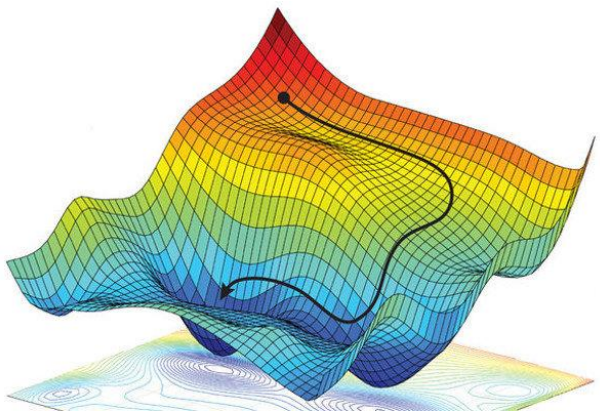


$$\begin{array}{ll} \min_x & f(x) \\ \text{s.t.} & x \in X \subset \mathbb{R}^n \end{array}$$

$$\nabla f(x) = \begin{bmatrix} \frac{\partial f(x)}{\partial x_1} \\ \vdots \\ \frac{\partial f(x)}{\partial x_n} \end{bmatrix}$$



ESE 415

Optimization

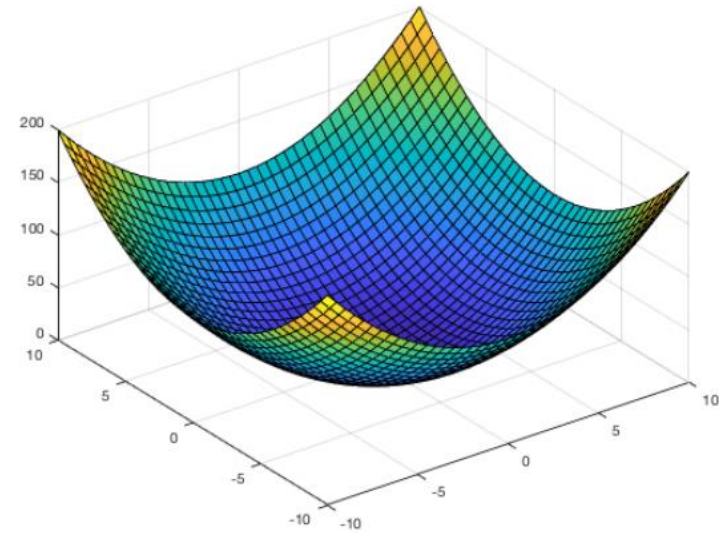
Instructor: Shen Zeng

ESE Department, Washington University in St. Louis

$$f(x + \alpha d) \approx f(x) + \alpha (\nabla f(x))^T d.$$

Mathematical Optimization in a Nutshell

$$\begin{array}{ll} \text{minimize} & f(\mathbf{x}) \\ \text{subject to} & \mathbf{x} \in \mathcal{X} \end{array}$$



- Optimization variable: $\mathbf{x} = (x_1, \dots, x_n)$
- Objective function: $f : \mathbb{R}^n \rightarrow \mathbb{R}$
- Constraint set: $\mathcal{X} = \{\mathbf{x} \mid \mathbf{h}(\mathbf{x}) = 0\} \cap \{\mathbf{x} \mid \mathbf{g}(\mathbf{x}) \leq 0\}$
- Optimal solution \mathbf{x}^* has smallest value of f among all vectors that satisfy the constraints

The goal in mathematical optimization is to find the smallest value of a function under certain constraints.

Optimization is finding extensive use everywhere

- Optimization ideas and principles can be found in almost all engineering domains nowadays. They are also deeply rooted in several mathematical and physical disciplines.
- Optimization is still an active research area with many open questions

The goal of **ESE 415** is to help you understand and apply the basics of optimization theory!

By the end of the semester, you should be able to:

- recognize and formulate problems as optimization problems
- derive and characterize optimal solutions
- develop code for optimization algorithms

Rough List of Specific Topics:

- optimality conditions
- convex sets and functions
- constrained and unconstrained optimization
- optimization algorithms
- analysis of optimization algorithms
- examples and applications

Organizational Matters

Anything relevant to ESE 415 can be found at the course website:

<https://wustl.instructure.com/courses/132084>

No fixed textbook requirement, but a list of recommended textbooks can be found at the above course website.



Grading policy:

- Homework: 35%. Biweekly; theoretical and computational exercises.
Late submission policy: deduction of 20 points per late day.
- Midterm Exam: 30%. Monday, **September 30, 2024**, taken during class hours.
- Final Exam: 35%. Tuesday, **December 17, 2024**, 10:30am-12:30pm, Lopata 101.

List of Assistants to the Instructor & Graders

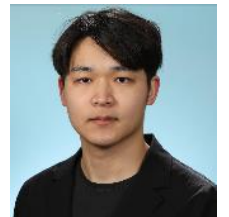
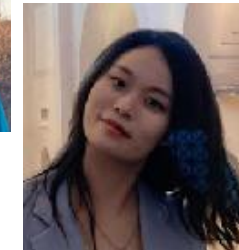
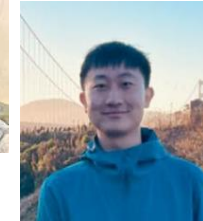
Head Assistant to the Instructor:

- Nia Hodges: h.nia@wustl.edu (2nd year Systems Science & Mathematics Ph.D. student, ESE)



Assistants to the Instructor & Graders:

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ESE 415 Office Hours:

Every **Tuesday, Thursday, and Friday** from **2:30-3:30pm** via Zoom (access through Canvas).

Brief history of optimization

Antiquity: Optimization questions pertaining to geometry

- Euclid (300bc): minimal distance between a point and a line
- Heron (100bc): light travels between two points through the path with shortest length when reflecting from a mirror

17-18th century: Analytical treatments via calculus

- Kepler (1615): Optimal dimensions of wine barrel
- Fermat (1636): derivative of function vanishes at extreme points
- Newton (1660s) and Leibniz (1670s) create mathematical analysis that forms the basis of calculus of variations*

*huge significance in mathematical and physical research areas

Brief history of optimization

19th century: the first optimization algorithms are presented

- Gauss & Legendre (~1800): least squares method
- Cauchy (1847): gradient method
- Gibbs shows that chemical equilibrium is an energy minimum

20th century: algorithmic research expands as computers become ubiquitous

- von Neuman and Morgenstern (1944): dynamic programming for solving sequential decision problems
- Dantzig (1947): simplex method for solving Linear Programs
- Kuhn & Tucker (1951): optimality conditions for nonlinear problems. Karush (1939): Similar conditions
- Nesterov (1983): accelerated gradient method
- Karmarkar (1984): polynomial time algorithm for Linear Programs;
creates a boom of interior point methods.
- **Modern era**: non-smooth analysis, stochastic optimization, optimal control of dynamical systems, ...