# Journal watch Journal Watch - IEEE Transactions on Signal Processing, Oct(2)/Nov(1) 2013

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November 9, 2013

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### Dynamic Dictionary Algorithms for Model Order and Parameter Estimation

### Authors: Christian D. Austin, Joshua N. Ash, and Randolph L. Moses

**Ohio State University** 

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- Low-order parametric model estimation from noisy models
- Dictionary based model estimation
  - Parameter space is discretized into K parameter samples
  - Evaluate a parametric function at a parameter sample for all measurements
  - Static dictionary
    - \* Estimates quantized to the sample spacing
    - Finer spacing leads to large dictionaries and higher inter-column correlations
    - ★ Dynamic dictionary sizes try to avoid the above issues
- Training free dynamic dictionary algorithms
  - Penalty based algorithm: repulsion penalty function that controls the parameter spacing
    - \* LASSO type objective function along with an extra penalty term,  $\mu g(\theta_{\mathbf{k}} \theta_{\mathbf{j}})$ , where g() can be  $1/\|z\|_2$  for example
  - Constraint based algorithm: Directly constrain the parameter distance
- Analysis shows that dynamic algorithms overcomes parameter estimation bias induced by quantization

## Improving Physical Layer Secrecy Using Full-Duplex Jamming Receivers

#### Authors: Gan Zheng, Ioannis Krikidis, Jiangyuan Li, Athina P. Petropulu and Bjorn Ottersten

Univ. of Luxemburg, Univ. of Cyprus, State Univ of NJ - Rutgers

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- Using Full-Duplex MIMO concepts in implementing PHY layer secrecy
- Practical approaches to PHY layer secrecy degrade the decoding capability of ED
  - Multiple Antenna at Transmitter Artificial noise (AN)
  - External help to jam the ED Cooperative Jamming (CJ)
  - No external sources Multiple transmissions from source (iJAM)
- Full-Duplex transmissions to jam the ED
  - Considered practical models of self-interference (LI model)
  - One Tx ant, one Rx ant and single ant ED
    - \* Closed form expressions for power allocation for the receiver
    - Full power not utilized (Saturation due to LI)
  - Destination with multiple antennas
    - \* System is no longer interference limited
    - Shown that optimal jamming covariance matrix is rank-1 and proposed algorithms to find it

## Convergence and Applications of a Gossip-Based Gauss-Newton Algorithm

#### Authors: Xiao Li and Anna Scaglione

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- Distributed optimization via diffusion
  - Combination of descent step with network diffusion steps
  - Distributed Gauss-Newton Algorithm for NLLS optimization
- Setup
  - Let  $g(\underline{x}) : \mathbb{R}^N \to \mathbb{R}^M$
  - Objective function is  $g^{T}(\underline{x})g(\underline{x})$
  - Multiple agents in the system. Each agent has access to partial information about g(x)
  - Compute <u>x</u> that minimized above objective function in a distributed fashion
- Each agent solves the problem "locally"
- Information required for computing descent direction is exchanged using gossiping
  - Time varying network graph considered
  - Each agent combines information from neighbors in weighted fashion and the weight matrix depends upon the topology during the exchange epoch
  - Static exchange model and Randomized exchange model considered
- The proposed GGN algorithm is applied for PSSE applications

## Enhanced Sparse Bayesian Learning via Statistical Thresholding for Signals in Structured Noise

#### Authors: Martin Hurtado, Carlos H. Muravchik and Arye Nehorai

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- Sparse reconstruction with multiple measurements (MMV) using Bayesian framework
- Apart from additive noise, a structured noise component is considered
  - Interference from other users (Clutter in radar, reverberation in active sonar etc.)
  - Modeled as Σ<sup>K</sup><sub>k=1</sub> Z<sub>k</sub>**u**<sub>k</sub>, where Z<sub>k</sub> are (known) regression matrices and **u**<sub>k</sub> ~ N(0, Σ<sub>u</sub>)
  - $\Sigma_u$  is unknown and treated as nuisance parameters and estimated from data
- Standard SBL framework
  - Component variances (hyper-parameters) are used to control which dictionary items are active
  - EM-algorithm is used to numerically compute these hyper-parameters
  - Pruning" to reduce the parameter space progressively
- A new statistical decision based pruning procedure is proposed
  - Test has CFAR like properties
  - Does not depend upon power of the signal, interference nor the noise

Nearest-Neighbor Distributed Learning by Ordered Transmissions Stefano Marano, Vincenzo Matta, and Peter Willett Univ. of Salerno, Italy; Univ. of CT

#### Dynamic Compressive Sensing of Time-Varying Signals Via Approximate Message Passing J. Ziniel, Phil Schniter Ohio State Univ.

Efficient Parametric Signal Estimation From Samples With Location

Errors Sumeet Kumar, Vivek K Goyal and Sanjay E. Sarma MIT

Semi-Blind Receivers for Joint Symbol and Channel Estimation in

Space-Time-Frequency MIMO-OFDM Systems Kefei Liu, Joao Paulo C. L. da Costa, H. C. So, and Andr L. F. de Almeida

Univ. of Hong Kong; Univ. of Brasilia; Federal Univ. of Ceara, Brazil

Decentralized Sparsity-Regularized Rank Minimization: Algorithms

and Applications M. Mardani, C. Mateos, G. B. Giannakis Univ. of Minnesota