



DV_2
W leaps over B

$$ISD + NR \rightarrow LM$$

$$\lambda \frac{\partial f}{\partial x} + (1-\lambda) \frac{\partial f}{\partial x} = -\lambda \nabla f$$

$$[\lambda \frac{\partial f}{\partial x} + (1-\lambda) \frac{\partial f}{\partial x}] \Delta x = -\lambda \nabla f$$

Nonlinear
Uncertainty
Aberrations
Case Studies
Direct Search
Leapfrogging
Take Away
Convergence
Penalty
Stochastic
Constraint
Levenberg
Encompassing
Analysis
Structural
Evolutionary
Metaheuristics

uncertainty on DV

$$\epsilon_{x^*} \approx \sqrt{\sum \frac{\partial^2 f}{\partial x^2} \epsilon_c}$$
 due to model
coefficient
values
 $J = \sum \text{desirables} - \sum \text{undesirables}$
 constraints



Surface & Path
Analysis

$$\left. \begin{aligned} COF_1 & \leq P(0, n) \\ u & \leq P(0, n) \end{aligned} \right\}$$

Engineering Optimization

Applications, Methods, and Analysis

R. Russell Rhinehart



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R. Russell Rhinehart

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Preface

Introduction

Optimization means seeking the best outcome or solution. It is an essential component of all human activities. Whether personal or professional, we seek best designs, best choices, best operation, more bang for the buck, and continuous improvement.

Here are some professional examples: Minimize work events that lead to injury while remaining economically competitive. Structure workflow to maximize return on investment. Design an antenna that maximizes signal clarity for a given power. Define a rocket thrust sequence to maximize height. Determine the number of parallel devices to minimize initial cost plus future risk.

Here are some personal examples: Seek the best vacation experience for the lowest cost. Minimize grocery bill, but meet desires for nourishment and joy of eating. Set the family structure for raising children that leads to well-adjusted, happy, productive outcomes, but keep within the limits of personal resources. Create a workout regime that leads to fastest and most attractive muscle development, with no injury, and in balance with other desires in quality of life.

Optimization is not just an intellectual exercise; although often, solving the challenge is as rewarding as completing a Sudoku puzzle. We implement the optimized decision. Accordingly, within any application it is essential to completely and appropriately assess the metrics that quantify “best.” If the description of what you want to achieve is not quite right, then the answer will also be wrong, which the implementation will reveal in retrospect. You want to get it right prior to implementation. So, part of this book is about development of the optimization objective.

After the objective is stated, we desire an efficient search logic to find the best solution, with precision and with minimal computational and experimental effort. So, other parts of this book are about the optimizer—the search logic, or algorithm.

Both aspects are essential, and I find that most books on optimization focus on the intellectually stimulating mathematics of the algorithms. So, I offer this book to provide a balance of essential topics to the application to guide user choices in structuring the objective, defining constraints, choosing convergence, choosing initialization, etc. Some will be disappointed that this book is not a compendium of every optimization algorithm conceived by mankind. However, others will value the application perspective.

Also, I find that most people using optimization as a tool did not have a course on it while in school. So, I have written this book in a style that I hope facilitates self-study by those who need to understand optimization applications while keeping it fit for use as a graduate-school course textbook.

Key Points

Here are a few essential aspects of optimization:

Point 1: Although optimization offers the joys of solving an intellectual puzzle, it is not just a stimulating mathematical game. Optimization applications are complicated, and the major challenges are the clear and complete statement of:

- 1) The objective function (OF—the outcome you wish to minimize or maximize)
- 2) Constraints (what cannot or should not be violated, or exceeded)
- 3) The decision variables (DV—what you are free to change to seek a minimum)
- 4) The model (how DVs relate to OF and constraints)
- 5) The convergence criterion (the indicator of whether the algorithm has found a close enough proximity to the minimum or maximum and can stop or needs to continue)
- 6) The DV initialization values
- 7) The number of starts from randomized locations to be confident that the global optimum has been found
- 8) The appropriate optimization algorithm (for the function aberrations, for utility, for precision, for efficiency)
- 9) Computer implementation in code Oh yes,
- 10) The mathematics of the optimization algorithm (understanding this is also important)

This book seeks to address all 10 aspects, not just the 10th.

Point 2: Do not study. Learning is most effective if you integrate the techniques into your daily life. You will forget the material that you memorized in order to pass a test. Since this book provides skills that are essential for both personal and career life, I want you to take the techniques with you. I want this book to be useful in your future. Although memorization and high-level mathematical analysis are both elements of the book, understanding the examples and doing of the exercises is more important. To maximize the impact of this material, you need to integrate it into your daily life. You need to practice it.

Oh, I see I omitted a comma in the first sentence of the paragraph above. It should have been “Do, not study.” Learn by doing. After you read a section and think you understand it, see if you can implement it. Of course, the comma “error” above was intentional to wake up curiosity about the message.

Point 3: Optimization is universal to all engineering, business, science, computer science, and technology disciplines. Although primarily written for engineering applications, this introductory book is designed to be useful for all those seeking to apply optimization in all fields.

Point 4: The implementation of optimization requires computer programming, which for many is an aggravation. To help the reader, I currently have, and plan to support, a website that offers to any visitor optimization software and examples. Visit www.r3eda.com. The “r3” in the address is my initials, and the appended “eda” means “enabling data analysis.” Seeking to maximize ease of use and accessibility, the programs are written as VBA macros for MS Excel. VBA is not the fastest-computing environment, nor does it have the best scientific data processing functions. However, it has been adequately functional for all of my applications, and if you need something better, the code can be translated. This book provides a VBA primer (Appendix F) for those needing the help in accessing and modifying the code. The programs on the r3eda site solve many of the examples in this book.

Book Aspirations

Readers should be pleased with their ability to:

- Understand and use the fundamental mathematical techniques associated with optimization
- Define objective functions, decision variables, models, and constraints for a variety of optimization applications
- Develop, modify, and program simplified versions of the more common optimization algorithms
- Understand and choose appropriate methods for:
 - Constrained optimization
 - Global optimization
 - Convergence criteria
 - Surface aberrations
 - Stochastic applications
- Understand diverse issues related to optimizer desirability
- Explore, contrast, and evaluate the performance of optimization algorithms and user choices of convergence criteria, numerical derivative estimation, threshold, constraint handling method, parameter values, etc. with respect to precision, user convenience, and other measures of optimizer desirability
- Apply optimization algorithms to case studies relevant to the reader's career
- Continue learning optimization methods from texts, reports, Internet postings, and refereed journal articles

Optimization is the name for the procedure for finding the best choices. “Procedure,” “best,” and “choices” are separate aspects, and the user must understand each to be able to appropriately define the application. And each aspect has a large range of options.

Procedure

This relates to the method used to find the optimum:

- In process or device design, for example, the choices could be the equipment specifications (type, materials, size), and the evaluation of best in the design could be to minimize capital cost with a constraint on reliability. With mixed continuous, discrete, and class variables as the choices, a direct search algorithm might be the best optimizer.
- Alternately, in scheduling a rocket thrust to reach a desired height, the stage choices might be height, best might be evaluated as minimizing either time or fuel use, and the appropriate algorithm might be dynamic programming.
- Another example is characterized as the traveling salesman problem in which the objective is to determine a sequence of locations to visit to minimize travel distance. Here the choice is the sequence, and the best sequence might be impacted by a priority of visits, expenses, wasted time, etc. The procedure might use the random keys method to convert a sorted list of rational numbers into the sequence.
- As a final contrasting example, in model-predictive control, the objective might be to minimize time to move a response to a set point while penalizing excessive manipulated variable moves while avoiding constraints; and the choices might be the future sequence of manipulations. If the penalties are quadratic, the appropriate algorithm might be a gradient-based procedure.

Best

Within optimization terminology, the definition of best for a specific application (and the method for calculating a value to quantify best) is variously termed the cost function or the objective function (OF). It is the function that returns a value representing an assessment of goodness. Best usually means minimize undesirable aspects and/or maximize desirable aspects, and the OF can represent a wide range of metrics related to economics, safety, time, resource conservation, quality, deviation, probability, etc. But best might mean to minimize a worst-case feature (min the max, or min–max), such as finding a path through mountains that minimizes the steepest ascent or finding a process design that minimizes the worst-case outcome (risk).

Defining the appropriate OF is situation specific, and often it is the key challenge in an optimization application. The user needs to clearly understand the complex situation and realize that a first statement of the OF usually embodies a superficial understanding. Subsequent analysis of the results will lead to an evolution of the OF. For example, a challenge might be to choose the best pipe diameter in a process design. A smaller diameter means a less expensive pipe and lower in-pipe inventory cost, but it means a larger pump. An initial OF choice might be to minimize capital. However, reconsideration from a business investment view might reveal that operating costs associated with pumping power and maintenance are also important issues, and perhaps net present value (NPV) is a right way to combine initial capital with future expenses. Then, reconsideration might bring understanding of the sensitivity of the optimum solution to uncertainty in the “givens,” which will lead to a refinement of the OF to represent the 95% worst case of the NPV in a Monte Carlo analysis, making it a stochastic function. Risk might then be perceived as an additional issue, and the OF might be split into a multi-objective version (risk and NPV) that provides a non-dominated set of solutions for a user to select a best for the particular situation. Finally, the user might realize that pipe comes in discrete diameter values and that the pipe diameter is not a continuous-valued number. This application might have evolved from an initial simple deterministic (textbook example) case to a complicated application, classified as mixed integer, stochastic, and multi-objective.

This book will address how to develop the OF and will show examples from a wide range of applications.

Choices

The choices a user has (you may call these inputs, decisions, degrees of freedom, or independent variables) to change things toward the best outcome are termed decision variables (DVs).

In regression DVs are the model coefficient values. In product design DVs could be polymer type, blend concentration, operating a process, color, or shape. In process design DVs could be the pipe diameters and pump sizes. In flying aircraft, the DVs would be the stick, throttle, and pedal positions. In control and scheduling, in operating a business, the DVs would be the future plan for both the timing and magnitude of the actions. Alternately, the DVs might be the coefficients in an equation that would define the future schedule for control actions. Again, there are many possibilities for how to choose the DVs; and the user choices impact efficiency of solution, the appropriate optimizer algorithm, and precision of solution. The book will also address such issues.

Organization

As with most books on engineering optimization, this one describes and develops many common algorithms. It starts with simple univariate (line) search approaches and progresses to multivariable and multiplayer approaches. I do not seek to cover every version, or every method. I use archetypical

examples of the many approaches, from which readers can grasp the concepts of other methods. Book topics include gradient based, Newton's, and blends such as Levenberg–Marquardt. They include surrogate function methods to characterize the “surface” such as successive quadratic. They include direct searches such as a simple heuristic cyclic, Hooke–Jeeves, and Nelder–Mead. They include multiplayer mimetic approaches of leapfrogging, particle swarm, and genetic algorithms. They include dynamic programming, in which the DVs are the states, and linear programming that takes advantage of certain structures. The book develops the basic techniques and addresses refinements that improve performance, such as quasi-Newton estimates of the Hessian elements, and grid refinement in dynamic programming.

The book provides a guide to match optimization procedures with features of the application such as discontinuities, flat spots, nearly flat spots, constraints, multiple optima, stochastic responses, parameter correlation, etc. Several sections discuss the issues that certain OF features create. Other sections are devoted to the analysis of the optimizers for precision, accuracy, global identity, work to converge, and robustness. Another section reveals sensitivity to user parameters such as contraction and expansion coefficients, thresholds, triggers, etc. A user needs to understand which optimizer is appropriate for which application and how to make the best choice of optimizer parameter values.

The book also addresses choices of convergence criteria that are appropriate for the application and for the optimizer. For example, in choosing thresholds on the DV as the convergence criteria (which is common practice), the user should use propagation of uncertainty to project the DV tolerance on the OF. As a contrasting example, in optimizing results of either experimental outcomes or a Monte Carlo stochastic simulation, the optimizer needs to stop when the noisy response is not making improvement relative to the noise amplitude.

The book is aimed at engineering applications, where optimization is essential for model development, product design, process and device design, dynamic system control, or system operation. However, the applications of optimization extend into all aspects of our lives from purchasing choices to investment choices, to career planning, and to dressing for a desired impact. The reader should be able to extend the guidance of the book to both personal and other professional decisions.

Rationale for the Book

Optimization is ages old. Prior to calculus, optimization was empirical, guided by heuristics and experience. Improvement was by a direct search, one that only uses the OF value and not the derivative information. The mathematics of calculus, however, created a new era, and Simpson (1740) extended Newton's root finding (1685) to the derivative of the function to find the optimum. Cauchy's sequential line search appeared in 1847. Modest technique progress continued through about 1944, at which time the power of the digital computer led to both practical applications and an explosion in the development of diverse techniques. In 1955 Levenberg blended “Newton's” with incremental steepest descent to spawn many approaches to using both the gradient and Hessian to guide sequential improvements in the trial solution. Advances continued to capitalize on computational power. Then the 1960s gave rise to mimetic multiparticle algorithms and multi-objective applications.

In that brief historical overview, gradient-based techniques replaced the precalculus era direct search techniques. Gradient-based techniques remain the mainstay of texts. However, the power of the digital computer is permitting new direct search techniques such as particle swarm, genetic programming, and leapfrogging to outperform gradient-based techniques on nonlinear and stochastic

applications with discontinuities—today’s relevant problems. One reason for the book is to promote the use of the new direct search techniques.

Most books on engineering optimization focus on the optimization algorithms. However, most users will not write the code; they will buy it. Of more need for a user is instruction on how to create an appropriate OF, how to choose DVs, how to identify and incorporate constraints, how to define convergence, and how to determine the number of independent starts needed to ensure that the global is found. This book seeks to fill in those application essentials.

I developed and used optimization throughout my initial 13-year career in the industry. However, my college preparation for the engineering career did not teach me what I needed to know about how to create and evaluate optimization applications. I recognized that my fellow engineers, regardless of their *alma mater*, were also underprepared. We had to self-learn what was needed. Recognizing the centrality of optimization to engineering analysis, I have continued to explore its application and technique development in my subsequent 30-year academic career.

This book is based on college and professional training courses that I’ve offered and is a collection of what I consider to be best practices in engineering optimization. It includes the material I wish I had known when starting my engineering career, and I hope the book is useful for the readers.

Target Audience

The examples and discussion presume basic understanding of engineering models, statistics, calculus, and computer programming. This book will have enough details, explicit equation derivations, and examples to be useful either as an introductory course or for self-study.

The book is aimed at a bachelors, or higher, graduate of engineering or a mathematical science (physics, chemistry, statistics, computer science), who has had an undergraduate course in calculus, mathematical models, statistics, and computer programming. However, upper-level undergraduates have been successful in my course. The reader could be either a student or a practicing engineer or scientist.

Presentation Style

In my experience, students cannot grasp the depth of one topic in isolation of the others. Depth in understanding two-dimensional (2-D) OF surface features is required to be able to relate to N -D issues. An initial understanding of the optimization algorithms is required to be able to set up the application OF and DVs. An understanding of the application is required to be able to choose the appropriate convergence criterion and thresholds. Accordingly, I start the book with elementary versions of each of the aspects of optimization in one-dimensional applications, demonstrate the whole of the applications on several case studies to reveal issues, then return to each item in more depth, and demonstrate the improvements of the second-level techniques in 2-D applications, discuss issues, and then extrapolate to N -D implementations.

I offer the reader with software (and access to computer code through my website www.r3eda.com) to execute key operations. Although there are many strong programming environments, the code is written in Excel VBA (Visual Basic for Applications), which is widely accessible. The book includes a listing of the code for the techniques.

A unique feature of the book is the “takeaway” sections associated with the chapters, which summarize the methods of choice using a practical, applications, utility perspective. This is intended as a user’s how-to book grounded in fundamentals, not as a math-analysis-of-the-fundamentals book. However, relevant properties of the optimization problems will be mathematically analyzed, the optimization algorithms will be developed from theory, propagation of uncertainty will be related to choices, and the book contains some proofs related to surface analysis and OF transformations.

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Other authors have provided books that have been very valuable to my understanding. I recommend these publications: Ravindran, Ragsdell, and Reklaitis, *Engineering Optimization—Methods and Applications*, Wiley, 2006; Beveridge and Schechter, *Optimization: Theory and Practice*, McGraw-Hill, 1970; Edgar, Himmelblau, and Lasdon, *Optimization of Chemical Processes*, McGraw-Hill, 2001; Snyman, *Practical Mathematical Optimization*, Springer, 2005; Hillier and Lieberman, *Introduction to Operations Research*, McGraw-Hill, 2001; Nocedal and Wright, *Numerical Optimization*, Springer-Verlag, 1999; and Rao, *Engineering Optimization: Theory and Practice*, 4th Edition, Wiley, 2009.

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