

Motivation

Operations Research in R
Stefan Feuerriegel

Outline

1 Introduction to Optimization

2 Motivation

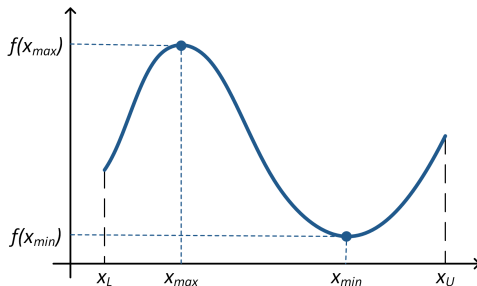
Outline

1 Introduction to Optimization

2 Motivation

Mathematical Optimization

- ▶ Optimization uses a rigorous **mathematical model** to determine the most efficient solution to a described problem
- ▶ One must first identify an **objective**
 - ▶ Objective is a quantitative measure of the performance
 - ▶ Examples: profit, time, cost, potential energy
 - ▶ In general, any quantity (or combination thereof) represented as a **single number**



Applications of Optimization

- ▶ **Management**

- ▶ Determining product portfolios
- ▶ Location planning
- ▶ Investments decisions

- ▶ **Game theory**

- ▶ Comparing players' strategies

- ▶ **Logistics**

- ▶ Finding optimal routes and schedules

- ▶ **Design decisions**

- ▶ Constructing processes, plants and other equipment

- ▶ **Operation**

- ▶ Adjustment to changes in environmental conditions, production planning, control, etc.

- ▶ **Mathematical modeling**

- ▶ Parameter estimation
- ▶ Model discrimination

Optimization Problem

Optimization is the **minimization or maximization** of a function subject to (s. t.) constraints on its variables

Notation: General form

$$\min_{\mathbf{x}} f(\mathbf{x}) \quad \text{s. t.} \quad h(\mathbf{x}) = 0 \\ g(\mathbf{x}) \leq 0$$

with

- ▶ $\mathbf{x} \in \mathbb{R}^n$ as the **variable, unknown or parameter**
- ▶ **Objective function** $f : D \rightarrow \mathbb{R}, D \subseteq \mathbb{R}^n$
- ▶ **Equality constraints** $h : D_h \rightarrow \mathbb{R}^l, D_h \subseteq \mathbb{R}^n$
- ▶ **Inequality constraints** $g : D_g \rightarrow \mathbb{R}^k, D_g \subseteq \mathbb{R}^n$

Properties of Optimization Problems

Objective	Linear, quadratic, non-linear, etc.
Constraints	Equality and inequality
Variable types	\mathbf{x} can be continuous, integer, mixed
Direction	$\min_{\mathbf{x}} f(\mathbf{x}) \Leftrightarrow \max_{\mathbf{x}} -f(\mathbf{x})$
Bounds	Lower $\mathbf{x}_L \leq \mathbf{x}$ or upper $\mathbf{x} \leq \mathbf{x}_U$
Dimension	One dimensional if $n = 1$, or multi-dimensional if $n > 1$
Optima	Isolated, local or global nature

Classification of Optimization Problems

- ▶ Linear Programming (LP)

- ▶ Objective function and constraints are linear
- ▶ $\min_{\mathbf{x}} \mathbf{c}^T \mathbf{x}$ s. t. $A\mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq 0$

- ▶ Quadratic Programming (QP)

- ▶ Objective function is quadratic and constraints are linear
- ▶ $\min_{\mathbf{x}} \mathbf{x}^T Q \mathbf{x}$ s. t. $A\mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq 0$

- ▶ Non-Linear Programming (NLP): objective function or at least one constraint is non-linear

- ▶ Integer Programming (IP): all variables are discrete

- ▶ Mixed Integer Programming (MIP)

- ▶ Continuous and discrete variables
- ▶ Problem can be linear (MILP) or non-linear (MINLP)

Classification of Optimization Problems

- ▶ **Dynamic Optimization:** solution is a function of time
- ▶ **Stochastic Optimization**
 - ▶ Model cannot be fully specified, but has uncertainties with confidence estimates
 - ▶ Optimize expected performance given uncertainty

Question

- ▶ What type is the following optimization problem?

$$\max_{x,y} 3x + y^2 \quad \text{s.t.} \quad x + y < 10 \text{ and } y \in \{1, 2, 4, 8\}$$

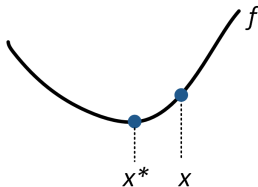
- ▶ MIP
 - ▶ MILP
 - ▶ MINLP
- ▶ Visit webpage with course quiz.

Optimal Solution

- x^* is a **global minimum** if $x^* \in D$ and

$$f(x^*) \leq f(x) \quad \text{for all } x \in D$$

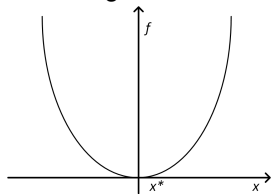
→ Global minimizers are desired, though often one has only local knowledge of f



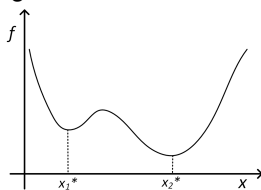
Optimal Solution

Examples of optimal solutions:

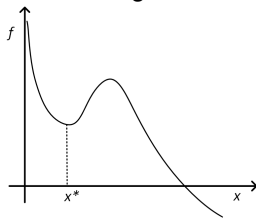
Isolated global solution



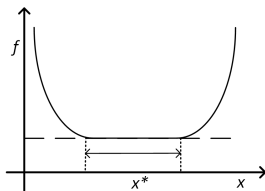
One global, two local solutions



A local but no global solution



Many non-isolated, global solutions



Optimization Procedure

Formulation and solution of optimization problems usually follows:

- 1 Analysis of environment to determine the variables of interest
- 2 Definition of optimality criteria as an objective function with (additional) constraints
- 3 Formulation as a mathematical model with degrees of freedom
- 4 Numerical optimization to find a solution
- 5 Verification of the solution through sensitivity analysis (with respect to the assumptions made in the problem formulation)

Outline

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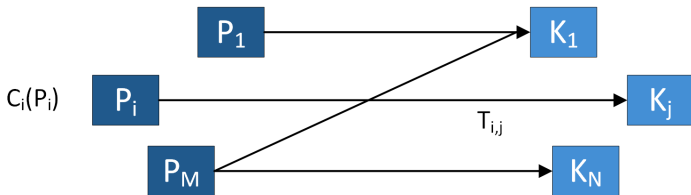
2 Motivation

Production Scheduling

Problem

- ▶ Given plants $i = 1, \dots, M$ where each manufactures P_i goods
- ▶ Each plant has a maximal output O_i
- ▶ Each plant manufactures at a capacity-specific cost $C_i(P_i)$, which gives the cost as a function of the production
- ▶ Each customer $j = 1, \dots, N$ requests C_j goods

Objective function: find the optimal production schedule such that the manufacturing and shipment costs are minimized



Portfolio Optimization

Problem

- ▶ Investor wants to **invest** money such that it **maximizes the investor's utility**
- ▶ Utility U depends on **daily return** μ and **risk** σ^2
- ▶ Given risk taking κ , then $U(\mu, \sigma^2) = \mu - \frac{\kappa}{2}\sigma^2$

Objective function: **Maximize** $U(\mu, \sigma^2)$ among a range of stocks s_1, \dots, s_N

\mathbf{c}^T	s_1	s_2	s_3
μ	0.2	0.5	0.1

Q	s_1	s_2	s_3
s_1	0.1	0.02	0.02
s_2	0.02	0.1	0.02
s_3	0.02	0.02	0.1

$$\Rightarrow \max_{\mathbf{x}} \mathbf{c}^T \mathbf{x} - \frac{\kappa}{2} \mathbf{x}^T Q \mathbf{x} \quad \text{s.t.} \quad x_i \geq 0 \text{ and } \sum_i x_i = 1$$

Portfolio Optimization in R

```
library(quadprog) # load necessary library

kappa <- 4 # set risk taking
# objective function
c <- c(0.02, 0.05, 0.01)
Q <- matrix(c(0.1,0.02,0.02, 0.02,0.1,0.02,
              0.02,0.02,0.1), nrow=3)
# constraints
A <- matrix(c(1,1,0,0, 1,0,1,0, 1,0,0,1), nrow=4)
b <- c(1, 0, 0, 0)

sol <- solve.QP(kappa/2*Q, c, t(A), b, meq=1) # call solver
sol$solution # ratio of stocks in portfolio

## [1] 0.2916667 0.4791667 0.2291667

sol$value # minimum value of objective function

## [1] 0.01729167
```


Outlook

1 Introduction to R

2 Advanced R

- ▶ Programming prerequisites
- ▶ Visualize optimization routines

3 Numerical Analysis:

- ▶ Mathematical prerequisites to derive and formalize optimization routines

4 Optimization in R:

- ▶ Use of built-in optimization routines