
Model-Driven Dynamic Control of Wireless Sensor Networks

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Overview

- ◆ Wireless sensor nets
 - evolution toward diverse, application-dependent deployment densities and node capabilities
 - a new technology for measuring environments
 - motivating application: **revolutionize understanding of environmental change**
- ◆ Project aim
 - forecast how altered climate and CO₂ will impact biodiversity and carbon storage in the biosphere
- ◆ Approach: dynamic sensor networks
 - Deeply embedded distributed intelligence for model inference and prediction

Outline

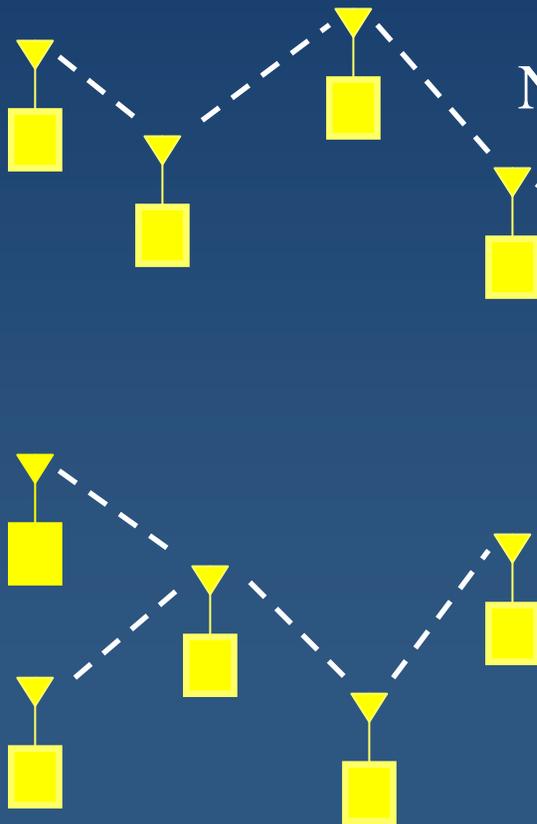
- Sensor net design and implementation
 - Node and software design
- Dynamic sensor networks
- Algorithms and software
 - Data service layer
 - Model-driven control
 - Research challenges
- Concluding remarks

Requirements – Environmental Sensing

- Minimal invasiveness
- Long battery life
 - Aggressive energy management
 - Target: > 12 months
- Scientific accuracy
- Support of a broad spectrum of probes
- Support transparent incremental deployment
- Scalable in network size and density
- Ease of installation and maintenance
- Support of internet connectivity via
 - Terrestrial (cellular)
 - Satellite
- Rugged, weatherproof packaging
- Low life cycle cost

Concept

Dense array of
energy-efficient
sensors



Multi-hop
Networking

Gateway
(Network
Interface)



Satellite
Link

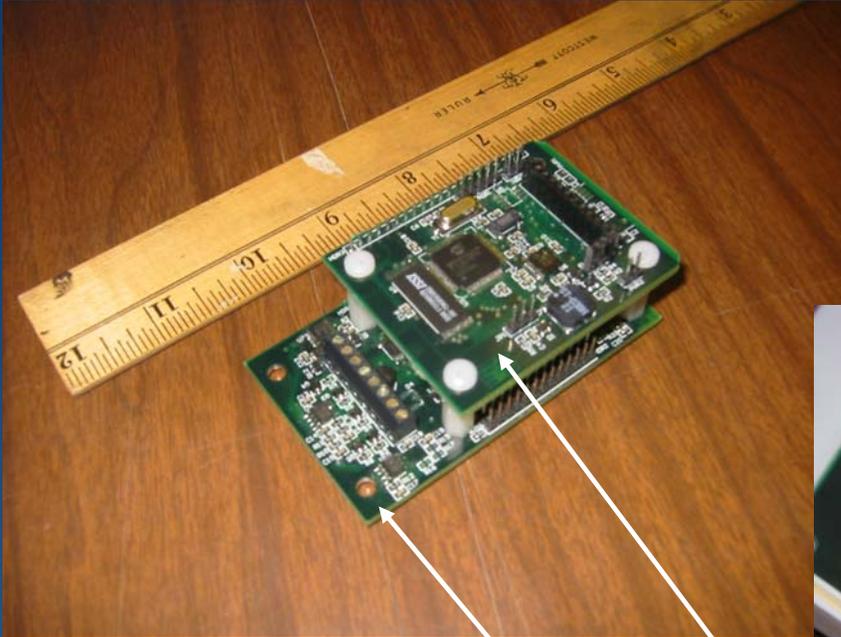
User
Community



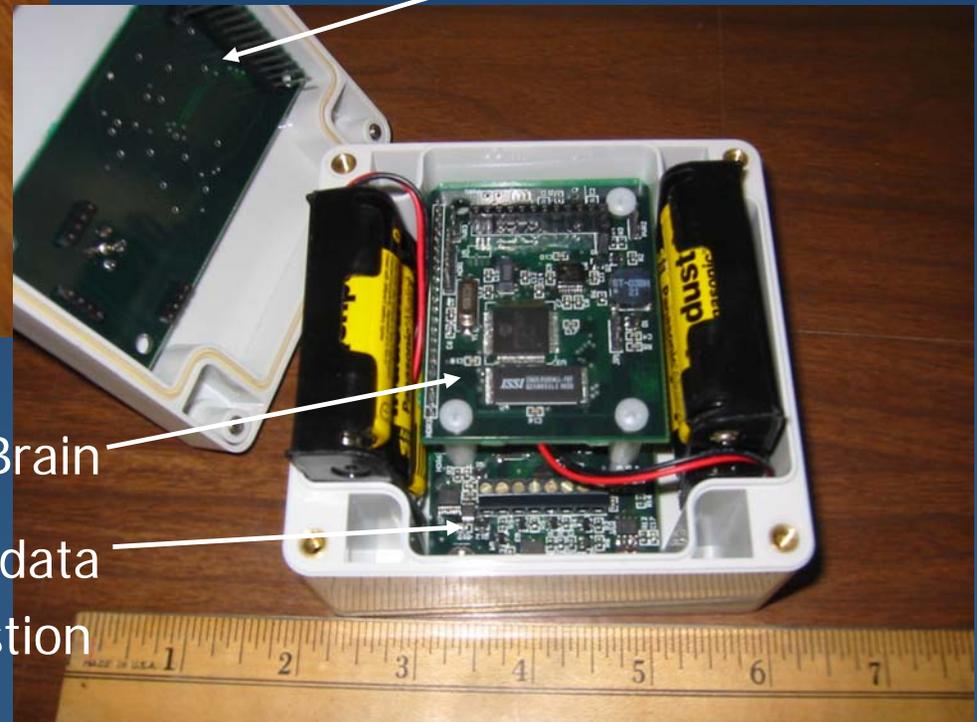
2nd Generation WiSARD

Modular hardware design

- ◆ Dual-processor architecture
- ◆ Three-board stack



Gateway nodes will use memory/time board in place of sensor data acquisition board



Radio

Brain

Probe data acquisition

G2 WiSARDNet Design

Communication and Networking

- ◆ 902 – 928 MHz ISM band
- ◆ Non-Coherent Binary FSK (NC-BFSK) modulation
- ◆ Slow time/frequency hopping spread spectrum via pseudo-random number generator
- ◆ CRMA radio channel sharing algorithm
 - Distributed control
 - Local information
 - Scalable

Self Organization

- ◆ Periodic search for new nodes
- ◆ On-demand search for lost nodes (under development)
- ◆ Can add, move, or delete nodes

Power Management

- ◆ Monitor power status
- ◆ Report battery voltage
- ◆ Adjustable radio transmit power (under development)

Scheduler

- ◆ Dynamic scheduling of communication
- ◆ Online-configurable sample rate (under development)

User Interface

- ◆ Command line from PC
- ◆ User selection of ID, sample rates
- ◆ On-line diagnostics



G2 WiSARD Capabilities

Built-in probe interfaces

- ◆ 12-bit A/D conversion
- ◆ 4 temperature channels
 - thermocouple
- ◆ 4 light (PAR) channels
 - photodiode
- ◆ 2 general purpose probe channels, two power outputs and two CCP modules (Capture/Compare/PWM)
 - Soil moisture
 - Decagon Ech2oprobe
 - Serial communication with intelligent probes
 - Sap flow (future)

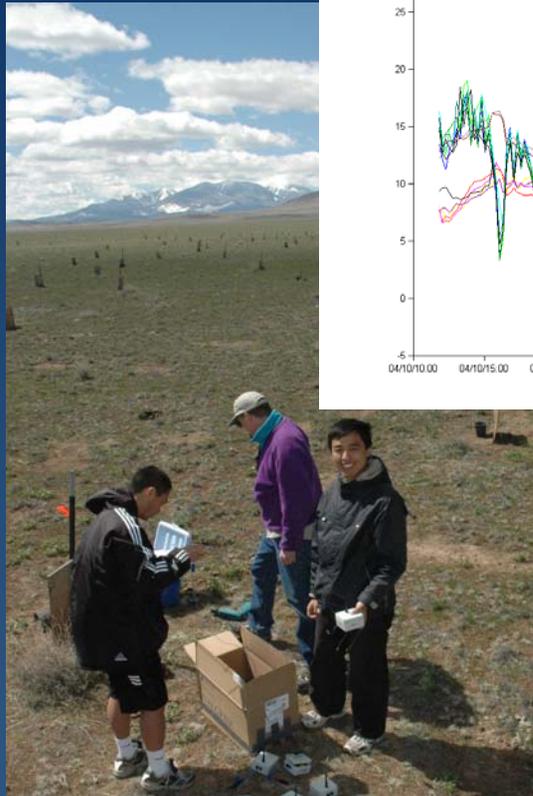
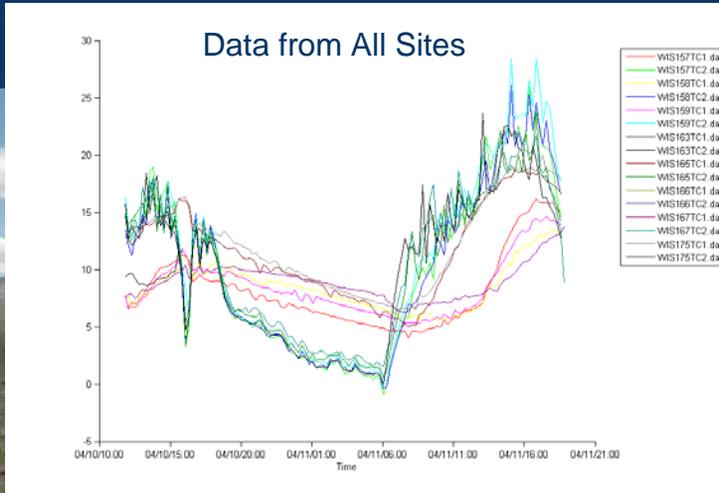
Interface for multiple additional intelligent probes

- ◆ One-wire bus

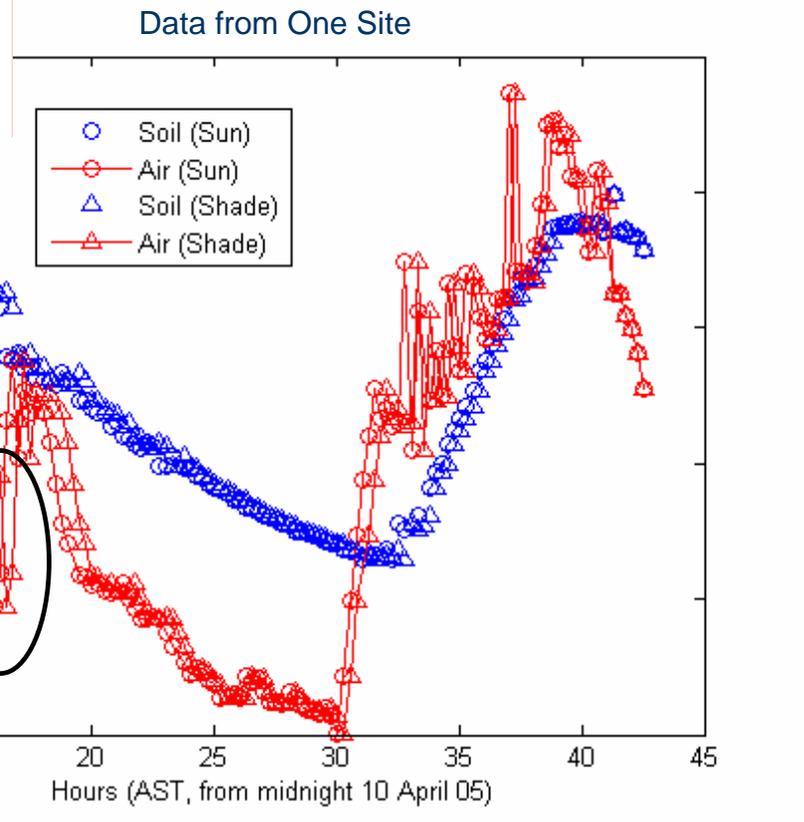
Provision for external energy supplies

- ◆ Supports autonomous switching between internal and external energy sources
- ◆ Battery-backed solar

Trial Deployment, April 2005: Grasslands site, C. Hart Merriam Elevational Gradient



Microburst and
brief cloud cover



A few lessons learned so far...

- ◆ Packaging is a challenge
 - Weatherproofing vs. probe interfaces
- ◆ Probe costs are significant
 - Probeset costs will exceed sensor node costs
- ◆ Deployment is time-consuming
 - 3D space
 - Truly “embedded” probes
- ◆ Requirements for deployment will vary by site
- ◆ Data correlation radii vs. transmission range
- ◆ Maintenance and QA
 - Probe models in non-stationary environments

Dynamic Sensing of Ecosystem Processes

- ◆ motivating application: **revolutionize understanding of environmental change**
 - forecast how altered climate and CO₂ will impact biodiversity and carbon storage in the biosphere
- ◆ challenge: endow network with sufficient explanatory power under significant energy consumption constraints
- ◆ tight coupling between the sensed and the sensors
 - sensors are even more deeply embedded in their environments
 - only sense and communicate when you need to

Application Properties

- ◆ data is highly heterogeneous: natural scales range from
 - meters to landscapes
 - seconds to years
 - ◆ systems are strongly non-linear and non-stationary
 - ◆ motivates **dynamic model-driven control** of
 - sampling
 - communication
 - estimation and prediction
 - model inference
- according to their relative values and costs:

integrate spatio-temporal sensing with modeling and prediction in an adaptive framework

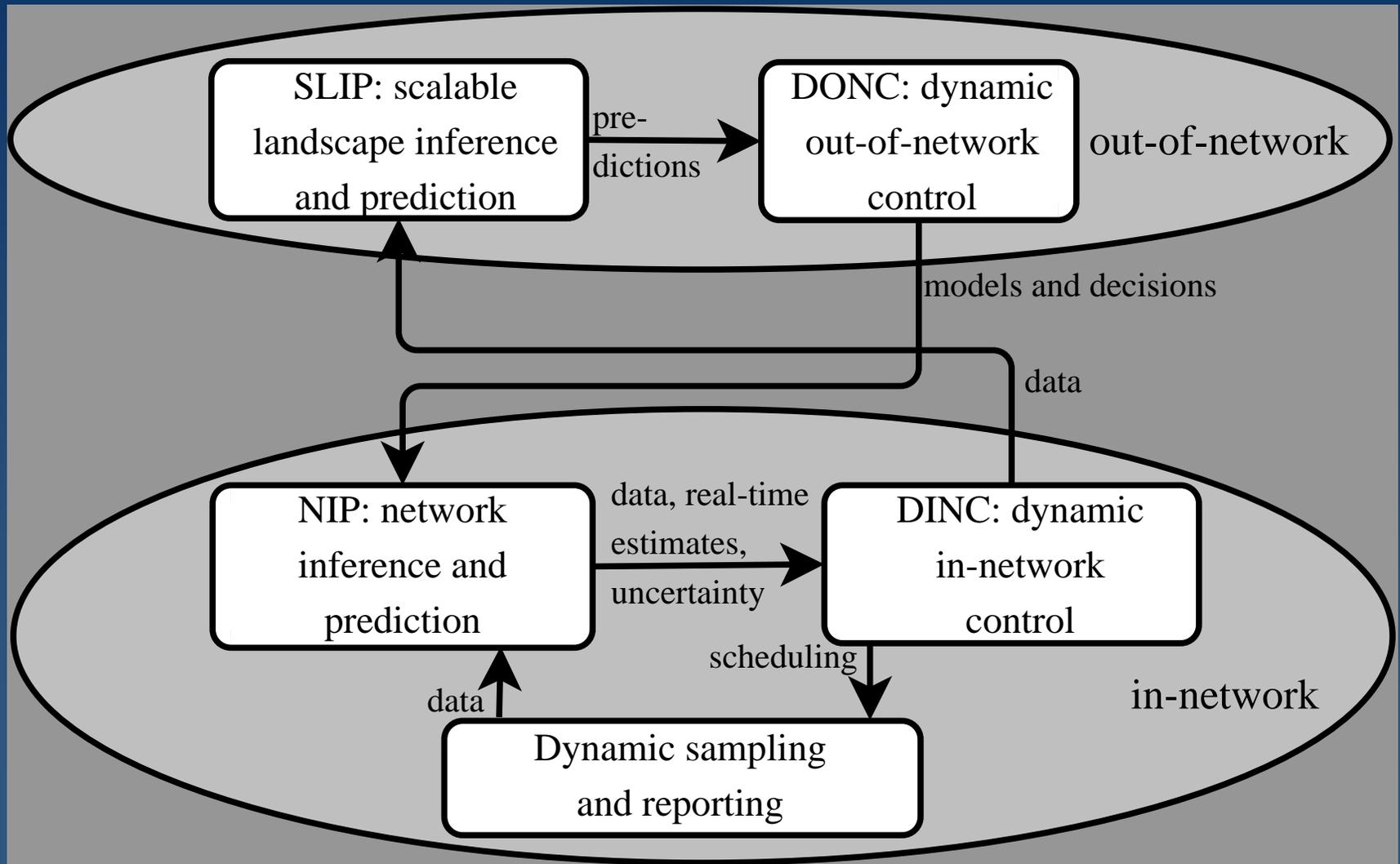
Architectural Properties

1. limited in-network resources
2. complementary out-of-network capability
 - relaxed energy constraints
 - massive processing power
 - latency

approach

- adaptive in-network joint estimation and coding (inner, fast control loop)
- supervisory out-of-network processing (outer, slow control loop) – model inference

Dynamic Sensor Networks: Components of Control



Model-driven sensor network activity

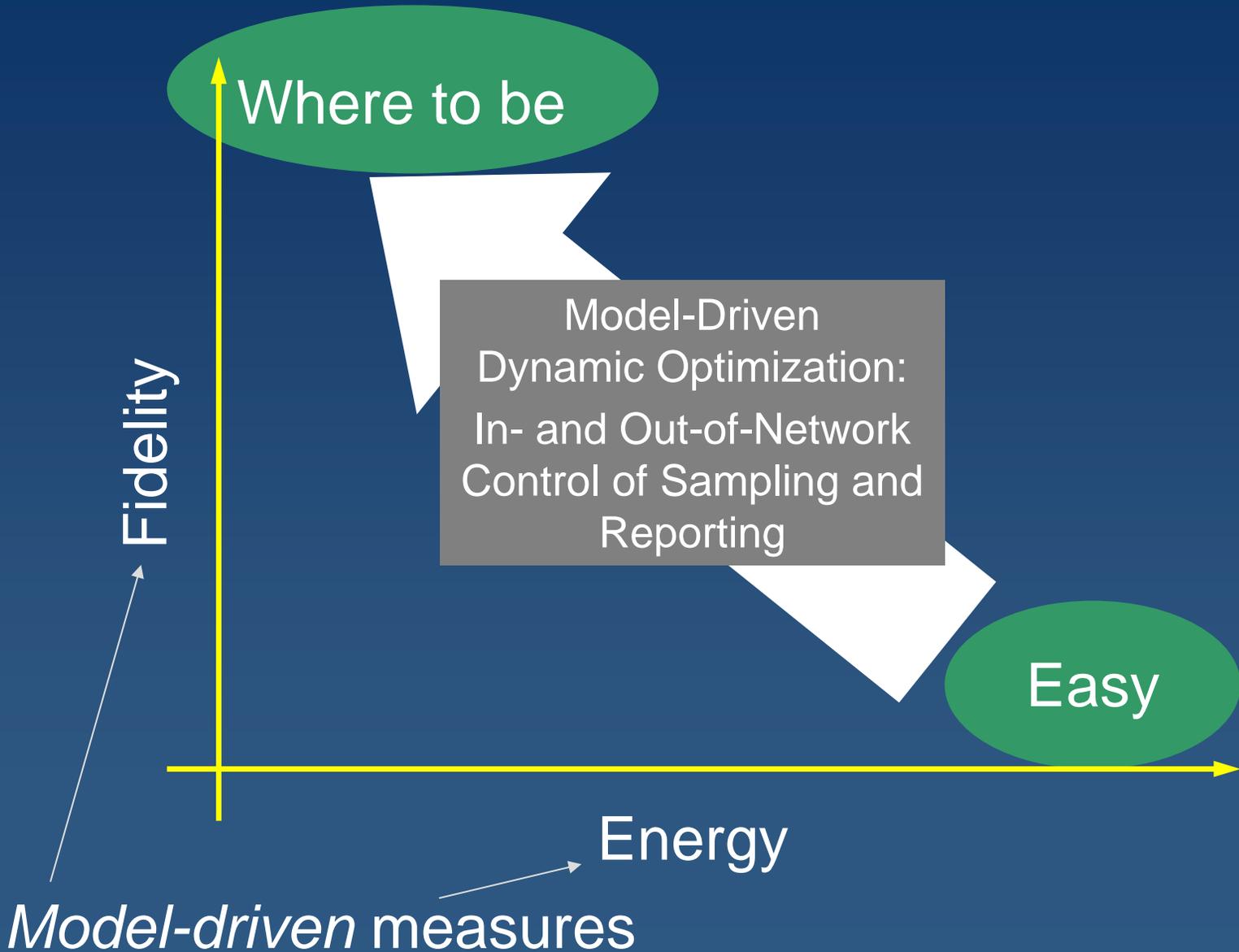
Modeling effort

- ◆ Model, measure, and control uncertainty as a function of cost
 - Hierarchical Bayes
 - MCMC

Data service component implementation

- ◆ coordinates the execution and adaptation of sub-plans and their interaction with WiSARDNet sensors and communication layer
- ◆ tags data with meta-data about sampling and measurement conditions
- ◆ provides multi-resolution data storage within the network

Goal: maximum predictive power *at ecological model level* for a given energy cost



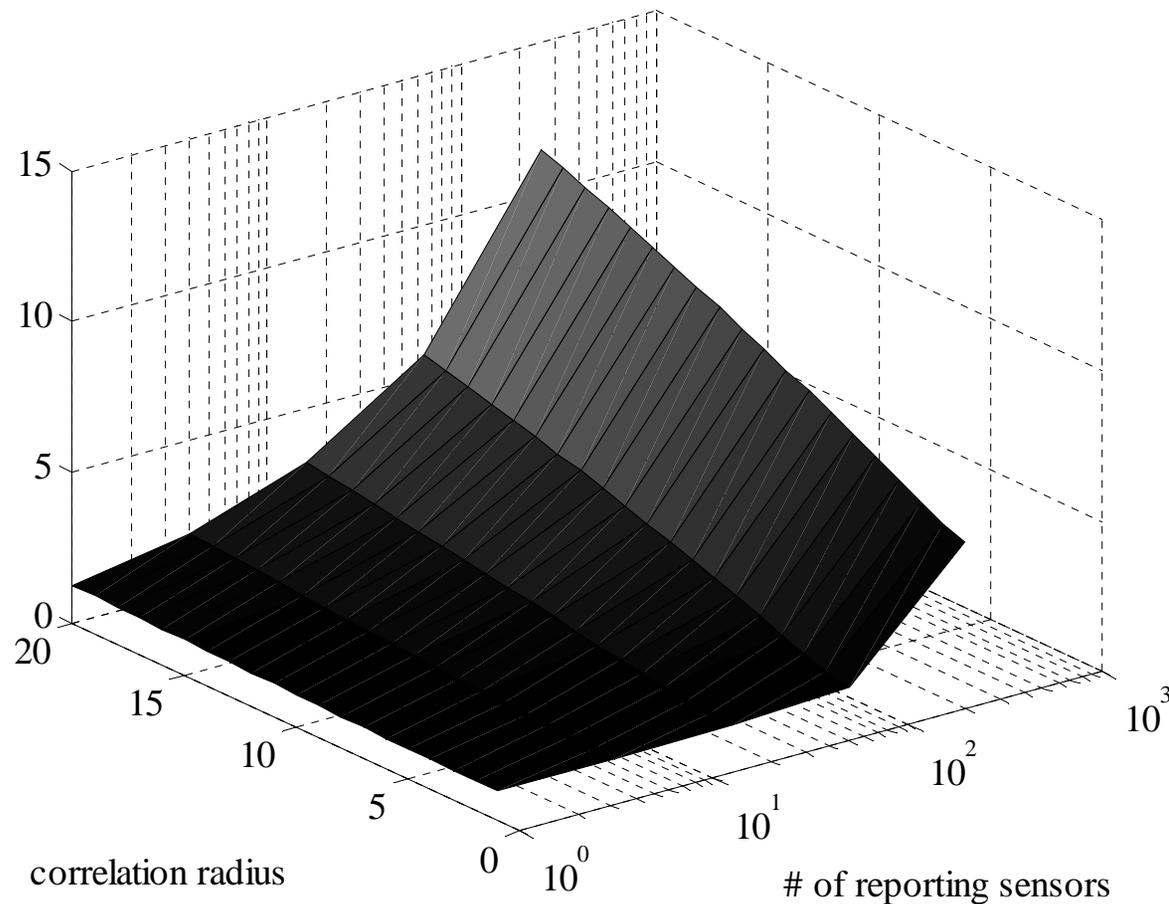
Working at the right level...

Example: multi-sensor time series of temperature.

Can view as

- ◆ streams of numbers
 - use general-purpose source codes (e.g., delta modulation)
- ◆ correlated spatio-temporal process
 - use a parameterized statistical model to drive adaptive space-time sampling and reporting
- ◆ high-level model input
 - sampling driven by the needs of, e.g., leaf efficiency/tree growth model; sampling rate may be high because of sensitivity of high-level model, even when dynamics appear slow

Example: Precision vs. # of reporting sensors: linear estimate of a sampled random field



Algorithmic challenges

- ◆ Finding good models to
 - trade fidelity (including latency) and cost
 - Explore power of dynamic algorithms
- ◆ Cross-layer run-time optimization framework
 - Multiple applications
 - Data services
 - Multi-resolution storage and communication
 - OS
 - Networking

Conclusion

- ◆ Can a dynamic sensor network deliver better results than fixed sampling and out-of-network modeling in the context of changing real-world environmental conditions?
- ◆ Understanding the biodiversity and carbon consequences of environmental change is a problem broad enough to encompass many of the types of challenges faced by DDDA systems.
- ◆ Results from these experiments will inform the design of accurate and energy-efficient production networks tailored to specific applications.
 - Micrometeorological sensing and nowcasting, pollution monitoring and environmental remediation, and public security/safety.