

Information Fusion: the Concepts, the Technology, the Community, and Modern Research Challenges

Dr. James Llinas

Research Professor (Emeritus), Executive Director
Center for Multisource Information Fusion
State University of New York at Buffalo, USA
llinas@buffalo.edu



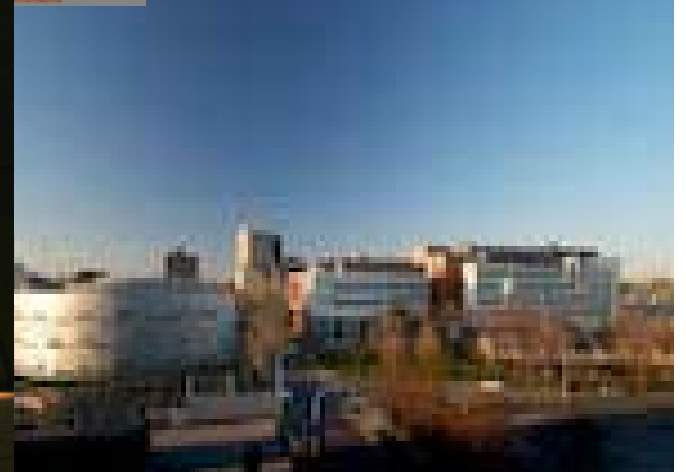
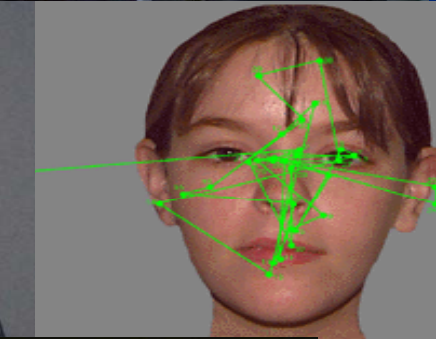
CMIF Location: Buffalo, New York, USA



About the “University at Buffalo”

- “University at Buffalo” aka State University of NY at Buffalo
- New York State’s largest and most comprehensive public university
- Member of the Association of American Universities
- Funded research activity in the range of US\$350M per year
- Selected Research Centers:
 - Center for Multisource Information Fusion
 - Natl Center for Geographic Information and Analysis
 - National Center for Ontological Research
 - Virtual Reality Laboratory
 - Center for Information Systems Assurance
 - Lab for Advanced Network Design, Evaluation and Research
 - Wireless and Networking Systems Lab
 - Semantic Network Processing Systems Research Group
 - The Center for Unified Biometrics and Sensors
 - Center for Computational Research (Supercomputing)
 - Center for Document Analysis and Recognition
- Current enrollment approx 30K+ students, 18K undergrad, 12K grad

Around the Campuses: Suburban-South-Downtown



CMIF OVERVIEW

- **Mission:** Information Fusion and related areas primarily but not exclusively for defense and homeland security applications
- **Basic and Applied Research in:**
 - Multiple-sensor and instrumented systems
 - Synergistic Human-Multisensor systems
 - Real-time Decision-making using Hierarchical Fusion
 - Graph Theory and Optimization for Level 2/3 Fusion
 - Multi-modal/spectral information environments (speech+text+imagery+RF sensor+human input)
- **Applications:**
 - Defense: Intelligence/Surveillance/Reconnaissance; Tactical Applications; Homeland Security
 - Non-Defense: Robotics; Conditioned-Based Maintenance; Medical; Transportation; Geology; Natural Disasters/Crisis Mgmt
- **History and Funding:**
 - Started in 1996 with Air Force Research Lab Contract
 - Typical funding activity ~US\$4M/year
- **Scholarly:**
 - Long-standing member of “JDL” fusion group and First President of Intl Society for Info Fusion
 - Extensive publishing by CMIF PI Team including books, JI papers, conference papers and review boards
 - “Critical Issues” Workshops—8 years
 - CMIF is unique in American Universities as a research activity focused on IF technology for DHS/DoD
 - Consortium development to include other universities (SU, RIT and PSU) and industrial partners and development of a Graduate-level pgm in Data Fusion

The Concepts

Everyday Data Fusion

Sound

Smell

Taste

Images

Touch

Pain

Balance

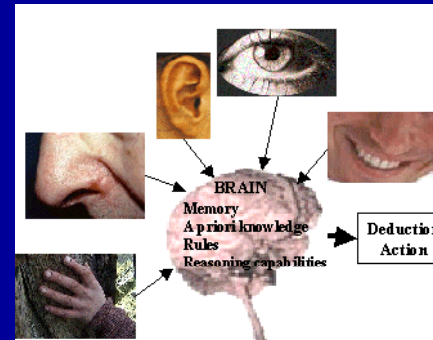
Temperature

Body Awareness
(Proprioception)

Sensing Association



Multinodal Fusion



Augmented Sensing



Robotic Multisensor Fusion

History of Information Fusion

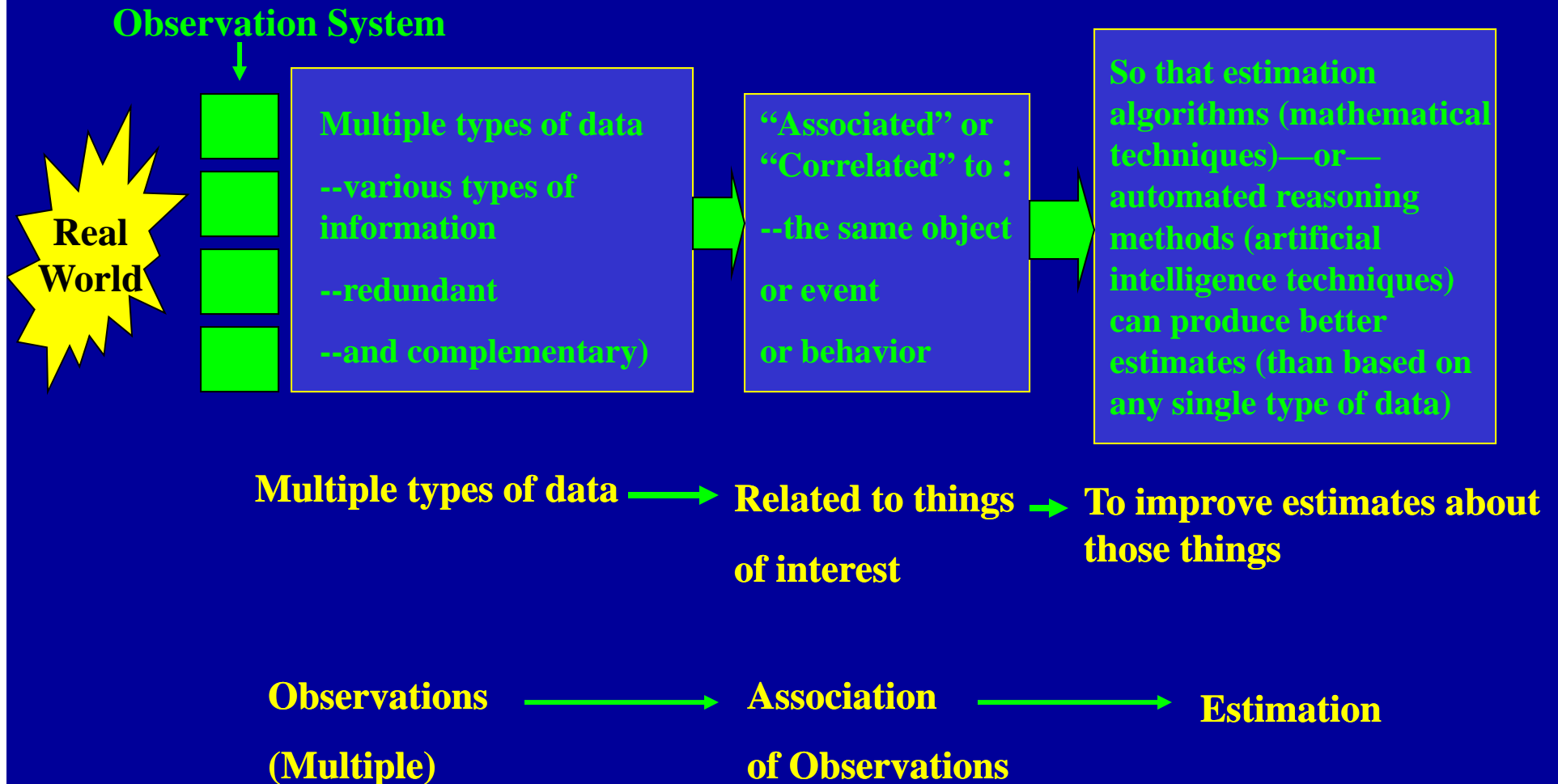
- **Dates to circa 1970's** —fairly young in the sense of technological history—a maturing technology/field of study
- **Driven by defense and intelligence needs**
 - Originally as a “data compression” device to digest huge amounts of sensed data as sensors advanced in capability (a “push” requirement)
 - Later as an important element for decision support (a “pull” requirement)
- **Matures to very broad range of application**
 - Robotics, medicine, imagery/remote sensing, intelligent transportation, conditioned-based maintenance, biometrics, etc

What is Information Fusion?

“Information fusion is an Information Process dealing with the:

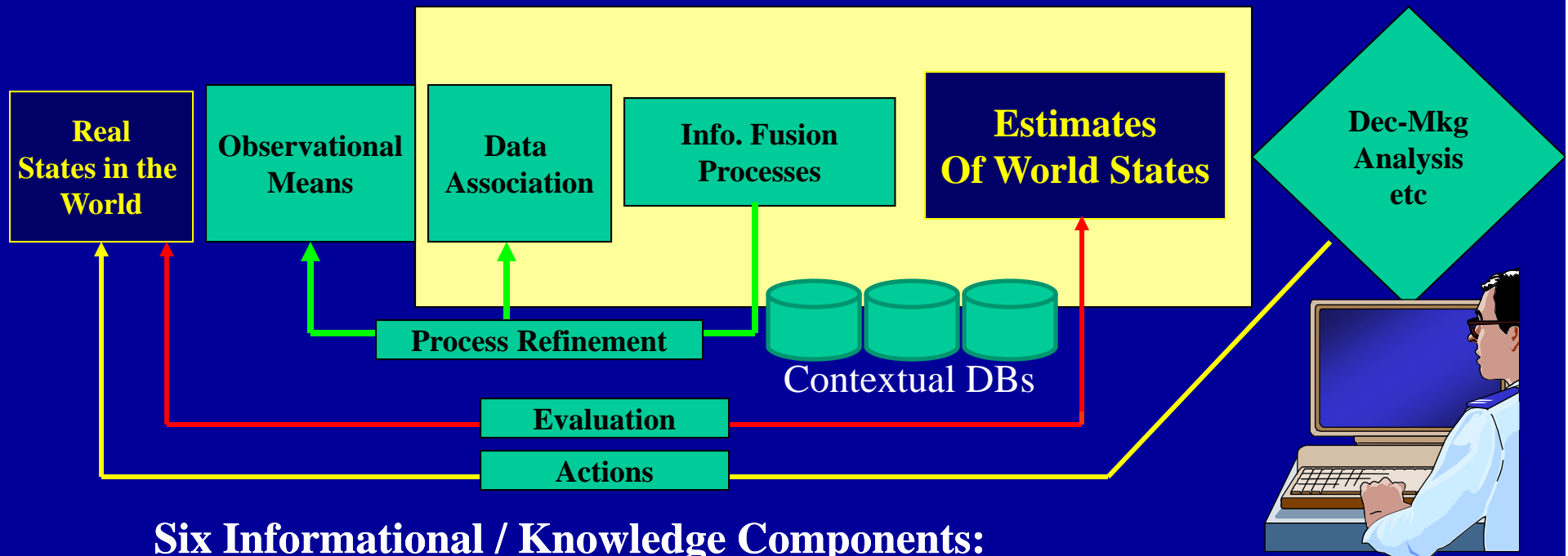
- {*Association, correlation, and combination*} of data and information from
- {*Single and multiple sensors or sources*} to achieve
- {*Refined estimates*} of parameters, characteristics, events, and behaviors for observed entities in an observed field of view
- It is sometimes implemented as a Fully Automatic process or as a *Human-Aiding process* for Analysis and/or Decision Support

Most Simply--



These Basic Ideas are Transferable to Many Types of Problems

Basic Role of Fusion: Adaptive, Recursive Estimation

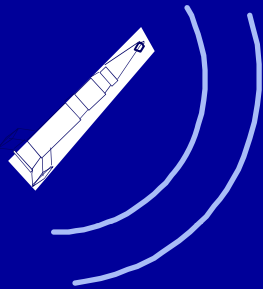


Six Informational / Knowledge Components:

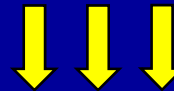
- **Observational Data**
- **A Priori Dynamic World Knowledge Model (Deductive)**
- **Contextual Information**
- **Runtime Learned Knowledge**
- **Tacit and Explicit Human Knowledge**
- **(Network) External Obsvns and Estimates**

Information Fusion Exploits Sensor/Source Commonalities and Differences

Unknown Moving Object



	DETECTION		KINEMATICS		CLASSIFICATION	
	CONFIDENCE	COVERT COVERAGE	RANGE	ANGLE	CLASS	TYPE
RADAR	FAIR	POOR	GOOD	FAIR	FAIR	POOR
EO/IR	FAIR	FAIR	POOR	GOOD	FAIR	FAIR
C3I	FAIR	GOOD	FAIR	FAIR	FAIR	FAIR



GOOD	GOOD	GOOD	GOOD	GOOD	GOOD
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Operational Benefits of Multiple Sensor Data Fusion

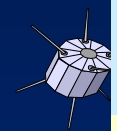
- **Multiple Sensors** →
 - Reliability
 - Improved Detection

- **Multiple Platform Sensors** →
 - Extended Coverage (spatial and temporal)
 - Improved Spatial Resolution

- **Diverse Sensors** →
 - Robustness (Weather/visibility, Countermeasures)
 - Improved Detection

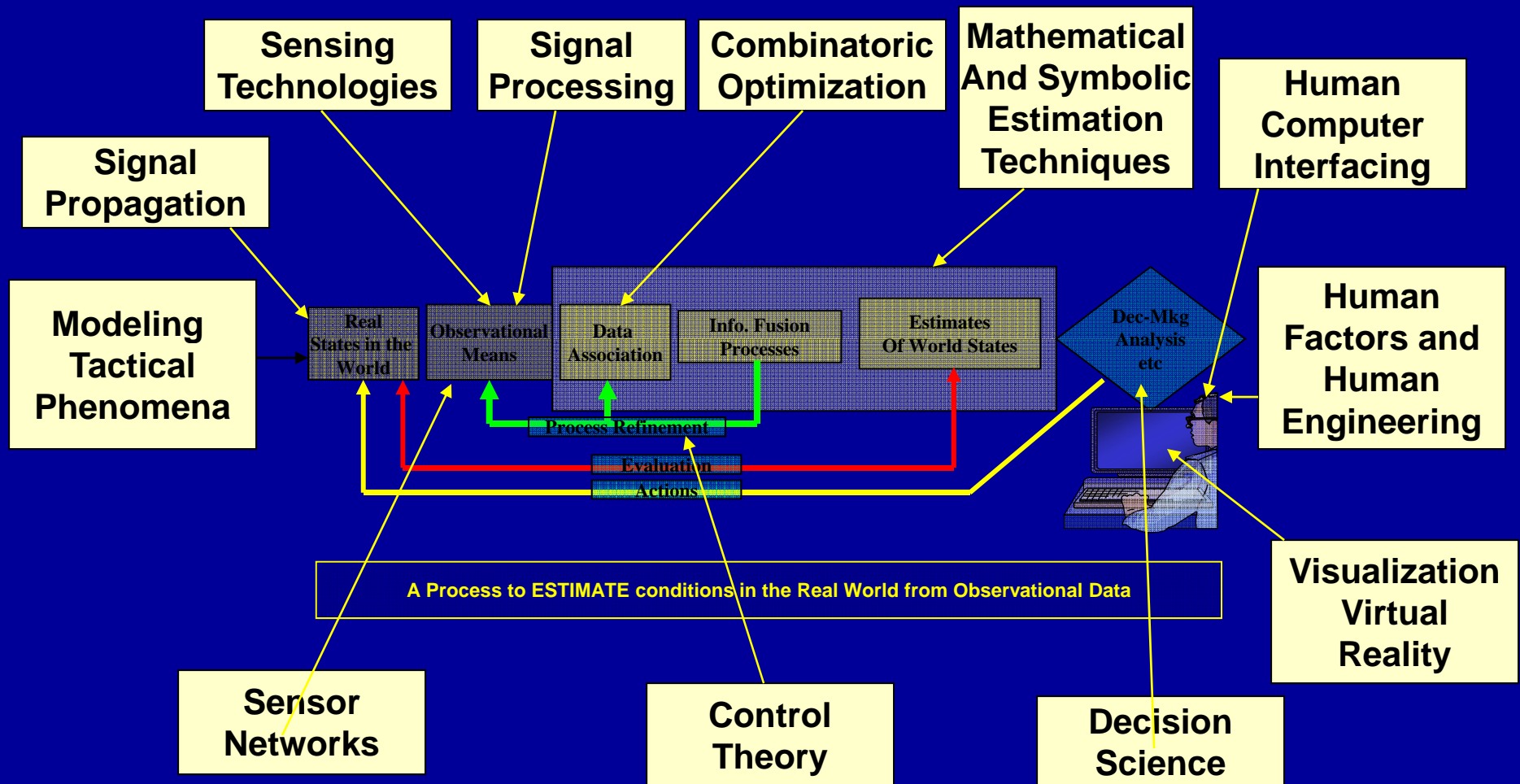
State Estimates of Reduced Uncertainty And Improved Accuracy

- Point and Standoff Sensors
- Data Sources
- Intel Sources
- Air Surveillance
- Surface Sensors
- Standoff Sensors
- Space Surveillance




The Technology

The Technology: Scientific Foundations of the Data Fusion Process



Broadly Multidisciplinary

Choices in Fusion Approaches

Fusion Approach	Nature of Process	
* No Fusion	* Best single-source approach	
* Non-fused but adaptive	* Phased application of single sources: --Multiresolutional --Cueing	
 <p>“Real” Fusion</p>	Synergistic; Adds Information, Reduces Uncertainty	
	--Limited Fusion	--Single Source + A Priori Info --Occasional in time,or --On demand
	--Moderate-Level Fusion	--Few Sources
	--Aggressive-Level Fusion	--”All Source”

Some Research Strategies

- **Estimation-process-centric**
 - Signal processing (eg detection fusion), intersource registration/alignment, estimation algorithms
 - Input: given; Output: mathematical estimate
- **System-centric**
 - Process architecture, standards, integration
 - Process mathematics
 - Process control, estimation/decision-making interdependencies, dynamic resource mgmt
 - Human-system design
 - Input: controllable; Output: usable by a human

Architecting Systems

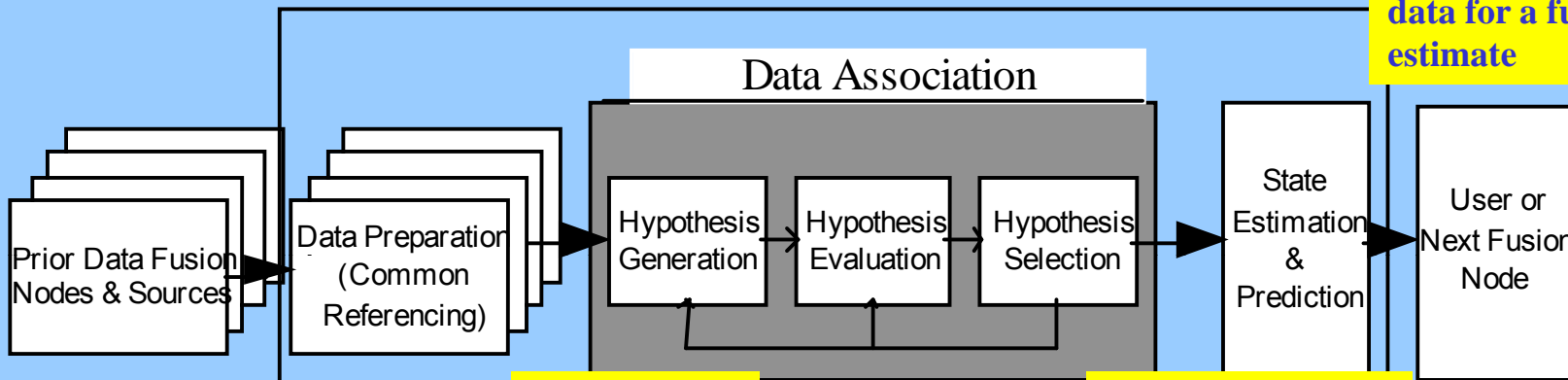
--Architectural Elements

--Dealing with Uncertainties

--Data Association, a core function

Data Fusion Tree Node

Data Fusion Tree Node



Now exploit the multiple observational data for a fused estimate

Things that can cause expected observations

How it is that observations are related to the entities or objects
(A notion of "closeness"—a "score")

Optimally assigning the observations to an estimation process which is estimating a parameter of interest for the entity/object

- Detect and resolve data conflicts
- Convert data to common time and coordinate frame
- Compensate for source misalignments

- Gating and generation of feasible and consistent association hypotheses
- Scoring of data associations
- Select, delete, or feedback data associations

- Estimate sensor/source misalignments
- Feed forward source/sensor status
- Aggregate states (blue, red)

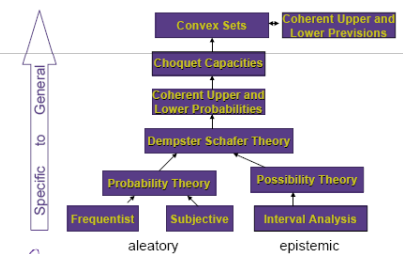
Architectural Elements

<p>Data Fusion Node</p>	<p>Common Representation for all Data Fusion Processes</p>	<p>The diagram illustrates the Fusion Node Paradigm. It shows a sequence of processes: SOURCES or prior fusion nodes feed into DATA ALIGNMENT, which then feeds into DATA ASSOCIATION, and finally into STATE ESTIMATION & PREDICTION. The output of this process is sent to a USER or next fusion node. Two external inputs are shown: SOURCE SENSOR STATUS and RESOURCE MGT CONTROLS, both pointing to the central processing block.</p>
<p>Data Fusion Network</p>	<p>Common Representation for all Data Fusion Architectures</p>	<p>The diagram shows a network of five sensors (Sensor 1 to Sensor 5) feeding into a series of Fusion Nodes (F). Sensor 1 and 2 feed into a first F node. Sensor 3 and 4 feed into a second F node. The outputs of these two F nodes feed into a third F node. Finally, the output of the third F node and the output of Sensor 5 feed into a final F node, which produces the network's output.</p>
<p>Integration of Data Fusion and Resource Management Networks</p>	<p>Common Representation for all Information System Architectures</p>	<p>The diagram illustrates the integration of data fusion and resource management. It shows three sensors (Sensor 1, 2, 3) and two resources (Resource x, Resource y). Each sensor and resource is connected to a pair of nodes: a Fusion Node (F) and a Management Node (M). The Management Nodes are interconnected, and the Fusion Nodes are also interconnected. A legend on the right defines F = Fusion Node and M = Management Node.</p>

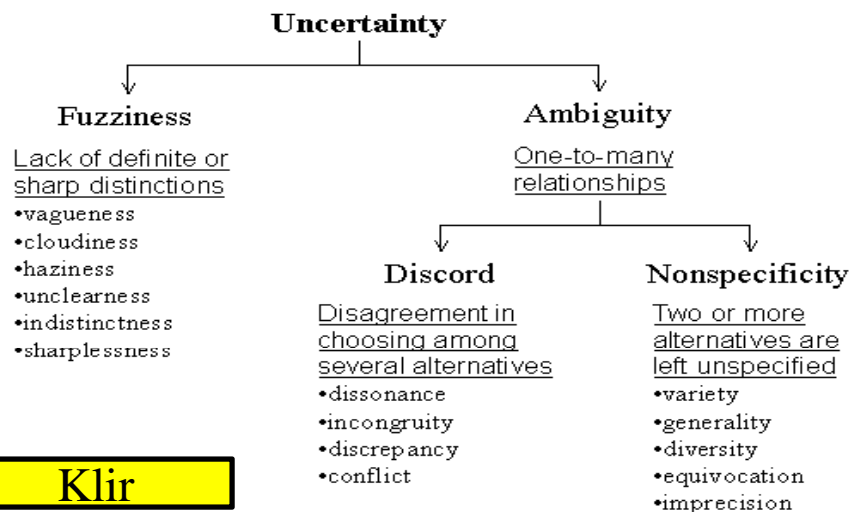
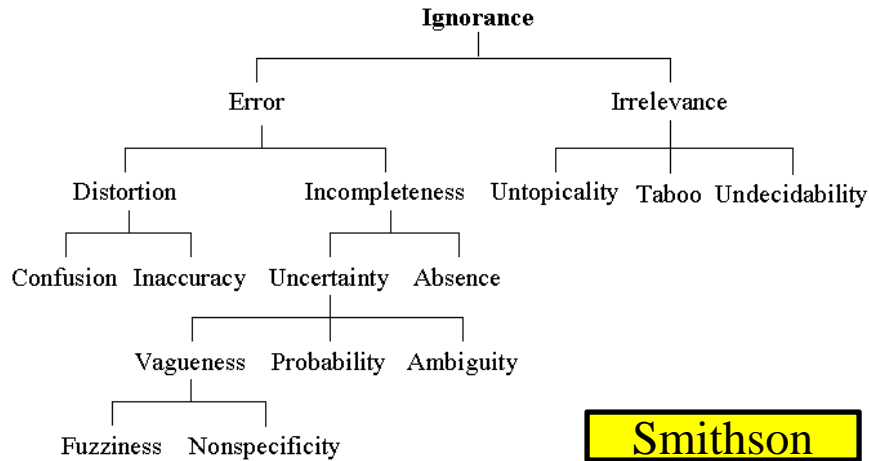
Dealing with Uncertainty

Second Order Uncertainty and Imprecise Probability

- Theoretical aspects of Second Order Uncertainty
 - Focus on Epistemic Uncertainty (limitations in knowledge)
 - Aspect of degrees of satisfaction of the Kolmogorov Axioms, especially the Additivity Axiom
 - Walley (1991) shows that imprecise probabilities satisfy the principles of coherence—relaxes need to satisfy Additivity
- Leads to range of alternatives, each of which satisfy “relaxed” Additivity Axioms

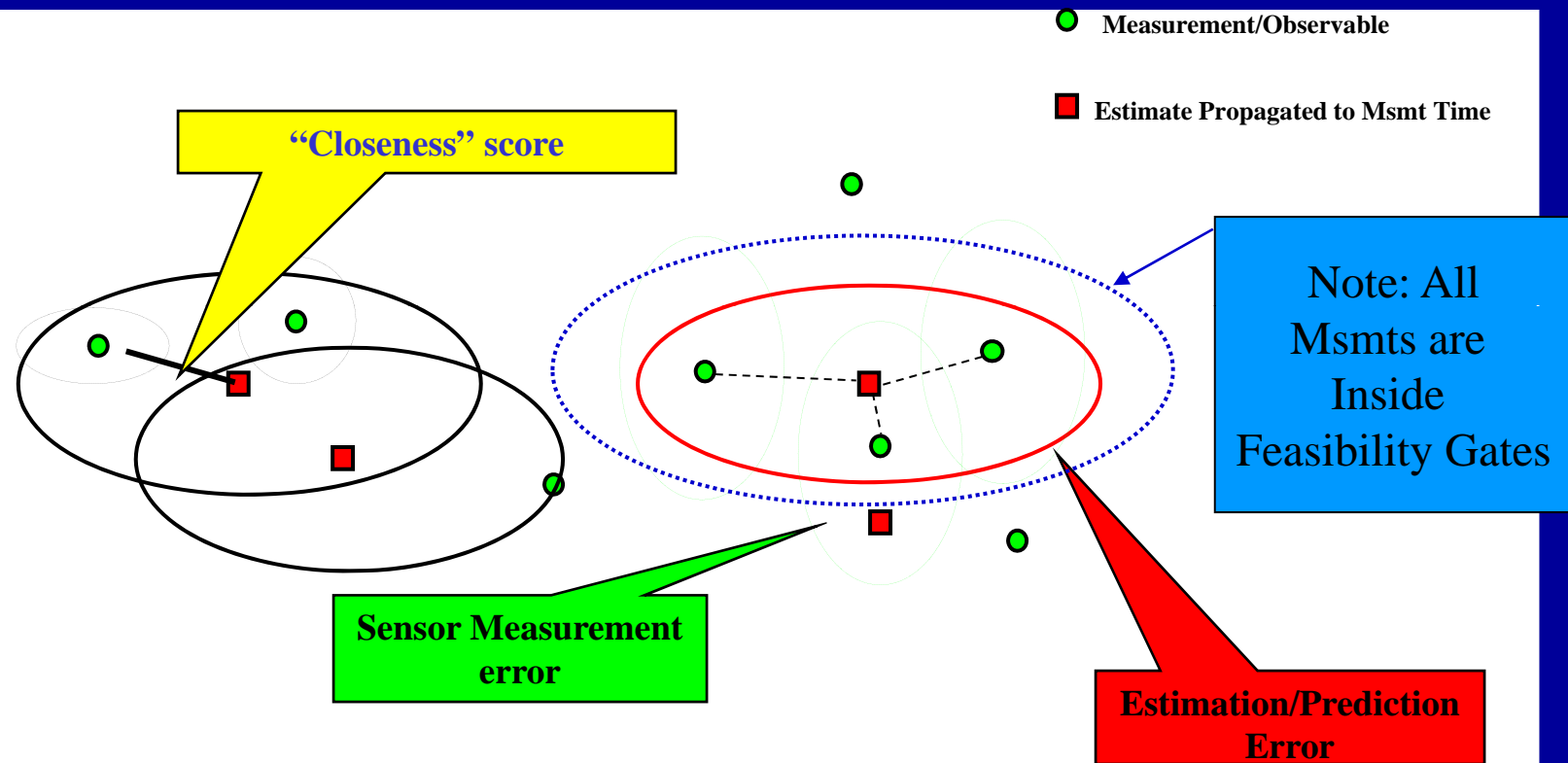


Most problems involve both Aleatory and Epistemic Uncertainties



Data Association Basics

--What measurement goes with what entity?

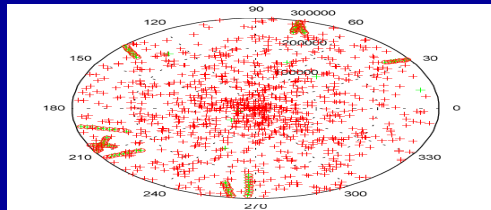


Leads to the formulation of a classic OR Assignment problem with usual repertoire of solutions

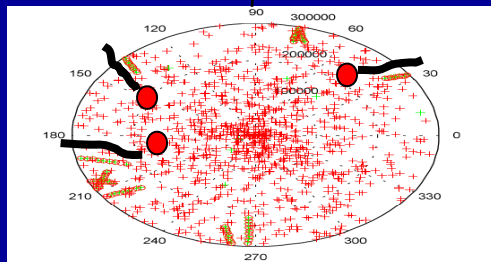
Association and Assignment Optimization

O	T _i	T _j	T _k	T _l	●	●	●
m _i							
m _j							
m _m							

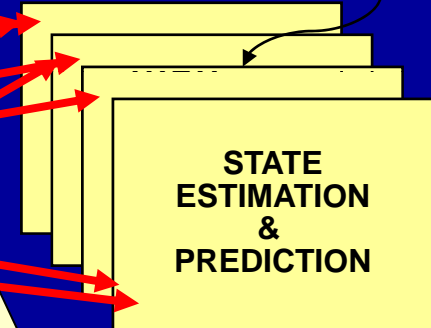
Need some type of Mapping that determines a good way to allocate Obsvns To Tracks



M Observations From N Sensors

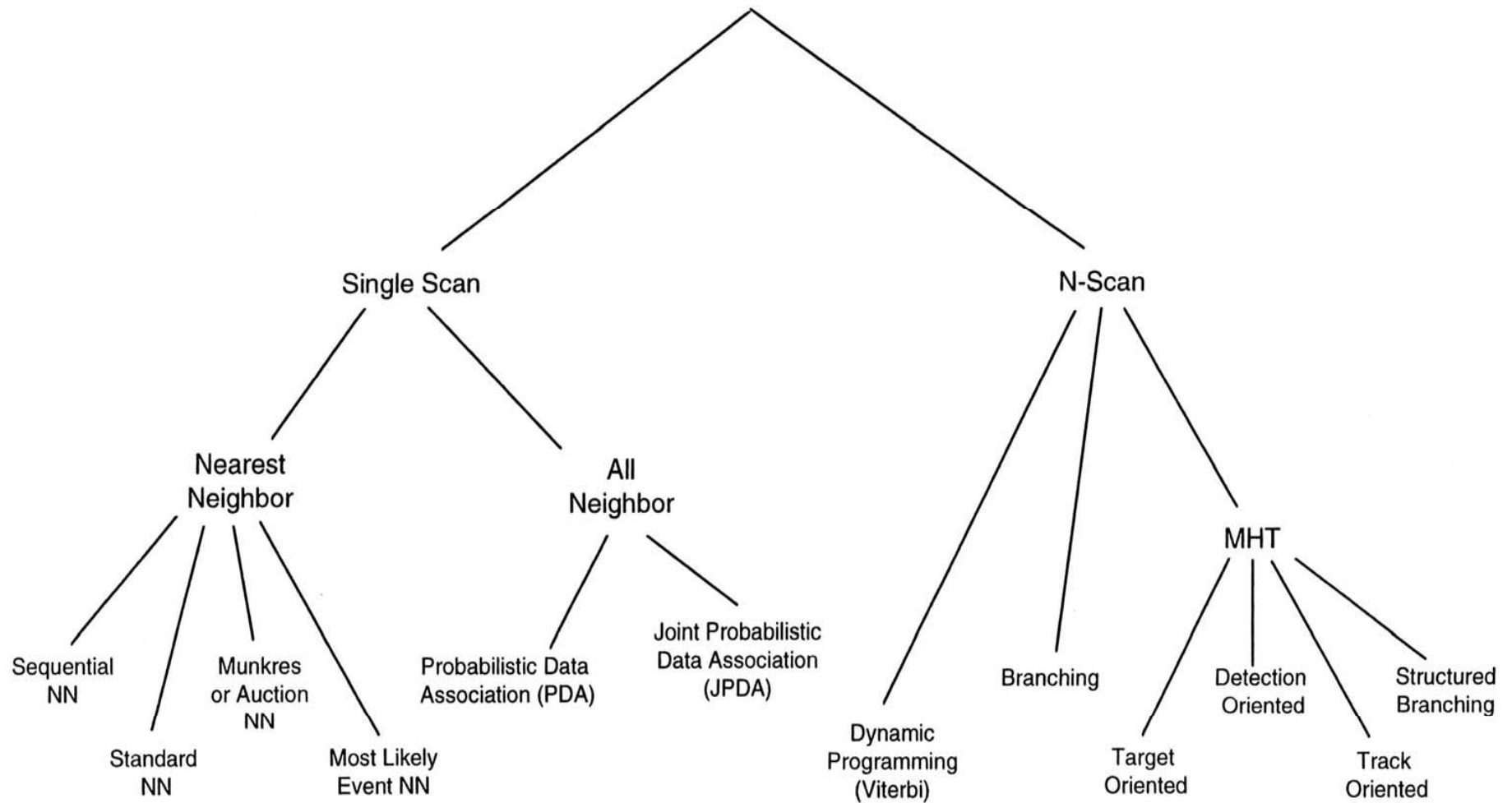


Tracks "T"



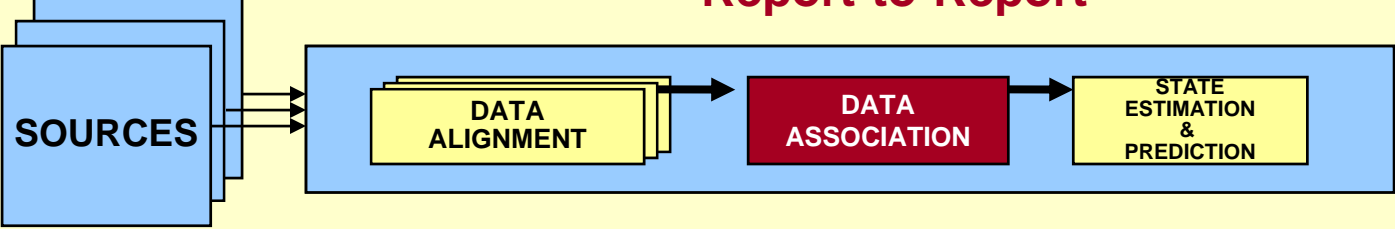
"Assigned" Observations Resulting from some "Best" way to decide which Observations should be "given" to each State Estimator

One Taxonomy of Assignment Problem Solutions



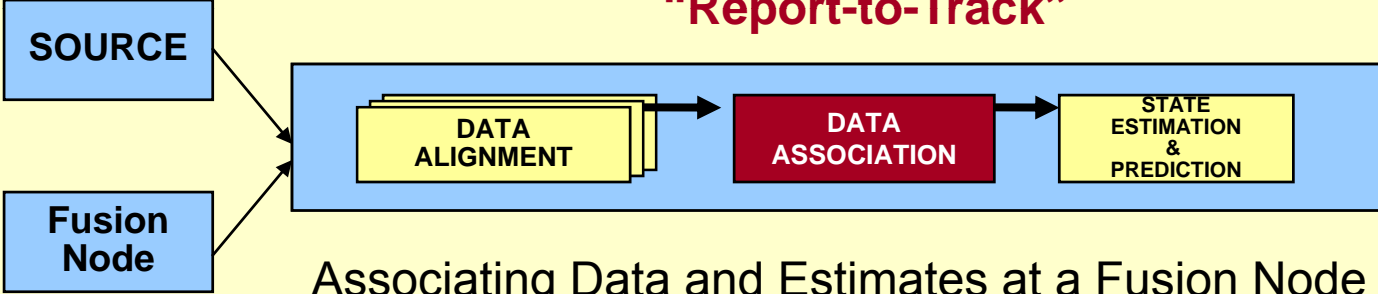
Categories of Data Association

“Report-to-Report”



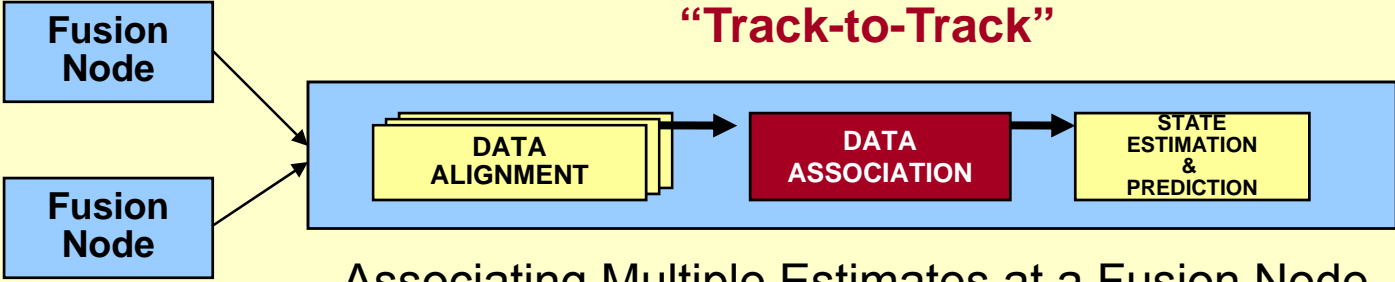
Associating Data from Multiple Sources at a Fusion Node
("Measurement-to-Measurement Association")

“Report-to-Track”



Associating Data and Estimates at a Fusion Node
("Measurement-to-Estimate Association")

“Track-to-Track”



Associating Multiple Estimates at a Fusion Node
("Estimate-to-Estimate Association")

The Community

The Early Community

- **Early 1980's**
 - Mostly US, UK, Australian, Canadian Defense related (in UK eg Royal Signals and Radar Establishment, DRA, etc before DERA)
 - **1985: First US “National Symposium on Sensor and Data Fusion”, NSSDF**
 - US only (attempts at NATO integration fail)
 - Classified
 - Ongoing today
 - **US: Joint Directors of Laboratories**
 - Aids in unifying terminology and concepts
 - **1990: First unified text published**

Second Generation

- **Early 1990's to early 2000's**
 - Still rather ad hoc through early 90's
 - Mid 90's sees evolving structure
- **Mid-late 90's**
 - IEEE Conference on Multisensor Fusion and Integration, from 1995 on
 - International Conference on Information Fusion, annually, from 1998 on
 - International Society of Information Fusion, established in 1999
- **Early 2000's**
 - International Journal of Information Fusion, 2000
 - Journal of Advances in Information Fusion, 2003
 - Many other conferences (e.g., in SPIE)
 - Textbooks begin to flow

Current Status

- **Community**
 - Stable but needing a broader technological view as capability for L1 matures and challenges of L2, L3 are addressed
 - Structured outreach required
- **Operating Domain**
 - Too defense-oriented; multi-domain outreach also required
- **Fusion process and concepts**
 - Need structured extension eg to distributed, networked case and L2, L3 processes
 - Control-theoretic aspects to be addressed
 - Frameworks for cost-effective development

Modern Research Challenges

One List – No particular order

- **Achieving scalability and robustness**
 - Beyond one-algorithm/process solutions
- **Structured, standardized strategies for contextual exploitation**
 - Extends, as a basic research topic, to hybrid deductive/inductive systems
- **Holistic strategies for distributed fusion processes**
 - Eg linking Information-sharing strategies with network fusion operations
- **Dealing with weak knowledge problems (world dynamics poorly understood)**
 - Second-order uncertainty, response-based balanced designs
 - Extends to the case of Situation Management
- **Overall Hard and Soft Fusion process designs and methods**
- **Improved techniques for Test and Evaluation, Metrics**

A Taste of the Hard + Soft Data Fusion Problem

The Network Enterprise, Irregular Warfare, and Fused Situational Awareness



**“Hard”
Sensor
Data**

Calibrated,
Precise

**“Soft”
Sensor/Contextual
Data**

Uncalibrated Human Observers/Uncalibrated Sources

Observations expressed in (inherently) ambiguous language

Extensive Data and Information-Sharing Enabled by Network Infrastructure Gives Rise to a New Challenge in Information Fusion:

“HARD” + “SOFT” INFORMATION FUSION

Some Distinctions in Hard and Soft Observational Data

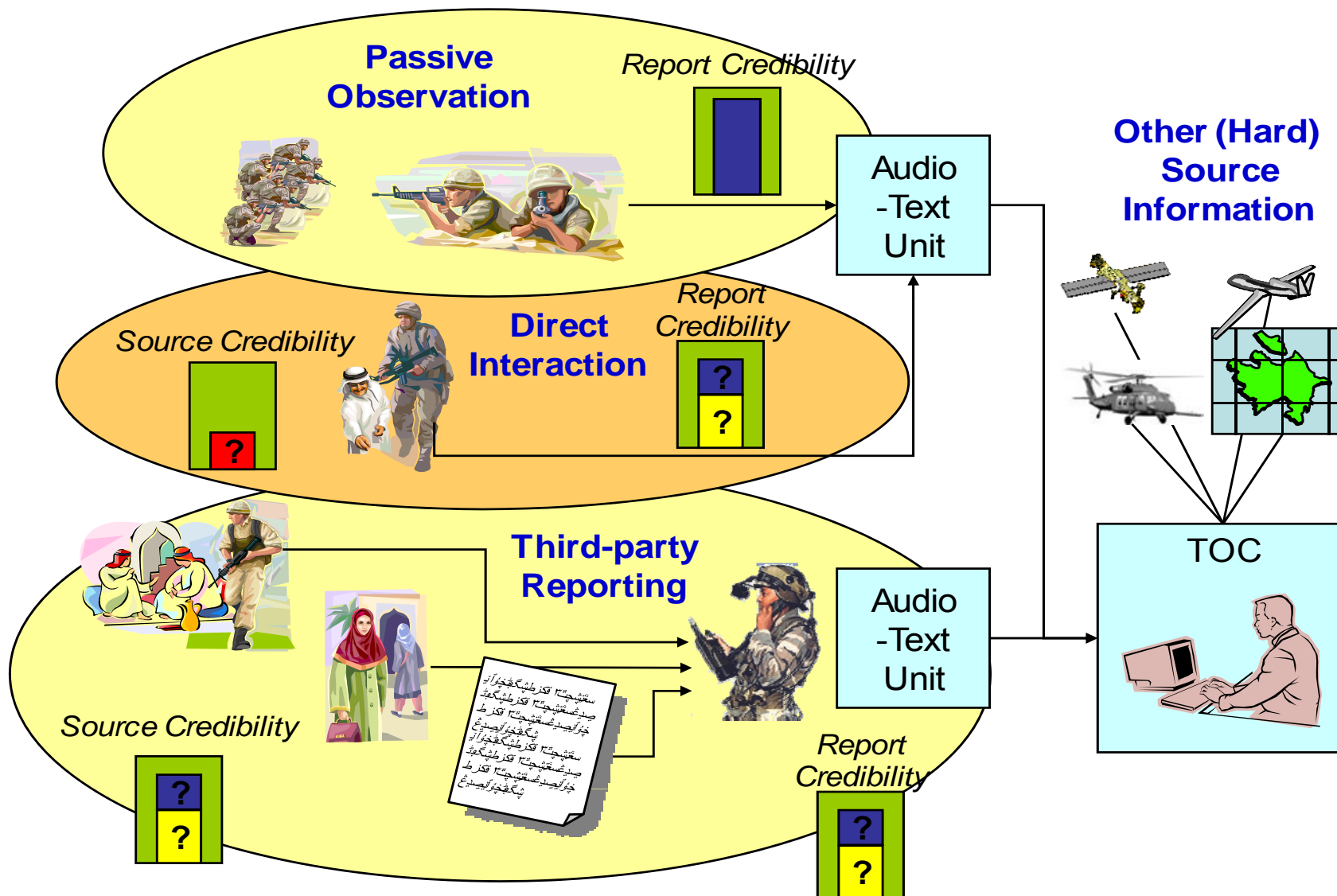
Data Characteristic	Hard	Soft	Remarks
Observation sampling rate	High	Low	Imputes requirements for adaptive, retrodiction-type processing (i.e. “Out-of-Sequence Measurement” type processing), as well as agile Temporal Reasoning
Semantic Content	Limited to specific, usually singular Entities	Can be conceptually broader than single Entities	Imputes requirements to design an automated Semantic Labeling process, coupled to a rich Domain Ontology
	Limited to Entity Attributes	Can include Judged Relationships	Requires ability to associate and infer at multiple levels of abstraction
Accuracy, Precision	Relatively high, good repeatability (Precision)	Broadly low accuracy in attributes, high at the conceptual level	Imputes requirements for robust Common Referencing and Data Association

Humans can also judge intangibles
--emotional state

Totally distinct from Hard Sensors
Philosophy: Relations not directly observable—require reasoning over properties of entities

This line of thought suggests *that relations are the result of a process* of some type of comparison, ie [Brower, 2001], “an act of reasoning”.

Categories of Human Input

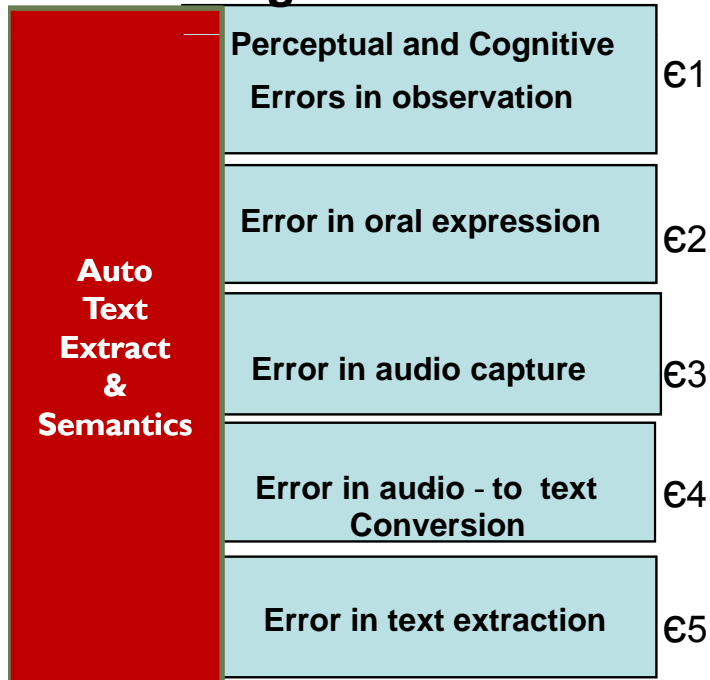


Source Characterization—very difficult to generalize

Soft Data

Real World Truth

Average Human Soldier



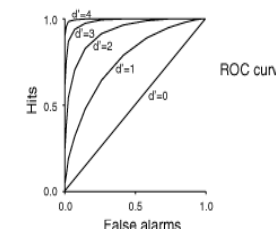
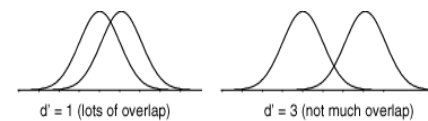
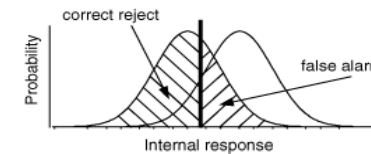
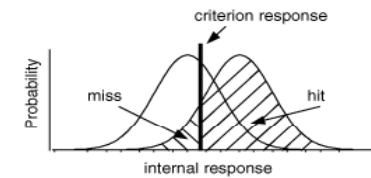
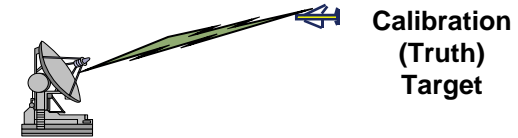
To Common Ref, Data Association

Some Obsvnl Data types qualified and Generalized

Some errors specific to obsvnl conds (need context)

Some errors will go unlabeled, unknown

Hard Data



Pd (Obs Params)

To Common Ref, Data Association

Initial Prototype Approach

