数学与系统科学研究院 计算数学所学术报告

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## 报告题目:

Nonconvex ADMM: Convergence and Applications (by Yu Wang, Wotao Yin, and Jinshan Zeng)

<u>邀请人</u>: 刘歆 副研究员

<u>报告时间</u>: 2015 年 12 月 16 日(周三) 上午 10:00~11:00

<u>报告地点</u>: 科技综合楼三层 **311**报告厅

## Abstract:

ADMM has been surprising us with numerical success on many non-convex optimization problems, which include, but not limited to, the minimization of Lq quasi-norm, Schatten-q quasi-norm, (0<q<1), SCAD, bi-linear, and bi-convex functions, as well as those subject to normalization and matrix manifold constraints.

We study when ADMM converges on minimizing a nonconvex, possibly nonsmooth, objective function, f(x 1,...,x p,y), subject to linear equality constraints that couple x  $1, \dots, x$  p,y, where p > 0 is an integer. Our ADMM sequentially updates the primal variables in the order x 1,...,x p,y, followed by updating the dual variable. We separate the variable y from x\_i's since y has a special role in our analysis. Our results provide sufficient conditions for this nonconvex ADMM to converge with two, three, or more blocks, as they are special cases of our model. By applying our analysis, we show, for the first time, that several ADMM algorithms applied to solve nonconvex models in statistical learning, optimization on manifold, and matrix decomposition are guaranteed to converge.

ADMM has been regarded as a variant to the augmented Lagrangian method (ALM). However, we present a simple nonconvex and no example to illustrate how ADMM converges but ALM (with a fixed penalty parameter) diverges.

欢迎大家参加!