



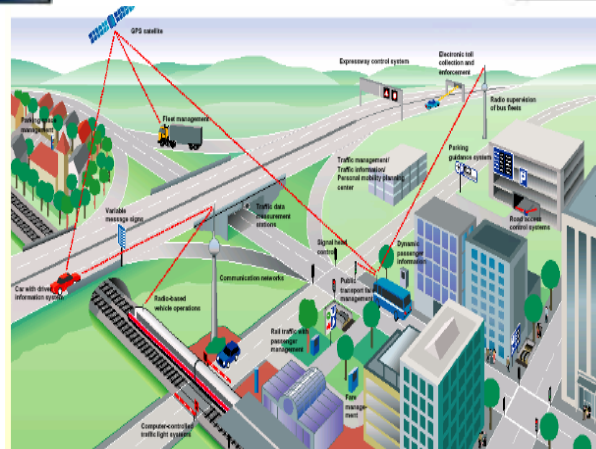
# **Overview of Systems, Control and Communications Research by Baras' Research Group**

**John S. Baras**

**Lockheed Martin Chair,  
The Institute for Systems Research and  
Electrical and Computer Engineering Department,  
University of Maryland College Park**

***July 8, 2016  
Presentation to ECE Dept. Advisory Board  
University of Maryland  
College Park, Maryland***

# Networked S-CPS: Ubiquitous Presence



**QUALCOMM**

Whiting Turner October 11, 2005

**CardioNet: Cardiac Monitoring Service -- Enabled by QUALCOMM's Wireless Network Management Services**

CardioNet

QUALCOMM

**Vehicle type: Cadillac XLR**  
Curb weight: 3,547 lbs  
Speed: 75 mph  
Acceleration: +20m/sec<sup>2</sup>  
Coefficient of friction: .65  
Driver Attention: Yes  
Etc.

**Alert Status: None**

**Alert Status: None**

**Alert Status: Multi-lane lane change**  
**Alert Status: Slowing vehicle ahead**  
**Alert Status: Passing vehicle on left**

**Vehicle type: Cadillac XLR**  
Curb weight: 3,547 lbs  
Speed: 75 mph  
Acceleration: +10m/sec<sup>2</sup>  
Coefficient of friction: .65  
Driver Attention: Yes  
Etc.

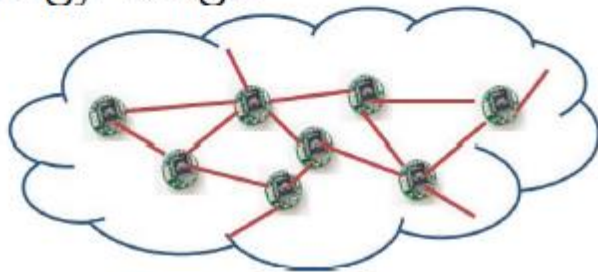
**Vehicle type: Cadillac XLR**  
Curb weight: 3,547 lbs  
Speed: 45 mph  
Acceleration: -20m/sec<sup>2</sup>  
Coefficient of friction: .65  
Driver Attention: No  
Etc.

**Alert Status: Passing Vehicle on left**

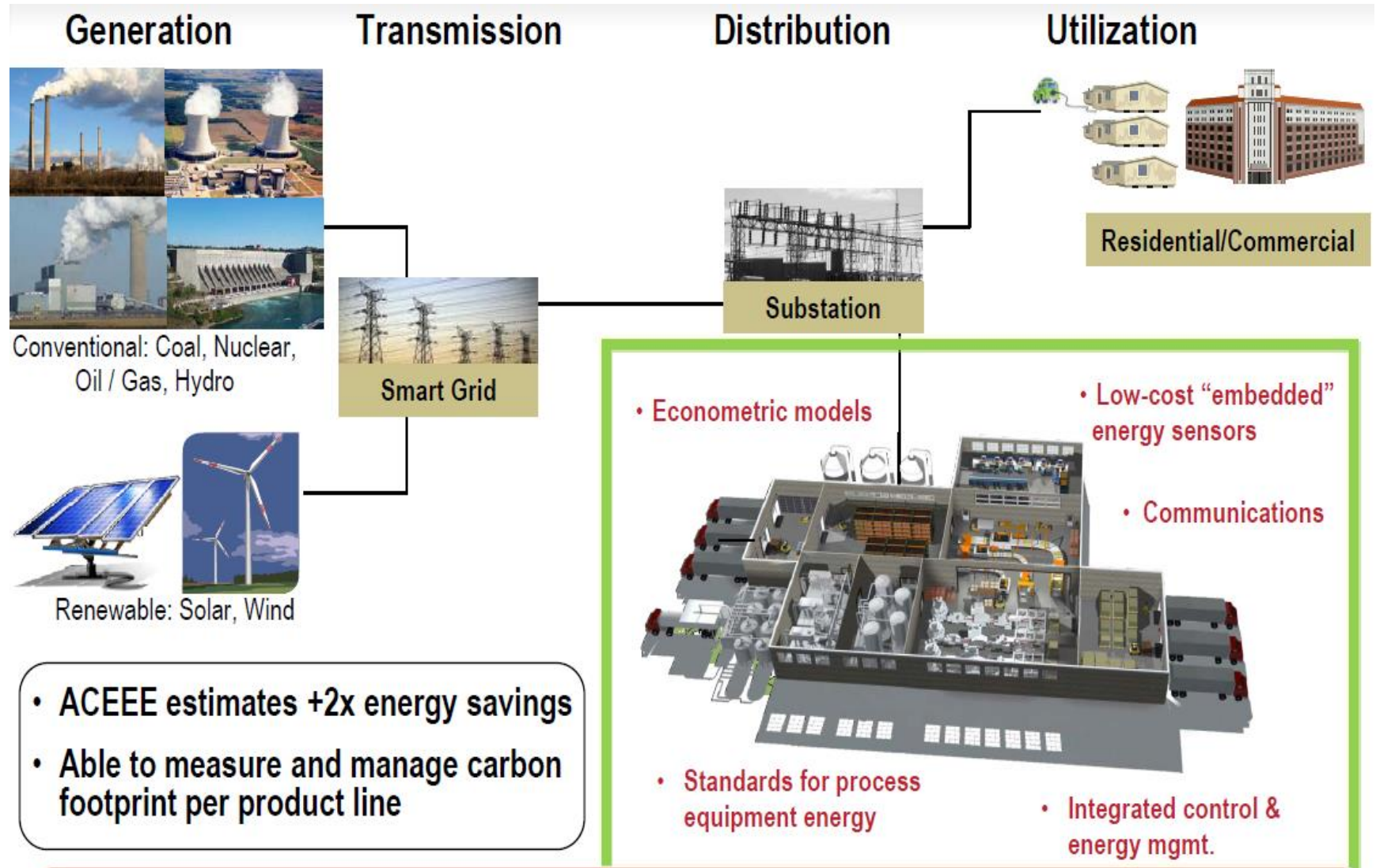
# Sensor Networks Everywhere

Wireless Sensor Networks (WSN) for infrastructure monitoring

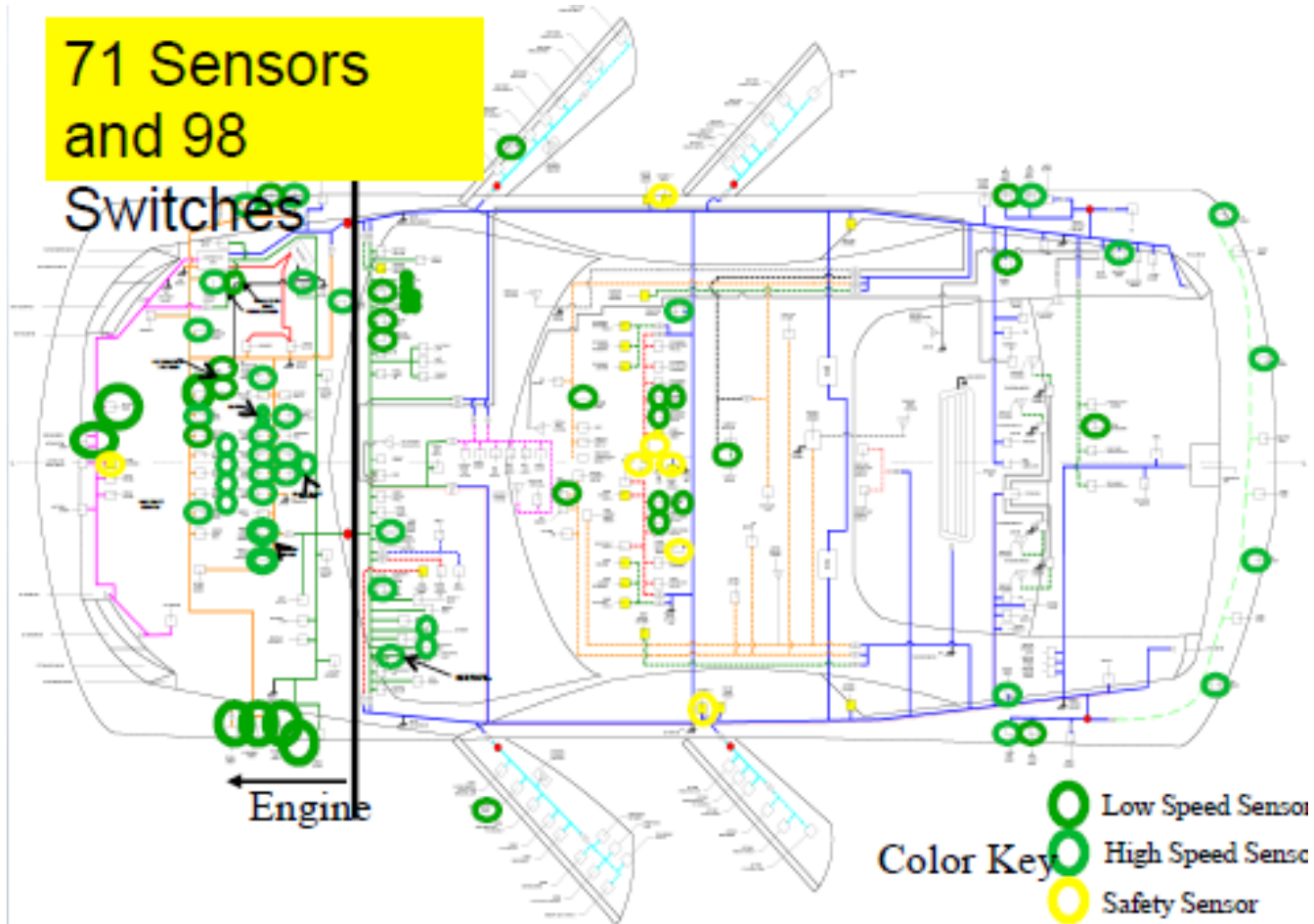
- Environmental systems
- Structural health
- Construction projects
- Energy usage



# Smart Grids in a Network Immersed World



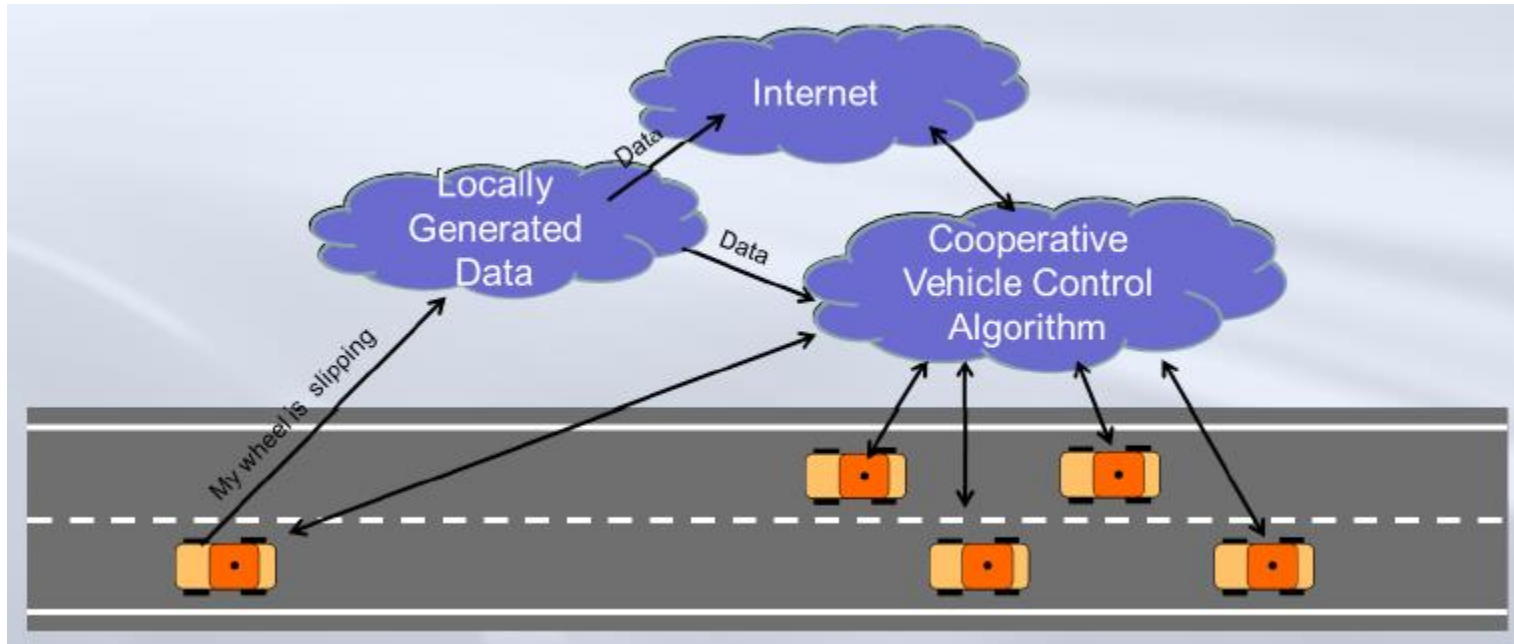
# Connected Cars: Internal



# Connected Cars: External



# Connected Cars: Cognitive and Collaborative



***Key Challenge:*** Humans

We are developing **novel frameworks** to include humans in this collaborative networked CPS environment

# A Network Immersed World: Swarms and the Cloud





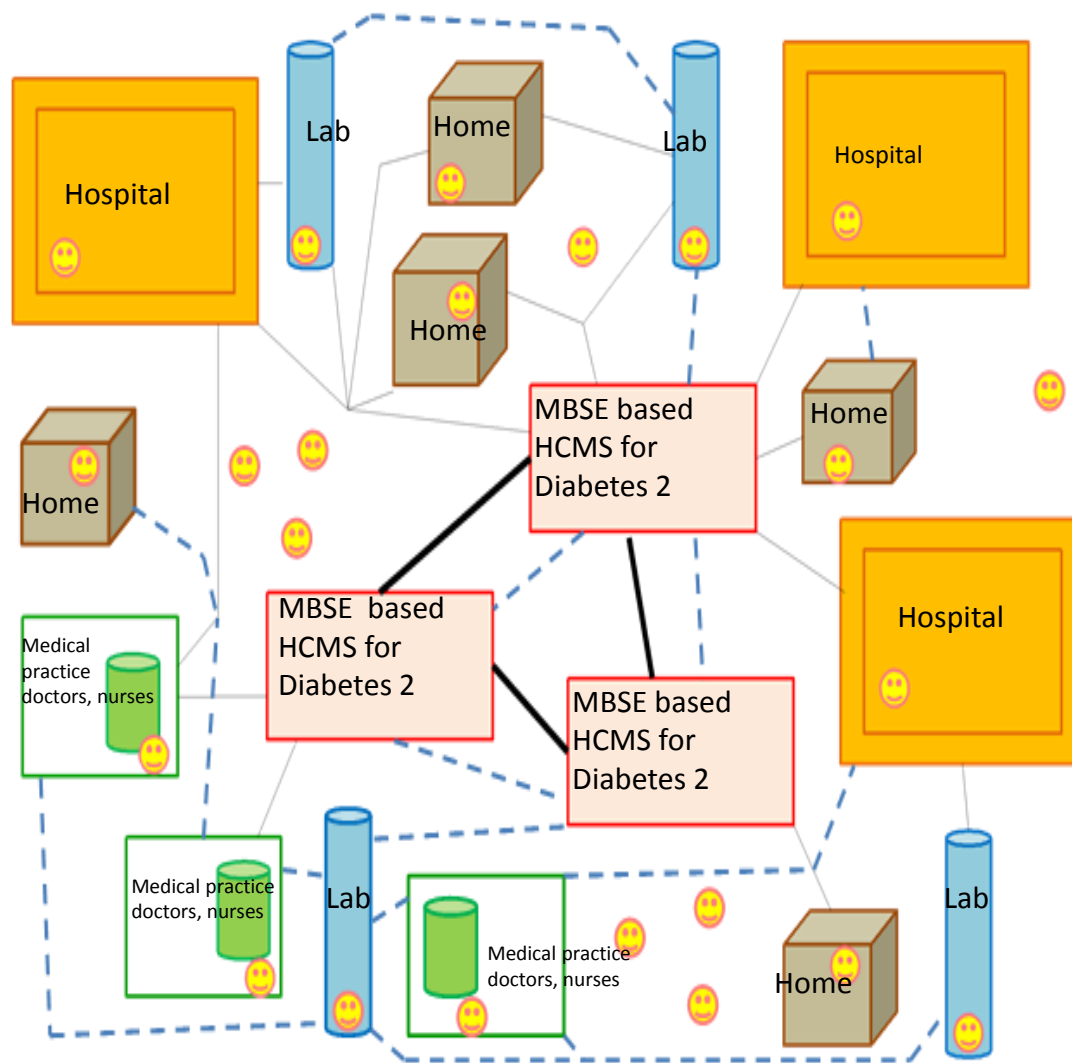
# Social Networks

- Online social network services (SNS)
  - Permeate our lives with tremendous popularity
  - Decision making via combining information from different sources
  - Benefits SNS-based applications
    - Recommender Systems
    - Online Ad targeting
- Trust relationships in SNS
  - People put different levels of trust on others in SNS
  - Important in decision making
    - People tend to accept suggestions from those they trust more



***Our work: Semiring-Based Trust Evaluation for Information Fusion in Social Network Services***

# MBSE based HCMS for Diabetes II and its functional connectivity



# Framework for MBSE: Key Challenges Addressed

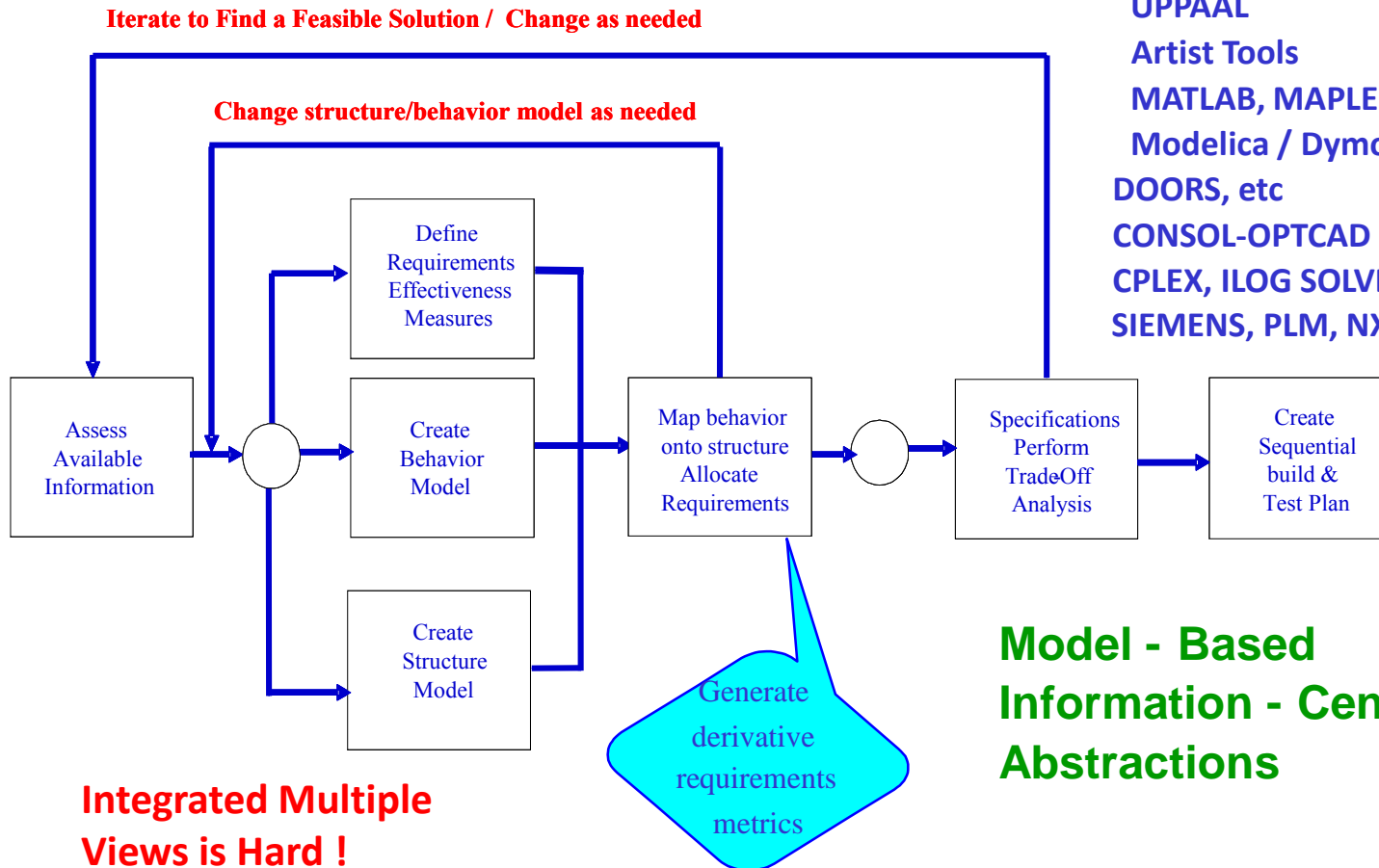
- Methodology to develop integrated modeling hubs (IMH) for CPS – multi-physics and cyber
- Methodology to link IMHs with design space exploration via multi-criteria tradeoff methods and tools
- Linkage to component databases
- Working on the last remaining challenge: requirements management
- Developed new methods and tools to handle complexity in design space exploration

# MODEL-BASED SYSTEMS ENGINEERING

## COMPONENTS -- ARCHITECTURE

**Integrated System Synthesis Tools & Environments missing**

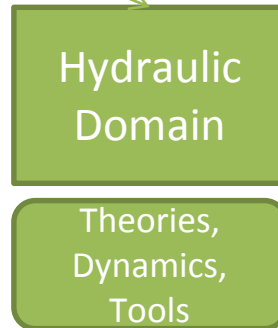
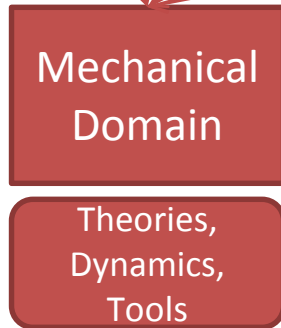
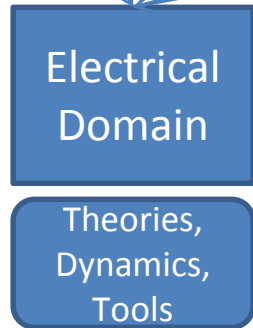
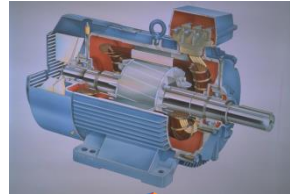
Model - based  
 UML - SysML - GME - eMFLON  
 Rapsody  
 UPPAAL  
 Artist Tools  
 MATLAB, MAPLE  
 Modelica / Dymola  
 DOORS, etc  
 CONSOL-OPTCAD  
 CPLEX, ILOG SOLVER,  
 SIEMENS, PLM, NX, TEAM CENTER



**Model - Based Information - Centric Abstractions**

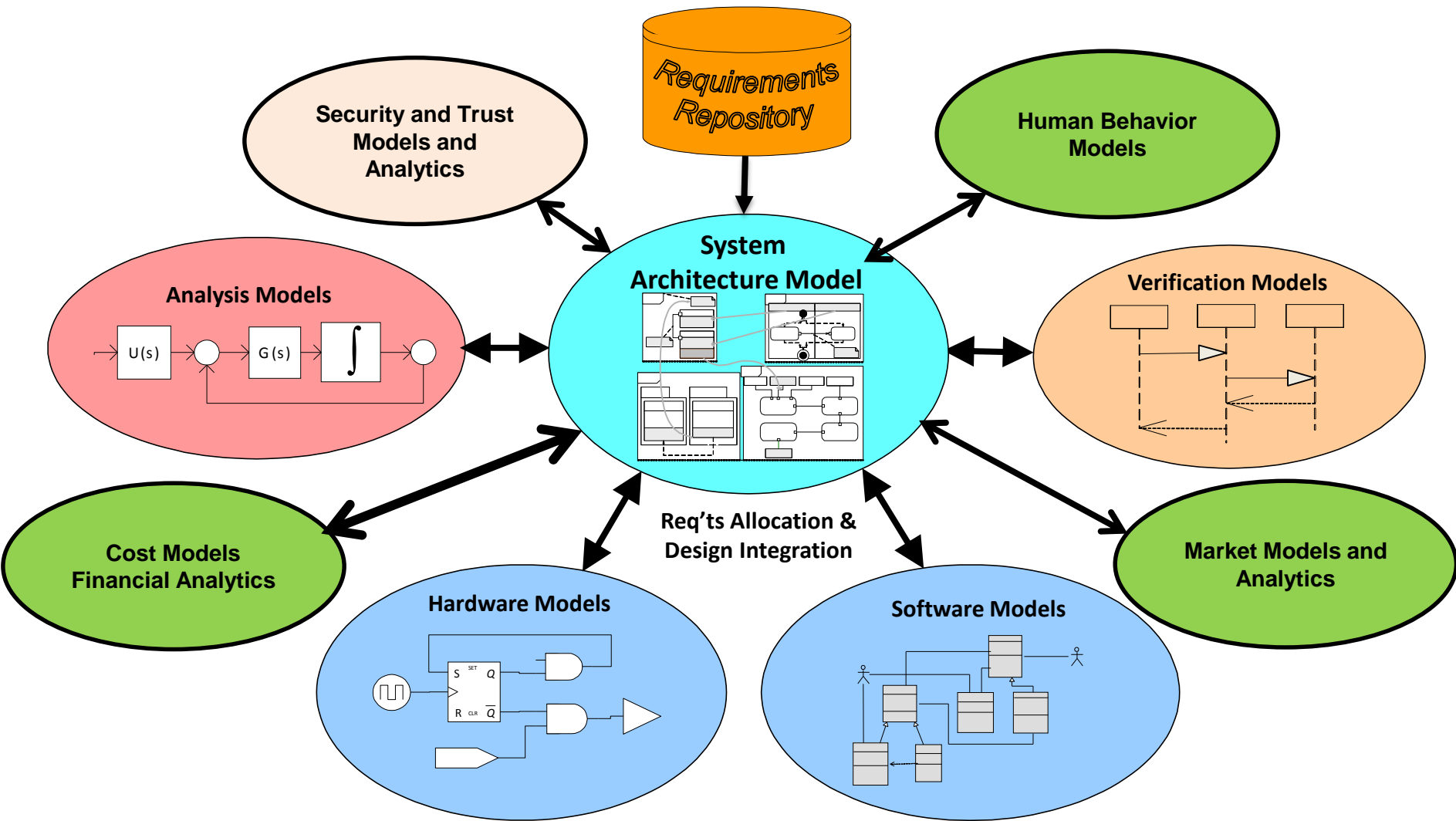
# Model Integration Challenge: Physics

Heterogeneity of Physics



**Physical components are involved in multiple physical interactions (multi-physics)**  
**Challenge: How to compose multi-models for heterogeneous physical components**

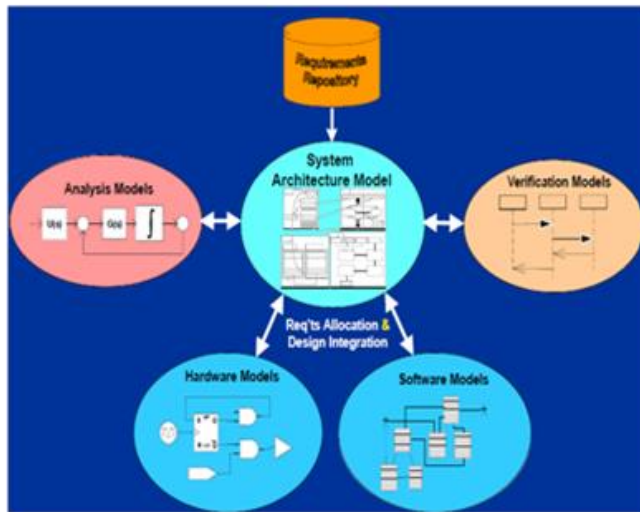
# Using *System Architecture Model* as an Integration Framework



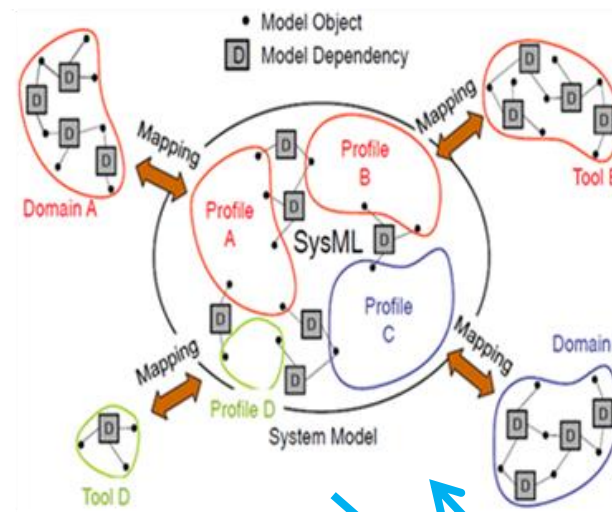
## The Challenge & Need:

Develop scalable holistic methods, models and tools for enterprise level system engineering

Multi-domain Model Integration via System Architecture Model (SysML)



System Modeling Transformations

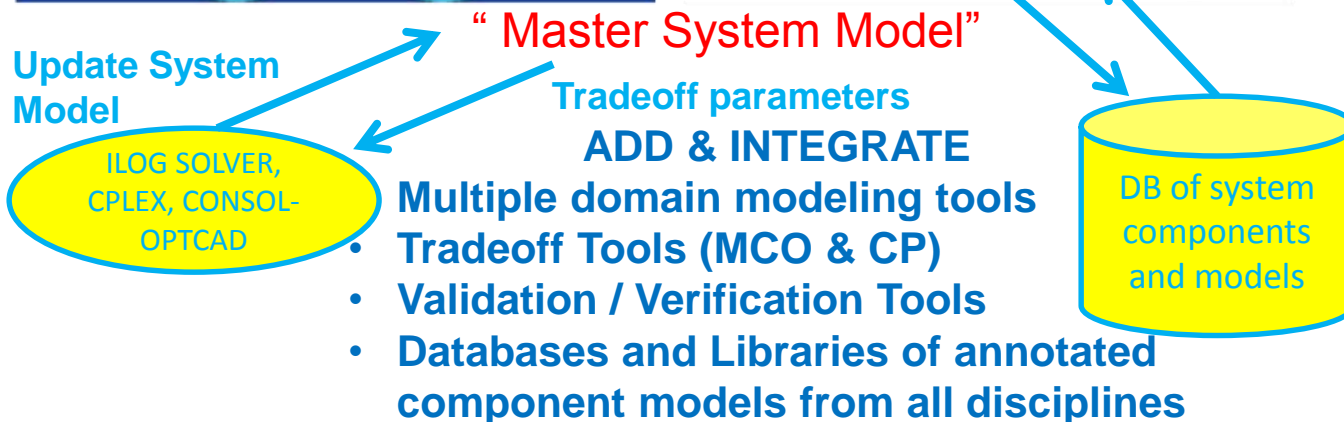


## BENEFITS

- Broader Exploration of the design space
- Modularity, re-use
- Increased flexibility, adaptability, agility
- Engineering tools allowing conceptual design, leading to full product models and easy modifications
- Automated validation/verification

## APPLICATIONS

- Avionics
- Automotive
- Robotics
- Smart Buildings
- Power Grid
- Health care
- Telecomm and WSN
- Smart PDAs
- Smart Manufacturing

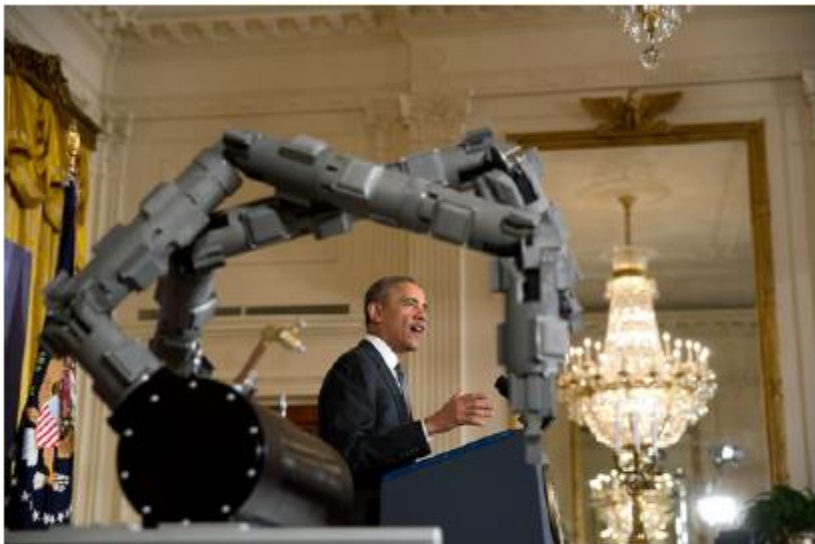


# Digital Manufacturing Design Innovation Institute (DMDII)



- Announced February 25, 2014, 2014 by President Obama

<http://www.whitehouse.gov/the-press-office/2014/02/25/president-obama-announces-two-new-public-private-manufacturing-innovatio>



President Barack Obama delivers remarks announcing two new public-private Manufacturing Innovation Institutes, and launches the first of four new Manufacturing Innovation Institute Competitions, in the East Room of the White House, Feb. 25, 2014. (Official White House Photo by Lawrence Jackson)

- Headquartered in Chicago, Illinois
  - Academic-Industry-Government “Mega Project” \$320M co-funding, 5 years
  - **Goal:** Revitalize manufacturing along the lines described in this lecture
- “Infinite number of virtual factories and an open-source manufacturing platform”



# Crowdsourcing Manufacturing



- **Google's Project ARA:** Smartphones are composed of modules (of the owner's choice) assembled into metal frames
- 
- A photograph showing a collection of disassembled smartphone components, including screens, keyboards, and camera modules, arranged on a light-colored surface.
- **Ubuntu Edge Project:** crowdsourcing the most radical smartphone yet “Why not look for the best upcoming tech and throw it together to stay ahead of the competition?”
  - **Crowdsourcing** the development and manufacturing of **small unmanned aerial vehicles**

# “Democratizing” Manufacturing

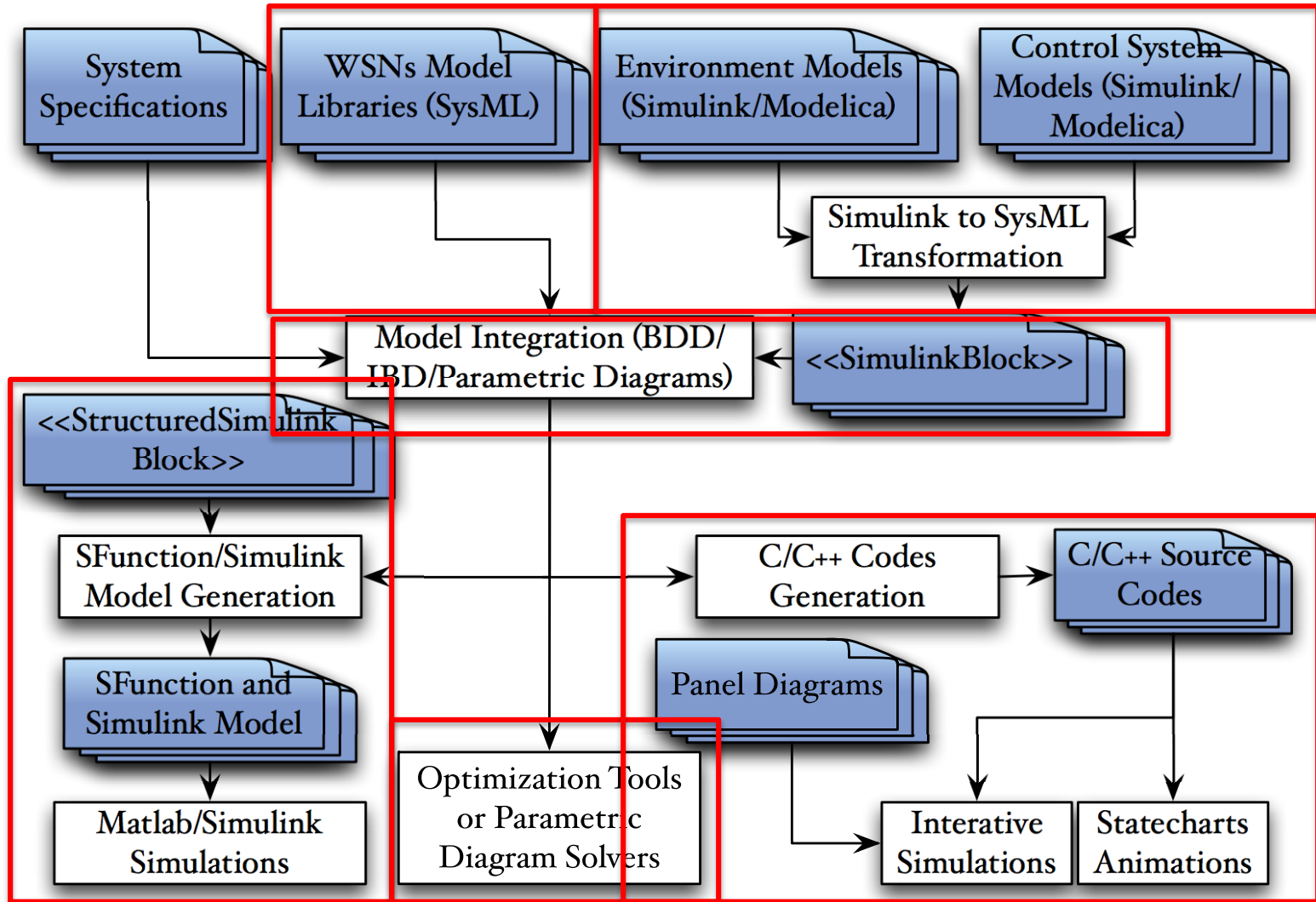


- **Goal:** Transforming more ordinary people to “makers” of products and services
- Helping small and medium size companies to manufacture products and services – **bridge the “gap”** from innovation, prototyping, to manufacturing

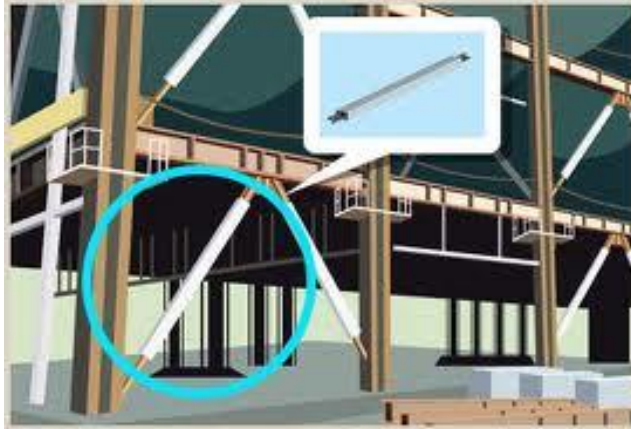


- General Electric (GE) opens manufacturing fab lab to spark ideas and participation in manufacturing through making
- Several companies have also opened up similar “open” labs: Ford etc.
- Several regional manufacturing centers (industry-university-government) are being established in various regions of USA
- “Industrial Internet” (USA) and “Industrie 4.0” (GE-EU) arrive

# MBSE for Sensor Networks



# CPS Architecture: Buildings



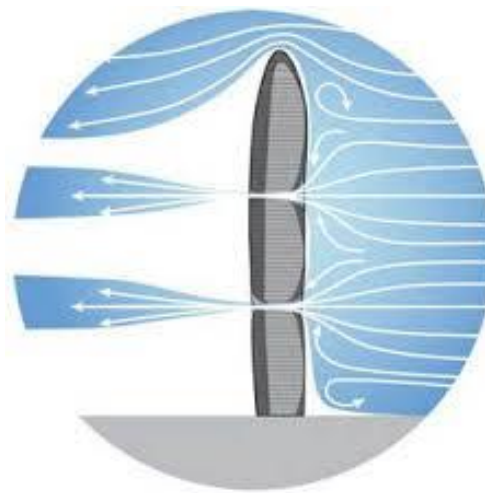
**Architecture for  
earthquake resistance**

Add computer controlled  
sensors, shock absorbers,  
material properties  
**CPS architecture?**

**Architecture for energy efficiency**



**Pearl River Tower Complex, Guangzhou**

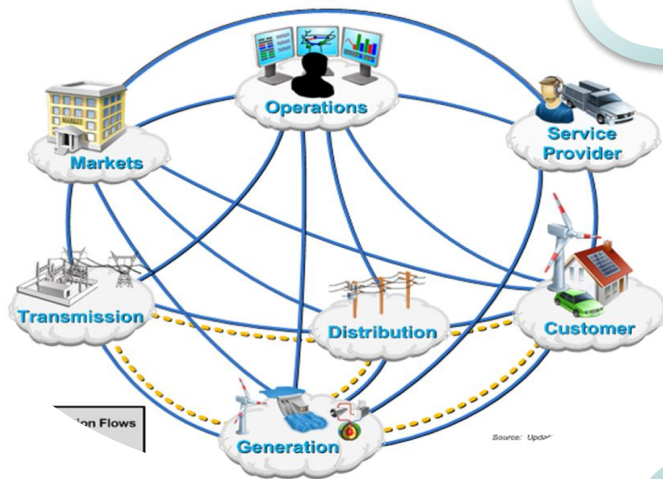
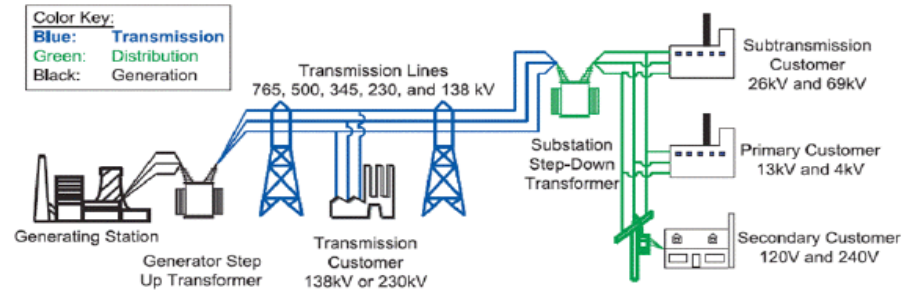


Add computer  
controlled  
sensing,  
HVAC, etc.

**CPS architecture?**

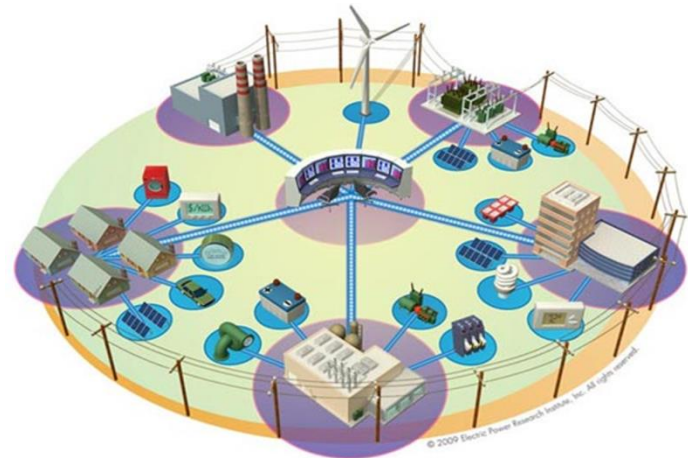
# Smart Grid - Microgrids Architecture

**Grid 1.0  
Legacy Grid**



**Grid 2.0  
Smart Grid**

**Grid 3.0  
Future Grid**



# CPS Architecture: Materials-Geometry-Controls

The 787 Dreamliner delivers:

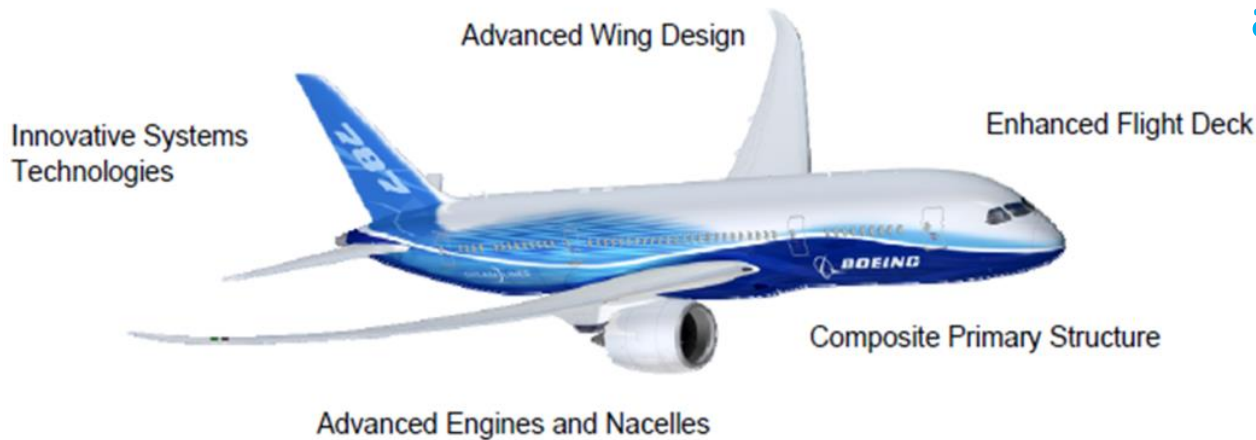
\*Relative to the 767

20%\* reduction in fuel and CO<sub>2</sub>

28% below 2008 industry limits for NO<sub>x</sub>

60%\* smaller noise foot print

## Architecture Logics, their Representation and Integration



*Composite wing – new control algorithms*  
*All-electric platform – new aircraft VMS*



*Smart suit – improve physical  
endurance & energy harvesting*

# Collaborative Autonomy



# Approach: Four Pillars

***The cognitive dialogue*** – a new architecture and formalism for cognitive systems

***A dynamic attention mechanism*** that works through a combination of signal processing and symbolic processing of prior knowledge

***The manipulation grammar*** and its associated *parser*

***A three-layer architecture*** involving dynamically interacting multi-graphs and heterogeneous internal world models



# Approach (cont.)

## Key problems:

Robots must make sense of cluttered audio-visual environments to execute autonomously and collaboratively tasks

Find and identify objects, tools, actions, based on multi-sensory input and prior knowledge

Represent and store prior knowledge

Search scene and knowledge in an efficient and organized manner

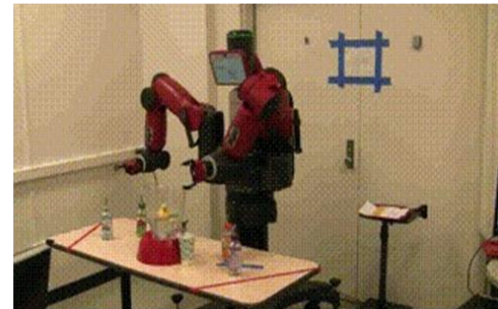
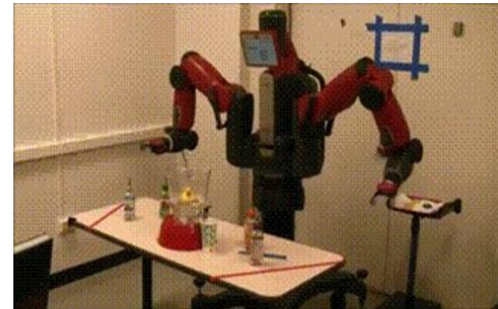
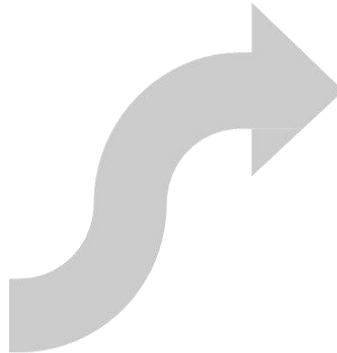
Humans utilize an elaborate attention system – need something similar, multimodal and adaptive

Need to learn, reason and communicate about objects, tools, actions

**Key principle of our approach: Task-driven integration of perception, control and language**

**Also essential for human-robot collaboration**

# Focus: Manipulation Actions



Manipulation actions



# Computer Vision

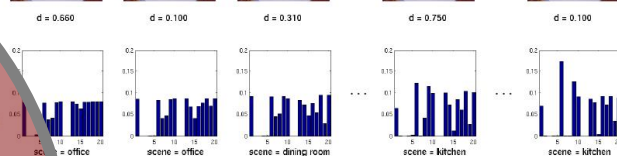
Expected Object  $O_k$ :

$O_1$  wall    $O_2$  person    $O_3$  books   ...    $O_{10}$  sink   ...    $O_{20}$  toilet   ...

Expected Location  $L_k$ :



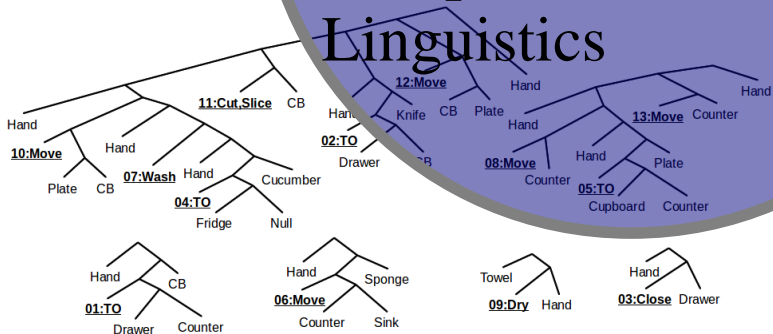
Sensory Module's Response  $(d_k, l_k)$



Reasoning Module's Belief  $P(S|d_{1:k}, l_{1:k})$  and  $S^*$

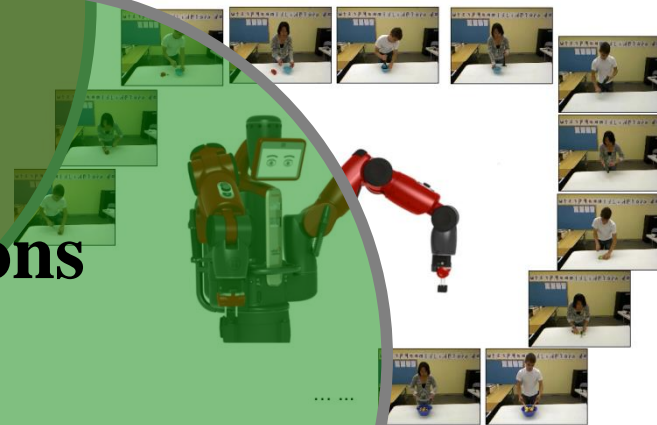
# Manipulation Actions

# Computational Linguistics

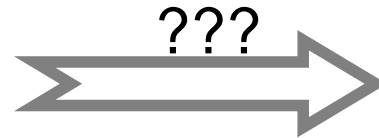


#Event	Action	Subject	Agent	From/at	To
01	Take Out	Hand	CB	Drawer	Counter
02	Take Out	Hand	Knife	Drawer	CB
03	Close	Hand	Drawer	Null	Null
04	Take Out	Hand	Cucumber	Fridge	Null
05	Take Out	Hand	Plate	Cupboard	Counter
06	Move	Hand	Sponge	Sink	Counter
07	Wash	Hand	Cucumber	Null	Null
08	Move	Hand	Plate	Counter	Null
09	Dry	Towel	hand	Null	Null
10	Move	Hand	Cucumber	Plate	CB
11	Cut Slice	Knife	Cucumber	CB	Null
12	Move	Hand	Cucumber	CB	Plate
13	Move	Hand	Plate	Counter	Null

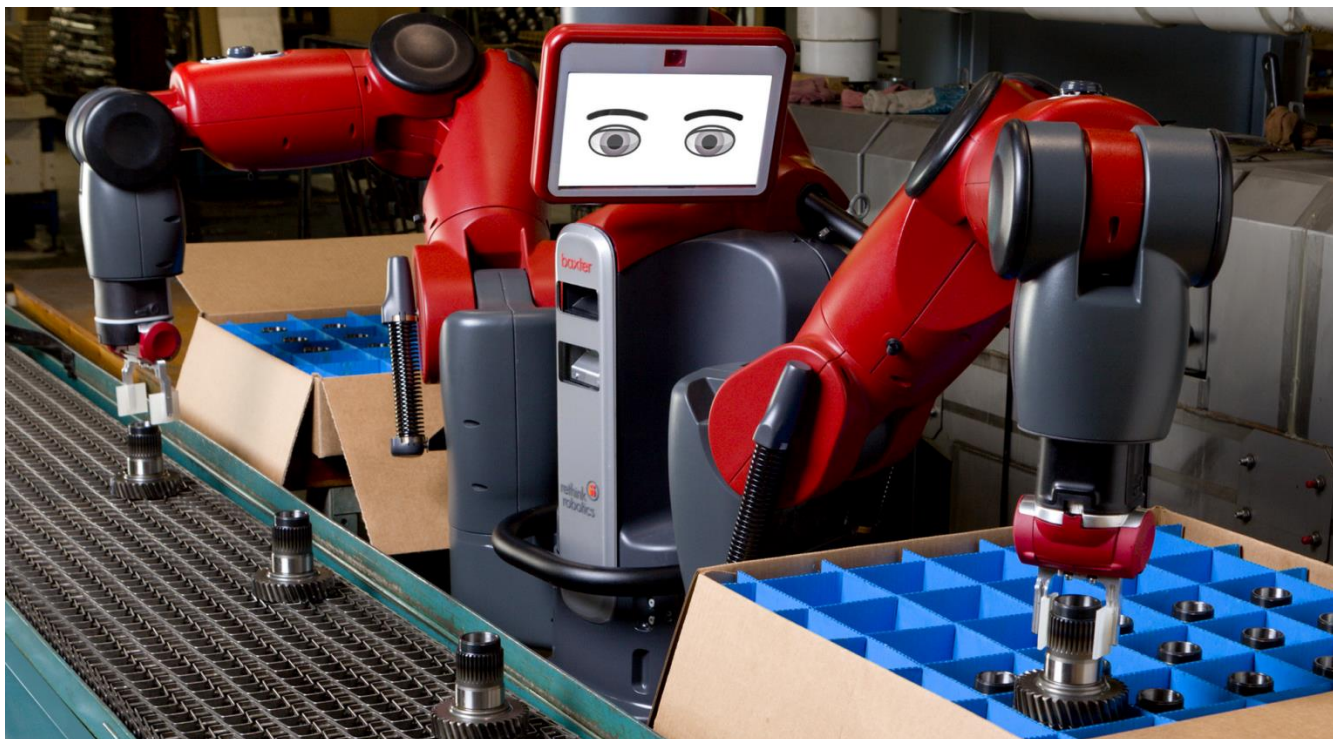
CB: Cutting Board; TO: Take Out



# Example 2: Robot Learning Manipulation Action Plans by “Watching” on-line Videos



# Example 1 : Learning Hand Movements from Markerless Demonstrations for Humanoid Tasks



- What tasks?
- How to learn tasks?
- Different situations?

# Our Approach

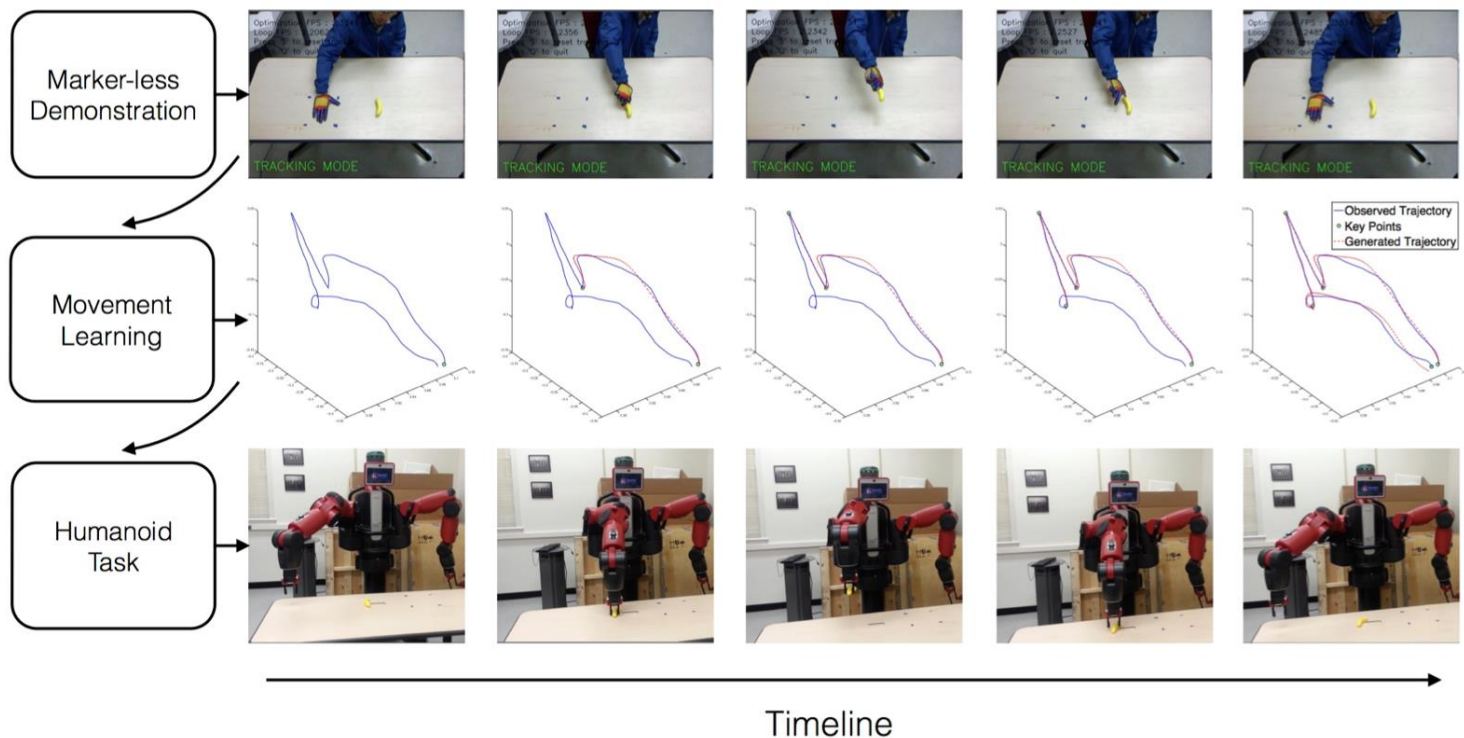


Fig. 1. Overview of learning hand movements for humanoid tasks. Placing task is an example shown here, and the drawing on hand in demonstration video is indicating hand tracker.



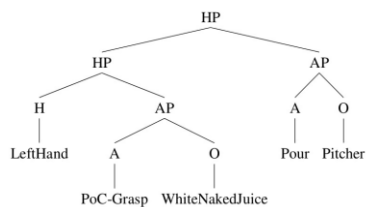
Robotic Execution



ongoing



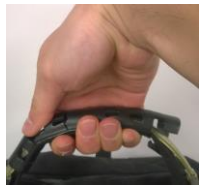
Action Sequencing



Juice	Pour	Pitcher
$N$		$N$
$NP$	$(AP \setminus NP) / NP$	$NP$
juice	$\lambda x. \lambda y. \text{pour}(x, y)$	pitcher
juice	$\rightarrow \text{mixed}(x, y)$	pitcher
	$AP \setminus NP$	
	$\lambda x. \text{pour}(x, \text{pitcher})$	
	$\rightarrow \text{mixed}(x, \text{pitcher})$	
	$AP$	
	$\text{pour}(\text{juice}, \text{pitcher})$	
	$\rightarrow \text{mixed}(\text{juice}, \text{pitcher})$	



Visual Grounding



Sensory Input



# From Logic/Semantics, to Timed Automata, to Action Execution

## Must link:

Abstract logical/semantic description of task

Timed automata representation of actions in a composable manner

Taking into consideration own kinematics constraints (embodiment)

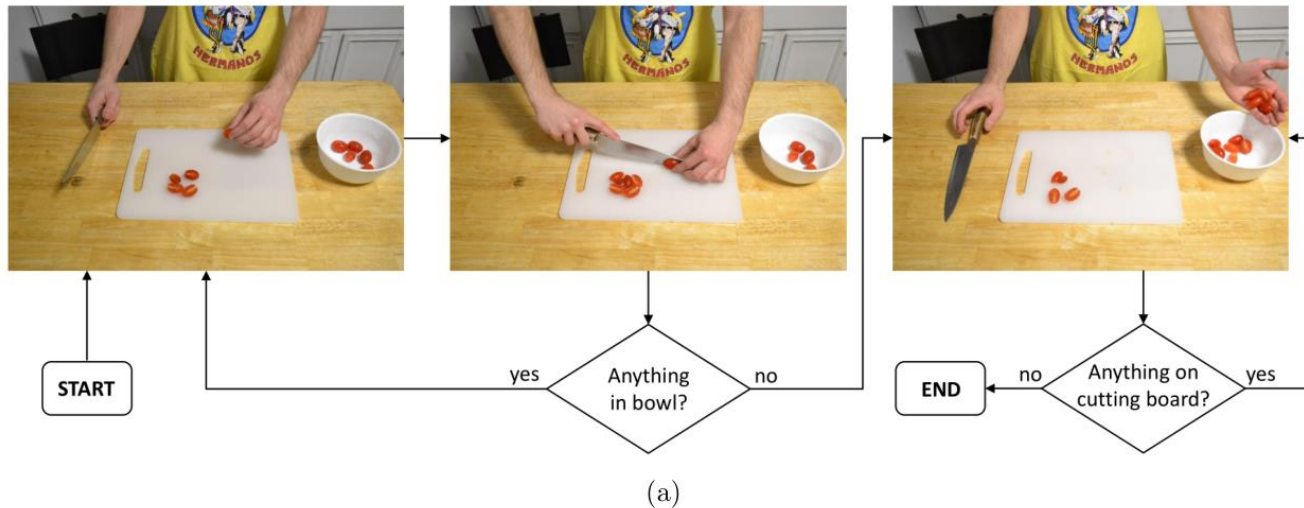
Can this be done in a principled, automatic, repeatable, verifiable manner?

**Challenges:** Role and form of learning, fast execution, role of task description and performance metrics, tolerance and uncertainty models



# Programming Language for Human Action

## A motivating example



```
for o in in(bowl) {  
  move(o, on(cutting_board))  
  cut(o, knife, 2)  
}  
// Objects have changed because of cut!  
for o in on(cutting_board) {  
  move(o, in(bowl))  
}
```

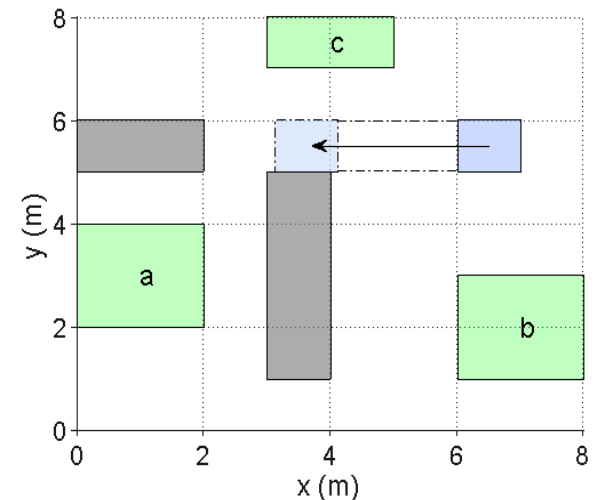
(b)

A simple action and its source code

# Motion Planning with Temporal Constraints



- Q: How to generate **trajectory/path** based on **temporal specifications** such as ordering between actions, repetition of tasks, safety of the motions?
- State of art: motion planning with temporal constraints without duration, such as Linear Temporal Logic (LTL).
- We have proposed two methods for **timed temporal logics**, such as Metric Temporal Logic(MTL) for motion planning problem:
  - An optimization based method
  - A timed-automata based method



Always visiting area a,b,c and stay there for at least 2s. Always avoiding obstacles

# Robotic Motion Planning Problem



**Given:** A dynamic workspace (environment),  
A **time constrained task** ( $\varphi$ ),  
A cost function.

**Objective:** Find the suitable control input such that the robot completes the **given task** and **minimizes** the cost function.

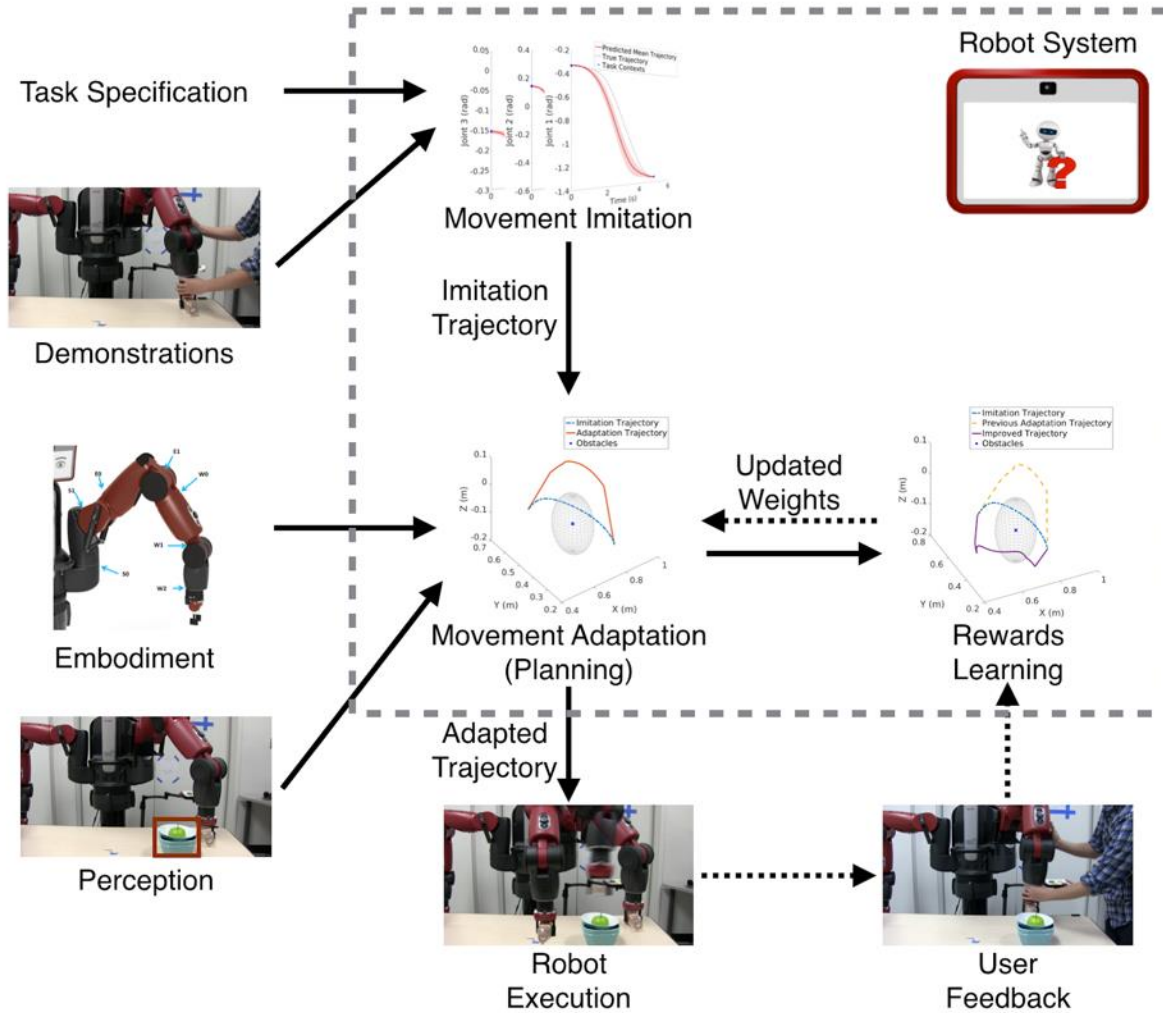
**Constraints:** Avoiding collisions with all **static and moving obstacles** in the workspace.

**Challenges/Innovations:** metric temporal logic, finite automata specs, uncertainties with mixed logical/numerical representations, automatic verification, **bridge the gap between action grammars and motion planning/controls independent of learning environment and platform execution**

# Learning to Plan Manipulation Task Execution in New Environments

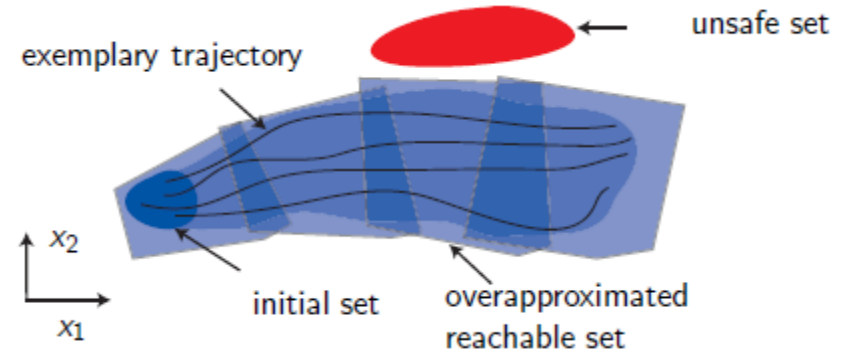
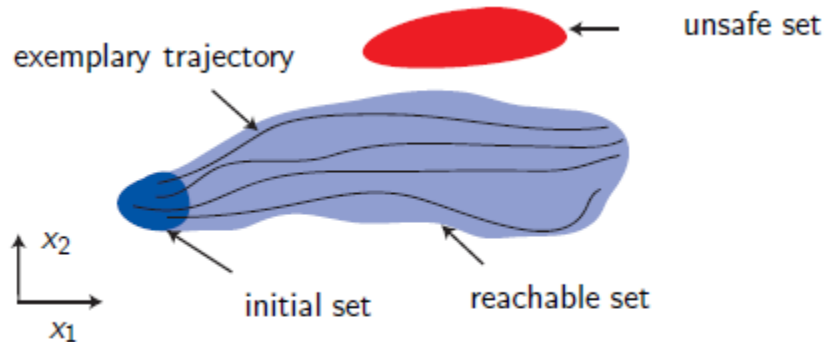
- **Learn** manipulator trajectory from demonstrations
- **Adapt** manipulator trajectory through planning with new constraints
- **Learn preferences** to adapt movement through feedback

# Proposed System

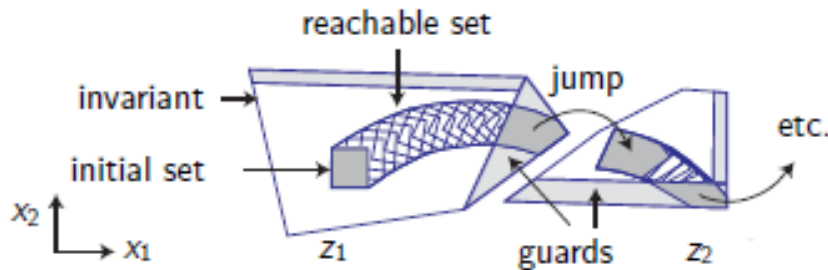


# Safety & Trust in Human-Robot Teams : Integrating Logic and Set-valued Analytics

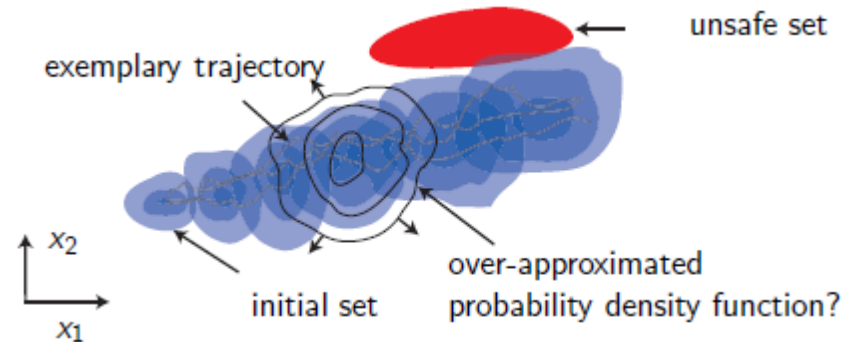
- **Space-time reachability analysis (now real time)**



## Hybrid systems



## Stochastic safety systems



- **Translate these to analytics: model checking, contracts, theorem proving, set valued -- Trust values? Metrics? Timed Languages?**
- **Roles? Role-based trust management?**

# Collision Avoidance via Reachable Sets

- We propose to use **reachable set** for collision avoidance
  - Reachable set of a dynamics is defined as set of states reachable from a **bounded initial set**, a **control set** and a **disturbance set**.
  - The control sets are then **synthesized collaboratively** so that the reachable sets of the UAVs have no intersection.
  - Existing studies in reachability literature exam the problem in a game theoretical setup such that other UAVs are treated as adversary. [I. Mitchell etc 2005] Commonly the collision avoidance is **collaborative**.
  - Efficient reachable set computation normally uses convex approximations such as **ellipsoids** [A. B. Kurzhansk 2000] and polytopes.

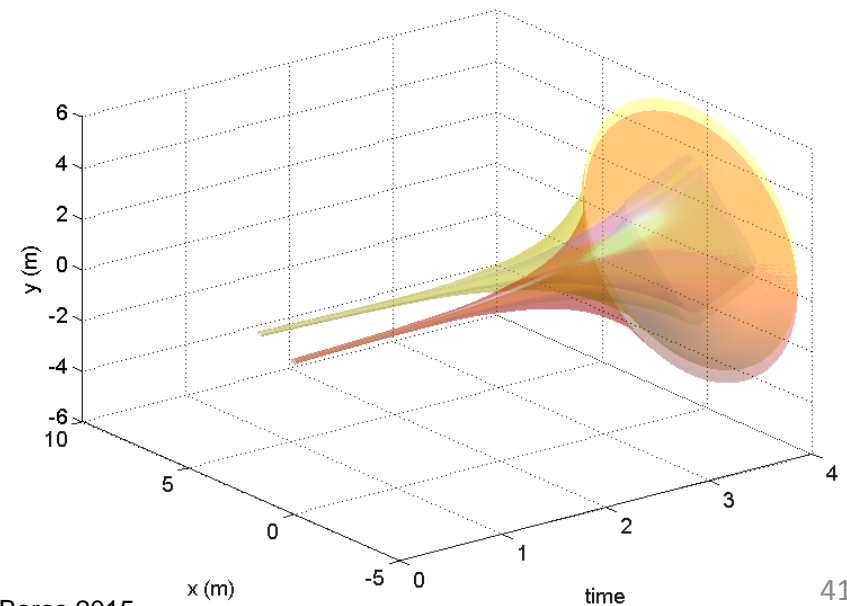
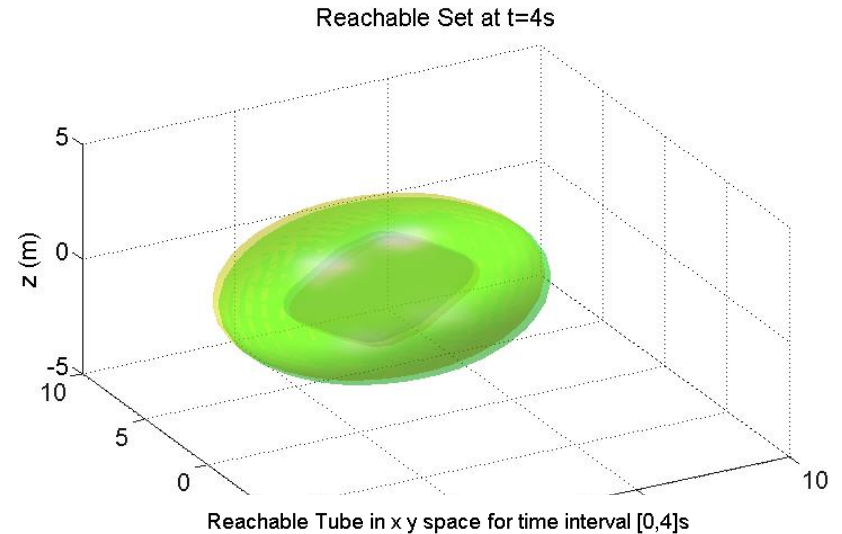
# Collision Avoidance Problem

- We seek a control set design for aircraft A and B such that by using **more constrained control sets** than their initial ones, **collision avoidance between sets** are guaranteed.
- Decompose the problem to two parts
  - First we seek a tighter control constraint set for aircraft B such that the reachable set **are far away from that of aircraft A** but at the same time the control set of B is **still large enough**.
  - At the second phase we seek a safe reachable tube for **aircraft A** so that the reachable tube will **be apart from the reachable tube of aircraft B** for at least the required separation.



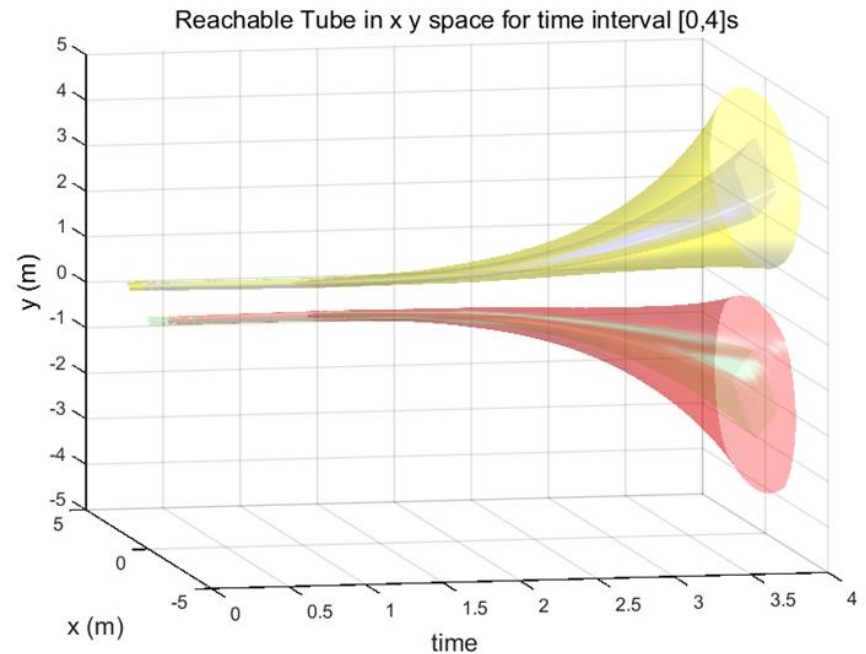
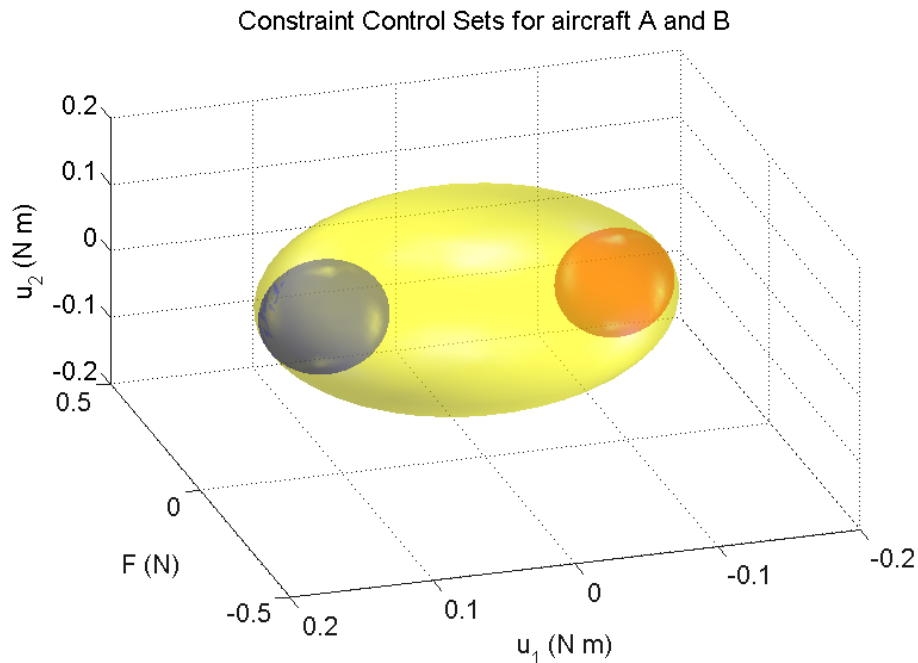
# Initial Reachable Set and Tube

- The top plot shows the reachable set of  $x$   $y$   $z$  location at time of collision. The two sets are overlapping.
- The darker colored ones in the center is the inner approximation of reachable set.
- The reachable tube of  $x$ ,  $y$  position only in the bottom plot shows same idea.



# Collision Avoidance Control Set Synthesis I

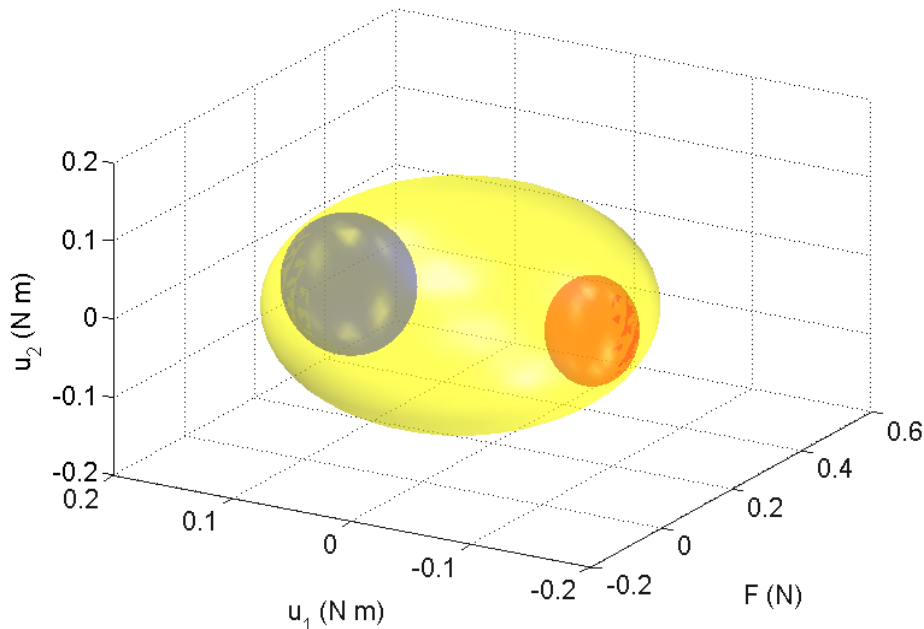
- If both UAVs have same priority, by tuning the scalarization factor, one can obtain control sets of similar size on the left.
- The reachable tube can be then visualized as the right figure.



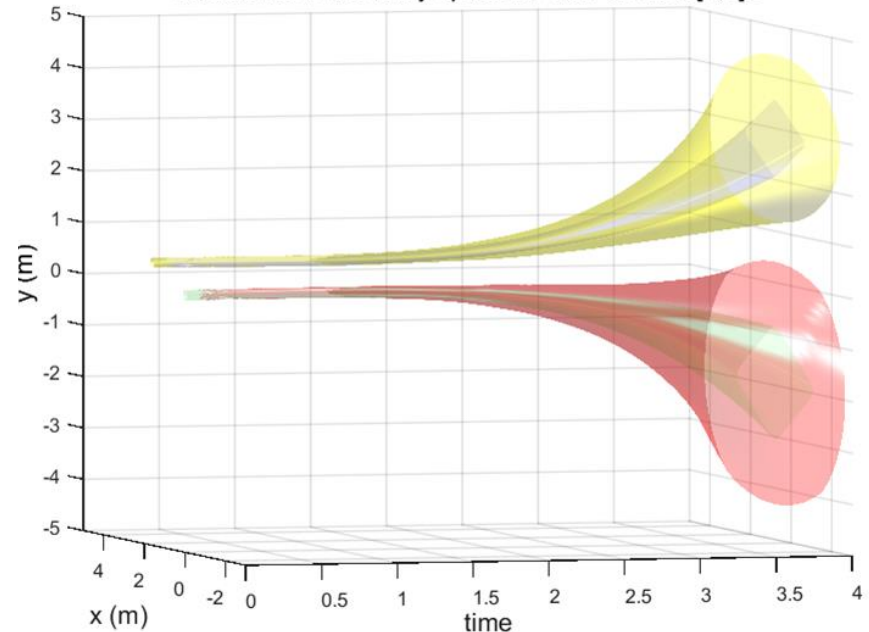
# Collision Avoidance Control Set Synthesis II

- If one of the UAVs has higher priority, it can have larger control set, so that it maintains more freedom comparing to the other one.
- The reachable tube can be then visualized as the one on the right

Constraint Control Sets for aircraft A and B

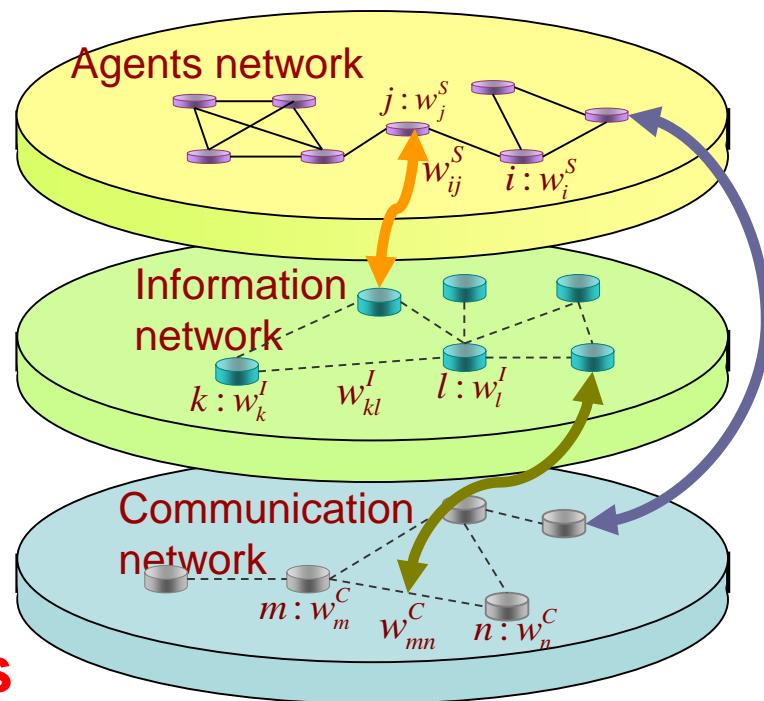


Reachable Tube in x y space for time interval [0,4]s



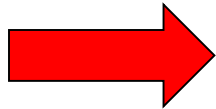
# Multiple Interacting Dynamic Multigraphs

- Multiple Interacting Multigraphs
  - *Nodes*: agents, individuals, groups
  - Directed graphs
  - *Links*: ties, relationships
  - *Weights on links*: value, strength
  - *Weights on nodes*: importance
- **Real-life problems: Dynamic time varying graphs, relations, weights**
- **Effects of connectivity topologies**
- **Taxonomies of multigraphs involved -- performance**
  - *Collaboration multigraph*: who collaborates with whom and when.
  - *Communication multigraph*: who communicates with whom and when
- Need for **different probability models**
- **Future: Dynamic goal oriented planning, re-planning**



- The nodes **gain** from collaborating
- But collaboration has **costs** (e.g. **communications**)
- **Trade-off: gain from collaboration vs cost of collaboration**

Vector metrics involved typically

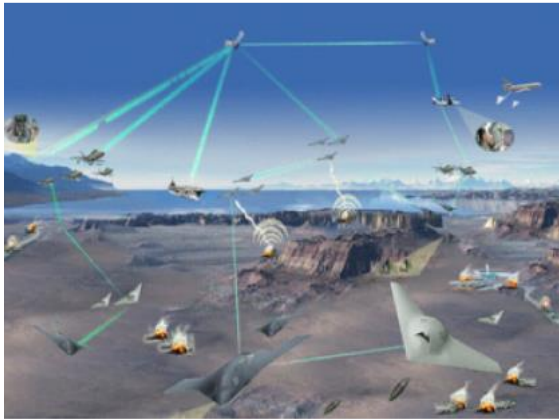


## Constrained Coalitional Games

- **Example 1: Network Formation** -- Effects on Topology
- **Example 2: Collaborative robotics, communications**
- **Example 3: Web-based social networks and services**
- **Example 4: Groups of cancer tumor or virus cells** • • •
- **Future:**

Introduce complex behavioral models, multiple-sensory perception, Language development and efficient communications, learning from collaboration, motifs, storing and recalling patterns, multiple internal models, complexity, trust in inference and control, composite trust

## The Challenge & Need: DoD Collaborative Autonomous Networked Human-Machine Systems



Heterogeneous, dynamic, multi - scale, rapid technology changes, rapid threat changes

Fig.1: MBSE process elements  
MODEL- BASED SYSTEMS ENGINEERING

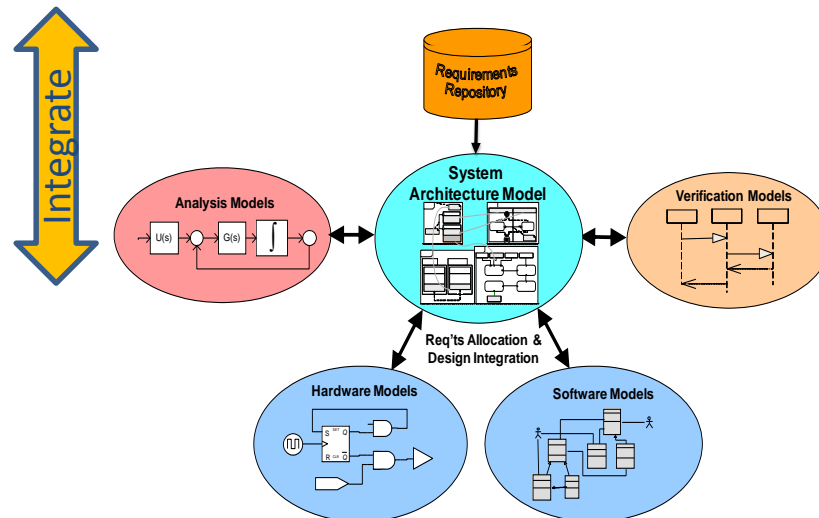
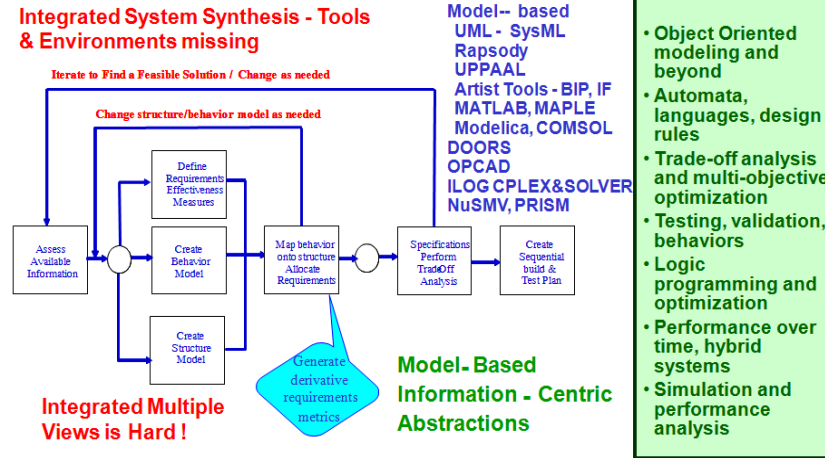


Fig. 2: Modeling and Analysis Tools Integration via **SysML** System Architecture Model

## ADD & INTEGRATE

- New modeling environments
- Network models and semantics
- Reasoning, Validation and Tradeoff Tools
- Databases and Libraries of component models from all disciplines

## BENEFITS

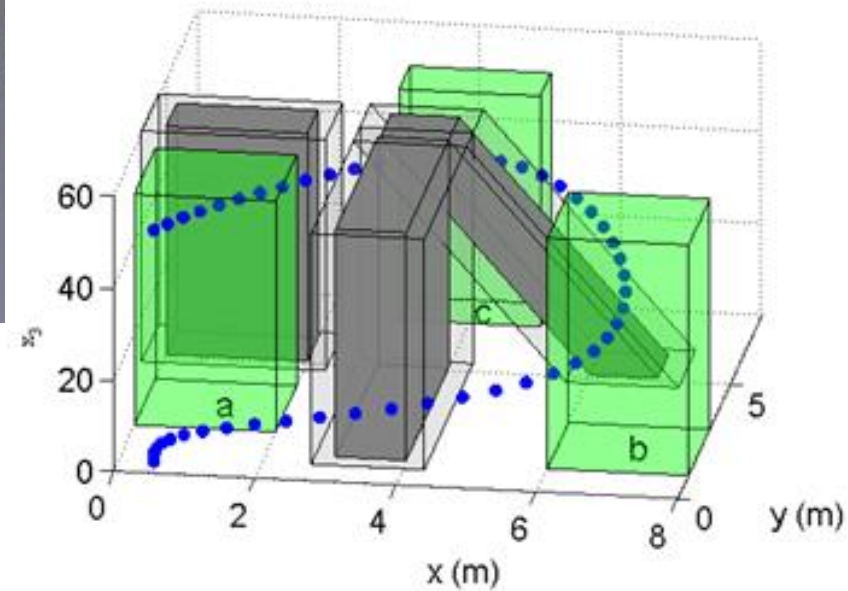
- Reduced cost and fielding time
- Modularity and re-use
- Increased agility in designing, modifying and fielding new systems

# Demo 1: Synchronized Flight of Small Unmanned Aerial Systems

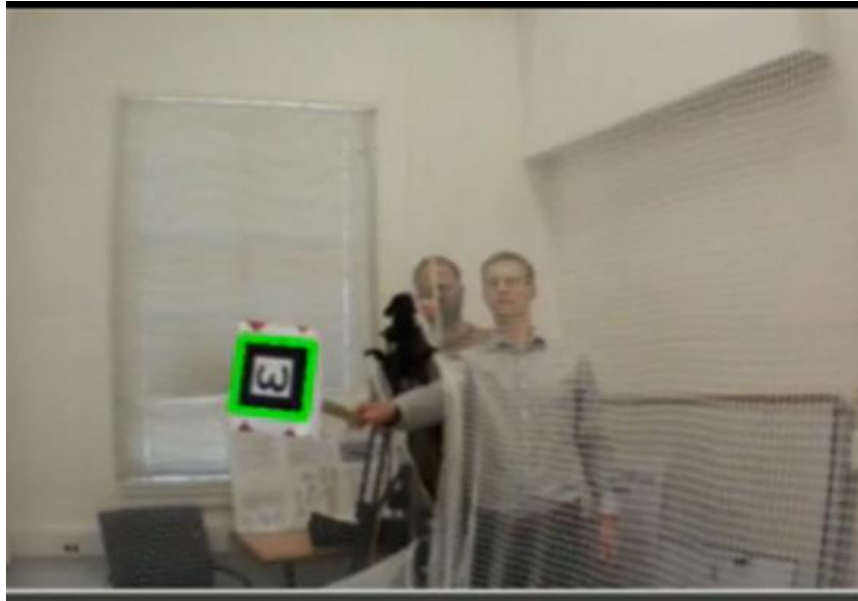


We follow **MBSE** (Model Based Systems Engineering) methods to create modular software

Our aircrafts use only basic onboard sensors and cameras, flying **without the aid** of motion tracking cameras that can be seen in many other experiments



# Demo 2: Small Unmanned Aircraft Following a Target



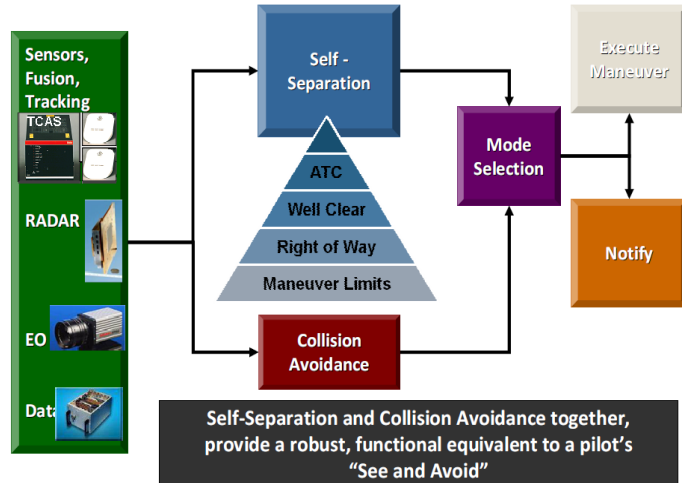
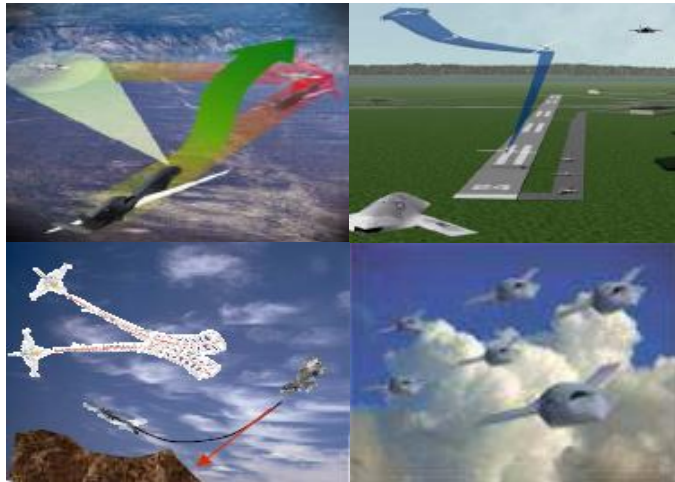
Our aircraft use a vision-based ROS package for the AR. The Drone aircraft automatically follow specific targets.

Our aircrafts use only basic onboard sensors and cameras, flying **without the aid** of motion tracking cameras that can be seen in many other experiments

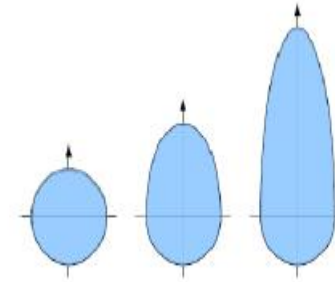




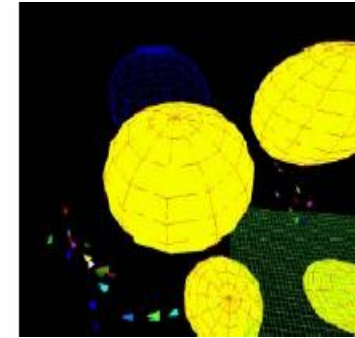
## The Challenge & Need: Cooperative Control Sense and Avoid Technology for Autonomous UAS in Dense Environments



(a)



(b)



(c)

**Fig.:** (a) Boid animation of birds in complex environments; (b) 'bubbles' of different shapes from slow to higher velocities; (c) diverse bubbles navigating obstacles to a goal

## APPROACH

- Biologically inspired control (swarms, birds)
- Control theoretic analytics
- Efficient and fast computations
- Aerodynamics
- Model predictive control of hybrid automata (switched dynamical systems) including temporal logic
- Formal safety verification
- Integrated modeling, simulation, synthesis, operations tool-suite for collaborating UAS

## OUTCOMES / APPLICATIONS

- Dynamic bubble shapes for varying safety constraints
- Guaranteed safe operation of UAS teams
- Biologically inspired high performance collaborative and safe control of UAS/UAV/UGV



# Distributed Coordination of Unmanned Underwater Vehicles (Baras)

- **Motivation:**

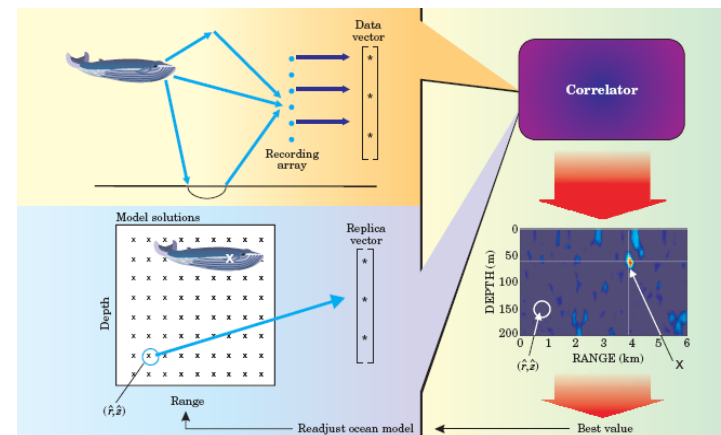
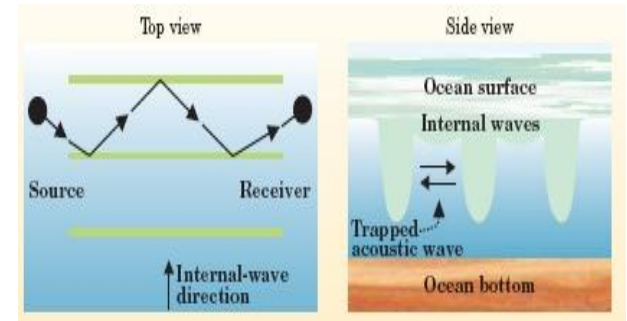
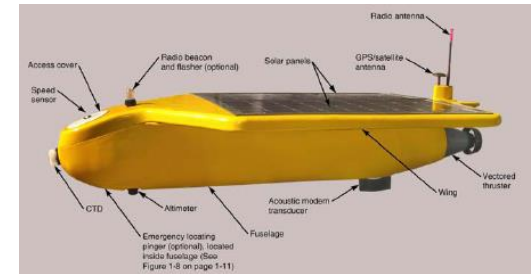
- Networks of underwater vehicles for sensing, ocean mapping and exploration, surveillance
- Cooperating heterogeneous sensing
- Hybrid acoustic and RF communications – avoid surfacing

- **Goals:**

- Adaptability to mission
- New communication schemes that explore idiosyncracies of underwater channel: multiple dynamic waveguides, trapping waves, ducts, multipath. Use predictive opportunistic comms employing on-line ocean channel predictor
- Distributed dynamic behavior-based control

- **Benefits:**

- Dynamic insertion and removal of mission elements during execution
- Sophisticated but energy efficient comms
- Longer collaborative, energy efficient missions

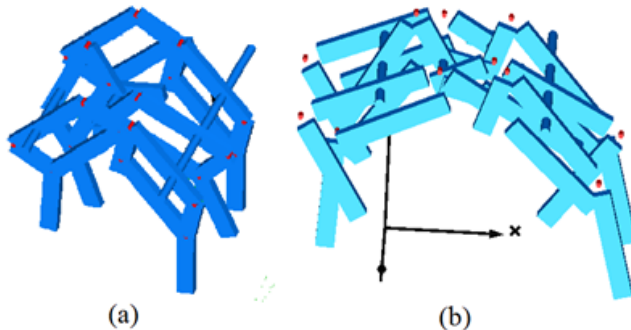


Use acoustic models to reduce comms requirements and increase efficiency



- Transcend areas of application: from space to micro robotics
- Include material selection in design
- Include energy sources, resilience, reliability, cost
- Include validation-verification and testing
- Use integrated SysML and Modelica environment
- Link it to tradeoff tools CPLEX and ILOG Solver
- Demonstrate reuse, traceability, change impact and management

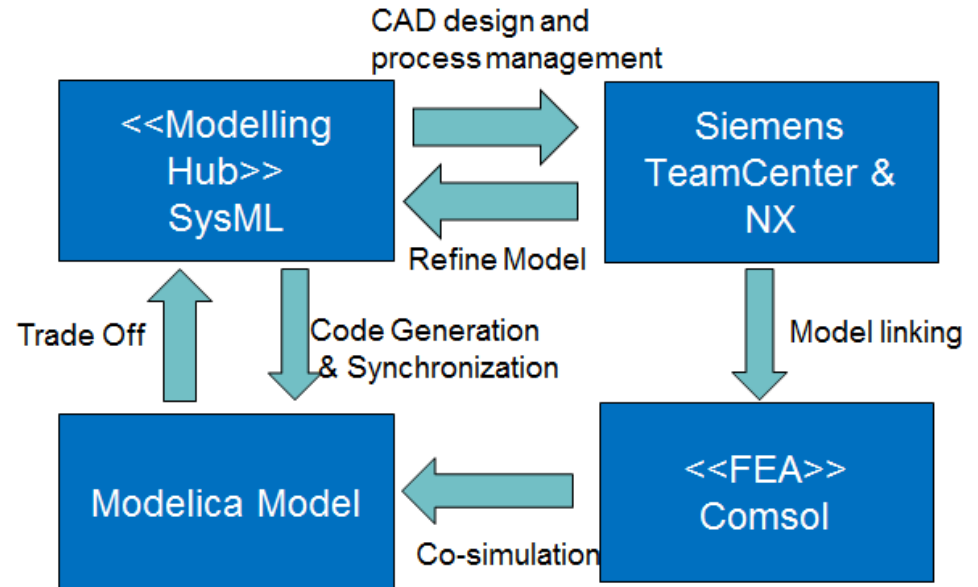
# Application to Microrobotics



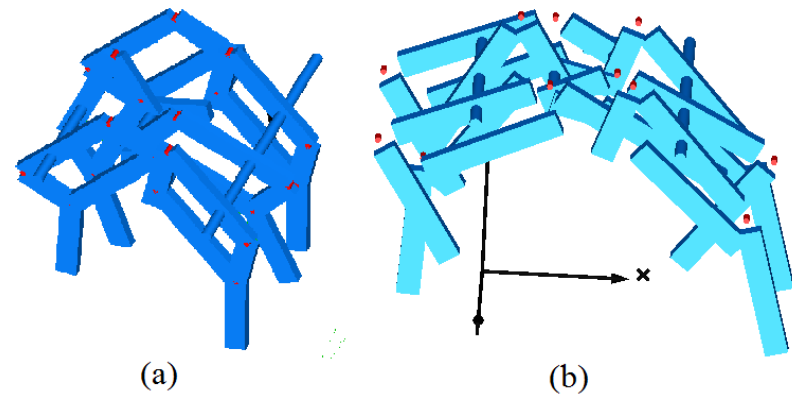
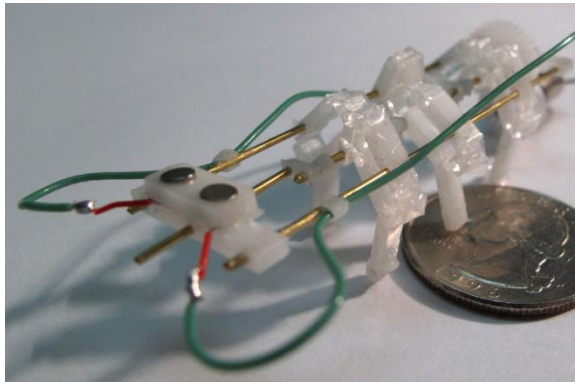
- Micro-robots design and manufacturing require **control algorithm** and **physical layer (material and geometry) co-design**.
- This insect-like robot is modeled in **Modelica** language using Differential Algebraic Equation.
- We are working on a **Model-Based Systems Engineering** approach to perform analysis, modeling and tradeoff for robotics and its **material** and **control** parameters.

## Siemens Tools Utilization

- Design and analysis CAD model at the design phase
- Guide requirement to implementation from CAD design to physical simulation



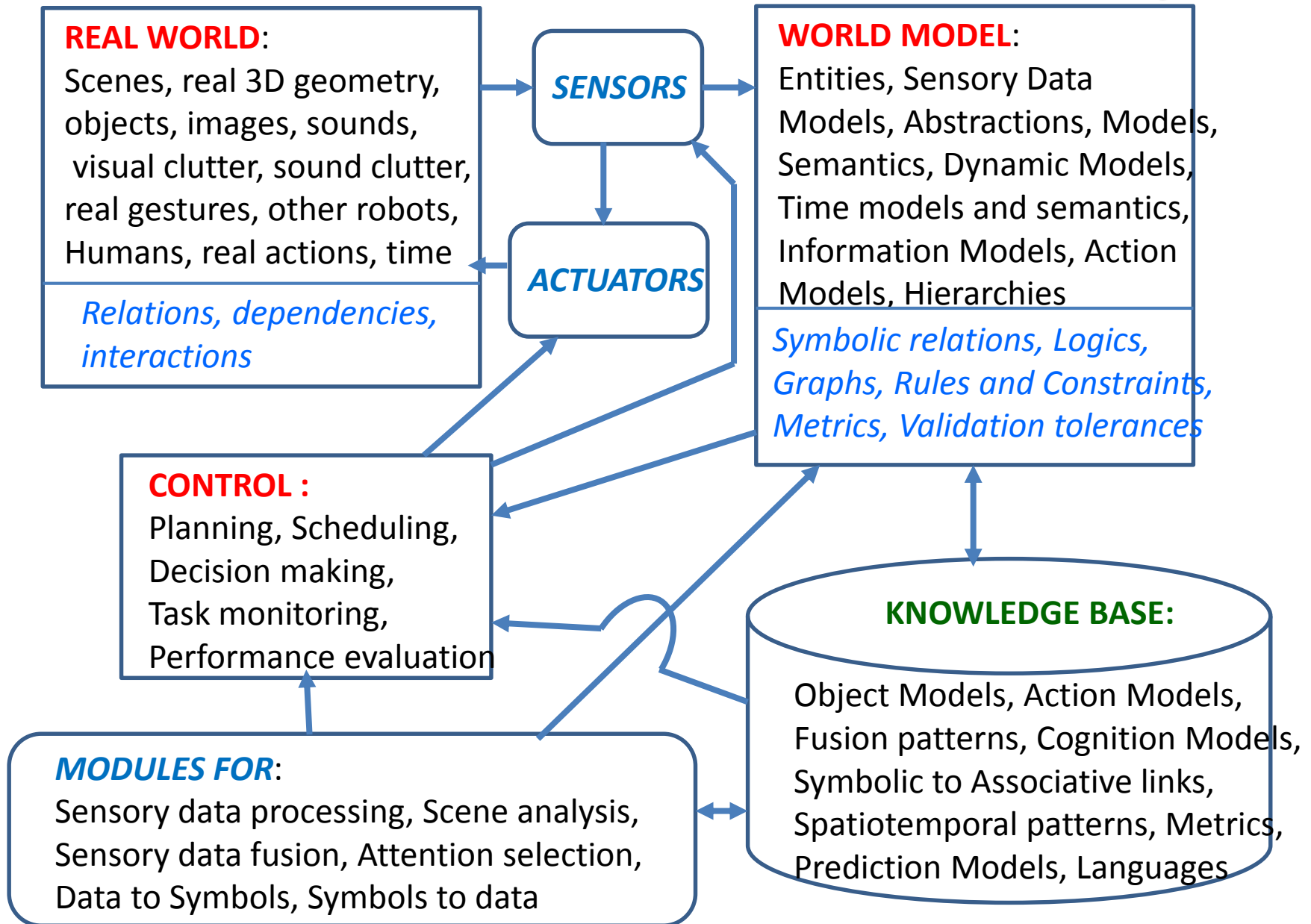
- The particular microrobots we are interested in are small insect-like robots with microfeatures, more specifically with flexible joints.



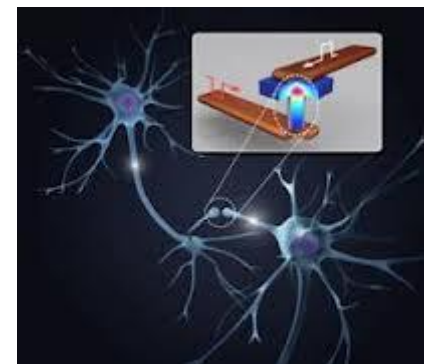
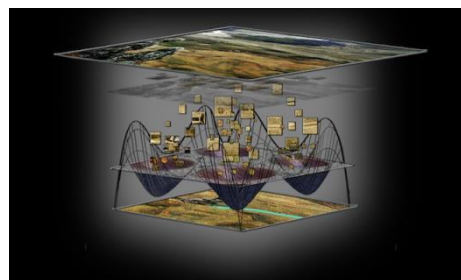
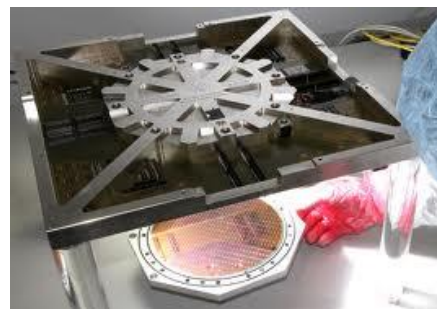
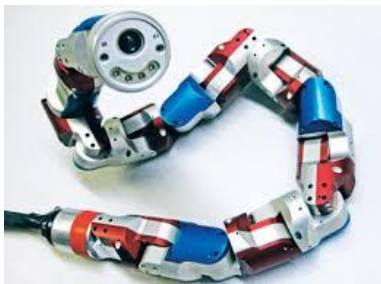
Real microrobot prototype on the left with Modelica DAE based model virtualized in Dymola on the right.

Dymola version has two distinct designs. (a) is the original design provided by D. E. Vogtmann, S. K. Gupta, and S. Bergbreiter [2012].

# Sensory Perception and Cognition: Internal Models for Collaborative Autonomy



# CPS Architecture: Perception- Cognition and Co-Robots



*The "pressure" of "P" on "C"*  
*The return of analog computation?*  
*Non-von Neumann Architectures?*  
*Physics of computation?*  
*Beyond Turing?*

*Cognition and knowledge generation from sensory perception –  
communicating with humans – collaboration  
Not just obeying commands – the inverse problem*



# Future “Smart” Homes and Cities

- UI for “Everything”
  - Devices with Computing Capabilities & Interfaces
- Network Communication
  - Devices Connected to Home Network
- Media: Physical to Digital
  - MP3, Netflix, Kindle eBooks, Flickr Photos
- Smart Phones
  - Universal Controller in a Smart Home
- Smart Meters & Grids
  - Demand/Response System for “Power Grid”
- Wireless Medical Devices
  - Portable & Wireless for Real-Time Monitoring



# Cars are Heavily Computerized: Electronics in Cars and Vulnerabilities



## UW/UCSD Work:

*Kosher et al., IEEE Symposium on Security and Privacy, '10*

- Reach CAN bus through diagnostic port

*Checkoway et al., USENIX Security, '11*

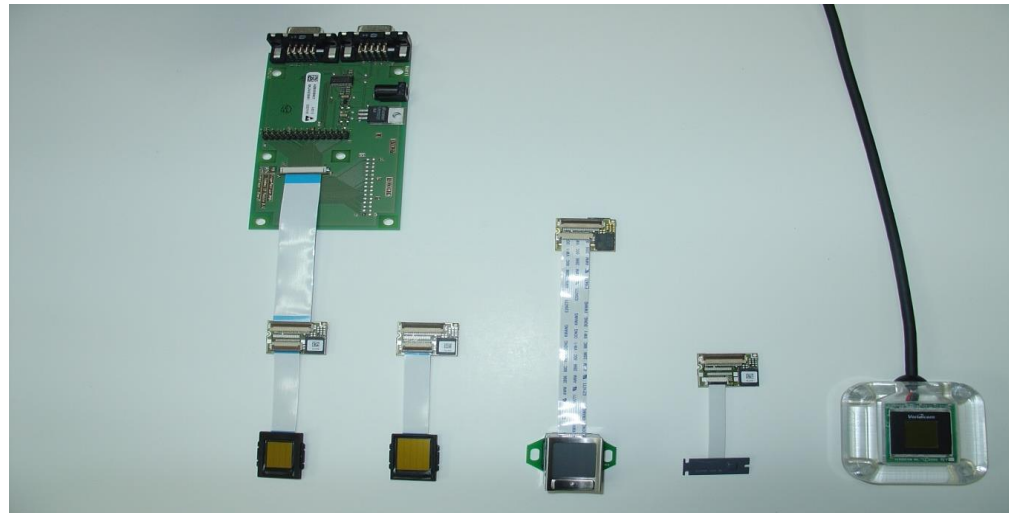
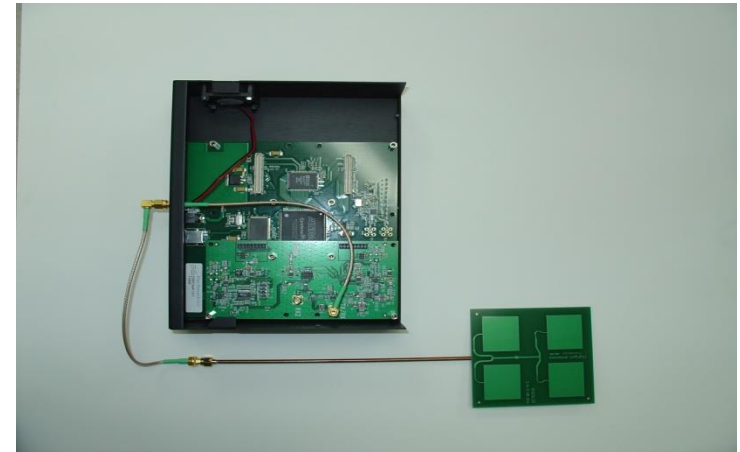
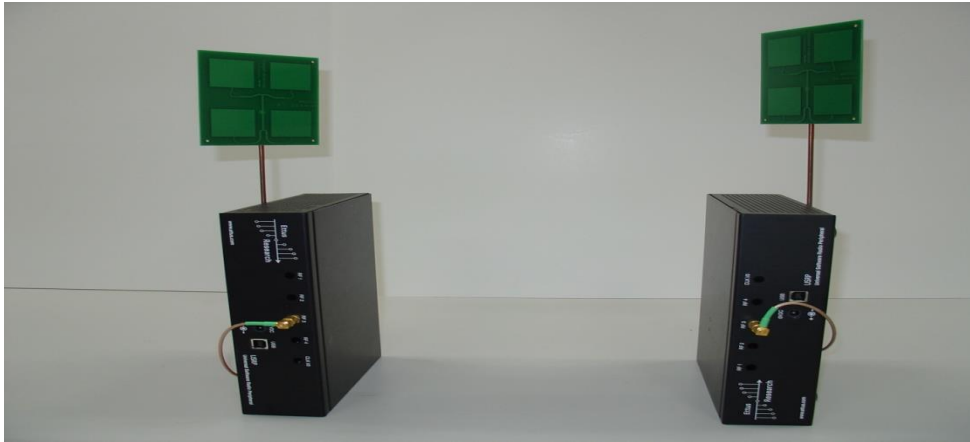
- Remote attacks
- Insert virus into computer system in mechanic shop
- Bluetooth
- Telematics unit
- CD player

# Physical Layer Authentication: Key Ideas and Challenges

- **Exploit characteristics (a.k.a. FINGERPRINTS) of physical layer (vastly ignored todate)**
  - Waveform, RF and hardware peculiarities ⇒  
    lead to ‘unshakeable’ fingerprints
  - Embed artificial and stealthy ‘fingerprints’
  - **Authenticate the device to the network and then the user to the device** ⇒ reduces attack risk (fewer times through the net)
- **Distribute assurance/trust function across software and hardware (increases difficulty to attacker significantly)**
  - Trusted computing platform – architecture modifications to allow multiple sources input (including biometrics)
  - TPM – MTM chip ‘add on’ to portable devices and TCN
  - Remote software attestation

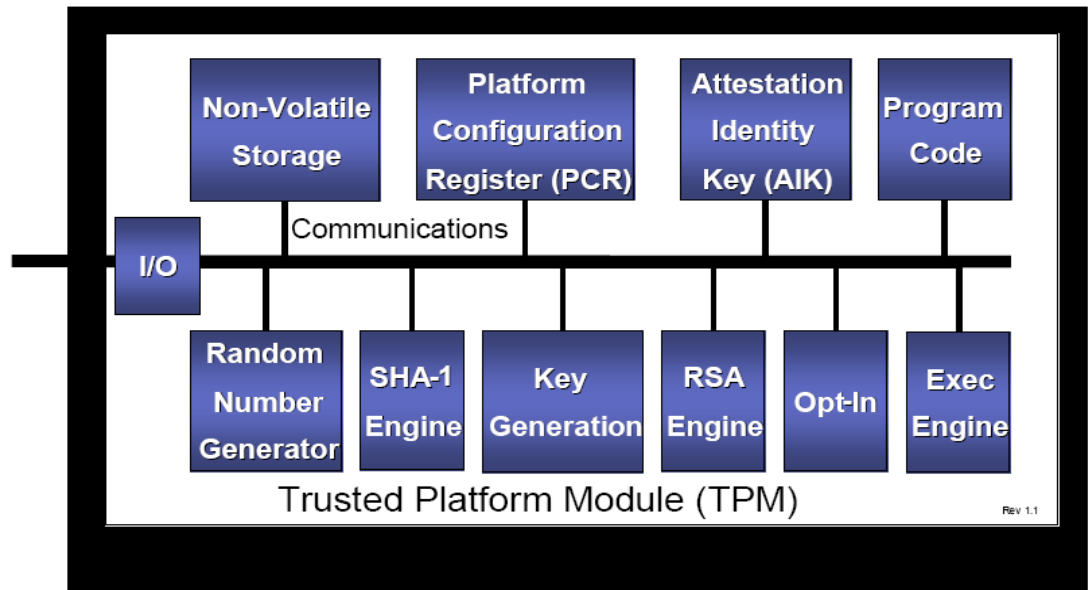
# Experimental Validation

Demonstrated Very Low Power Authentication is Feasible



# Trusted Computing

- **Trusted Platform Module technologies (TPM, MTM, TCN)**
  - A secure hardware
  - Protects the integrity and confidentiality of data with hardware support
  - Performs integrity measurements and reports them, thus attesting for the software running in the device
- **Provides a way to**
  - Understand the state of the platform,
  - Evaluate the state
  - Make a decision if the platform is appropriate for the task



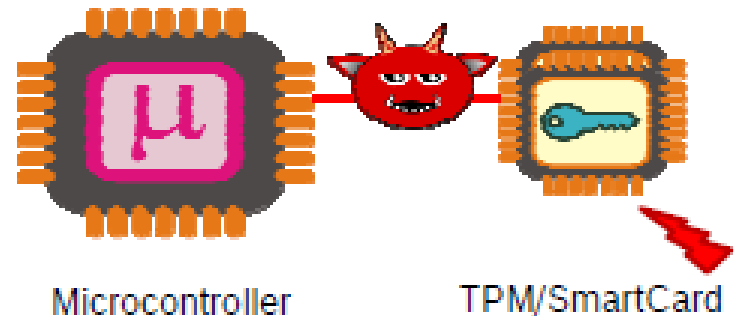
Source: TCG Architecture Overview, <http://www.trustedcomputinggroup.org>

# New Ideas: Hardware-Based Security

## Using an external TPM?

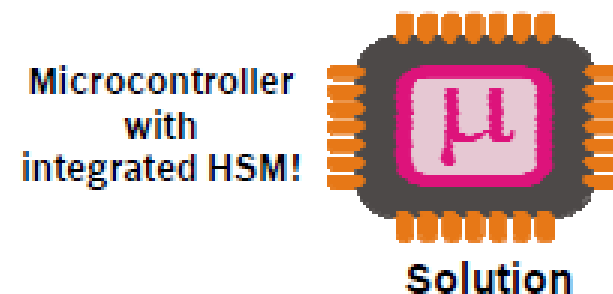
→ Initial idea: Use an existing component-of-the-shelf like a TPM or SmartCard as root-of-trust

- But...



→ Cost, PCB area,

→ Quality requirements, availability of suitable components (e.g. temperature range) and

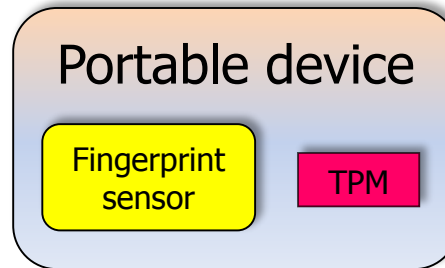


→ Sensitivity to valid attacks

- Reset attack (TPM is reset, manipulated  $\mu$ C continues operation)
- Data exchange between  $\mu$ C and TPM not protected

# Security Integration on the Portable Device

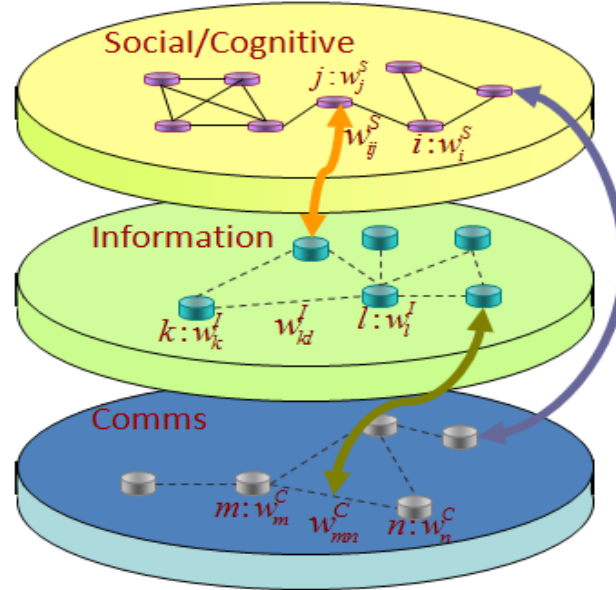
- The TPM/MTM is incorporated in the device



- Biometric information
  - protected in the TPM or
  - stored in the device but encrypted with keys that are managed by the TPM
- Hardened security encourages the use of the device
- **Challenges:**
  - (a) How to use informative time varying pieces of the biometric**
  - (b) Develop anti-spoofing techniques using the sensor signature**
  - (c) System integration and validation of the various fingerprints and physical layer techniques**
  - (d) Proof methods that security is improved – Information theoretic methods**

## The Challenge & Need:

- Composite trust in distributed sensing and control systems (DSCS)
- Security and Trust-aware DSCS algorithms
- Universal compositional security
- Performance, security, energy, tradeoffs
- Vulnerability analysis and resilient system architectures



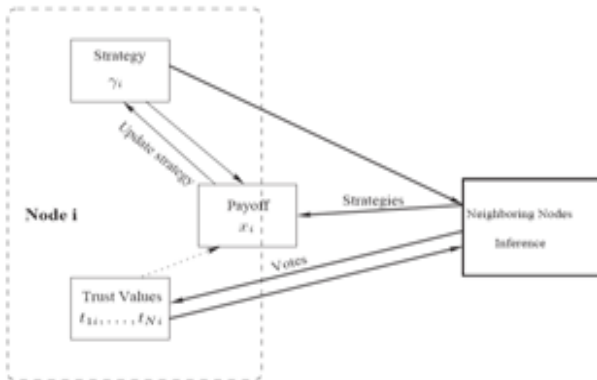
**Fig. 1:** Social (human agent) networks supported by technological networks

## APPROACH

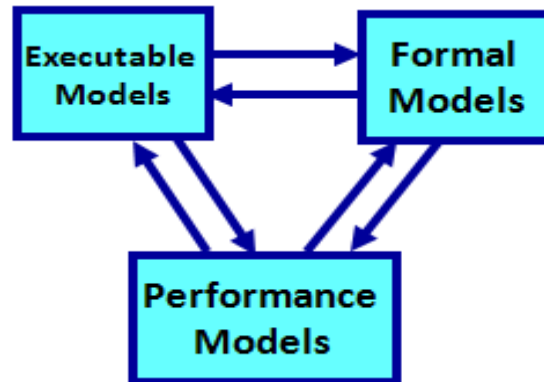
- Security and trust aware network utility maximization
- Weighted multi-graphs
- Multiple ordered semi-rings,
- Physical layer security and authentication for universal compositional security
- Network game theory
- Distributed hybrid systems

## APPLICATIONS

- Wireless communication and sensor networks
- Safety critical aircraft management systems
- Web-based social nets
- Power grid, smart grid, SCADA
- Smart buildings
- High integrity reputation and recommendation systems
- Resilience and robustness in the presence of adversaries



**Fig. 2:** Effects of trust on collaborative distributed control/operations (Baras 2005)



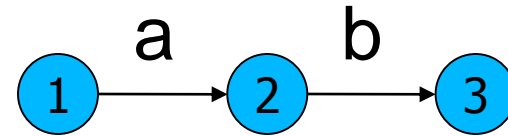
**Fig. 3:** Linked component-based executable, formal, performance models



# Trust Semiring Properties: Partial Order

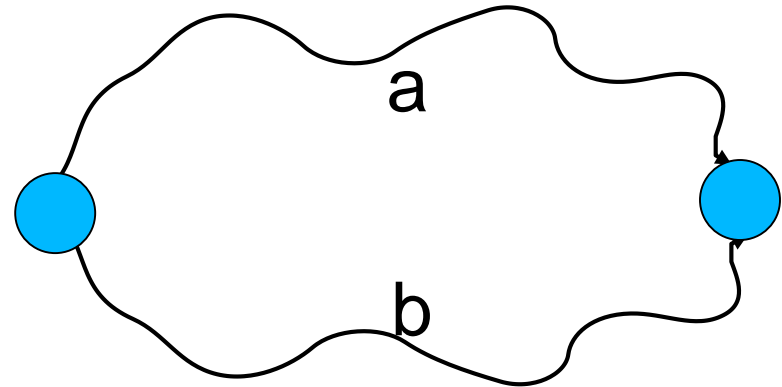
- Combined **along-a-path weight should not increase** :

$$a \otimes b \leq a, b$$



- Combined **across-paths weight should not decrease** :

$$a \oplus b \geq a, b$$



- Path interpretation

$$t_{i \rightarrow j} = \bigoplus_{\text{path } p: i \rightarrow j} t_{i \rightarrow j}^p$$

- **Linear system interpretation**

$$t_{i \rightarrow j} = \bigoplus_{\text{User } k} t_{i \rightarrow k} \bigoplus W_{k \rightarrow j}$$

$$\vec{t}_n = W \otimes \vec{t}_{n-1} \bigoplus \vec{b}$$

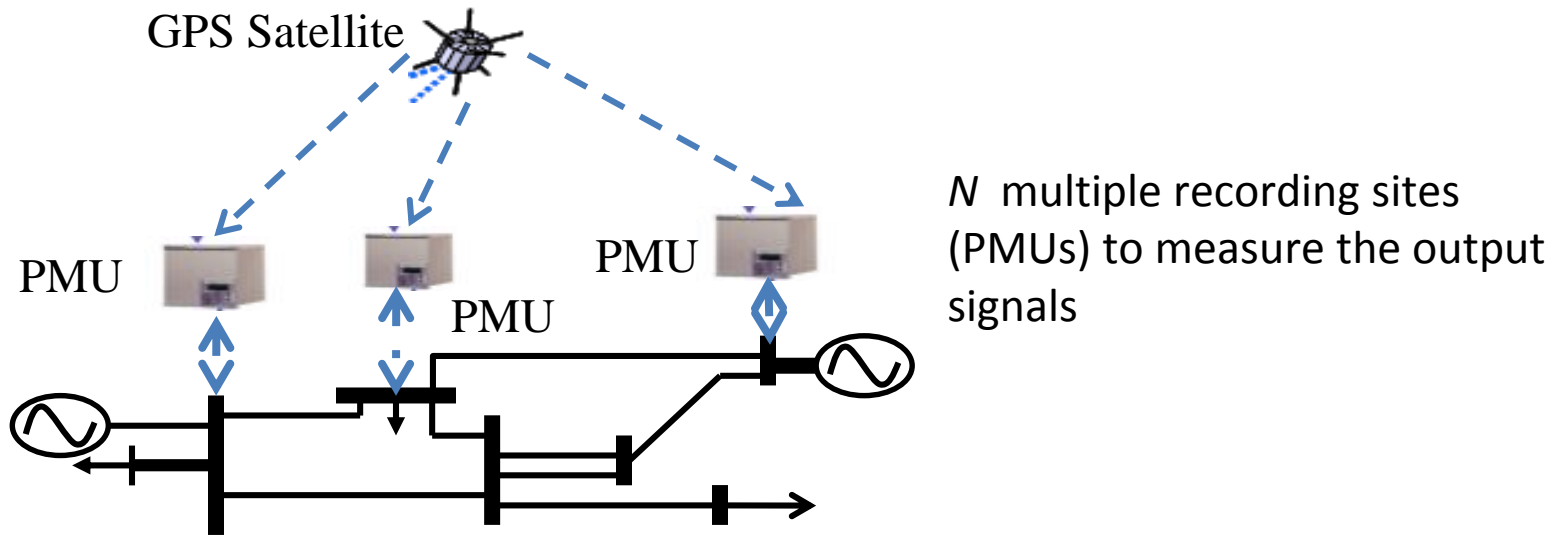
Indicator vector of pre-trusted nodes

- Treat as a **linear system**
  - We are looking for its **steady state**.

# Power Grid Cyber-security

- Inter-area oscillations (modes)
  - Associated with large inter-connected power networks between clusters of generators
  - Critical in system **stability**
  - Requiring **on-line** observation and control
- Automatic estimation of modes
  - Using currents, voltages and angle differences measured by PMUs (Power Management Units) that are distributed throughout the power system

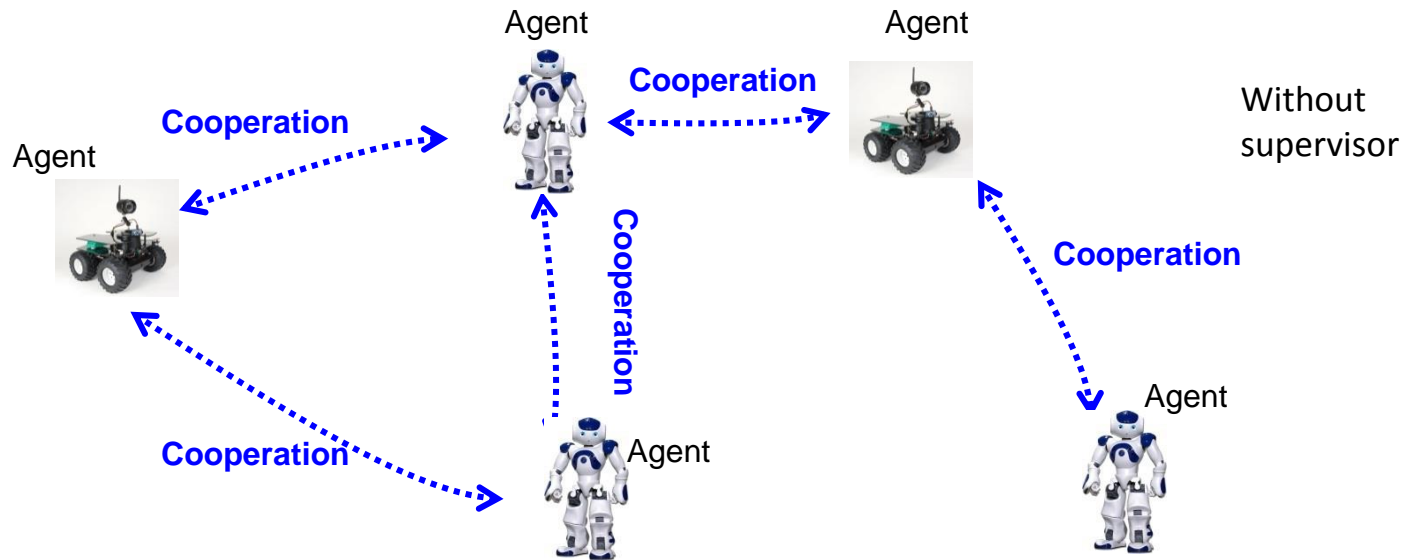
# Distributed Estimation



- To compute an accurate estimate of the state  $x(k)$ , using:
  - **local measurements**  $y_j(k)$ ;
  - information received from the PMUs in its **communication neighborhood**;
  - confidence in the information received from other PMUs provided by the **trust model**

# Consensus with Adversaries

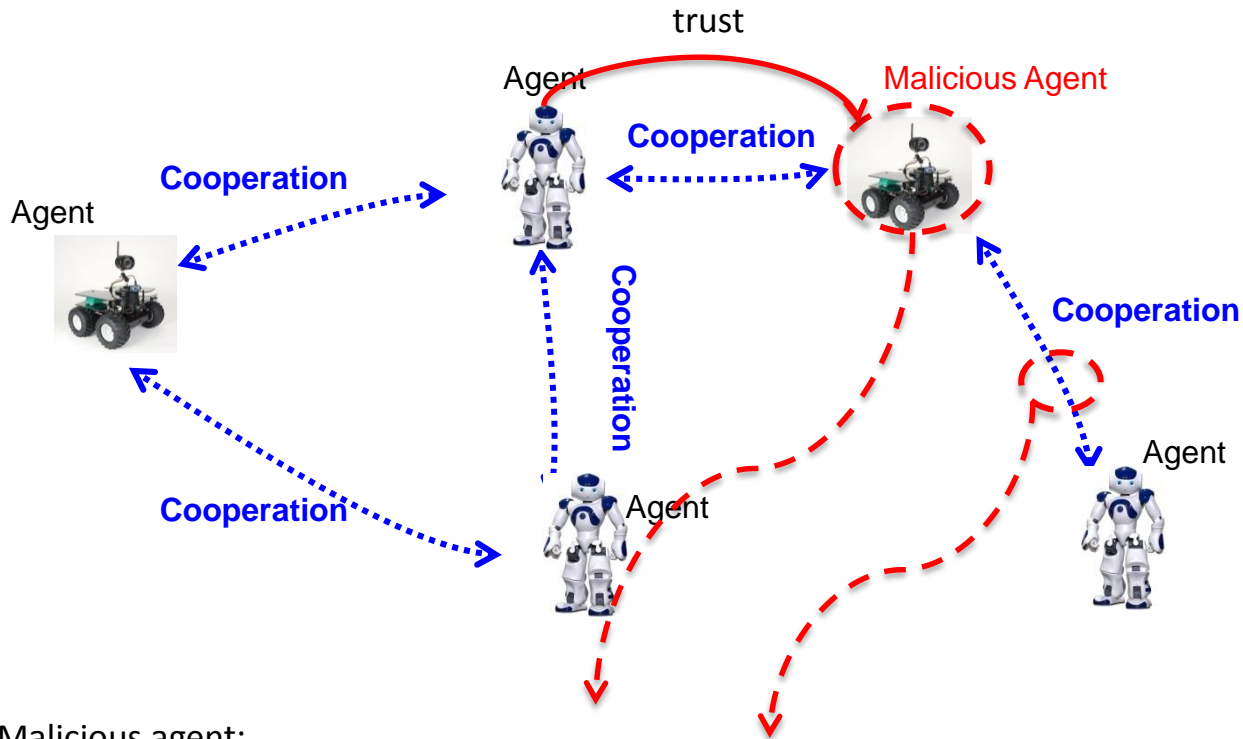
- Solve the problem via detecting adversaries in networks of low connectivity.
- We integrate a **trust evaluation mechanism** into our consensus algorithm, and propose a two-layer hierarchical framework.
  - Trust is established via **headers (aka trusted nodes)**
  - The top layer is a super-step running a **vectorized consensus algorithm**
  - The bottom layer is a sub-step executing our **parallel vectorized voting scheme**.
  - Information is exchanged between the two layers – they **collaborate**
- We demonstrate via examples solvable by our approach but not otherwise
- We also derive an upper bound on the number of adversaries that our algorithm can resist in each super-step



- Distributed sensor fusion. Goal: all agents reach consensus on ML estimate.
- Distributed coordination. Goal: all agents reach decision on same direction (location)

[1] Xiao, Lin, Stephen Boyd, and Sanjay Lall. "A scheme for robust distributed sensor fusion based on average consensus." *Information Processing in Sensor Networks*, 2005. IPSN 2005. Fourth International Symposium on Information Processing in Sensor Networks, 2005. IEEE, 2005.

[2] Jadbabae, Ali, Jie Lin, and A. Stephen Morse. "Coordination of groups of mobile autonomous agents using nearest neighbor rules." *Automatic Control, IEEE Transactions on* 48.6 (2003): 988-1001.



Malicious agent:

- Multiparty secure computation
- Link Jam & Noise Injection:

[4] Garay, Juan A., and Rafail Ostrovsky. "Almost-everywhere secure computation." *Advances in Cryptology—EUROCRYPT 2008*. Springer Berlin Heidelberg, 2008. 307–323.

[5] Khanafar, Ali, Behrouz Touri, and Tamer Basar. "Consensus in the presence of an adversary." *3rd IFAC Workshop on Distributed Estimation and Control in Networked Systems (NecSys)*. 2012.

[5] Pasqualetti, Fabio, Antonio Bicchi, and Francesco Bullo. "Consensus computation in unreliable networks: A system theoretic approach." *Automatic Control, IEEE Transactions on* 57.1 (2012): 90-104.

# Problem Formulation (cont.)

- Without considering failures, for certain nodes, the consensus problem in distributed control can be solved by simply iteratively calculating weighted averages of nodes' neighboring states.
  - Network of agents modeled by directed graph  $G(k) = (V; E(k))$   
 $V$  denotes the set of nodes and  $E(k)$  the set of edges at time  $k$   
 $N_i(k) = \{j \mid e_{ij}(k) \in E(k), j \neq i\}$  set of neighbor nodes of  $i$   
“can hear from at time  $k$ ”.  $N_i^+(k) = N_i(k) \cup \{i\}$
  - Nodes' states (decisions, beliefs, opinions, etc.) evolve in time according to the dynamics:

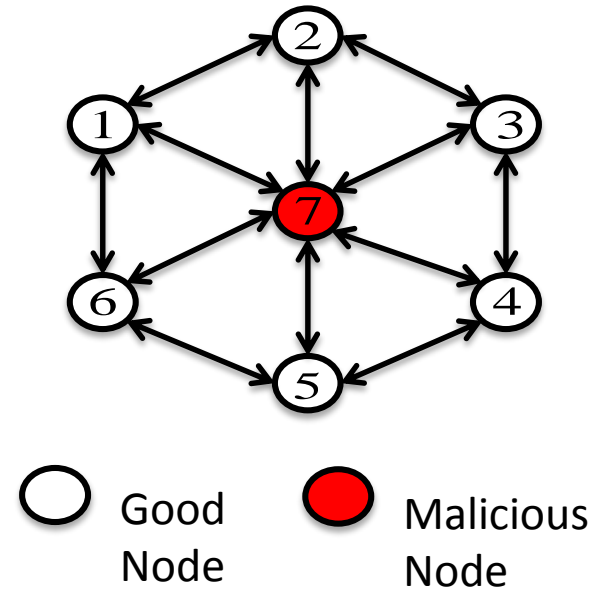
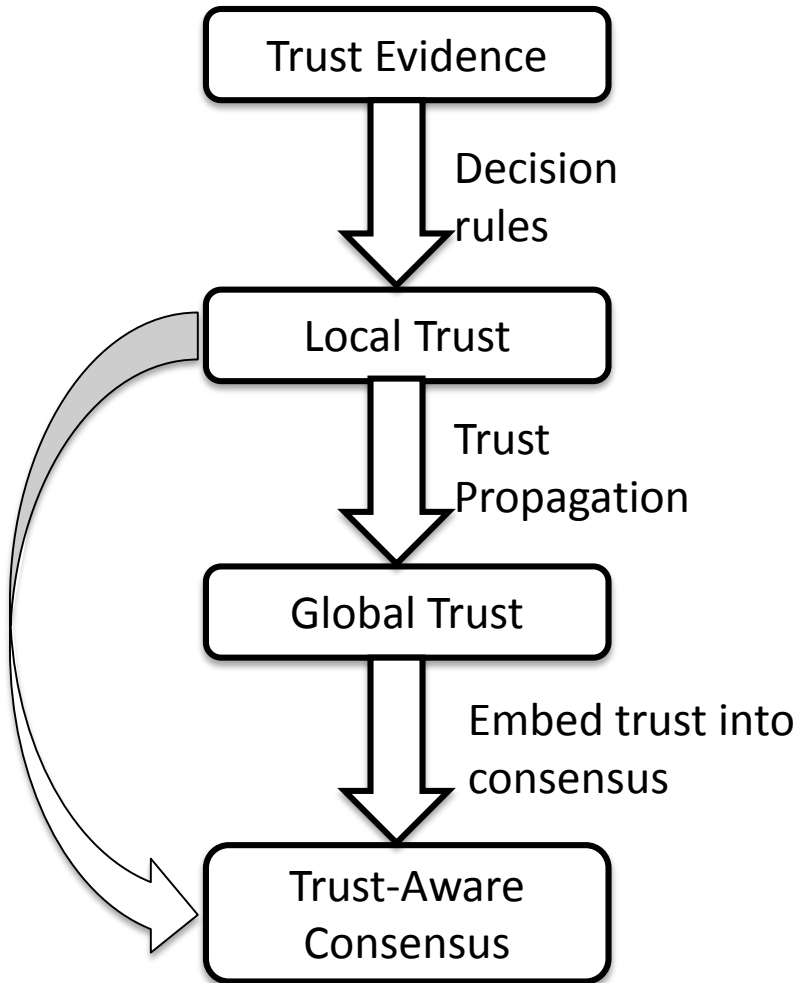
$$x_i(k) = \sum_{j \in N_i(k)} w_{ij}(k)x_j(k-1) + w_{ii}(k)x_i(k-1)$$

$X(k) = \{x_1(k), x_2(k), \dots, x_N(k)\}^T$   $N$ -dimensional vector of nodes' states at time  $k$ .

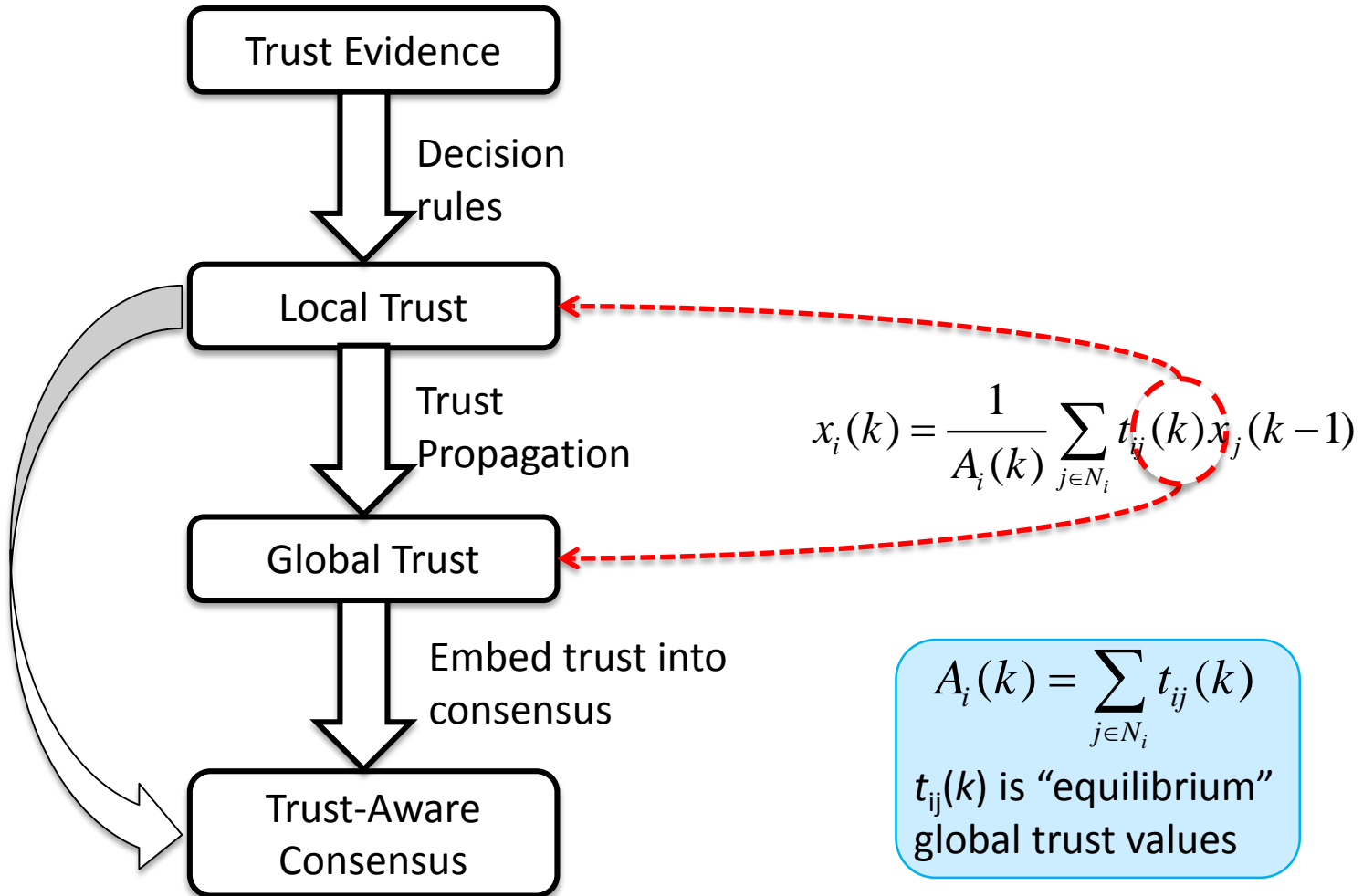
$W(k)$  is the updating matrix (weight matrix) at time  $k$ , rows sum to 1.



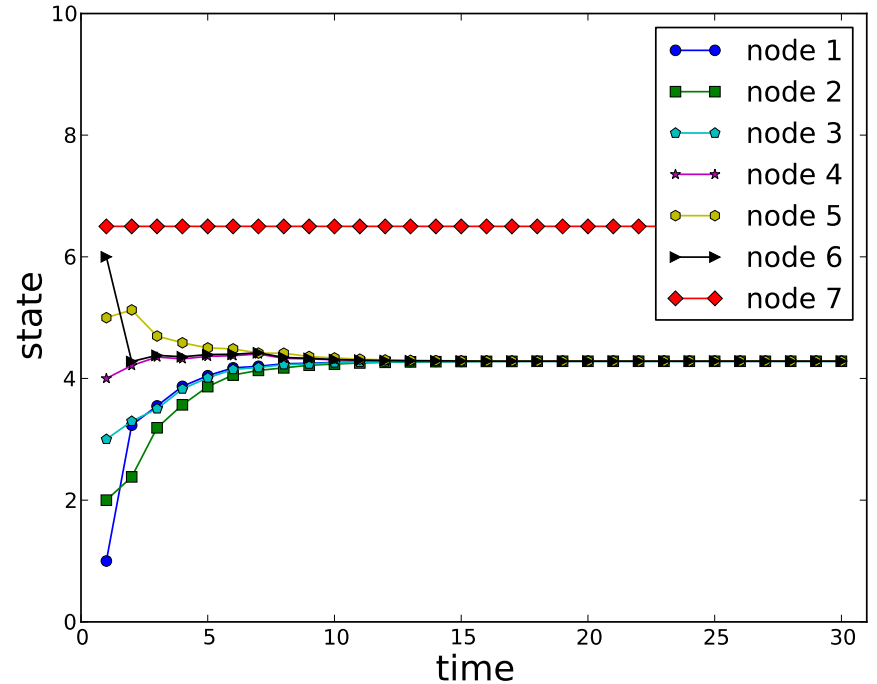
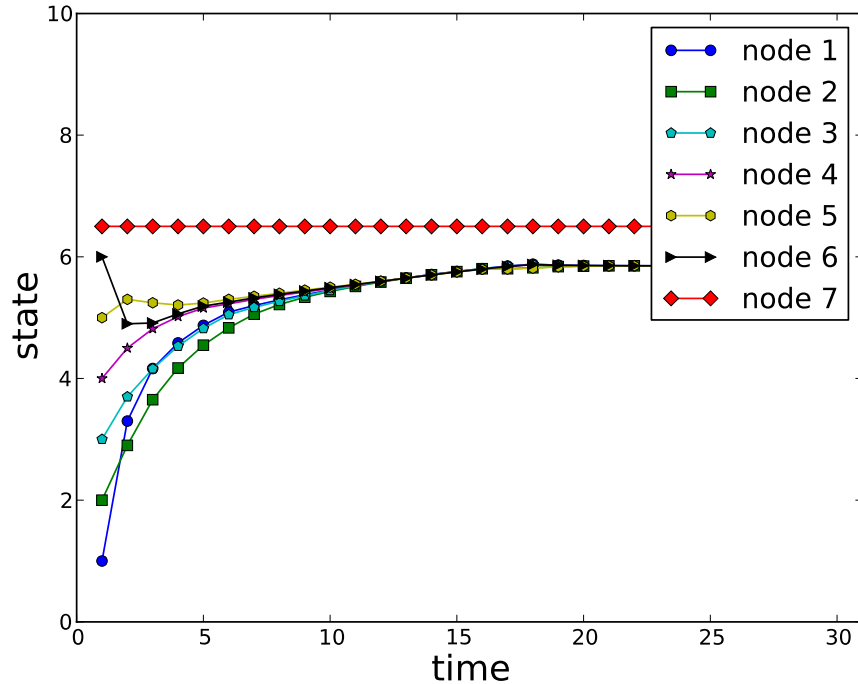
# Trust-Aware Consensus



# Trust-Aware Consensus



# Simulations



Adversary outputs constant message. Figure on the left has no trust propagation. Figure on the right has trust propagation.



# Joint Content Delivery and Wireless Network Optimization

## Existing System Design

### Social Network

Predict “rewards”.

“Big Data”: various ML models.

Slow in training, fast in computing.

Asynchronous, centralized.

### Wireless Network

Schedule resource for delivery.

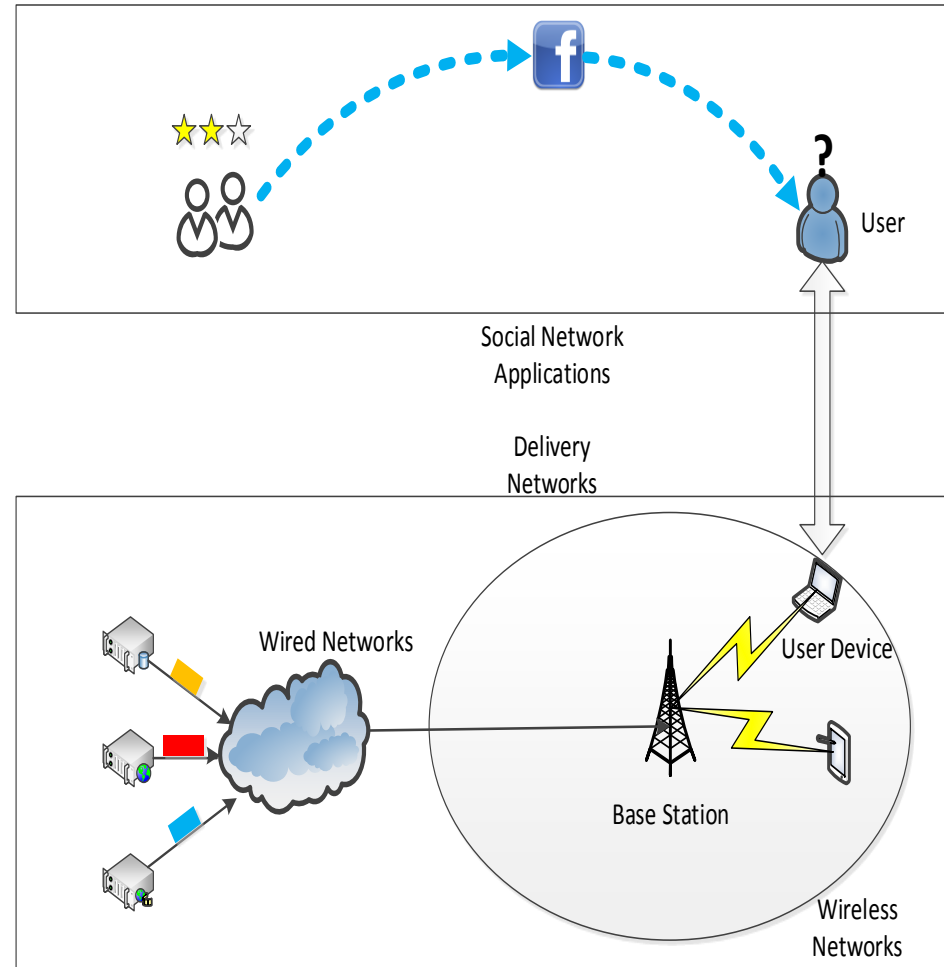
Randomness of channel.

Time-variant.

Synchronized, distributed.

Different metrics/utilities:

- Ads: number of views (= ad payout).
- Videos: time spent.
- General: user satisfaction.



# Three Scenarios/Problems

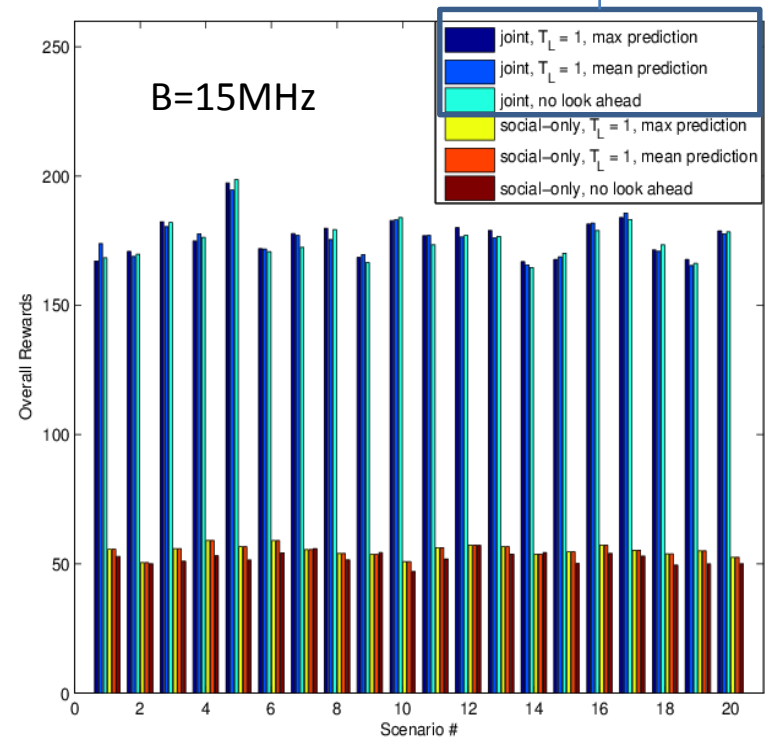
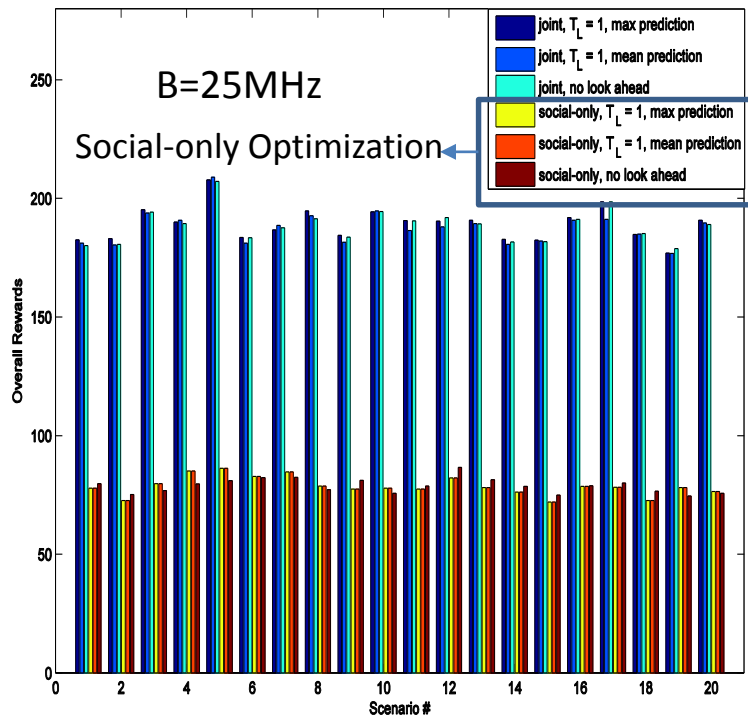
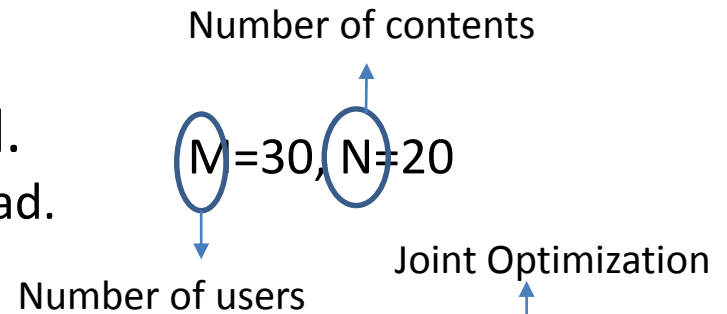
- Single base station, time-invariant reward.
  - Basic problem.
  - Establish foundation framework.
- **Multiple** base stations, time-invariant reward.
  - Many system configurations.
- Single base station, time-**variant** reward.
  - Time-variance specifically due to social dynamics.

# Comparison

	<b>Traditional</b>	<b>Joint Optimal</b>
<b>What to deliver?</b>	Social optimal	Joint optimal
<b>How to deliver?</b>	Unicast	Multicast
<b>Fragmentation?</b>	Packet	Content package

# Simulation Results – Overall System Rewards

Significant joint optimization gain.  
 Myopic scheduling is sufficiently good.  
 No significant improvement for look-ahead.



*Thank you!*

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*Questions?*