

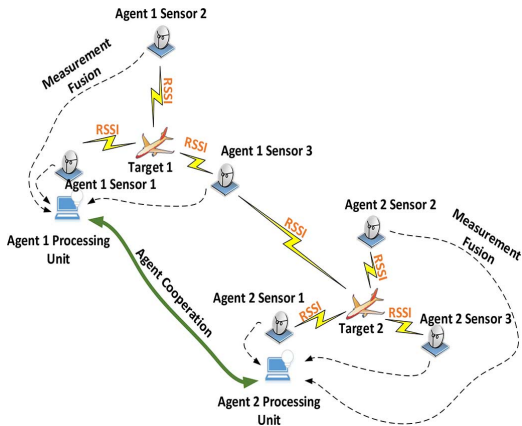
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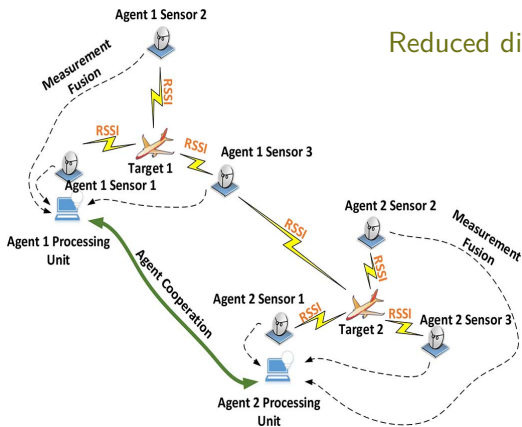
RSSI-Based Multi-Target Tracking by Cooperative Agents Using Fusion of Cross-Target Information

Jonathan P. Beudeau, Mónica F. Bugallo, and Petar M. Djurić



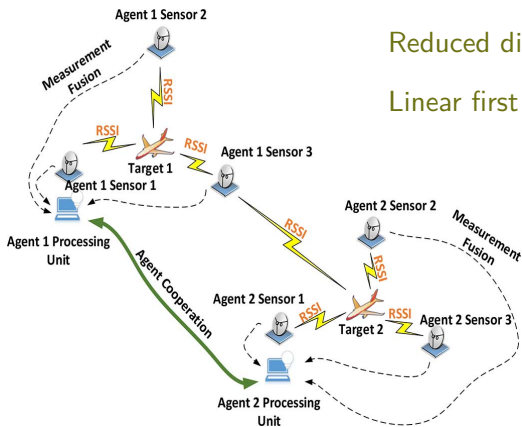
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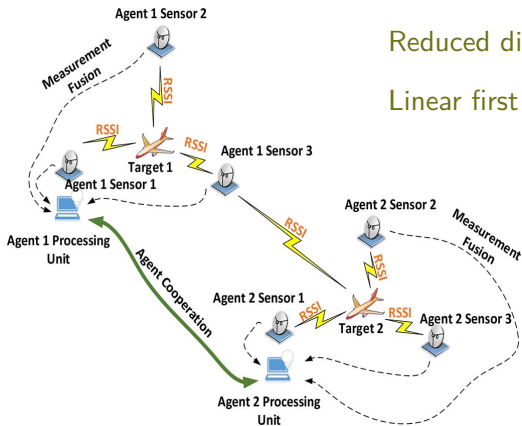


Reduced dimensionality

Linear first order state evolution model

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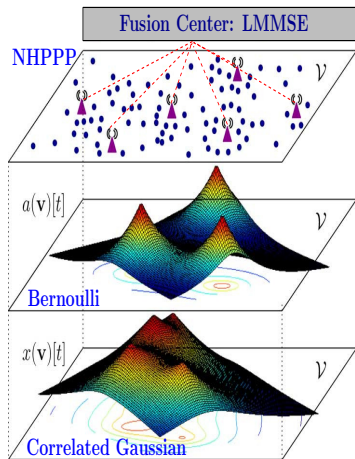
Linear first order state evolution model

Particle filter

Unscented Kalman filter

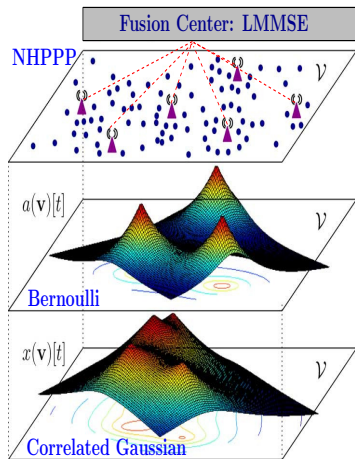
Optimized Random Deployment of Energy Harvesting Sensors for Field Reconstruction in Analog and Digital Forwarding Systems

Teng-Cheng Hsu, Y.-W. Peter Hong, and Tsang-Yi Wang



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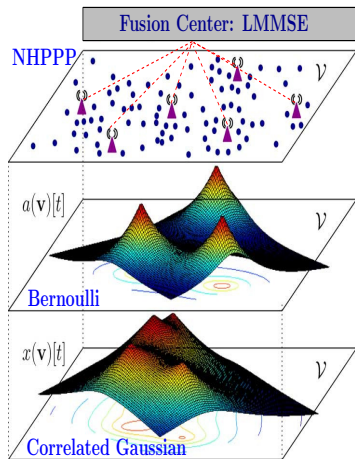


Optimal sensor densities and the energy thresholds for field partition

Minimize MSE upper bound with a constraint on mean number of sensors

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Analog Forwarding:

Condensation method for solving GP

Digital Forwarding:

Alternating optimization algorithm

Block-Sparsity-Induced Adaptive Filter for Multi-Clustering System Identification

Shuyang Jiang and Yuantao Gu

$$d_n = \mathbf{x}_n^T \mathbf{s} + v_n$$

Diagram illustrating the system identification equation $d_n = \mathbf{x}_n^T \mathbf{s} + v_n$. The terms are labeled as follows:

- Input**: Points to \mathbf{x}_n^T
- Noise**: Points to v_n
- Observation**: Points to d_n
- System**: Points to \mathbf{s}

Sparsity model: Simplified Ising model

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Analysis under Gaussian assumption:

- ▶ Steady-State Performance
- ▶ Optimal Group Partition Size
- ▶ Faster convergence than l_0 -LMS

Orthogonal Matching Pursuit With Thresholding and its Application in Compressive Sensing

Mingrui Yang, and Frank de Hoog

$$f = \phi a, \|a\|_0 = k$$

Initialization: $r_0 := f, x_0 := 0, \Lambda_0 := \emptyset, s := 0.$

while $\|r_s\|_2 > t\|f\|_2$ **do**

Find an index i such that

$$|\langle r_s, \phi_i \rangle| \geq t\|r_s\|_2.$$

Update the support:

$$\Lambda_{s+1} = \Lambda_s \cup \{i\}.$$

Update the estimate:

$$x_{s+1} = \arg \min_z \|f - \Phi_{\Lambda_{s+1}} z\|_2.$$

Update the residual:

$$r_{s+1} = f - \Phi_{\Lambda_{s+1}} x_{s+1};$$

$s = s + 1;$

end while

Significantly **LESS** complexity

NO performance degrading
under suitable choice of threshold

Other Papers

- ▶ **Joint Channel Estimation and Data Detection in MIMO-OFDM Systems: A Sparse Bayesian Learning Approach**
 - ▶ R. Prasad, C. R. Murthy, and B. D. Rao
- ▶ **Capacity Analysis of One-Bit Quantized MIMO Systems With Transmitter Channel State Information**
 - ▶ J. Mo and R. W. Heath
- ▶ **A Novel Decomposition Analysis of Nonlinear Distortion in OFDM Transmitter Systems**
 - ▶ L. Yiming and M. O'Droma
- ▶ **Distributed Kalman Filtering With Quantized Sensing State**
 - ▶ D. Li, S. Kar, F. E. Alsaadi, A. M. Dobaie, and S. Cui