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Sparse Regression at Scale: Convex and Mixed Integer Programming Perspectives

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Via Zoom (session information will be e-mailed to subscribers)
Virtual Reception After Colloquium

ABSTRACT

Many fundamental high-dimensional statistics (aka sparse learning) estimators can be naturally expressed as discrete optimization problems, posing computational challenges. Recently, mixed integer programming methods have been shown to be promising candidates in formulating and solving (to optimality) small/moderate instances of the best-subset selection problem for instances much larger than what was considered possible in the statistics community. This sheds interesting insights into the statistical properties of best-subsets. Furthermore, compared to specialized L1-based methods, current high-performance integer programming methods rely on commercial solvers that are harder to deploy at scale in an open-source (non-academic) environment. To this end, I will discuss (a) methods to compute approximate solutions for best-subset type problems at scales comparable to current L1-based methods; and (b) tailored branch and bound methods to solve these problems to optimality. These algorithms employ techniques from first order methods in convex optimization and integer programming.