# Journal Watch (ArXiv Watch (Machine Learning/Information Theory)

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## Social Teaching: Being Informative vs. Being Right in Sequential Decision Making (arXiv:cs.IT)

#### Authors: Joong Bum Rhim and Vivek K Goyal MIT

- Framework of sequential decision making with social teaching
  - Effect of agent's decision/action on subsequent agents
  - Compared to prior work, non-herding framework has been considered
  - Previous agent can be considered as *advisers* for the next agent
- An agent has access to (unbounded) private signal, biased (incorrect) prior probability and access to previous agent's decision (No access to the prior probabilities of other agents)
- Each agent uses decision rule optimized for her own belief (in contrast to adjusting decisions to minimize Bayes risk for last agent)
- The interest is in last agents decision (and the rule)

#### Conclusions

- Each agent updates it's belief based on previous agents action
- It is the probability of each hypothesis conditioned on the decision of previous agents
- A counter intuitive conclusion is that having a wrong belief is not always bad !!
  - For N=2, the optimal belief update is not equal to true prior probability

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• Best Advisers are necessarily open-minded people

Structure estimation for discrete graphical models: Generalized covariance matrices and their inverses (arXiv: stat.ML)

> Authors: Po-Ling Loh and Martin Wainwright UC Berkley

- For Gaussian graphical models, zeros in the inverse of covariance matrix indicate missing edges in the graph structure
- For Non-Gaussian this is not generally true
- In this paper, this linkage of inverse of covariance matrix with graph structure is explored and some results are proven for certain graph structures in context of exponential family of distributions
  - Multinomial distribution has been treated mainly but many important results generalize to exponential family
  - Instead of covariance matrix an augmented covariance matrix is considered (i.e. include higher order interaction terms)
  - One result is that, for a tree graphs the inverse of generalized covariance is always graph-structured
  - Some results are also proved for arbitrary graphs that are augmented with graph triangulations
- This explains that why the graphical selection methods like graph-Lasso ( that are derived based on Gaussian Models) also work for other distributions

## Distributed Sparse Signal Recovery For Sensor Networks (arXiv:cs.IT)

Authors: Stacy Patterson, Yonina C Eldar, Idit Keidar Technion, Haifa, Israel

- Distributed algorithm for sparse signal recovery
- Setup
  - Each sensor is making a compressive measurement
  - Individual measurement matrices are not known to different sensors
  - Emphasis is on reducing the communication amongst different nodes
- Based on Iterative Hard Thresholding

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$$\mathbf{x}_{t+1} = \mathcal{T}_k(\mathbf{x}_t - \alpha \mathbf{A}^T(\mathbf{b} - \mathbf{A}\mathbf{x}_t))$$

- Communication cost is reduced based on solutions to the distributed top-K problem
  - Send the highest component to the central node which decides when it has received top K items
- Caution: Not a complete work yet, lot of items left unexplained !

## Low-rank Matrix Completion using Alternating Minimization (arXiv: stat.ML)

Authors: Prateek Jain, Praneeth Netrapalli, Sujay Sanghavi MSR India, UT Austin

- Theoretical analysis of Alternate Minimization algorithm used for low-rank matrix sensing (recovery) and completion methods
- Recovery with matrix sensing
  - We have *d* measurements of type:  $b_i = tr(\mathbf{A}_i^H \mathbf{M})$
  - Need to recover *M* from *b*,  $rank(M) = k \ll m$  and with  $d \ll mn$
  - Find **X** s.t.  $\mathcal{A}(\mathbf{X}) = b$ , rank(**X**)  $\leq k$
  - Let  $\mathbf{X} = UV^T$ , then  $\min_{U \in \mathbb{R}^{m \times k}, V \in \mathbb{R}^{n \times k}} \|\mathcal{A}(UV^T) b\|_2^2$
  - The algorithm finds approximate solution to above by alternating between finding an optimal *U* and *V* by keeping the other fixed
  - Note that U and V are much smaller than X
- Assuming that *linear map* A satisfies equivalent RIP, theoretical guarantees have been provided for convergence of above algorithm

#### Optimal Classification in sparse Gaussian Graphic Model

Authors: Yingying Fan, Jiashun Jin, and Zhigang Yao USC, CMU, EPFL

### On the Diversity-Multiplexing Tradeoff of Unconstrained Multiple-Access Channels

Authors: Yair Yona, Meir Feder Tel-Aviv Univ

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