

Nonlinear Programming

Theory, Algorithms, and Applications

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Washington, D.C.**

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To Ginny first,

then to

Jessica

Rachel

Callie

and

John

Preface

The primary purpose of this book is the exposition of algorithms for solving nonlinear programming (optimization) problems. This book also contains the theory of nonlinear optimization, mostly concerned with the characteristics of optimal points. An important subsidiary aim is the presentation of some optimization models of real world problems that can be solved by nonlinear programming methodology. After working in the field of algorithm development for several years, I was disturbed by the number of bad mathematical models of serious problems that were considered in the literature. In 1972 I initiated a course, "Applications of Linear and Nonlinear Optimization Theory," in the Department of Operations Research at the George Washington University in order to teach mathematical modeling, that is, how to do it and its limitations. This is the origin of the discussion in Chapter 1 of characteristics of real world problems that prevent their representation as optimization models. The students are given a structured way in which to analyze application papers and to assess their relevance in solving the underlying problems. The material in Chapter 1, the models that precede each major section of the book, and papers gathered from the published literature constitute the subject matter of the applications course.

The bulk of the material can be used for a traditional course in nonlinear programming. There is a rigorous development of first- and second-order optimality conditions. Algorithms are presented for optimization problems in order of increasing complexity, that is, for single-variable problems, unconstrained n -variable problems, linearly constrained problems, and, finally, nonlinearly constrained problems. For the latter two categories, the components of the algorithms are presented in general and then it is shown how classical methods fit into the general framework. This is a departure from the usual way of presenting algorithms—by author, with each subalgorithmic detail treated as though it were intrinsic to the method, whereas in fact it is usually just a replaceable part.

Some of the newer algorithms, multiplier (augmented Lagrangian) methods, exact penalty function methods, and sequential linearization techniques are

explained with the use of an idealized exact penalty function created by the movement of a particle under different forces. It is my contention that many algorithms can be motivated through the use of simple physical models, and this approach is used in the development of Newton's method for unconstrained and constrained optimization problems.

This book contains several topics not usually covered in books on nonlinear programming. The future of nonlinear programming lies in the integration of numerical techniques for solving systems of equations. Some preliminary attempts to do this and the necessary background for understanding the current literature are contained here.

There is no lack of algorithms for solving nonlinear programming problems, as evidenced by the material in this book and in its references. Probably the single most important thing, in addition to the primitive nature of the use of matrix methods, preventing the automatic solution of nonlinear programming problems is the lack of a computationally oriented way of representing nonlinear functions. Chapter 3 is a summary of a new approach using "factorable" functions to provide the interface between computer-coded algorithms and the algebraic representation of nonlinear programming problems. This new approach has a great variety of implications for other aspects of nonlinear programming. One of these is evidenced in Chapter 18 on solving nonconvex programming problems.

The field of nonlinear programming is vast, and no attempt has been made to include all the developments to date or in progress. It would require a book of several volumes to do this. In some cases, long proofs of important theorems are omitted. Many of the topics covered are included because of my special research interests. Basic or pioneering works are mentioned at the ends of chapters, where a synthesis of the many contributing works is contained.

Much of the research that contributed to this book was sponsored by the Army Research Office, the Office of Naval Research, the Air Force Office of Scientific Research, the Department of Energy, and the National Bureau of Standards. Their help is appreciated.

Many individuals have contributed to portions of the book, and their contributions are noted at the appropriate places. I wish to thank Professor William H. Marlow, Director of the Institute for Management Science and Engineering, for creating an environment in which basic and applied research could be carried out.

Professor Stephen M. Robinson was particularly helpful in the early stages of this book: the material relating to the classical Newton's method in Chapter 7 and several other contributions are due to him and are greatly appreciated.

GARTH P. MCCORMICK

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December 1982*

Use of Cross Referencing in This Book

Within each section important items, such as theorems, equations, and definitions, are numbered sequentially. The numbers of the chapter and section appear at the top of each page; thus 6.3 at the top of a page indicates that the material appearing there is from Section 3 of Chapter 6. When reference is made to an item within the same section, only the item number is used. If the referenced item is in the same chapter but in a different section, both the section and item numbers are used; thus [7.5] means item 5 of Section 7 of the same chapter. When the item referenced is in a different chapter, all three numbers are used; thus equation [2.7.5] indicates equation 5 of Section 7 of Chapter 2.

Notation

x	$(x^1, \dots, x^n)^T$, an $n \times 1$ vector of variables
E^n	Euclidean n space
$f'(x_0)$	The $1 \times n$ derivative vector of the function f evaluated at x_0 , that is, the $1 \times n$ vector whose j th component is $\partial f(x_0)/\partial x^j$
$\nabla f(x_0)$	The $n \times 1$ gradient vector of f at x_0 , that is, the $n \times 1$ vector whose j th component is $\partial f(x_0)/\partial x^j$ [note that $f'(x_0) = \nabla f(x_0)^T$]
$f''(x_0)$	The $n \times n$ Hessian matrix of f at x_0 , that is, the matrix whose i, j th element is $\partial^2 f(x_0)/\partial x^i \partial x^j$
$\nabla^2 f(x_0)$	The $n \times n$ Hessian matrix defined above
$A^\#$	A pseudo-inverse of the matrix A ($AA^\#A = A$)
A^+	The Penrose–Moore generalized inverse of the matrix A

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I

Basics

Chapter 1 has enough material to provide the basis for a course in the applications of nonlinear programming. An introduction to simple nonlinear programming problems using as a basis the classical diet problem begins the book in Section 1.1. Section 1.2 contains a discussion of the characteristics that must be present in real world problems in order for them to be legitimately modeled as optimization problems. In Section 1.3 is an elementary discussion of the mathematical characteristics of optimization problems. With this background a moderately sophisticated reader should be able to understand published nonlinear programming models and much of the algorithmic and theoretical material in this book.

In order to provide the reader with a complete understanding of the proofs of theorems and some of the more complicated algorithms, Chapter 2 contains the necessary mathematical background: functional analysis, convexity, and linear algebra.

Chapter 3 contains a discussion of factorable functions. This is a new topic oriented toward providing a computationally valuable way to represent complicated functions of several variables.

The Nature of Optimization Problems

1.1 GENERAL REMARKS

The study of nonlinear programming begins with algebra. It begins with the notion that it is useful to express a relationship between measurable attributes symbolically rather than writing down all or some of the possible associated values. The relationship between the area of a triangle and its height and base can be expressed by the examples in Table 1.1.1.

Use of algebra simplifies this task by providing a shorthand way of expressing the relationship. Let A denote the area of a triangle, let x denote the base length, and y the height. Then $A = xy/2$.

This is an example of an explicit functional relationship. Such relationships are useful as models of the real world and are helpful in making decisions. The question of how many pounds of grass seed to buy for a triangular plot with a base of 20 ft and a height of 30 ft can be answered by computing the area and knowing the recommended coverage per square foot.

Even this simple mathematical model of the situation is subject to criticism. No plot of ground is actually triangular and slope considerations have not been taken into account. The uneven surface of the ground may require more grass seed in some places, less in others—better to get an experienced gardener, let him look at it, and buy the quantity he suggests. In the absence of an experienced gardener, though, most people would accept the basic model as an aid in deciding on how much grass seed to buy.

Indeed, all mathematical models can be challenged. It is not difficult to present arguments stating that models oversimplify the complexity of the real situation, both with respect to the operating laws of cause and effect and also their inclusiveness; still it is useful in many cases to abstract the important elements of a real world problem and model them. The hard question for the model builder is whether or not the *important* elements of a real world problem can be quantified.

Mathematical model building is not an activity that is carried on by just a few abstractly oriented persons. Most people implicitly develop mathematical

Table 1.1.1 Associated Values of the Area, Height, and Base Length of a Triangle

Length of Base (in.)	Height of Triangle (in.)	Area of Triangle (in. ²)
0.2	7.0	0.7
5.6	10.0	28.0
1.25	4.4	2.75

models and make decisions based on rather simple assumptions. Questions of how long to cook a roast, how much fertilizer to apply to a crop, how long it will take to drive a car from point *A* to point *B*, at what speed to take a picture, and the number of calories to eat in a day are all answered in some form by the use of a mathematical model.

One difference between a good and a bad model is whether or not the proper functional relationships are used. Consideration of the rate of heat transfer and the shape of a roast implies that the cooking time should not vary directly with the number of pounds. The average speed of a car on a trip is not the average of the speeds on its segments. A crop's yield per acre as a function of the fertilizer applied follows an exponential form.

Mathematical models arise from a variety of sources and needs. Tolstoy resorted to the theory of integral calculus to explain how the aggregate of Russian souls could combine to defeat Napoleon.

Mathematical model building has often come under attack, but like the praise of its advocates, the attacks are many times undifferentiated. Obviously in some circumstances model building is relevant, and working with a model can help one predict, suggest controls, or simply visualize phenomena that would not be possible to understand intuitively from the mass of detail in the real world situation. In other cases the models are constructed (it is sometimes felt) to avoid looking at the relevant complexities; these models fail to describe the real world situation accurately. There are many situations that are impossible to model and where functional relationships do not exist. In this book critical questions will be asked whenever models are presented, whether or not development of the model is warranted. The reader should be critical, but should also look for situations where models are warranted.

As a simplification, it can be stated that optimization problems are posed in one of two modes: the descriptive or the prescriptive (normative). As a simple example of the former and to illustrate in concrete terms what an optimization problem is, consider the following.

[1] Example. Among all the triangles with a base of 8 in. and an area of at least 12 in.², find that one with the smallest perimeter.

In setting up an optimization problem the “variables” or “unknowns” of the problem must be specified. Often there is a choice among the quantities to

consider as variables. In this example enough variables are needed to completely specify a triangle. Since the base is fixed, knowing the values of any two other independent quantities of the triangle will specify it. Let x denote one side of the triangle and let y denote the height. These are the *variables* of the optimization problem. Other quantities that are fixed inputs to the problem are called *parameters*.

There are an infinite number of values of x and y for which the associated triangle has an area of at least 12 in.². It is intuitively clear that for some values, call them \bar{x} and \bar{y} , the perimeter of the triangle is smallest. The requirement that the triangle have an area of at least 12 in.² is stated algebraically as $8y/2 \geq 12$ or, more simply, $y \geq 3$. From Figure 1.1.1 arises another natural constraint, that $x \geq y$ hold.

A functional relationship must be derived by relating the problem variables and parameters to the perimeter of the triangle. From two applications of the Pythagorean theorem (referring again to Figure 1.1.1) the length of the third side is

$$\left\{ y^2 + \left[8 - (x^2 - y^2)^{1/2} \right]^2 \right\}^{1/2}.$$

The perimeter to be minimized is then

$$8 + x + \left\{ y^2 + \left[8 - (x^2 - y^2)^{1/2} \right]^2 \right\}^{1/2}.$$

Combining this information results in the following optimization problem: find values (\bar{x}, \bar{y}) that

$$[2] \quad \underset{(x, y)}{\text{minimize}} f(x, y) = 8 + x + \left\{ y^2 + \left[8 - (x^2 - y^2)^{1/2} \right]^2 \right\}^{1/2},$$

subject to the constraints that

$$[3] \quad y \geq 3$$

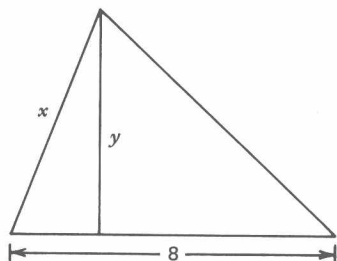


Figure 1.1.1 Geometry of triangle example.