E1 260 Optimization for Machine Learning and Data Science

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Course information

Instructor

Sundeep Prabhakar Chepuri email: spchepuri AT iisc.ac.in

Class schedule

Tuesdays and Thursdays 2.00 – 3.30 pm (Online via MS Teams)

(Pls. be on mute when you don't have to ask anything)

Attendance is mandatory (watch recorded videos only for revising!)

Course objective

- > Cover optimization techniques suitable for problems that frequently appear in the areas of *data* science, machine learning, communications, and signal processing.
- Focus on the *computational*, <u>algorithmic</u>, and <u>implementation</u> aspects

Grading and course requirements

- Prerequistite: Basic linear algebra, probability, and knowledge of a programming language like Python (Google CoLab) to conduct simulation exercises.
- ➤ This is a 3:1 course (expect about 3 hours of work each week, almost)
- Five assignments (problem and programming set): 10% each

> Two projects: 20% each

Final Exam (take-home 24 hour exam): 10%

Grading	%
5 x HW	50
Project 1	20
Project 2	20
Final Exam	10
Total	100

> No mid-term exam

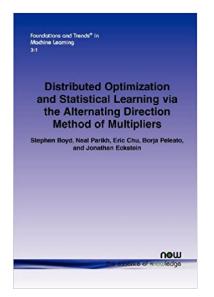
Sessional (50%)	Final (50%)
HW 1,2,3, Project 1	HW 4,5,Project 2 and final exam

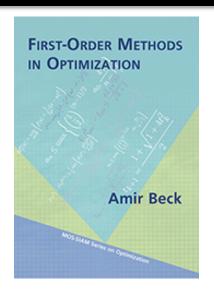
References

Optimization for Machine Learning Lecture Notes CS-439, Spring 2021

> Bernd Gärtner, ETH Martin Jaggi, EPFL

> > May 4, 2021





and papers....

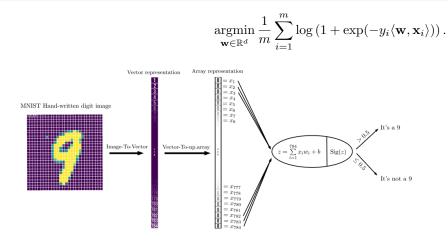
Unconstrained optimization problem

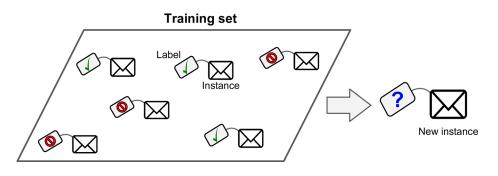
minimize
$$f(\mathbf{x})$$

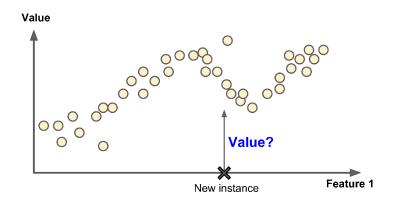
subject to $\mathbf{x} \in \mathbb{R}^d$

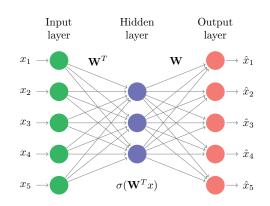
- $ightharpoonup f: \mathbb{R}^d o \mathbb{R}$ is the objective or the cost function. The value $f(\mathbf{x})$ is the cost of the decision \mathbf{x}
- > x is the decision variable

Unconstrained optimization problems



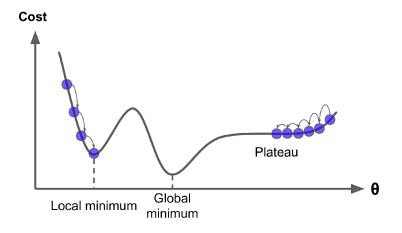


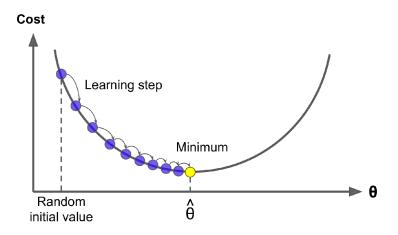




Theory of convex functions

2	24/08/2021	Theory of convex functions
3	26/08/2021	Theory of convex functions
4	31/08/2021	Theory of convex functions
4	02/09/2021	Theory of convex functions





Unconstrained optimization

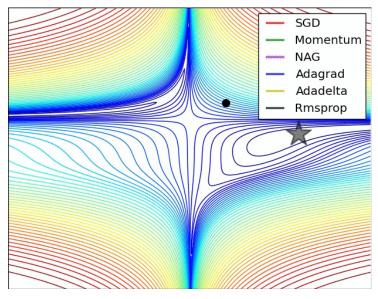
5	07/09/2021	Gradient descent for unconstrained problems
	09/09/2021	Gradient descent for unconstrained problems
6	14/09/2021	Accelerated gradient method
O	16/09/2021	Accelerated gradient method
7	21/09/2021	Proximal gradient descent
7	23/09/2021	Mirror descent
8	28/09/2021	Subgradient methods
O	30/09/2021	NO CLASS
9	05/10/2021	Stochastic Gradient Descent
9	07/10/2021	SGD and friends
10	12/10/2021	SGD and friends
10	14/10/2021	Project 1 presentations
11	19/10/2021	HOLIDAY
11	21/10/2021	SGD and friends

Many APIs, libraries are available

optimizer = keras.optimizers.RMSprop(lr=0.001, rho=0.9)

In this course:

- We look inside these algorithms
- Study why, when, and how fast they work



Constrained optimization problem

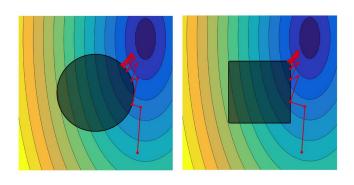
12 26/10/2021 Projected gradient descent
28/10/2021 Frank-Wolfe

minimize $f(\mathbf{x})$ subject to $\mathbf{x} \in \mathcal{C}$

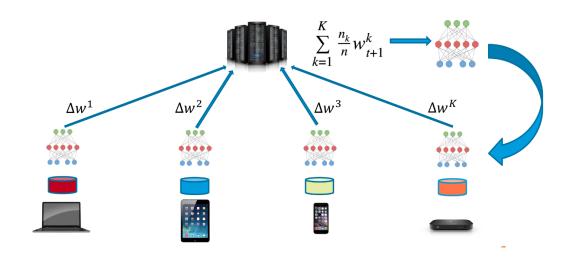
- $ightharpoonup f: \mathbb{R}^d \to \mathbb{R}$ is the objective or the cost function. The value $f(\mathbf{x})$ is the cost of the decision \mathbf{x}
- > x is the decision variable
- $\triangleright \mathcal{C}$ is the feasible set

Theorem: The Frank-Wolfe method using standard step sizes $\gamma_k=2/(k+1)$, $k=1,2,3,\ldots$ satisfies

$$f(x^{(k)}) - f^* \le \frac{2M}{k+2}$$



Distributed optimization: ADMM and federated learning



14	09/11/2021	ADMM
14	11/11/2021	ADMM
15	16/11/2021	Federated Learning
10	18/11/2021	Federated Learning

Submodular optimization

Optimization of a set function

minimize
$$F(X)$$
 subject to $X \subset \{1, 2, \cdots, n\}$ and $|X| \leq K$



16	23/11/2021	Submodular optimization
10	25/11/2021	Submodular optimization

Non-convex optimization



minimize rank(**M**) subject to $m_{ij} = z_{ij}$ for all $(i, j) \in \Omega$

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Customer 3	•	2		4	•	•	•	•	2	
Customer 4	3	•		•	•	•	•	•	•	•
Customer 5	5	5	•	•	4	•		•	•	•
Customer 6	•	•	•	•		2	4	•	•	•
Customer 7	•	•	5	•		•		3	•	•
Customer 8	•	•	•	•	•	2	•	•	•	3
Customer 9	3	•	•	•	5	•	•	5	•	•
Customer 10	•	•	•	•	•	•	•	•	•	•

47	30/11/2021	Non-convex optimization
17	02/12/2021	Non-convex optimization

Week	Date		Project/HW announcement	Due dates
1	10/08/2021	Introduction to the course : logistics, optimization in ML and data science, applications		
	12/08/2021	Mathematical background		
2	17/08/2021	Mathematical background		
_	19/08/2021	HOLIDAY		
3	24/08/2021	Theory of convex functions		
ŭ		Theory of convex functions		
4	31/08/2021	Theory of convex functions	HW1	
•	02/09/2021	Theory of convex functions		
5	07/09/2021	Gradient descent for unconstrained problems		
ŭ		Gradient descent for unconstrained problems	Project 1	
6		Accelerated gradient method	HW2	HW1
	16/09/2021	Accelerated gradient method		
7	21/09/2021	Proximal gradient descent		
·		Mirror descent		
8		Subgradient methods		
		NO CLASS		
9		Stochastic Gradient Descent		HW2
· ·	07/10/2021	SGD and friends		Project 1
10		SGD and friends	HW3	
		Project 1 presentations		
11	19/10/2021			
		SGD and friends		
12		Projected gradient descent		HW3
		Frank-Wolfe	HW4	
13		NO CLASS	Project 2	
	04/11/2021			
14	09/11/2021			
	11/11/2021			
15		Federated Learning		
		Federated Learning	HW5	HW4
16		Submodular optimization		
		Submodular optimization		
17		Non-convex optimization		
		Non-convex optimization		HW5
18	07/12/2021	NO CLASS		Project 2