Introduction to the Course

Optimization Techniques (ENGG*6140)

School of Engineering, University of Guelph, ON, Canada

Course Instructor: Benyamin Ghojogh Winter 2023

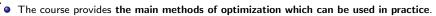
Introduction of the Instructor and Students

Let us know each other by introducing ourselves!

Feel free to let us know (if you would like):

- Your name
- Your major
- Whether you are studying MEng, MASc, or PhD
- What it your goal for taking this course? What are your expectations from this course?
- How much you know about optimization?
- How was your Christmas vacation? :)

Introduction of the Course



- (a) We start with **preliminaries** including sets, norms, functions, local/global minimizer, classical methods derivatives, gradient, Jacobian, Hessian, convexity of sets, and convexity of functions.
 - We introduce the standard problems (e.g., convex problem, linear programming, quadratic programming, semidefinite programming, etc).
 - We cover linear programming (the simplex algorithm) and integer linear programming for continuous and discrete linear problems, respectively.
 - We introduce the Karush-Kuhn-Tucker (KKT) conditions along with the Lagrangian function and the method of Lagrange multipliers.
 - Ne cover unconstrained and constrained first-order optimization which are gradient methods.
 - The unconstrained and constrained second order optimization techniques, including the interior-point method, are covered in order to be able to solve all convex optimization problems (this method also works fairly well on nonconvex problems).
 - We cover important metaheuristic methods such as genetic algorithm, particle swarm optimization, and simulate annealing.
 - (a) [If time allows,] we will also go through distributed optimization (such as ADMM) in order to solve complex multivariate optimization problems.

The Tentative Schedule of Weeks

Unconstrained an constrained first-order optimization: gradient descent, line-search, steepest descent, backpropagation and neural networks, stochastic gradient descent, proximal mapping, proximal gradient method.

projected gradient method

Topics:

Week 1 - 2	,		Week 8 - 9	
Topics:	٠	<u>Preliminaries</u> (sets, norms, functions, gradient, Jacobian, Hessian, etc). Convexity of sets and functions. Introduction to standard optimization problems.	Topics:	Unconstrained and constrained second-order optimization: unconstrained Newton's method, equality constrained Newton's method, interior-point and barrier methods, quasi Newton's methods
Week 2 - 3			Week 10	
Topics:		Linear Programming, the Simplex method, and Integer linear programming.	Topics:	Metaheuristic optimization: genetic algorithm, particle swarm optimization, simulate annealing, etc.
Week 4 - 5			Week 10	
Topics:		Karush-Kuhn-Tucker (KKT) conditions, the Lagrangian function, dual variables, primal and dual feasibility, the dual problem, and the method of Lagrange multipliers	Topics:	If time allows: distributed optimization; alternating optimization, dual decomposition methods, augmented Lagrangian method, Alternating Direction Method of Multipliers (ADMM).
			Week 11 - 12	
Topics:		The midterm exam.	Topics:	Group final project presentations.
Week 6 - 7				

Course Materials

- Lecture notes will be provided to you.
- Our tutorial paper: [1] [Link]
- Additional resource for interested students: Convex Optimization I, and II) at the YouTube channel of Stanford University. [Link]
- Additional books:
 - Boyd's textbook (focusing on convexity and the second-order methods) [2] [Link]
 - ► The books of Yurii Nesterov (mostly focused on first-order optimization):
 - ★ Introductory lectures on convex optimization: A basic course [3] [Link]
 - ★ Lectures on Convex Optimization [4] [Link]
- Research articles in the literature

Course's Websites

- Classes will be online via Zoom. The link will be provided in the CourseLink.
- Discussion chats and questions will be in Microsoft Teams group of the course. The students will be added to the Teams group.
- The course's website is: [Click here] https://bghojogh.github.io/pages/uoguelph/engg-6140-w23/
- I will probably upload the videos of the classes to my YouTube channel [Click here].

 I will eliminate personal information of students (such as when they introduce themselves) in the videos.

Course Evaluation

- Assignments (20%): Assignments will be posted on CourseLink along with the due dates. They are performed individually. We will probably have three assignments.
- The midterm exam (30%): Date will be around week 6. Details to be discussed in class.
- Course project (40%)
 - Date: Week 6 11
 - More details will be discussed in class. Report will be electronic submission due in Coursel ink
 - The number of people in each group will be announced in the class.
 - Pick a topic after 6 weeks.
 - Send me the title and objectives to approve it.
- Group Presentation (10%):
 - Date: Week 11 12
 - During class time
- Bonus points: participation in class, participation in the discussions, asking questions, and answering questions.

Course's Goal

- Don't worry much about your marks!
- Focus on understanding the materials of the course.
- Our goal is to learn the important practical optimization techniques, so you can use them in both your **industrial projects** and **academic research**.
- About theory and practice:
 - ▶ We will learn **some theory** to understand why these methods work.
 - ▶ We will also learn how to use the methods in practice for practical usage.

Ask Questions!

- Please ask questions whenever you do not understand something.
- Let the class be discussion-based. I do not want to be the sole speaker. We are gonna learn all together.

References

- [1] B. Ghojogh, A. Ghodsi, F. Karray, and M. Crowley, "KKT conditions, first-order and second-order optimization, and distributed optimization: Tutorial and survey," arXiv preprint arXiv:2110.01858, 2021.
- [2] S. Boyd, S. P. Boyd, and L. Vandenberghe, *Convex optimization*.

 Cambridge university press, 2004.
- [3] Y. Nesterov, *Introductory lectures on convex optimization: A basic course*, vol. 87. Springer Science & Business Media, 2003.
- [4] Y. Nesterov et al., Lectures on convex optimization, vol. 137. Springer, 2018.