Accepted Manuscript

On the Flexibility of Block Coordinate Descent for Large-Scale Optimization

Xiangfeng Wang, Wenjie Zhang, Junchi Yan, Xiaoming Yuan, Hongyuan Zha

 PII:
 S0925-2312(17)31270-5

 DOI:
 10.1016/j.neucom.2017.07.024

 Reference:
 NEUCOM 18712



To appear in: Neurocomputing

Received date:19 January 2017Revised date:30 April 2017Accepted date:10 July 2017

Please cite this article as: Xiangfeng Wang, Wenjie Zhang, Junchi Yan, Xiaoming Yuan, Hongyuan Zha, On the Flexibility of Block Coordinate Descent for Large-Scale Optimization, *Neurocomputing* (2017), doi: 10.1016/j.neucom.2017.07.024

This is a PDF file of an unedited manuscript that has been accepted for publication. As a service to our customers we are providing this early version of the manuscript. The manuscript will undergo copyediting, typesetting, and review of the resulting proof before it is published in its final form. Please note that during the production process errors may be discovered which could affect the content, and all legal disclaimers that apply to the journal pertain.

On the Flexibility of Block Coordinate Descent for Large-Scale Optimization

Xiangfeng Wang¹, Wenjie Zhang¹, Junchi Yan^{1,2}, Xiaoming Yuan³ and Hongyuan Zha^{1,4}

¹ Shanghai Key Lab for Trustworthy Computing, School of Computer Science and Software Engineering, East China Normal University, Shanghai, China ² IBM Research - - China

³ Department of Mathematics, Hong Kong Baptist University, Hong Kong, China
 ⁴ School of Computing, Georgia Institute of Technology, Atlante, USA

Abstract

We consider a large-scale minimization problem (not necessarily convex) with non-smooth separable convex penalty. Problems in this form widely arise in many modern large-scale machine learning and signal processing applications. In this paper, we present a new perspective towards the parallel Block Coordinate Descent (BCD) methods. Specifically we explicitly give a concept of so-called two-layered block variable updating loop for parallel BCD methods in modern computing environment comprised of multiple distributed computing nodes. The outer loop refers to the block variable updating assigned to distributed nodes, and the inner loop involves the updating step inside each node. Each loop allows to adopt either Jacobi or Gauss-Seidel update rule. In particular, we give detailed theoretical convergence analysis to two practical schemes: Jacobi/Gauss-Seidel and Gauss-Seidel/Jacobi that embodies two algorithms respectively. Our new perspective and behind theoretical results help devise parallel BCD algorithms in a principled fashion, which in turn lend them a flexible implementation for BCD methods suited to the parallel computing environment. The effectiveness of the algorithm framework is verified on the benchmark tasks of large-scale ℓ_1 regularized sparse logistic regression and non-negative matrix factorization.

Keywords: Block Coordinate Descent, Large-Scale Optimization, Jacobi, Gauss-Seidel

Preprint submitted to Elsevier

Download English Version:

https://daneshyari.com/en/article/6865316

Download Persian Version:

https://daneshyari.com/article/6865316

Daneshyari.com