



Optimising Shared Awareness in Communication Constraint Sensor Networks

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Outline

- › **Applications**
 - › Task Force, Cooperative Intelligent Cars, Camera Network

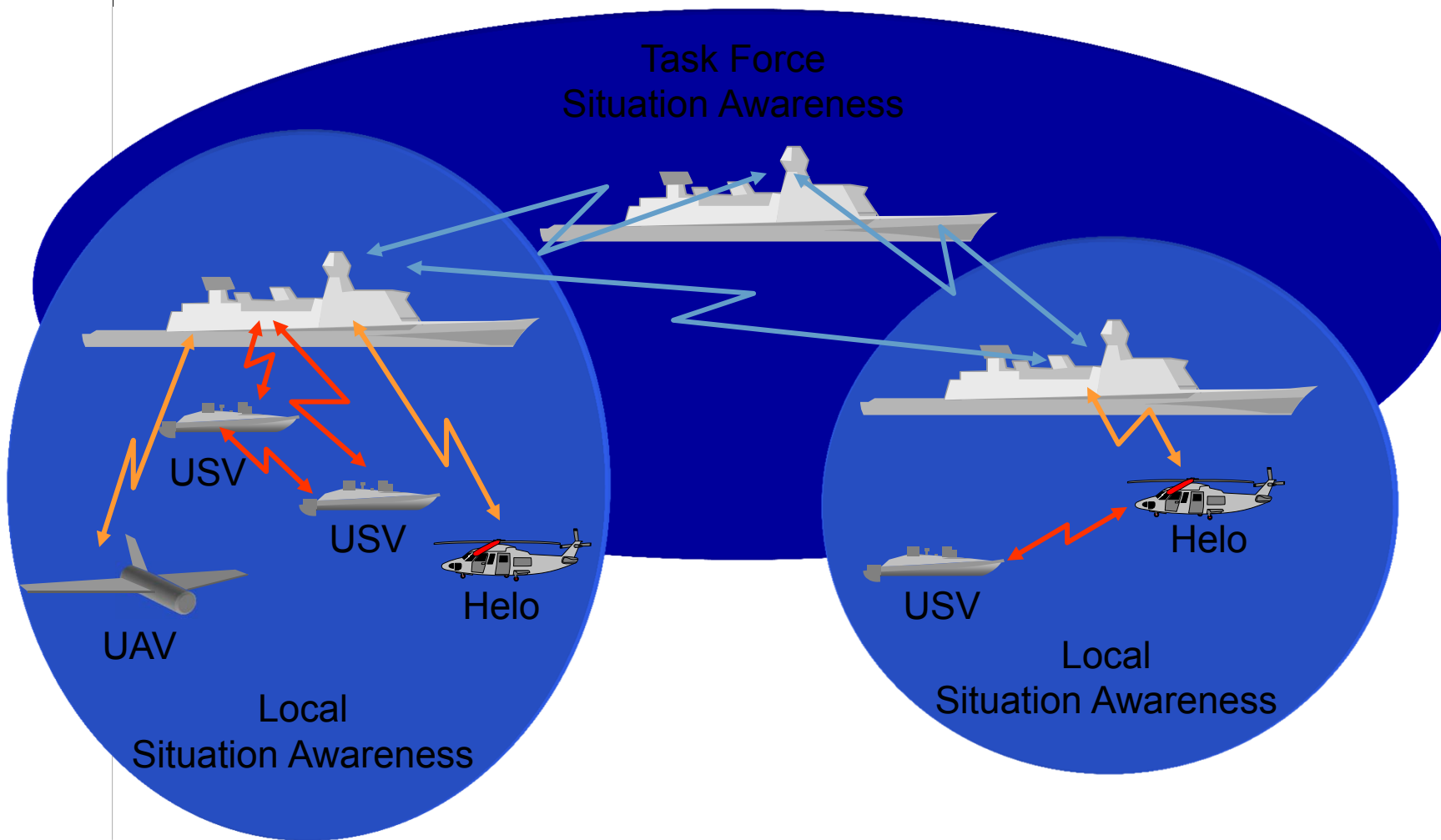
- › **Intelligent Systems**
 - › Challenges, Hybrid systems, Adaptivity

- › **Modelling Intelligent Systems**
 - › Decomposition, Adaptivity: Self-organisation and Self-optimisation

- › **Optimisation by balancing value and cost of communication**
 - › Utility and Value of information, Cost of communication



Situation Awareness in a Task Force



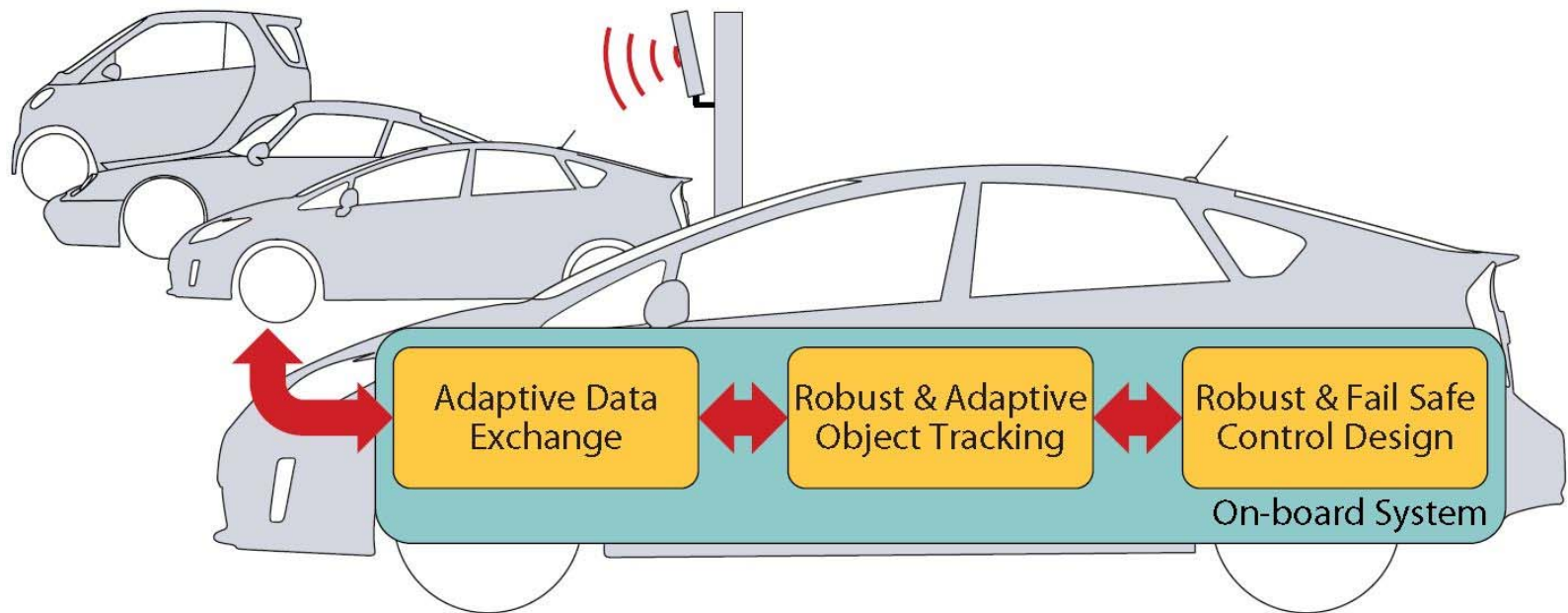


Situation Awareness in a Task Force



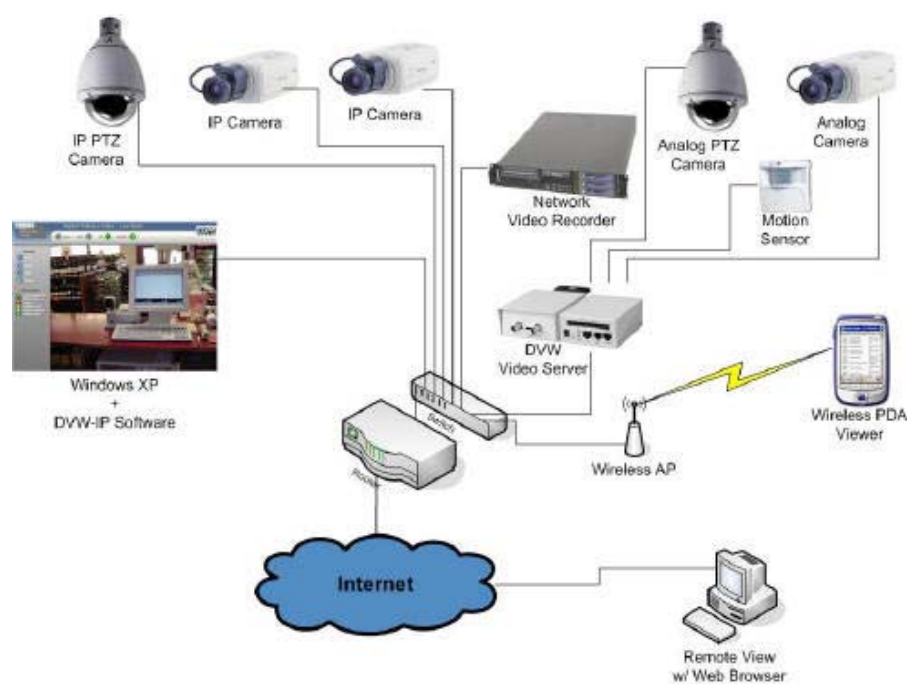


Situation Awareness in Cooperative Intelligent Cars



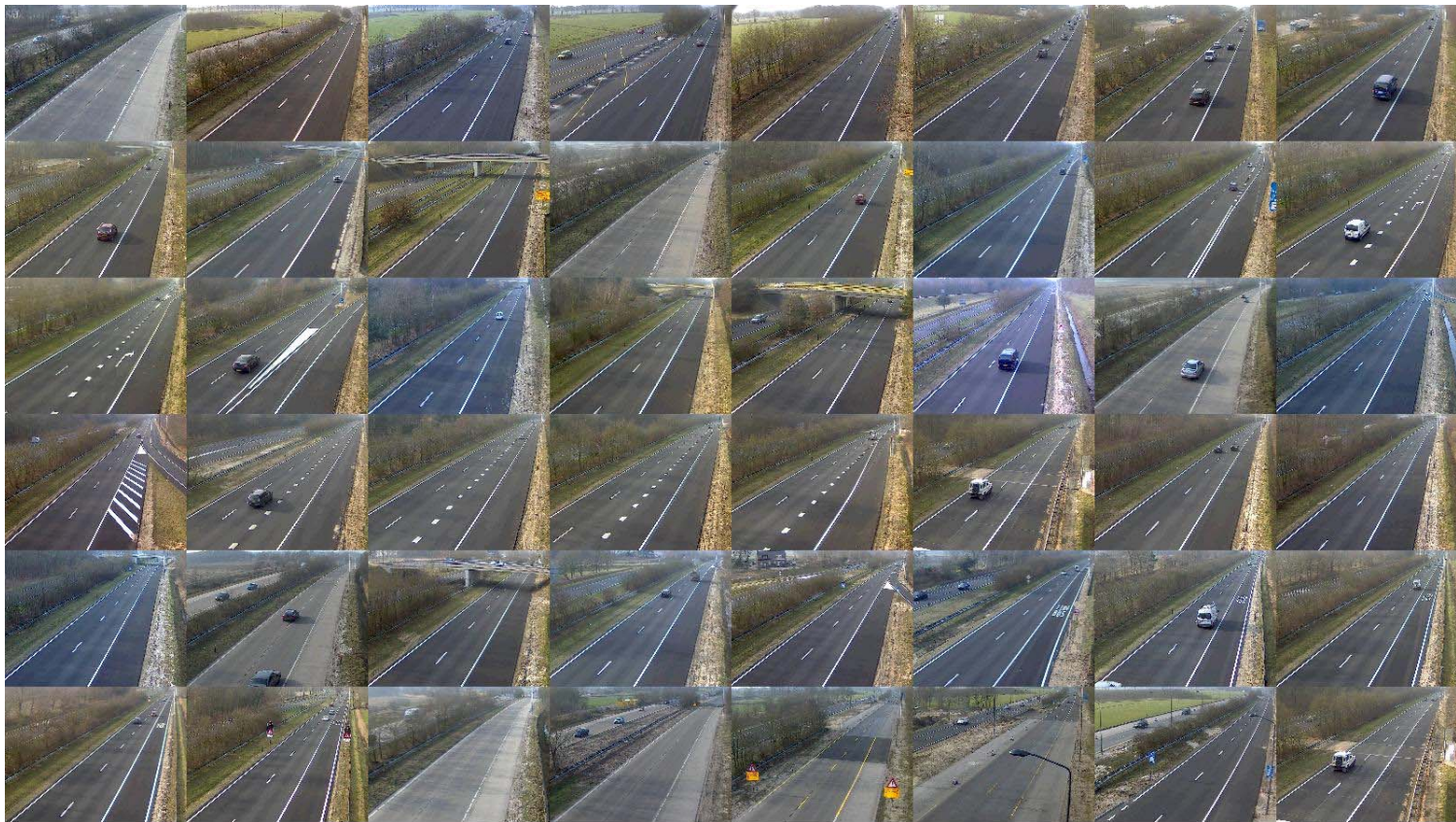


Situation Awareness in Camera Network





Situation Awareness in Camera Network





Why optimisation?

- › Information needs may depend on the situation
- › Constraints in Communication may depend on the situation
- › This requires some intelligent behaviour of the system:
 - › What is my goal?
 - › What is my current state?
 - › What can I do?
 - › What is my plan?



Intelligent Systems: Challenges



- Semantic interoperability
- Common system model
- No Design time thinking
- Clear/quantified goals
- Focus on efficiency
- Hammer only for nails

Breugel: Tower of Babel (1563)



Intelligent Systems: symbolism - connectionism

- › **Symbolism:** system exhibits intelligent behavior by manipulating symbols with formal rules
 - › E.g. Chess computer, rule based reasoning

- › **Connectionism:** system exhibit intelligent behavior without storing, retrieving, or otherwise operating on structured symbolic expressions;
 - › Pattern matcher, dynamic neural network

- › **Hybrid systems:** some mix, intermediate form or full integration of both views.



Intelligent Systems: Situated Adaptivity

› What?

- › System that can adapt to the situation in such a way that it becomes more efficient in reaching its implicit or explicit goals.

› How?

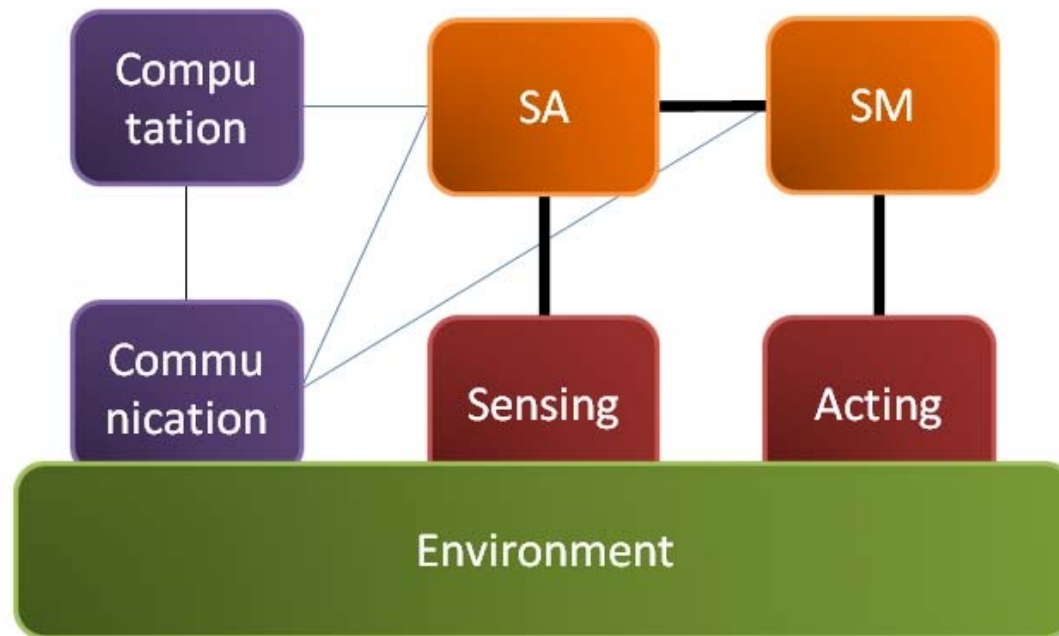
- › Internal management functions that can either reorganise or optimise core system functions (Hybrid approach)

› Why?

- › If systems need to perform in a wide variety of situations adaptivity is more efficient.

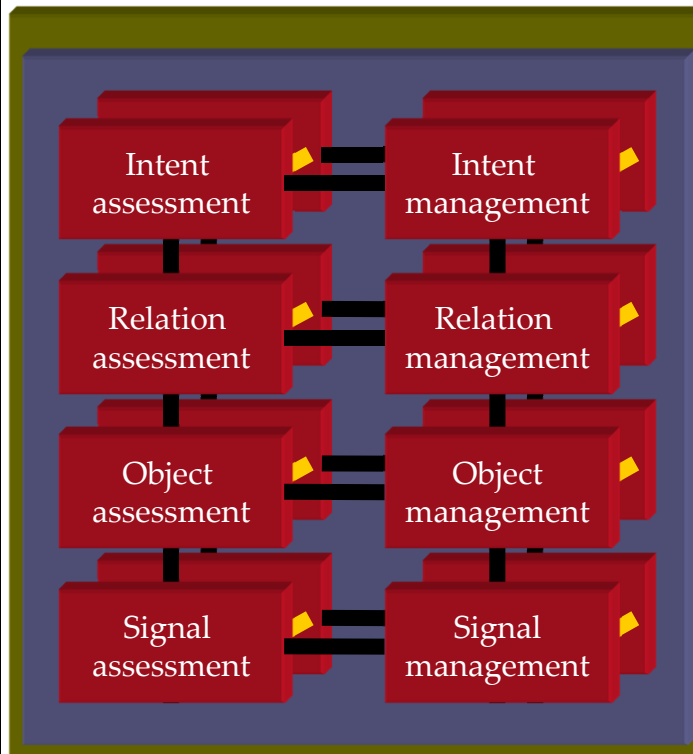


System modeling: Basic functions



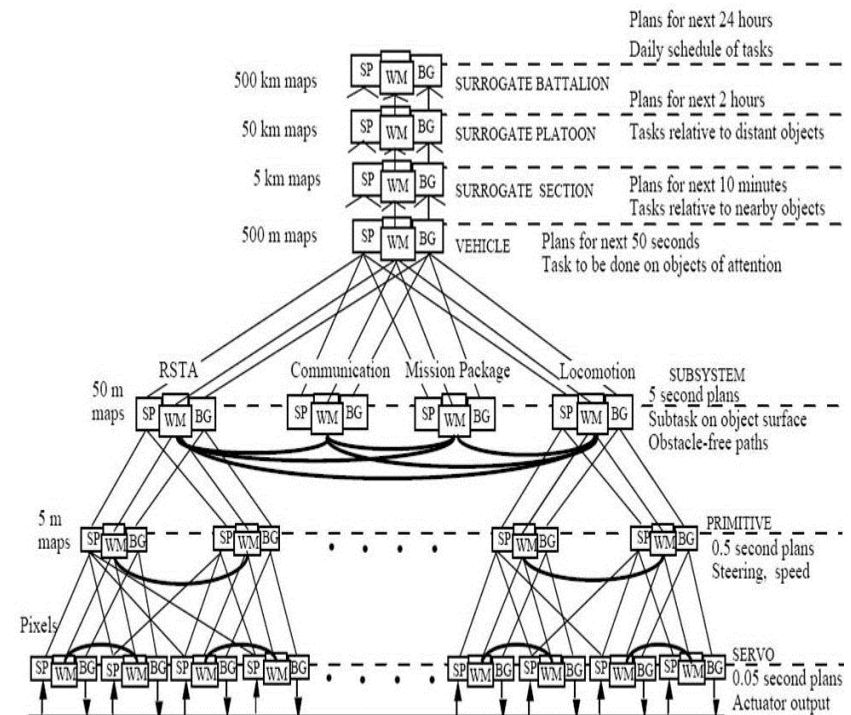


System modeling: Hierarchy



JDL

situation abstraction

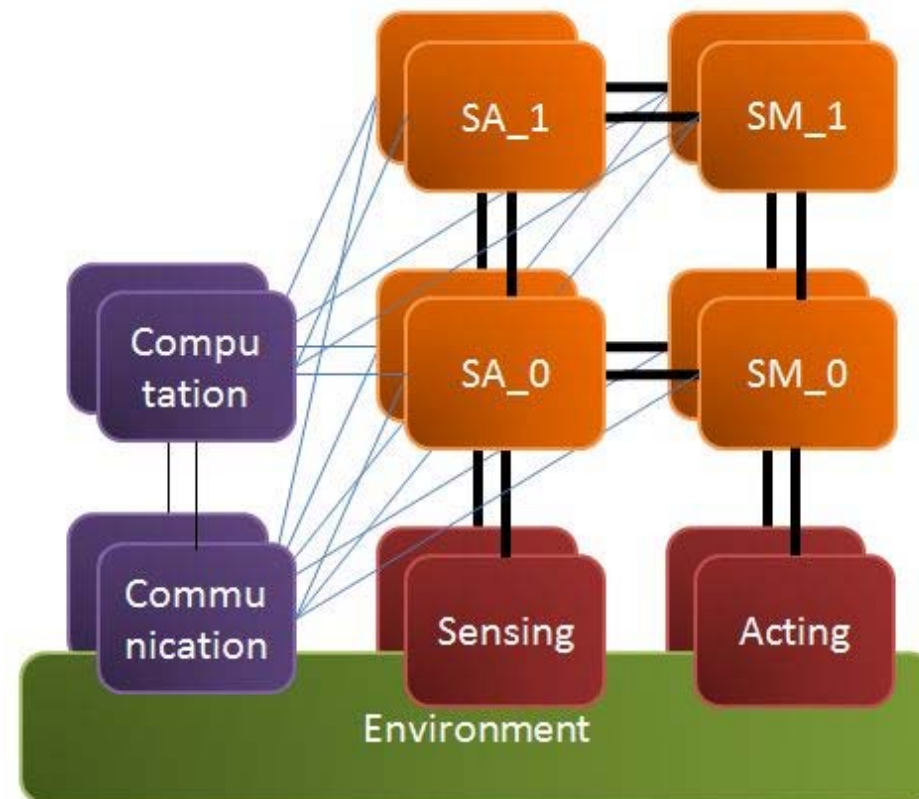


4D/RCS

& space time abstraction



Distributed intelligent systems with hierarchy



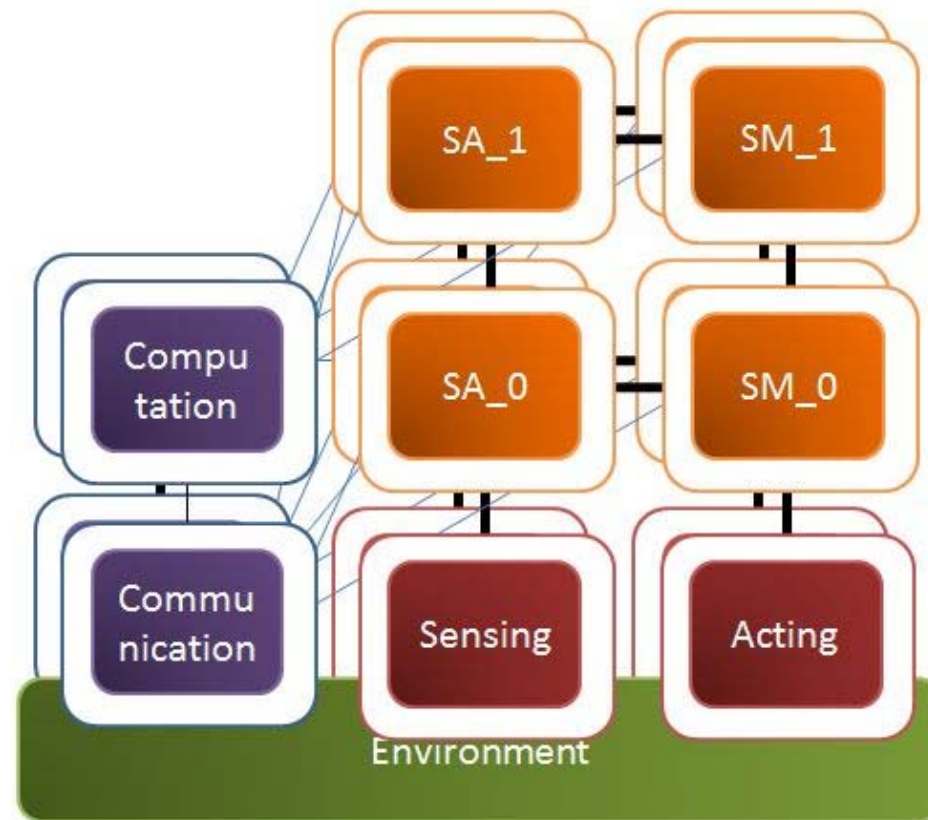


System modeling: Adaptive systems

- › **Service oriented architectures**
- › **Multi agent systems**
- › **Dynamic Bayesian networks**
- › **4D/RCS**

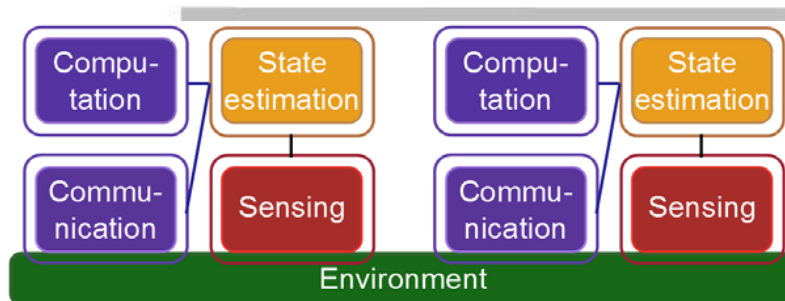


Adaptivity: self-organisation & self-optimisation

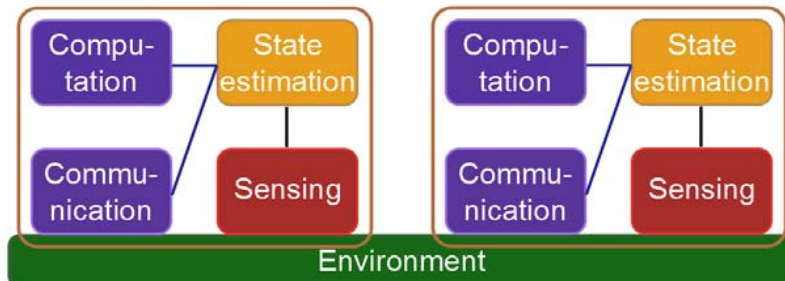




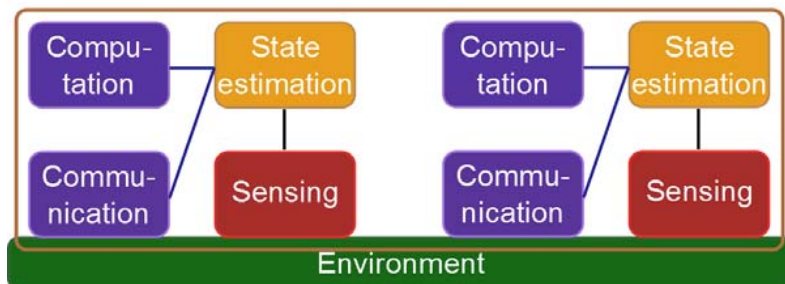
Centralised or decentralised adaptivity



A setup of two nodes, where each core-functionality/resource is surrounded by a supportive shell-functionality for self-organisation & self-optimization



A setup of two nodes, where the core-functionalities/resources are surrounded by a supportive shell-functionality per node for self-organisation & self-optimization

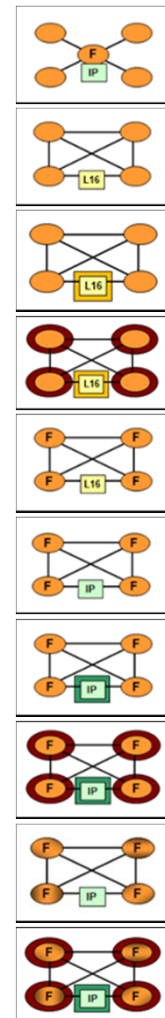


A setup of two nodes, where the core-functionalities/resources of both nodes are surrounded by a centralized shell-functionality for self-organisation & self-optimization



SA concepts

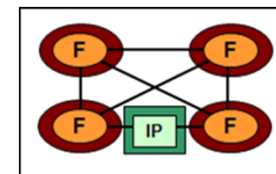
- › Concept 1.1: Centralised measurement fusion
- › Concept 2.1: Link 16
- › Concept 2.2: Link 16 enhanced
- › Concept 2.3: Link 16 smart evaluation
- › Concept 2.4: Link 16 track fusion
- › Concept 3.1: Decentr. measurement fusion
- › Concept 3.2: Decentr. measurement fusion smart network
- › Concept 3.3: Decentr. measurement fusion smart network smart eval.
- › Concept 3.4: Decentr. measurement fusion non-identical algorithms



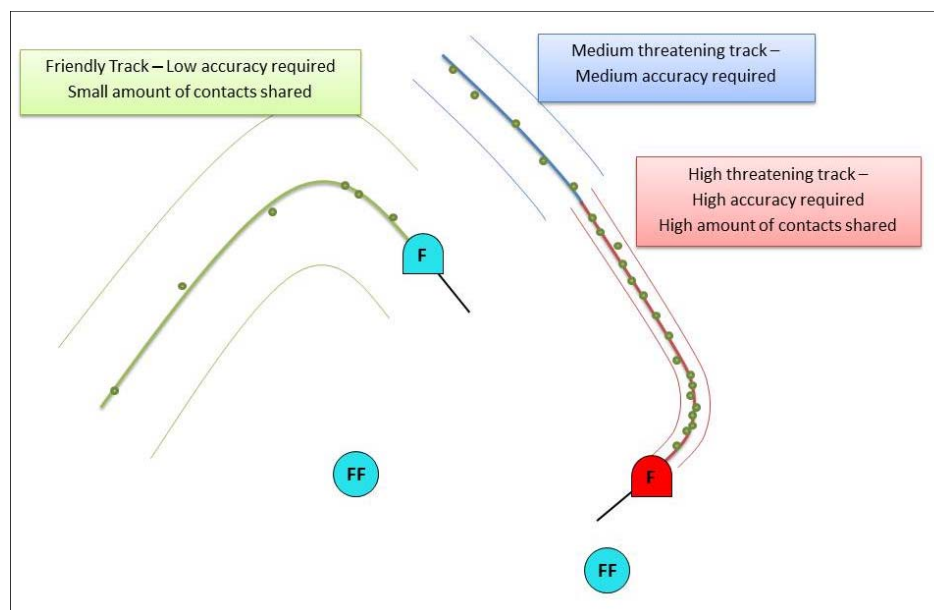


Concept 3.3

Decentr. measurement fusion smart network smart eval.

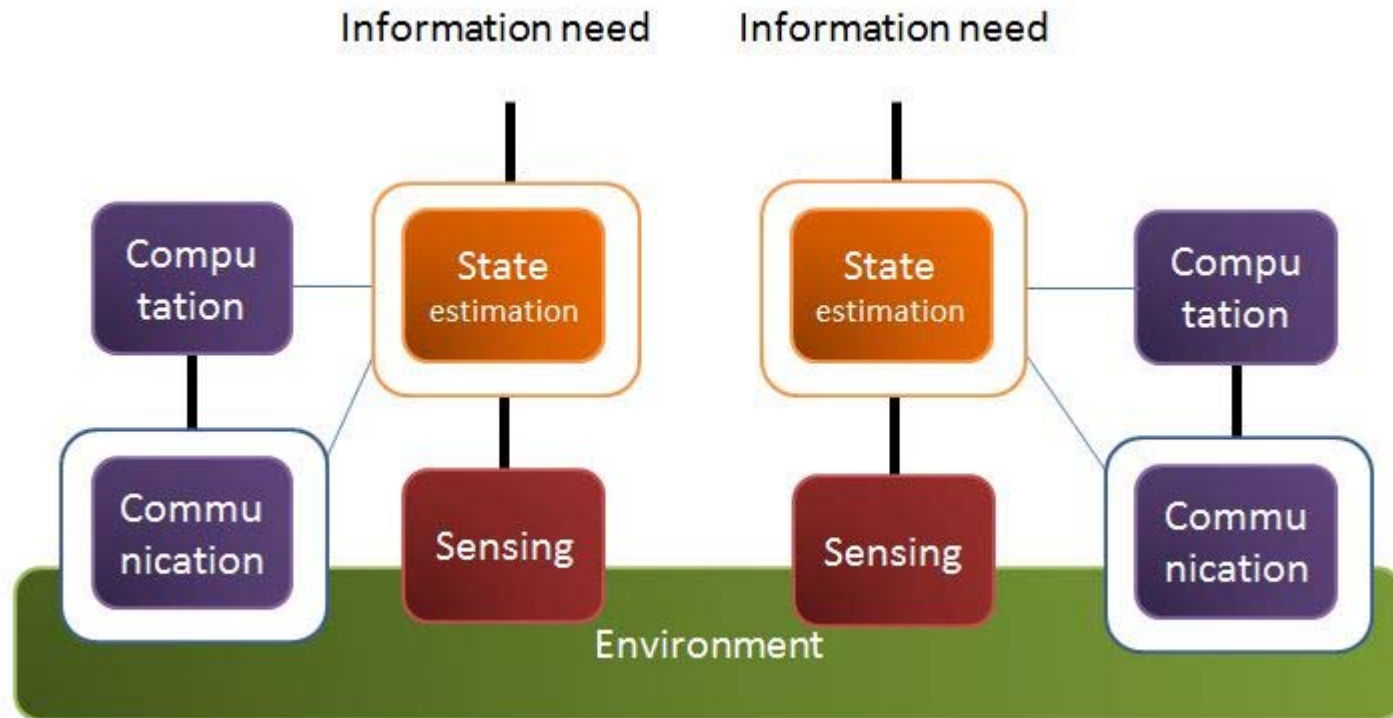


› Utility Function: accuracy,





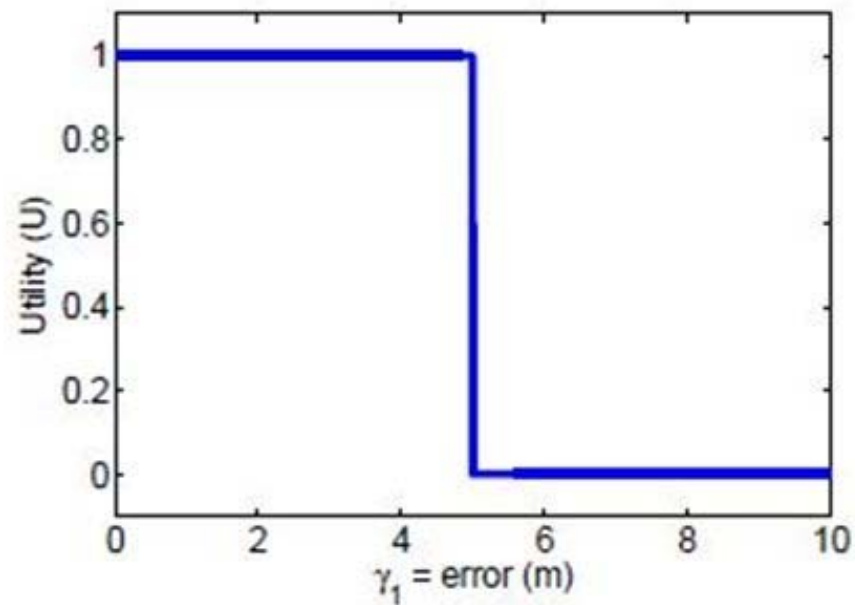
Self-optimisation with communication constraint





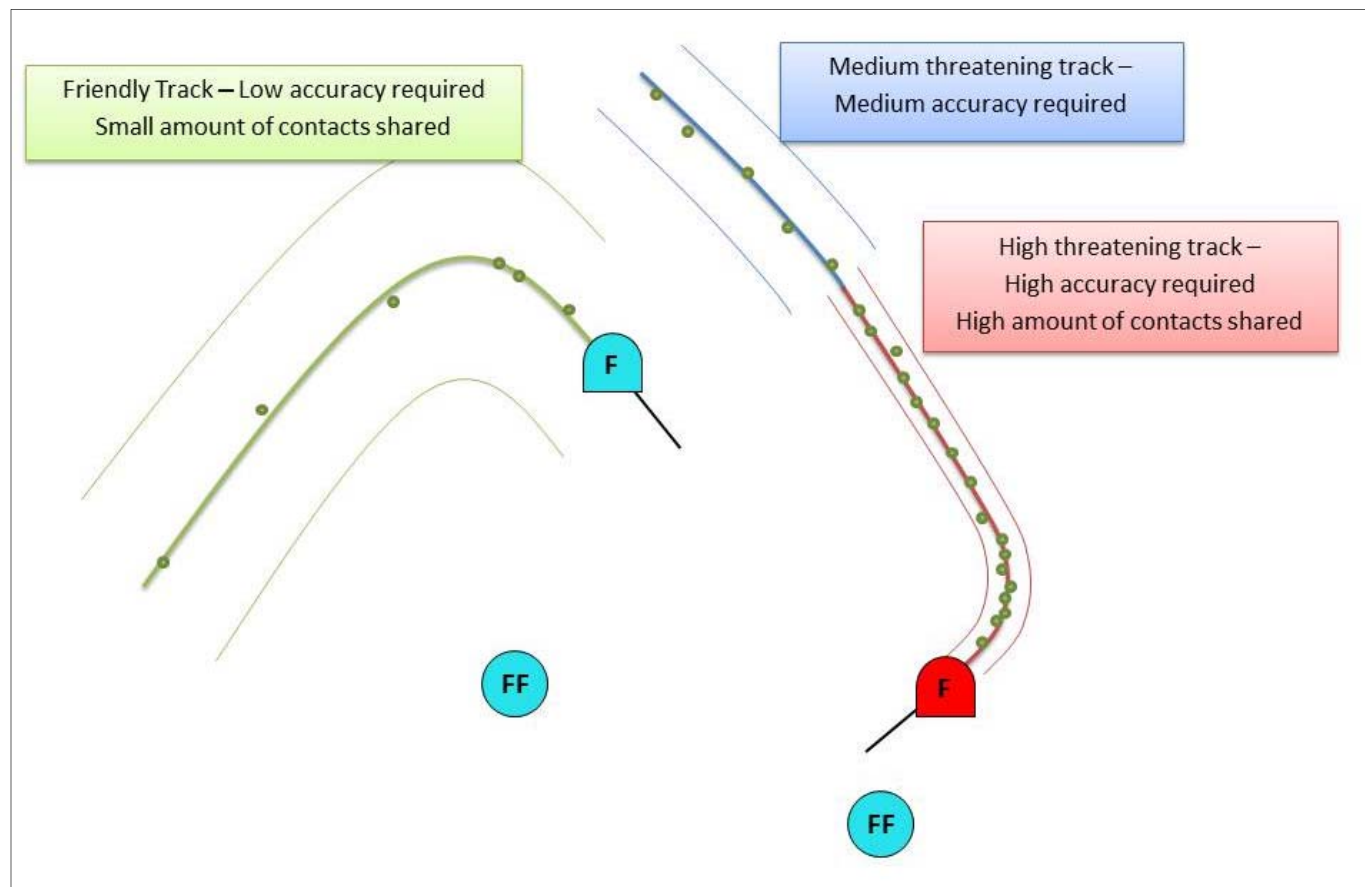
Balancing value of information and cost of communication

Information need: Utility function (an example)





Utility function can be object specific

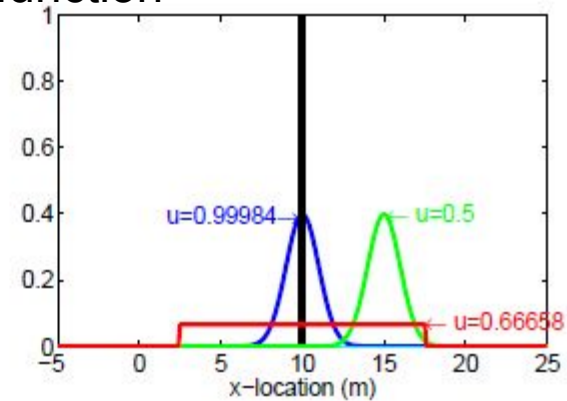
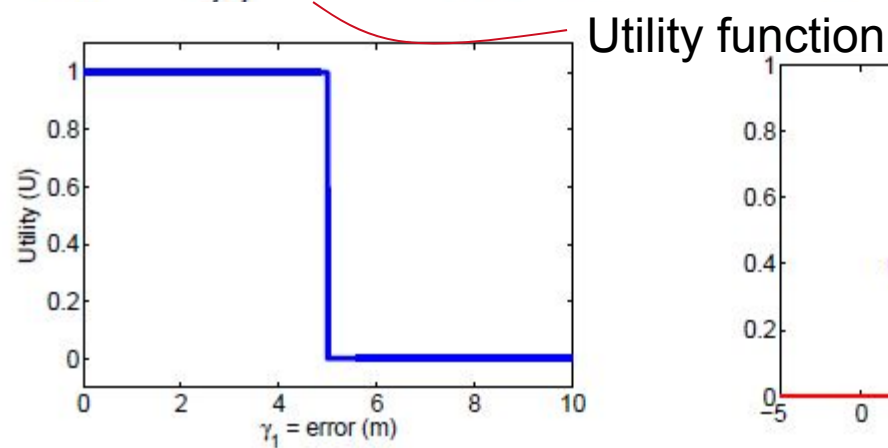




The Value of New Information

Utility function with respect to a reference state

$$U(\Gamma(\hat{p}_r, \hat{p})) = \iint U(\Gamma(\vec{x}_r, \vec{x})) \hat{p}(\vec{x}) \hat{p}_r(\vec{x}_r) d\vec{x} d\vec{x}_r.$$



$$\hat{v}(\hat{X}, \mathcal{Z}) = U(\Gamma(\hat{X}_r, \hat{X} | \mathcal{Z})) - U(\Gamma(\hat{X}_r, \hat{X})).$$

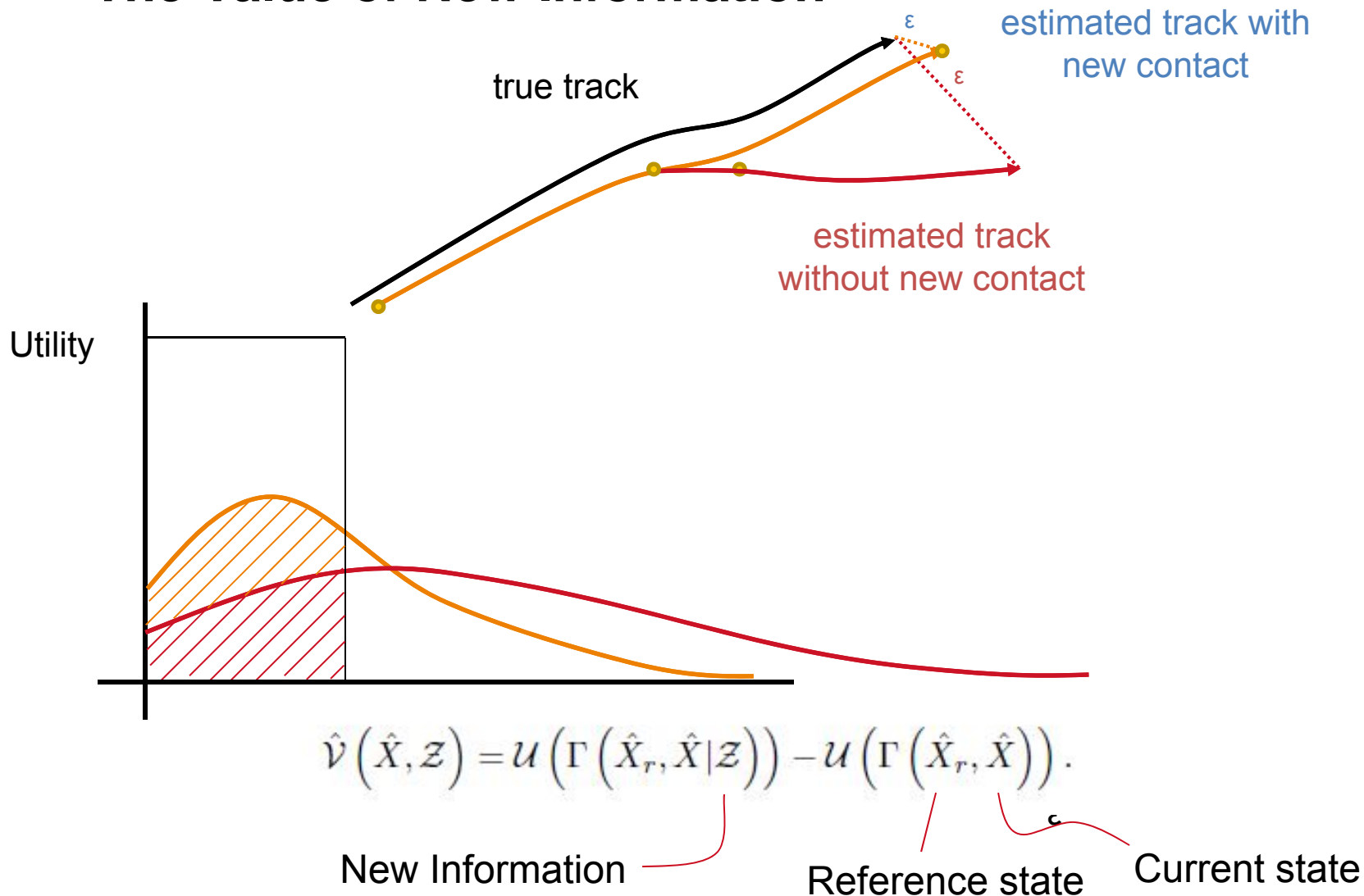
New Information

Reference state

Current state



The Value of New Information





Cost of communication

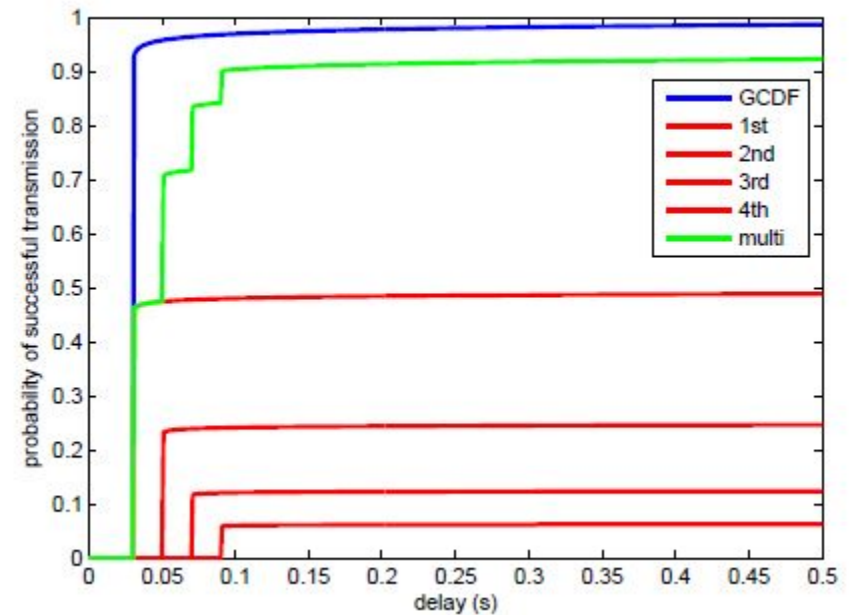
Expected delay distribution

$$\hat{P}(t, \mathcal{Z}, \mathcal{S}) = \sum_n^N p(\mathcal{S}, n) \prod_{a=1}^{|\mathcal{S}|} \mathcal{G}(t - \Delta t_{\text{det}}(a, n)).$$

Expected cost of communication

$$\hat{c}(n, \mathcal{Z}, \mathcal{S}) = Fn \frac{TB \sum_{a=1}^{|\mathcal{S}|} P(a)}{T_{\text{tot}} P_{\text{tot}} B_{\text{tot}}},$$

$$\hat{C}(t, \mathcal{Z}, \mathcal{S}) = \sum_{n=1}^N \hat{c}(n, \mathcal{Z}, \mathcal{S}),$$





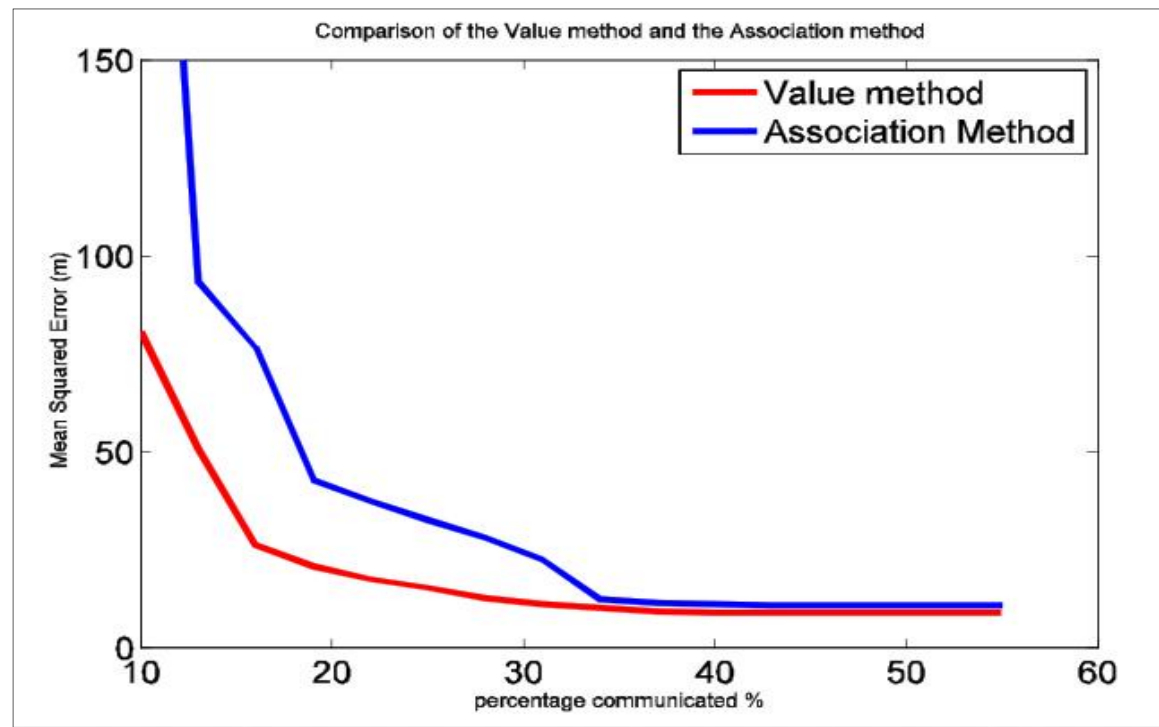
Expected Reward Function

$$\hat{\mathcal{R}}(\hat{X}, \mathcal{Z}, \mathcal{S}) = \hat{\mathcal{V}}(\hat{X}, \mathcal{Z}) - \hat{\mathcal{C}}(t, \mathcal{Z}, \mathcal{S}).$$

$$\hat{\mathcal{R}}(\hat{X}, \mathcal{Z}, \mathcal{S}) = \sum_{n=1}^m P(\Delta t_n, \mathcal{Z}, \mathcal{S}) \left(\hat{\mathcal{V}}(\hat{X}, \mathcal{Z}) - \hat{\mathcal{C}}(\Delta t_n, \mathcal{Z}, \mathcal{S}) \right) - P_f \sum_{n=1}^m \hat{\mathcal{C}}(\Delta t_n, \mathcal{Z}, \mathcal{S}).$$



The Value of new Information





Applications next year

- › Cooperative Adaptive Cruise Control
- › Dynamic Map of traffic including floating car data
- › Task Group Awareness for Command and Control
- › Awareness in autonomous underwater, surface and air vehicles
- › Camera Network for Surveillance
- › Smart Grids



Questions ?



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