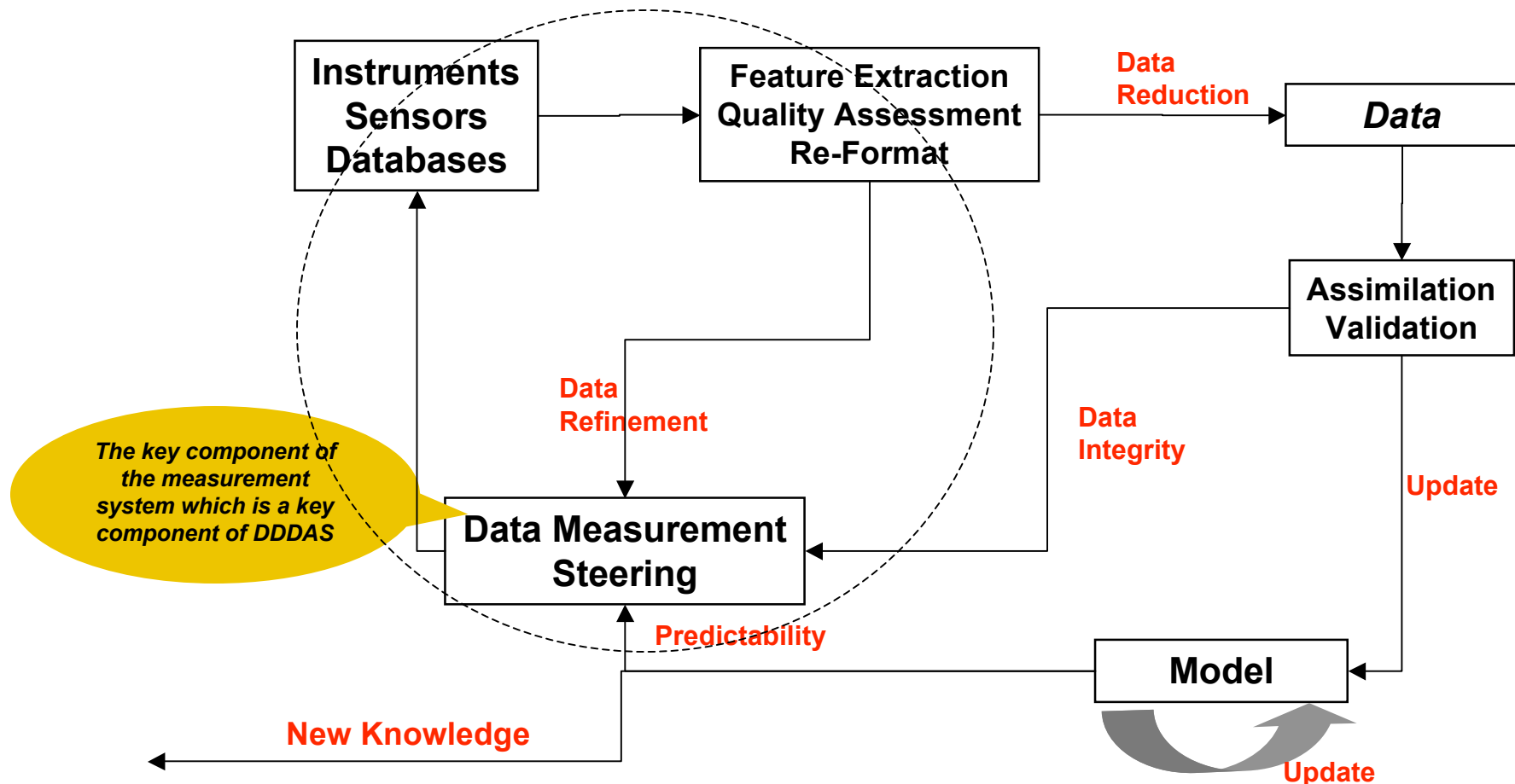


Measurement Systems

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+ ~10 others

Measurement system within DDDAS Framework



- Application / system software / algorithms impact all aspects of this process
- **Issue:** Need to be able to accurately predict the **value** of the measurements to perform the measurement redeployment

Data Measurement Redeployment

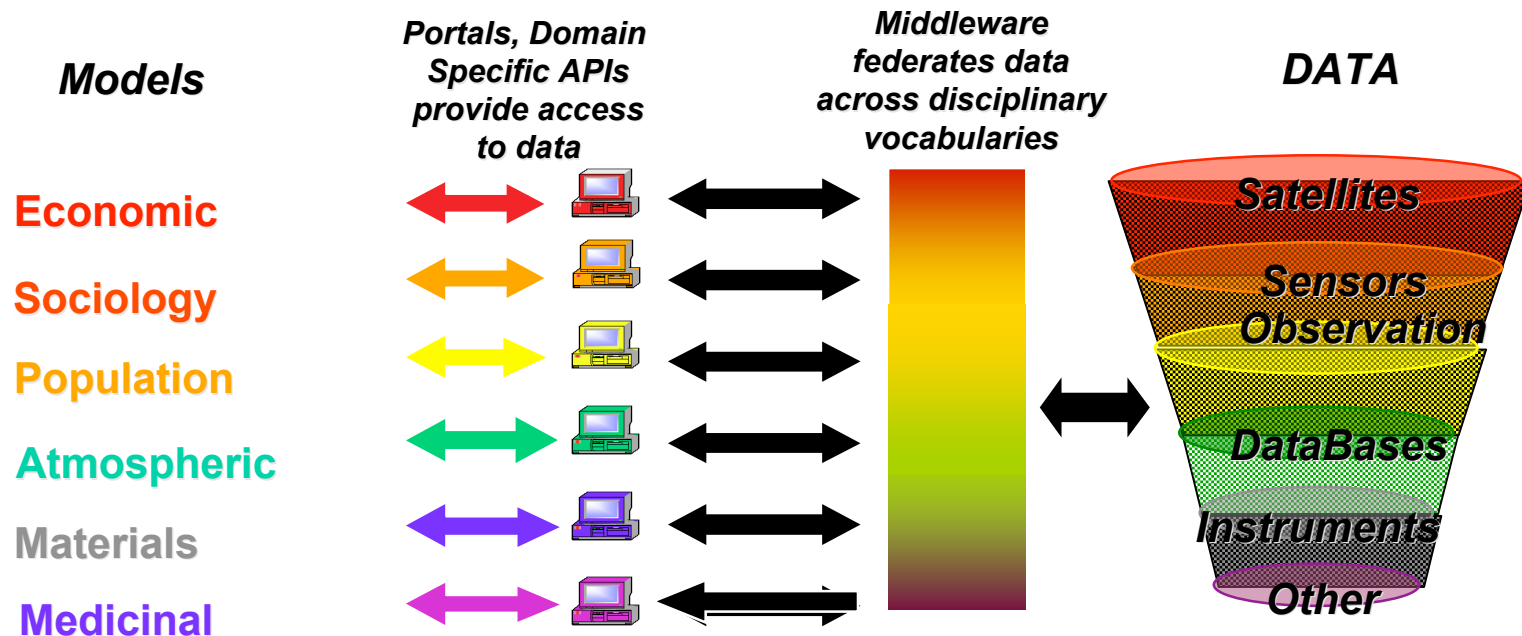
- Process depends on sensor types available and the application
- Different types of sensor selection choices
- **Static:** e.g., change sensor properties
 - Examples:
 - Many distributed structural sensors
- **Mobile:** e.g., change where measurements are taken and on what length/ time-scales
 - Mobile sensor networks (communication): UAVs, UUVs, Bouys
 - Distributed on-demand measurements, information consensus
 - Self-organizing, redeployment
 - Examples:
 - Aircraft flight plans loosely coupled to weather modeling → tighter coupling
 - River environmental sensing (passive deployment) → active deployment
- **Adaptive:** e.g., change what types of measurements are taken
 - Focus analysis of a camera image to different regions
 - Adaptive optics
 - Examples:
 - Real time 2D MRI slice → 3D
 - Fixed frequency spectrometers → different frequency ranges
 - Fixed grid → Adaptive grid (B. Plale)
 - When/where/which sensors to adapt
 - Adapt to what and at what price?

Redeployment Issues

- When data is collected, to sense or not to sense, what to sense?
 - How will we use the models/data to make these decisions?
- For some applications, it is not obvious what to measure
 - Can be a very indirect process for some complex systems (biological)
 - Need a good model of what the sensor does and how it interacts with the system to assess the value of the measurement → can we even write $y=h(x,t)$?
 - How can we use the data to directly improve this process.
 - What algorithms are used? (nonlinear estimation)
- For what reason – what is the objective / goal of the sensing?
 - To obtain new knowledge, for closed-loop control, to improve the model quality, to reduce the uncertainty in the loop, or to improve the system observability
 - Optimization
 - What are the cost/benefit of deployments - Use of finite resources (battery) - QOS of the sensor network (high BW over short periods or low BW over long periods)
- When - How dynamic is the system?
- Where to measure?
 - Measurement allocation process – analyze the utility of moving sensors to specific locations
 - Danger (e.g., Fire tracking)
 - Cost – retrievable? (e.g., Ocean)
- How to measure - Feedback to enhance the robustness of the measurement process
- How automate this process?
 - Can we prove that the process is stable and/or show that it is improving the model?

Issues 2: Limiting/Enabling Factors

- *Cost of measuring system*
 - *Cost of device/static vs mobile*
- *Power*
 - *e.g., battery life; reliability in extreme conditions; bandwidth of data collection limitations*
- *Uniformity of Interface to measurement process*
 - *protocols; standards*



Issues 2: Limiting/Enabling Factors

- Interdisciplinary nature from theory to expt - education/training need for more specialized education opportunities at the interface of disciplines
- Improvements in measurement systems:
 - Power, e.g., battery life; reliability in extreme conditions
 - collection bandwidth
 - Design/robustness of expt systems
- Improvements in means/methods of collecting data
 - Focus in regions of relevancy
 - Controlling sampling
- Assess the validity of data
 - Error bars; time stamps; quantification of uncertainty
 - Uniformity of data across instruments/sensors of the same type.
 - Sensitivity/quality of same data as measured across different types of instruments/sensors/databases
 - Failure indicators
 - Capacity to collect; percentage of noise

Issues 2: Limiting/Enabling Factors

- 'Features' in data
 - When to include data as 'rare' events (for training) and
 - when to assume it is a fault of the measuring device
 - How to recover from 'feature' events
- Decision as part of the measurement system:
 - New component offered by DDDAS
 - Feedback loop is the decision/prediction of what data is needed to improve model
 - Need for more data
 - Where to collect
 - Error correction
 - Quantification of uncertainty
 - Quality decision as a function of application type
 - fast/cheap/lower quality
 - slow/expensive/higher quality