November 2020 / VOL. 108 / NO. 11

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#### SPECIAL ISSUE

#### OPTIMIZATION FOR DATA-DRIVEN LEARNING AND CONTROL

Edited by U. A. Khan, W. U. Bajwa, A. Nedić, M. G. Rabbat, and A. H. Sayed

#### **1869** A General Framework for Decentralized Optimization With First-Order Methods

By R. Xin, S. Pu, A. Nedić, and U. A. Khan

|INVITED PAPER| This article presents a general framework for distributed firstorder methods, for minimizing a finite sum of functions, that is applicable to both undirected and directed graphs. Such problems have found a significant interest in control, signal processing, and estimation, and more recently in large-scale data science and machine learning problems.

#### 1890 Decentralized Zeroth-Order Constrained Stochastic Optimization Algorithms: Frank-Wolfe and Variants With Applications to Black-Box Adversarial Attacks

By A. K. Sahu and S. Kar

**|INVITED PAPER|** This article presents an overview of the recent work in the area of distributed zeroth-order optimization, focusing on constrained optimization settings and algorithms built around the Frank–Wolfe framework.

#### **1906** Stochastic Quasi-Newton Methods

By A. Mokhtari and A. Ribeiro

**|INVITED PAPER|** This article discusses recent developments to accelerate convergence of stochastic optimization through the exploitation of second-order information and shows applications in the context of predicting the click-through rate of an advertisement displayed in response to a specific search engine query.

#### **1923** Primal–Dual Methods for Large-Scale and Distributed Convex Optimization and Data Analytics

By D. Jakovetić, D. Bajović, J. Xavier, and J. M. F. Moura

**(INVITED PAPER)** This article focuses on the augmented Lagrangian method (ALM), where a constrained optimization problem is solved with a series of unconstrained subproblems, with respect to the original (primal) variable, while the constraints are controlled via dual variables.

## **1939** Distributed Optimization, Averaging via ADMM, and Network Topology

By G. França and J. Bento

**INVITED PAPER** This article reviews recent research quantifying the influence of the network topology on the convergence behavior of distributed methods and further explores the connections between the alternating direction method of multipliers (ADMM) and lifted Markov.

#### DEPARTMENTS

#### **1863** SCANNING THE ISSUE

Optimization for Data-Driven Learning and Control By U. A. Khan, W. U. Bajwa, A. Nedić, M. G. Rabbat, and A. H. Sayed

#### 2083 SCANNING OUR PAST

The Bell Versus Gray Telephone Dispute: Resolving a 144-Year-Old Controversy By B. L. Brown

#### 2097 FUTURE SPECIAL ISSUE/SPECIAL SECTIONS



On the Cover: This month's cover image illustrates the use of optimization as a tool for efficiently processing big data for learning and control applications.

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#### SPECIAL ISSUE: Optimization for Data-Driven Learning and Control

#### **1953** Distributed Optimization for Robot Networks: From Real-Time Convex Optimization to Game-Theoretic Self-Organization By H. Jaleel and J. S. Shamma

**INVITED PAPER** This article presents a collection of state-of-the-art results for distributed optimization problems arising in the context of robot networks, with a focus on two special classes of problems, namely, real-time path planning for multirobot systems and self-organization in multirobot systems using game-theoretic approaches.

#### **1968** Variance-Reduced Methods for Machine Learning

By R. M. Gower, M. Schmidt, F. Bach, and P. Richtárik

|INVITED PAPER| This article discusses stochastic variance-reduced optimization methods for problems where multiple passes through batch training data sets are allowed.

#### **1984** Scaling-Up Distributed Processing of Data Streams for Machine Learning

By M. Nokleby, H. Raja, and W. U. Bajwa

|INVITED PAPER| This article reviews recently developed methods that focus on distributed training of large-scale machine learning models from streaming data in the compute-limited and bandwidth-limited regimes, with an emphasis on convergence analysis that explicitly accounts for the mismatch between computation, communication, and streaming rates, and that provides sufficient conditions for order-optimum convergence.

#### 2013 Advances in Asynchronous Parallel and Distributed Optimization

By M. Assran, A. Aytekin, H. R. Feyzmahdavian, M. Johansson, and M. G. Rabbat |INVITED PAPER| This article focuses on asynchronous parallel and distributed methods for large-scale optimization problems in machine learning, where the processors may maintain an inconsistent view of the optimization variables.

#### 2032 Time-Varying Convex Optimization: Time-Structured Algorithms and Applications

By A. Simonetto, E. Dall'Anese, S. Paternain, G. Leus, and G. B. Giannakis

**|INVITED PAPER|** This article reviews a broad class of algorithms for timevarying optimization with an emphasis on both algorithmic development and performance analysis.

#### 2049 Graph Learning Under Partial Observability

By V. Matta, A. Santos, and A. H. Sayed

**INVITED PAPER** This article examines the network tomography problem and considers the question: How much information can one glean about the underlying graph topology by observing the behavior of certain distributed optimization methods over the graph nodes?

#### 2067 Accelerated First-Order Optimization Algorithms for Machine Learning

#### By H. Li, C. Fang, and Z. Lin

**INVITED PAPER** | This article provides a comprehensive survey of accelerated first-order methods with a particular focus on stochastic algorithms and further introduces some recent developments on accelerated methods for nonconvex optimization problems.

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