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Integration of Scientific Information Management and **Environmental Research**

Susan G. Stafford

James W. Brunt

William K. Michener

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Integration of scientific information management and environmental research

Susan G. Stafford, James W. Brunt and William K. Michener

A new dynamic between science and technology is changing the management and analysis of environmental information. As a result of the expanding scope of environmental research, developers of environmental databases must now address difficult and diverse issues: the wide variety of data being considered for inclusion; the size and complexity of databases being collected; their design, development and utilization; the different scales of data collection; increasingly sophisticated analytical requirements; and the integration of data from various sources and disciplines. Increased attention to spatial variability, scale and the integration of basic and applied research to address societal issues has several implications for the management of scientific information. These new research directions have created a need for rapid and easier data analysis; timely, broad-scale, high-resolution data; new analytical approaches; better sampling resolution in space; and a shift in focus from data to information to knowledge. This will require the use of remote sensing, geographic information systems (GIS), spatial statistics and other methods. Several challenges are discussed for future environmental information management and analysis systems: distributed analytical environments, database management systems, integrated GIS and remote sensing in the time domain, user interfaces, visualization software and knowledge discovery, improved spatial sampling resolution and standardization.

Introduction

Historically, environmental research has been conducted as small-scale studies involving one or a few investigators in a single discipline and funded for relatively short periods. Recently, increased understanding of small-scale environmental patterns and processes, coupled with burgeoning interest in landscape, regional and global patterns and processes, has led to the development and funding of studies addressing broad-scale, long-term questions. For example, the Long-Term Ecological Research Program (LTER), initiated by the National Science Foundation in the early 1980s, supports long-term investigations that could not be addressed effectively in the normal 1-3-year funding cycle of most US granting agencies (Franklin et al., 1990). These and related programmes generally involve experiments and monitoring lasting years, decades or

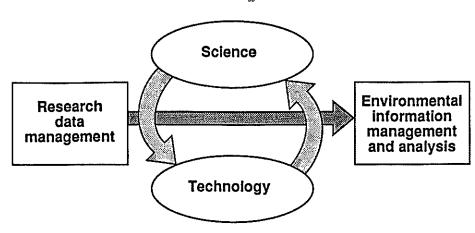


Figure 1.1 A new dynamic between science and technology has forced evolution of the way in which environmental information is now managed and analyzed.

centuries (e.g., Harmon's (1992) 200-year study of log decomposition), and they frequently include many scientists from several disciplines.

As a result of the expanding scope of environmental research, developers of environmental databases must now address difficult and diverse issues: the wide variety of data, both spatial and non-spatial, being considered for inclusion; the size and complexity of databases being created; their design, development and utilization; the different spatial and temporal scales of data collection; increasingly sophisticated analytical requirements; new systems (e.g., Gophers) to provide information about and access to databases; and the integration of data from various sources and disciplines. Concomitantly, attention to management of environmental data within scientific organizations has increased. The intensified focus on effective management of environmental data is based on the premise that data, like people, machines and capital, are a significant organizational resource and product (Michener, 1986).

Coupled with new research directions in the environmental sciences and with improved technologies, data management will evolve far beyond its traditional scope (Figure 1.1). Data management within scientific organizations historically has been driven by the need to support specific tasks (e.g., data entry, archiving, security and quality assurance), many of which may be viewed simplistically as scientific custodial services. The importance of fully integrating data management into the research process – beginning with the design of data format, collection and documentation; following through in data collection, quality assurance, analysis and interpretation; and concluding with synthesis, review and publication – has been emphasized in previous workshops (Stafford et al., 1986; Gorentz, 1992). We see data management within scientific organizations necessarily expanding and evolving to emphasize the timely and effective transformation of data into information and provision of that information to scientists, managers, policy makers and the public. We anticipate that this

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evolution will lead to the emergence of 'scientific information management' as a discipline. The focus on 'information' will ensure that both management and research are fundamental components of this activity.

In the following discussion, we examine how both the environmental sciences and computational technologies are changing and the implication of these changes for scientific information management. Finally, we describe challenges and research needs in scientific information management.

New Directions in Environmental Research

Increased attention to spatial variability and scale (Levin, 1992) and the emergence of co-directed research (integrated basic and applied research to address societal issues) in environmental science (Sharitz et al., 1992; Swanson and Franklin, 1992) have several implications for the management of scientific information.

Spatial Variability and Issues of Scale

Characteristic scales define the space and time over which processes can be detected, as well as the characteristic dimensions of geographic phenomena. A phenomenon may therefore appear homogeneous at one spatial scale, but heterogeneous at another (Nellis and Briggs, 1989). Many natural systems show hierarchical organization, with nested patterns and processes occurring over a wide range of characteristic spatial and temporal scales (Allen and Starr, 1982; O'Neill et al., 1988; Bradshaw, 1991). Concepts embedded in hierarchy theory highlight the importance of identifying appropriate scales for investigation, and recognizing that new patterns and processes may emerge as the scale of observation is varied (Allen and Starr, 1982). For example, climatic fluctuations over relatively long periods (decades) and large areas (106 m²) are associated with changes in phenology and population dynamics that produce broad-scale patterns in plant communities. At smaller scales, wind and lightning associated with localized thunderstorms may be responsible for many of the patterns observed in forest stands and ecosystems (Gosz et al., unpublished data).

To address phenomena that depend on spatial pattern, we must find ways to better quantify patterns of variability in space and time and to understand the causes and consequences of pattern and how patterns vary with scale (Levin, 1992). Achieving this will require the use of remote sensing, spatial statistics and other methods.

Co-directed Research

Co-directed research may be defined as the integration of basic and applied research to better address issues with management, policy and societal impli-

cations. Foci of co-directed research include natural resource management and sustainable development, environmental monitoring, environmental risk

analysis and restoration ecology.

Natural resource management has received much attention recently. In the Pacific Northwest, for example, a new paradigm for forestry is being defined that focuses on forest management primarily for ecosystem structure, diversity and function, rather than for timber and commodity production (Forest Ecosystem Management Assessment Team, 1993). The International Geosphere Biosphere Programme (Rosswall et al., 1988) and the Sustainable Biosphere Initiative of the Ecological Society of America (Lubchenco et al., 1991; Gosz, this volume) are focusing the efforts of the environmental research community on developing the knowledge and technologies for supporting

more sustainable development.

Better management of our natural resources requires that we develop environmental monitoring programmes that allow us to identify and track the condition of specific resources at appropriate spatial and temporal scales. For example, the United States Environmental Protection Agency is launching a multi-year, multi-hundred-million dollar programme to assess status and trends of the earth's natural resources. This programme, the Environmental Monitoring and Assessment Program (EMAP), will focus initially at a national level and eventually expand to global proportions (Environmental Protection Agency, 1991). The United States Geological Survey has begun a more modest programme, the National Water Quality and Assessment Program (NAWQA), to monitor the nation's water resources (Hirsch et al., 1988). Similarly, many new strategies for management of forest resources in the Pacific Northwest may be viewed as broad-scale experiments, the outcome of which can be followed through long-term monitoring.

The success of new resource management strategies and their effects on ecosystems and industrial development cannot be assessed without precise methods of analysing environmental risk. The relatively new field of environmental risk analysis seeks to develop a robust capability for predicting ecosystem risk that can be used for management, policy development and validation of ecosystem models (Whyte and Burton, 1980).

Research in restoration ecology (Jordan et al., 1987; National Research Council, 1992) focuses on applying fundamental ecological principles to the restoration of natural ecological patterns and processes in degraded habitats.

Implications of New Research Directions

These new research directions have created a need for rapid and easier data analysis; timely, broad-scale, high-resolution data; new analytical approaches; better spatial sampling resolution; and a shift in focus from data to information to knowledge. Recent events like the Forest Summit in Portland, Oregon (Forest Ecosystem Management Assessment Team, 1993) and the

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easier data pproaches; to infor-Portland, i) and the Four-Corners virus outbreak in New Mexico (Parmenter et al., 1993) are good examples of the need to alter our trajectory from a data-rich to an information-rich science. The Forest Summit, for example, required panels of experts to integrate and synthesize vast amounts of data into meaningful information within 60 days in order to produce scientifically credible options for policies guiding management of forest and aquatic resources in the Pacific Northwest. Rapid analysis of data from disparate sources was critical in relating the Four Corners virus outbreak to the increasing population densities of rodents.

In order to address issues of scale and spatial variability, data are being collected over a broader spectral range, at finer spatial sampling resolution, and at shorter sampling intervals (Wickland, 1991). Most data used in analysing the spatial dependence of surface and atmospheric phenomena are obtained by satellite- and aircraft-mounted, multispectral sensors. Environmental scientists are increasingly employing geographic information systems (GIS) and remotely sensed data to examine questions at multiple spatial and temporal scales. Data density, which depends jointly on spatial and temporal resolution, is a technical consideration. Continuous increases in computing and storage capacities have shifted studies toward smaller temporal and spatial scales. As a result, processes operating at relatively fine spatial and temporal scales can now be investigated with finer spectral resolution over increasingly large areas (Davis et al., 1991).

According to one estimate, the amount of data in the world doubles every 20 months (Frawley et al., 1992). Earth-observing satellites launched in this decade will generate one terabyte of data every day – more than all previous missions combined. The Earth Observing System (EOS) will create roughly the equivalent of the entire 17-year Landsat archive every two weeks (Marshall, 1989). Similarly, the federally funded Human Genome project will store thousands of bytes for each of the several billion genetic nucleotide bases (National Research Council, 1988). Moreover, scientists are loath to discard any data for fear that a scientist in the future may need exactly that data set to test a specific hypothesis (French et al., 1990; Silberschatz et al., 1991). As a result, the gap between data generation and data understanding is growing (Frawley et al., 1992), and our ability to gather and store data is much more advanced than our ability to manage, analyse and interpret them.

The new analyses needed for examining spatially correlated data require abandoning many of the traditional experimental designs developed primarily for agricultural experimentation (e.g., focusing on Type I and Type II errors). New experimental designs, broad-scale, spatially explicit sampling programs, geostatistical tools, and other methods that will allow us to better identify patterns and link them with processes must be developed. The extensive databases generated by EOS and other programs highlight the need for better tools that will support time-series analyses of spatial data.

Evolving Technologies Applicable to New Directions in Environmental Research

Several technologies that have been developed and improved during the past decade are now being applied increasingly to environmental research. Major advances include greatly increased desktop computing power, new analytical approaches and software, new database designs and the proliferation of remote sensing and GIS.

Advances in Computer Technology

The 1980s saw a dramatic convergence in capabilities of mini-, micro- and mainframe computers. The introduction in 1981 of the IBM personal computer (PC), based on the Intel 8086 computer chip, propelled microcomputers into a high profile. IBM extended PC capabilities initially with the Intel 80286, a 16-bit processor, and then the Intel 80386, a 32-bit processor, providing faster data access and combining multiple instruction executions in clock time (Faust et al., 1991). Minicomputers and workstations have achieved mainframe speeds, and distributed processing strategies via networked PCs have fundamentally altered computing environments. In the short time between the development of the VAX 11/780 minicomputer in the late 1970s and the present, we have gone from a realizable speed of 1 million instructions per second (MIPS) to about 50 MIPS on current-generation single-CPU workstations.

The rate of development of new computer hardware tends to be underestimated, whereas that of software improvements usually is overestimated. Nevertheless, according to recent projections for hardware development (Frank et al., 1991), processing speed will approximately double each year; prices for main memory will decrease about 50 per cent biannually; hard-disk access times will increase very slowly; the capacity of hard disks will increase as prices decrease; and communication networks for the exchange of large volumes of data will proliferate.

In the near future, we anticipate faster communication networks with speeds of 100 megabits/second (Mb s⁻¹); faster CPUs, with 500-1000 MIPS; more efficient devices that can store terabytes; higher resolution monitors; and higher level programming languages, both visual and object-oriented. Networks will allow complete transparency among machines world-wide, and desktop access to supercomputers will become common. By the late 1990s, the specifications for a personal GIS workstation may be a CPU with 500-1000 MIPS; 500 megabytes of main memory; 5 gigabytes of storage space on hard disks and an additional 50 gigabytes on optical disk; a workstation screen 2000 × 2000 pixels; and a communication device with a transfer rate of 100 Mb s⁻¹.

Development of software for these hardware platforms is expensive and timeconsuming and lags significantly behind hardware improvements. Popular 'new' programming languages such as C and Pascal are almost 20 years old. Current operating systen years ago to circ technologies. U invented to red 1986).

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operating systems (VMS, MS-DOS and UNIX), also were developed 10-20 years ago to circumyent obstacles made obsolete by rapidly changing hardware technologies. UNIX, for example, with its cryptic short commands, was invented to reduce typing when input was commonly by teletype (Winkler,

Although software development lags far behind, extensible software systems promise relief. Visual programming environments like Khoros (Rasure et al., 1990), and database management systems (DBMS) and analytical systems like S+ (Becker et al., 1988) allow application building and customization for specific projects.

Analytical Approaches

New emphasis on spatial scale and structure has increased characterization and modelling by spatial statistics (Robertson, 1987). Spatial variation can be modelled in a variety of ways by covariance, variograms or fractal analysis. Meteorologists and oceanographers are able to use the covariance or the powerspectrum approach (the Fourier transform of the covariance), since they can overcome some of the inherent limitations of these techniques by assuming stationarity of their data (Davis et al., 1991). Semi-variograms have been used to characterize spatial variability in topography (Dubayah et al., 1989), solar radiation (Dubayah et al., 1990), surface albedo (Webster et al., 1989), spectral vegetation indices (Davis et al., 1989), soil properties (Robertson et al., 1988) and vegetation (Pastor and Broschart, 1990). Precipitation (Lovejoy and Schertzer, 1985), soil properties (Burrough, 1983a, 1983b) and land use (Milne, 1991) have recently been analysed with fractals. Explicit modelling of spatial statistical structure should improve the interpolation of point data, optimize sampling designs for field collections and contribute to our understanding of underlying processes (Davis et al., 1991).

New research directions have also created the need for better resolution in spatial sampling in order to understand phenomena at relevant spatial scales. For example, in order to predict spotted owl habitat, Ripple et al. (1991) decomposed landscape pattern by scale (coarse to fine), using the wavelet transform. Phenomena were matched to their appropriate scale of investigation: nest tree to the tree level; thermal cover to the stand level; prey base and oldgrowth patch connectivity to the landscape level; and conifer cover to the subregion level. This technique has numerous potential applications in landscape ecology, where analysis may focus on several scales of pattern simultaneously.

Remote Sensing and GIS

Remote sensing and GIS technology, which support the computerized capture, management, display and analysis of spatial data (Thapa and Bossler, 1992),

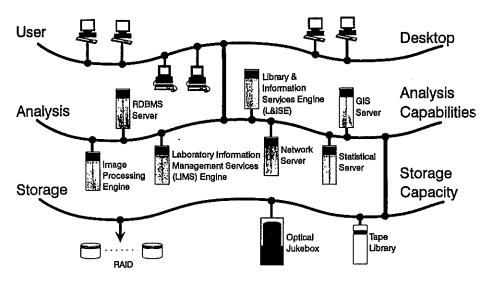


Figure 1.2 Analytical and storage capabilities accessible from the future user's desktop.

Note: RAID: Redundant Array of Inexpensive Disks.

have opened new frontiers in our ability to monitor and manage natural resources – from mapping soil mycorrhizal mats (Griffiths et al., 1993) to analysing regional phenomena (Burke et al., 1991). GIS technology has migrated to smaller machines with ever-increasing processing power and ever-decreasing costs (Juhl, 1989). Unfortunately, although many view GIS as a panacea solving data volume and management problems, more important problems still underlie GIS, including the lack of adequate DBMS software to manage scientific data effectively and the lack of sophisticated statistical software and other analytical tools within the GIS environment.

Challenges for Future Environmental Information Management and Analysis Systems

Distributed Analytical Environments

A primary challenge in this area is creation of a highly interactive metacomputer composed of workstations with graphics accelerators, high-speed networks, terabyte data vaults and supercomputers that can address multidisciplinary 'grand challenge' problems. In the future, scientific organizations will likely use multiple file and computer servers with ease from their desktop computers. File servers may be dedicated to one or more specific tasks (e.g., statistics, a DBMS, library support, or image processing; Figure 1.2). Hierarchical storage servers will provide transparent access to huge amounts of optical disk and tape storage previously maintained off-line.

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nputer works, plinary ely use s. File BMS, servers torage Organizations may have heterogeneous DBMSs differing in their capabilities and structure and dispersed over several sites. The heterogeneity, distribution and sheer number of databases in the public and private domains make it imperative to share access to these databases while maintaining autonomy of individual systems. Thus, new techniques are needed that support the interoperability of autonomous databases without requiring their global integration (Scheuermann et al., 1990).

In the system envisioned for the future, a user submits a database query, completely unaware of and unconcerned with the location and format of the data. The software automatically locates the data across any number of DBMS, processes them and returns the answer. Security, error correction and rollback are independent of user, DBMS, vendor and location. Multimedia, hypertext and hypermedia technologies will be increasingly used for creating, organizing and disseminating on-line information (Mohageg, 1992) and for combining text and graphics with working models and imagery.

The complex nature of spatial, as compared with record-oriented, data complicates data distribution and communication for distributed GIS. Network technology ultimately will develop to the point where intermachine communication is no longer an issue. The challenge will be in providing software to utilize this type of network efficiently. Thus, scientists in Albuquerque, New Mexico, will be able to interactively access soils data stored on the United States Geological Survey server in Denver, Colorado, and climate data stored on the National Climate Data Center server in Asheville, North Carolina.

Database Management Systems

The promise of a DBMS lies in its ability to provide timely and consistent data, to enable users to access those data directly without technical assistance and to evolve easily to meet changing user requirements (Curtice, 1986). Environmental research at the project level requires integrated computational services that provide access to sophisticated database, statistical and numeric operations and to technical text support (Conley et al., 1986). These operations could be supported collectively by a scientific DBMS; however, most commercial DBMS software is record-oriented and directed specifically at business solutions.

If the next-generation database applications are to meet the needs of environmental scientists, they are unlikely to resemble current business-oriented DBMSs (Silberschatz et al., 1991). Recently, considerable research has focused on developing general-purpose DBMS prototypes that present some advantages over relational DBMS (RDBMS) in representing complex (i.e., spatial) data, which do not lend themselves easily to representation or efficient management within a record-oriented system. One such prototype is POSTGRES (for post Ingres) (Pollock and McLaughlin, 1991). Additional research in this area focuses on extending current RDBMS technology and developing object-

Table 1.1 Advanced database management system (DBMS) strategies with potential for scientific information management.

| Extended relational: | incorporates new features in the relational model, such as inheritance, user-defined functions and data types, and procedures for data types and nested relations. |
|------------------------|---|
| Object-oriented: | accesses a complex object as a single concept and performs operations on that object in the DBMS, rather than in an external application program (Dittrich, 1986); supports inclusion of 'unconventional' data types, such as text, voice, graphics or CAD (computer-assisted drafting) (Scheuermann et al., 1990). |
| Extensible: | utilizes a 'toolkit' approach to tailor DBMS internals (e.g., storage structures, query optimizers) to meet application requirements that are inadequately served by existing DBMSs; supports easier customization of DBMSs for scientific applications. |
| Logic-based languages: | support access to relational RDBMS by providing a consistent interface between the tools and the RDBMS (Hessinger, 1987); e.g., Structured Query Language (SQL). |

oriented and extensible DBMSs and logic-based languages (Table 1.1) (Shoshani, 1990). The overall challenge will be integrating the best of each of these techniques to manage scientific data.

Integrated GIS and Remote Sensing in the Time Domain

Overcoming the technological obstacles associated with jointly processing image and GIS data, such as converting between vectors and rasters and jointly displaying and updating digital imagery and maps, has already entailed much work. Continued efforts in this area should lead to the evolution of integrated geographic information systems (IGIS) (Davis et al., 1991). Since not all physical processes can be mapped directly from remotely sensed data or existing maps, an important role of IGIS will be the generation of such information by appropriate modelling based on more than one information source (Risser, 1986). The coupling of satellite measurements with other spatial data has tremendous potential for describing earth surfaces, predicting future conditions and validating physical representations produced solely through remote sensing or GIS. One key impediment to total integration of GIS and remote sensing is that current GIS technology does not adequately represent the time domain, whereas remote sensing samples in both space and time. Causal relationships among events or entities are embedded in temporal information, and data therefore should be interpreted in a temporal context (Snodgrass and Ahn, 1986).

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Errors associated with remotely sensed and GIS data can significantly impact the confidence with which decisions are made (Lunetta et al., 1991). GIS provides new information without simultaneously establishing its reliability. Uncertainty is propagated and transformed each time a conceptual or physical model is constructed in GIS processing (Bedard, 1987). Lack of concern for error in spatial databases reflects the lack of a standard framework for analysing the propagation of modelling error (Lanter and Veregin, 1992).

As these technologies continue to improve, the potential of IGIS analysis will come to be limited only by our conceptual understanding of the phenomena under investigation and their representation in spatial databases (Davis *et al.*, 1991).

User Interfaces

The improvement of interfaces within and among image processing, GIS, DBMS, expert systems, models and statistical software will facilitate analyses significantly. To meet the needs of environmental scientists, future systems will emphasize sharing data and tools among disparate software environments, rather than computing power. More powerful and more efficient languages and user interfaces are needed that can lead less computer-literate users systematically through common data management and information processing without requiring them to master the intricacies of the DBMS and the computer operating system. Ideally, the further development of higher level languages will allow the investigator to focus on the problem at hand (Peled, 1987).

There is also a strong need for better management of critical information associated with models and model input and output - that is, model metadata. This includes keeping good records of model evolution and of the lineage of data as they are used to produce publications or as input to other models or analyses that may cascade to further models. The bottom line is clearly linking the data with the models. Intuitively designed user interfaces can help expedite this.

Visualization Software and Knowledge Discovery

Visualization tools have been developed and primarily utilized by physical scientists. One of the challenges for environmental researchers will be the use of such tools for extracting information from the massive volumes of data being collected. Visualization software will be a required part of the environmental scientist's toolkit.

Knowledge discovery, the use of computers to extract implicit and potentially useful information from data, carries potential for tackling some of the enormous data analysis challenges of the next generation of environmental scientists. Knowledge discovery is expensive even with today's computers and requires additional knowledge regarding the form and content of data.

Visualization software, knowledge discovery algorithms and expert systems will necessarily be used increasingly to facilitate the processing of vast amounts of data into the information needed by scientists, managers and policy makers.

Improved Spatial Sampling Resolution and Standardization

In most environmental research, data are collected in plots of 1 m² or less (Kareiva and Anderson, 1988; Brown and Roughgarden, 1990). Few studies entail collection of pattern- and process-oriented data on the ground that can be readily related to the scale of data sampled by Landsat Thematic Mapper (30 m \times 30 m) or SPOT (10 m \times 10 m, panchromatic; 20 m \times 20 m, multispectral). Thus, there is currently a wide discrepancy between the spatial scales of data collected by remote sensing and those collected on the ground that can be used for verification and validation. This gap likely will decrease as scientists take advantage of new airborne remote sensing systems that sample data at the submetre scale.

Nevertheless, several impediments to scaling up from the plot to broader scales likely will remain. For example, several networks (including the LTER Network, Global Change Sites funded by the National Park Service, and National Oceanic Atmospheric Administration Marine Sanctuaries) have been established to examine ecological patterns and processes at the scale of the continental USA. Certain attributes of primary production are examined at most or all of the sites; however, specific parameters being measured, temporal and spatial resolution, field and laboratory protocols, and data management and analysis generally differ significantly among sites. In addition, numerous biomes within any given network are generally not sampled, and spatial replication within biomes is rare or nonexistent. Expanding individual networks to encompass more of the inter- and intra-biome variability, as well as integrating individual networks into a 'network of networks', could facilitate broad-scale research by ensuring better spatial coverage and encouraging data comparability and information exchange (Bledsoe and Barber, 1993). We need to focus attention for intersite comparison at the information stage, however, in part because the big differences among biomes and ecosystems may make standardized data collection much less useful than standardized approaches to information assembly and management. Spatial and temporal variability, standardization of monitoring programmes, laboratory and analytical protocols, and data and information management practices likely will continue to challenge us.

Conclusion

The virtual explosion in computing power and accessibility has significantly revolutionized the scientific environment. Managing scientific information has gone far beyond a mere 'custodial service'. Users of scientific databases work in a

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nificantly ation has work in a great variety of environments, from a single personal computer or workstation to supercomputers to heterogeneous networked systems. Network technology ultimately will develop to the point where intermachine communication is no longer an issue, and the potential for analysis will be limited only by our conceptual understanding and our ability to frame the questions.

We face challenges in monitoring and modelling global change. Of these, two interrelated problems are key: (1) combining remote sensing and GIS, and (2) determining the nesting of observations required to bridge scales between short-term, fine-scale and long-term, broad-scale measurements and process models (Rosswall et al., 1988). Tight linkages between research and application will be critical to monitoring the environmental and social consequences of global change (Davis et al., 1991).

Because large and diverse data sets, documents and images are becoming common in both business and science, the gap between business and scientific software is closing. DBMS software that will be useful to both business and scientific applications should result. A properly instituted and supported DBMS can increase the productivity of environmental researchers and enhance the quality of environmental science (Risser and Treworgy, 1986), reducing the time scientists devote to learning details of hardware and software. This level of DBMS for a large project must be supported with a great deal of forethought and ongoing commitment.

Technologically advanced societies frequently cannot transform new ideas into successful products quickly (Strome and Lauer, 1977). Acceptance of technology in any field is complex, and decades often elapse before practical applications of research are achieved. Barriers to communication among engineers, environmental scientists, computer scientists, policy makers and others further delay product development.

A sociology of information management is developing that is as critical as the hardware and software issues. Because bridges must be built between researchers and data managers, training a new breed of scientists and scientific information managers will be a prerequisite to the conscientious efforts required to close these gaps. We anticipate the emergence of 'scientific information management' as a discipline with both a management and a research component, emphasizing the timely and effective transformation of data into information and knowledge for scientists, managers, policy makers and the public.

Acknowledgement

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