optimize the sensing process. The key idea is that the system learns spatio-temporal relationships among the measurements made by sensor nodes, and uses this knowledge to optimize the sensing (i.e. whether, when, where and at what fidelity level should a sensor measurement be made) for energy consumption and position of fixed sensors for a required level of overall application-sensing task. The process of learning the spatio-temporal relationships can be based on a variety of approaches: for example, modeling sensors as Gaussian processes and capturing the relationships among them in terms of covariances, or modeling the relationships among sensor values using nonparametric statistical models (Batalin *et al.*, 2004; Schoellhammer, 2008). Common to the different approaches is the ability to predict to some level of confidence the value of a sensor measurement based on sensor measurements at other points in the space–time sensor continuum. Such model-based approaches to sensor data acquisition are in their relative infancy, but promise a general framework to enable the better design and efficiency of sensor networks.

Technological advances in sensors, sensor data logging and communication, and software management of sensor networks will continue to provide transformative potential for new and innovative avenues of ecological research in ways previously not possible. The challenges for the community of ecological researchers, engineers and specialists in software science will be to maintain two-way avenues of communication to continue to design and deploy new technologies for sensor networks. A key step will be the development of programs with field training and sensor network curricula to familiarize the next generation of scientists with these emerging tools.

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