

# Optimization in 15 minutes

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# What is Optimization and Mathematical Programming?

- Optimization is any process of improving
  - Optimize your writing process
  - Optimize your cooking
  - Optimize your PhD
- Mathematical Programming is a sub-field which deals with problems with a specific form:

$$\begin{array}{ll} \text{minimize} & f_0(x) \\ \text{subject to} & f_i(x) \leq 0 \\ & g_j(x) = 0 \end{array}$$

# Mathematical Programming

- Q: Why is everything in the form:
  - minimize  $f(x)$
  - subject to  $f_i(x) \leq \bullet$
  - $g_j(x) = \bullet$
- Convenient form to see the structure of the problem
- 50 years in convention
- This form covers most design problems, except:
  - Problems that require a trade-off analysis
  - Multi-objective problems
  - (These problems can usually be formulated to a sequence of "minimize" "subject to" problems)

# Different types of Optimization Problems

$$\begin{aligned} &\text{minimize} && f_0(x) \\ &\text{subject to} && f_i(x) \leq 0 \\ &&& g_j(x) = 0 \end{aligned}$$

	domain	$f_0$	$f_i$	$g_i$	Optimality condition
Linear Programming	Continuous, convex	linear	linear	linear	x
Geometric Programming	Continuous, convex	posynomial	posynomial	linear	x
Quadratic Programming	Continuous, Convex	Quadratic, convex	linear	linear	x
Convex Programming	Continuous, Convex	convex	convex	linear	x
Nonlinear Programming	General, connected	general	general	general	
Integer Programming	Integer	general	general	general	

# Linear Programming

$$\begin{array}{ll} \text{minimize} & c^T x \\ \text{subject to} & Ax \leq b \\ & Gx = h \end{array}$$

- Everything is linear above
- Easiest problem to solve (aside from unconstrained):
  - Can solve very large problems (sometimes 1 million +)
  - Simplex methods or interior point methods
- Software:
  - MATLAB linprog, CVX, Ip\_solve (for large problems)

# Quadratic Programming

$$\begin{array}{ll} \text{minimize} & x^T A^T A x + b^T x \\ \text{subject to} & Gx \leq h \end{array}$$

$$\begin{array}{ll} \text{minimize} & \|Ax - b\|^2 \\ \text{subject to} & Gx \leq h \end{array}$$

- Can solve large designs
- Most cases are from fitting problems
  - Fitting data ( $\|Ax - b\|^2$  is a measure of error)
  - Optimal sensor locations
- Software:
  - CPLEX, LOQO, Mosek, Matlab fmincon

# Geometric Programming

$$\begin{array}{ll} \text{minimize} & f_0(x) \\ \text{subject to} & f_i(x) \leq 1 \\ & x \geq 0 \end{array}$$

- $f_0$  and  $f_i$  are posynomials:

$$f(x) = \sum_k c_k \prod_j x_j^{\alpha_{jk}}, \quad c_k > 0, \quad \alpha_{jk} \in \mathbb{R}$$

- A change of variables  $x = e^y$  makes the problem a convex optimization problem
- Software:
  - CVX, GGPLab, Mosek,

# Examples of Posynomials

- Gate delay:  $D_i(x) = x_i^{-1} \sum_{j \in \text{fo}(i)} x_j$
- Area:  $A(w, l, h) = wlh$
- Opamp currents:  $I_7 = \frac{W_5 L_8}{L_5 W_8} \Rightarrow \begin{cases} W_5 L_8 L_5^{-1} W_8^{-1} I_7^{-1} \leq 1 \\ W_5^{-1} L_8^{-1} L_5 W_8 I_7 \leq 1 \end{cases}$
- Opamp, biases, communications theory, wire-sizing
  - see Boyd et. al, A Tutorial on Geometric Programming, Hershenson & Boyd, etc.



# Integer / Discrete Programming

$$\begin{array}{ll} \text{minimize} & f_o(x) \\ \text{subject to} & f_i(x) \leq \delta \\ & x \in \{0, 1, 2, \dots\} \end{array}$$

- Generally very difficult to solve exactly
  - NP Complete class
- Solvers use heuristics (usually not exact!):
  - Branch and Bound
  - Sequential rounding
  - SAT
- Usually get good results, but not **optimal** results

# A History of Mathematical Programming

**1940.** Computers emerge Dantzig invents Linear Programming (LP) and the Simplex Algorithm to solve LP

**1960's- 1970's.** Most theoretical results and algorithms are developed

**1979.** Khachiyan shows that LP's are polynomial time

**1984-1990s.** Interior Point Algorithms developed (reliable algorithms for mathematical programming); Emergence of fast and affordable computers

**1990's.** Convex Optimization is a hot topic

**2000's.** Robust Optimization is a hot topic

# What problems can I solve in a couple hours? (a **very** rough guide)

Exact Optimum:

Unconstrained Smooth Convex Optimization	~10-100 million
Linear Programming	~1 million
Convex Smooth Optimization	~10,000 variables
Convex non-smooth	~1,000~1 million variables
Non-convex continuous	10~1,000 variables
Integer Programming	10~100 variables

Approximate Solutions:

- Pretty much anything!
  - with varying degrees of success

# What was not covered in this talk

- o Algorithms to perform mathematical programming
- o Modeling real world problems as a convex programming problems
- o Re-formulating problems for faster solving
- o Statistical optimization
- o Integer Programming
- o Convex analysis

# References

- Convex Optimization and Interior Point Methods
  - Convex Optimization, Boyd, Vandenberghe
- Nonlinear Programming Algorithms
  - Numerical Optimization, Nocedal and Wright
- First-order methods and Lagrangian Methods
  - Nonlinear Programming, Bertsekas
- Pre-1990 Optimization
  - Introductory Lectures on Convex Optimization, Nesterov
- Geometric Programming
  - “A Tutorial on Geometric Programming”, Boyd et al
- Stochastic Programming
  - “A Tutorial on Stochastic Programming”, Shapiro and Philpott

# References

- Geometric Programming for analog design
  - Hershenson, Boyd (assorted papers)