# NETWORK FLOW PROGRAMMING

Paul A. Jensen J. Wesley Barnes

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### **PREFACE**

Today, both theoretical and applied interests in the use of network flow programming are experiencing an expansion unrivaled by that of any other optimization technique. Perhaps the most important reasons for this expansion are the fresh advances in network computational methods that allow analysts to solve problems of enormous size formerly unassailable by any other approach. Networks have been used in innumerable applications to represent such things as inventory systems, river systems, distribution systems, precedence ordering of events, flowcharts, and organization charts. In fact, the network representation is such a valuable visual and conceptual aid to the understanding of the relationships between events and objects that it is used in virtually every scientific, social, and economic field.

In this book, we present a synthesis of the more important techniques, both recent and traditional, that are related to network flow programming. The level of discussion is easily within the reach of first-year graduate or advanced undergraduate students. The two background requirements are a reasonable facility with a general-purpose programming language, such as Fortran IV, and an understanding of Linear Programming that is consistent with a good undergraduate text in Operations Research. The population possessing this background includes most persons holding undergraduate degrees in Computer Science, Engineering, Operations Research, Applied Mathematics, and Business Administration. However, the most important and largest segment of our intended audience is the practitioner.

With this audience in mind, we do three things. First, we emphasize concepts rather than theoretical proofs. Such proofs abound in the literature, and at the end of the book an extensive bibliography is provided which covers the theoretical underpinnings of the subject. Second, we include in the text a coherent

computational package of efficient algorithms that will solve any of the various network programming problems addressed in the book. Third, we take great care to present both the relationships between Linear Programming and Network Flow Programming and the interrelationships that join the various network flow programming problems and their suggested techniques of solution.

In the process of achieving these three objectives, we have formulated a reasonably complete exposition of the total spectrum of single commodity network flow programming problems. This presentation ranges from basic topics such as network storage techniques and handling representations to such advanced topics such as stochastic and generalized networks.

The principal strengths of this book can be summarized as follows:

The description of the various network flow programming problems in a consistent and unified notation. This makes evident the strong relationships between the problems and the relationships to the underpinning linear programming theory.

The modular nature of the manipulation and optimization algorithms and programs. Algorithms are presented in small packages that are easily understood by the student. Complex operations are performed by collections of modules. Modularization also provides for easy substitution of algorithms for experiments that test alternative computational approaches.

Emphasis on the computational aspects of algorithms. The recent emergence of network flow programming is primarily due to the computational procedures used to store and manipulate networks and network components in the computer. The serious student or user of network programming cannot neglect this aspect.

The method of presentation of the algorithms. We present algorithms at three levels: a rigorous flowchart using notation related to the computer programs (the form of the flowcharts is particularly compact and easily understood), a parallel English language description, and an example.

The scope of the material covered. We cover all of the important single commodity network flow programming problems. This includes extensive discussion of generalized networks and networks with separable nonlinear cost functions. In particular, this gathering of subjects is an especially useful one not currently available in existing texts.

Although most of the material in this book derives from previously published work, some appears here for the first time. The results and algorithms given here are immediately applicable to a wide variety of "real-world" problems. The computer codes, which we designed to be especially useful for teaching, can be obtained from us for a nominal handling charge. In addition, a solutions manual containing detailed answers to all exercises in the book is available to instructors and practitioners from John Wiley on written request.

There is ample material for a two-semester course, with some augmentation—perhaps with case studies. The first semester would comprise the material of Chapters 1 through 7. The second semester would deal with the more advanced material in the remaining chapters and the augmentations selected by

the instructor. A particularly able class should be able to cover Chapters 1 through 10 in a single concentrated semester. These comments are based on our more than five years of experience teaching this material.

We express our thanks to the great number of graduate students in the Mechanical Engineering Department's Operations Research graduate program at the University of Texas at Austin who gave us assistance in the development of this book. Finally, special thanks are due to Margaret Jensen, who typed the numerous revisions of the manuscript.

Paul A. Jensen J. Wesley Barnes

# **CONTENTS**

CHAPTER 1	NETWORK FLOW MODELS	1				
1.1	Introduction	1				
1.2	Relationships between Network Flow Programming Problems	3				
1.3	Specializations of the Pure, Linear, Minimum Cost Flow Problem	4				
1.4 The Network-with-Gains Model						
1.5	A Preview of Things to Come	8				
1.6	Historical Perspective	9				
1.7	Exercises	10				
CHAPTER 2	MODELING APPLICATIONS OF NETWORK					
	PROGRAMMING	11				
2.1	Introduction	11				
2.2	Slack Node Parameters	11				
2.3	The Pure, Linear, Minimum Cost Flow					
	Problem—Example Applications	12				
2.4	The Transportation Problem—Example Applications	23				
2.5	The Assignment Problem—Example Applications	28				
2.6	The Shortest Path Problem—Example Applications	34				
2.7		39				
2.8	Networks with Gains—Example Applications	42				
2.9	Historical Perspective	48				
	Exercises	50				

#### xii CONTENTS

CHAPTER 3	FORMALIZATION OF NETWORK MODELS	59		
3.1	Network Notation	59		
3.2	3.2 Two Useful Transformations			
3.3	3.3 The Algebraic Network Model			
3.4	The Linear Programming Model for the Pure			
	Minimum Cost Problem	69		
3.5	Terminology from Graph Theory	73		
3.6	The Expanded and Marginal Networks	76		
3.7 Networks with Nonlinear Arc Costs				
3.8	Solution Algorithms for the Pure Minimum Cost Flow			
	Problem	83		
3.9	Limitations of Network Flow Programming	85		
3.10	Historical Perspective	86		
	Exercises	87		
CHAPTER 4	NETWORK MANIPULATION ALGORITHMS	91		
4.1		91		
4.2	Network Representations	92		
4.3	Method of Algorithm Description	97		
4.4	Reading and Storing the Network	99		
4.5	Tree Representation	103		
4.6	Equivalent Algorithms Using a Preorder Traversal List	116		
4.7	Flow Manipulation Algorithms	121		
4.8	Historical Perspective	123		
	Exercises	123		
CHAPTER 5	THE SHORTEST PATH PROBLEM	125		
5.1	Introduction	125		
5.2	Representation as a Minimum Cost Flow Problem	126		
5.3	All Admissible Arc Costs Positive	129		
5.4	Some Arc Costs Negative with No Negative Cycles	132		
5.5	Negative Cycles	136		
5.6	A Nonbasic Algorithm	138		
5.7	A Dual Shortest Path Algorithm	139		
5.8	Historical Perspective	142		
	Applied Exercises	144		
	Theoretical Exercises	145		
CHAPTER 6	THE MAXIMUM FLOW PROBLEM	147		
6.1	Problem Statement	147		
6.2	Physical Interpretation of the Dual	148		
6.3	Theoretical Results	150		
6.4	Basic and Nonbasic Algorithms	153		

	CONTENTS	xiii
6.5	Flow Augmenting Algorithms	154
6.6	Historical Perspective	164
. 0.0	Applied Exercises	165
	Theoretical Exercises	166
	Theoretical Exercises	
CHAPTER 7	PURE MINIMUM COST FLOW PROBLEMS	167
7.1	The Maxflow Method to Obtain a Primal Feasible	
	Solution	168
7.2	The Artificial Arc Method to Obtain a Primal Feasible	
	Solution	170
7.3	A Nonbasic Primal Algorithm	175
7.4	A Basic Primal Algorithm	186
7.5	A Dual Node Infeasible Algorithm	194
7.6	Historical Perspective	201
	Exercises	202
CHAPTER 8	THE OUT-OF-KILTER ALGORITHM	203
8.1	Network Model	207
8.2	Linear Programming Model	208
8.3	200707	210
8.4		214
8.5		219
8.6		223
8.7	Historical Perspective	227
0.7	Exercises	227
CHAPTER 9	NETWORK MANIPULATION ALGORITHMS FOR	
CHAFTER 9	THE GENERALIZED NETWORK	231
9.1	Introduction	231
9.2		233
9.3	Linear Programming Model	234
9.4	the control of the co	241
9.5	Representation of Basis Networks	242
9.6	Flow for Generalized Networks	254
9.7	The Node Potentials	268
9.8	Historical Perspective	275
	Exercises	275
CITA INTER 40	GENERALIZED MINIMUM COST FLOW	
CHAPTER 10		281
10.1	PROBLEMS The Generalized Shortest Path Problem	281
10.1 10.2	All Arc Costs Positive—All Gains Less than or Equal	201
10.2	to One	283
	to One	

#### xiv CONTENTS

10.3	10.3 Negative Arc Costs—Arc Gains Greater than One					
10.4	10.4 Dual Generalized Shortest Path Algorithm					
	10.5 Generalized Minimum Cost Flow Algorithms					
10.6	10.6 The Flow Augmentation Approach					
10.7	A Generalized Primal Minimum Cost Flow Algorithm	315				
10.8	Historical Perspective	333				
	Exercises	334				
CHAPTER 11	THE CONVEX MINIMUM COST FLOW					
	PROBLEM	339				
11.1	Convex Cost Functions	339				
11.2	The Character of Solutions	343				
11.3	Physical Network Flow Problems	344				
11.4	Cost Functions That Depend on Random Variables	351				
11.5	Piecewise Linear Approximation	355				
11.6	Implicit Piecewise Approximation	357				
11.7	Historical Perspective	363				
	Exercises	364				
CHAPTER 12	CONCAVE COSTS	367				
12.1	Applications	368				
12.2	Notation	371				
12.3	Exhaustive Enumeration	372				
12.4	Implicit Enumeration	376				
12.5	Lower Bound	378				
12.6	The Implicit Enumeration Algorithm	381				
12.7	Example	385				
12.8	Historical Perspective	388				
	Exercises	389				
	REFERENCES	391				
	INDEX	403				

## **ALGORITHMS**

	IAPTER 4		4	FPATH2	160	4	MFLOG	266
1	READ	101	5	LABEL2	160	5	FLOWG	266
2	ORIGS	102				6	DUAL	271
3	TERMS	102	CH	APTER 7		7	CYCLE	273
4	ORIG	103	1	PHASE1	168			
5	TERM	103	2	ARTIFIC	172	CH	APTER 10	
6	TPATH	107	3	STARTM	174	1	DSHRTG	283
7	ROOT	108	4	PRIMAL1	175	2	PSHRTG	286
8	DELTRE	111	5	SHORTM	178	3	SMNSPG	287
9	ADDTRE	113	6	DSHRTM	178	4	DULSPG	302
10	TRECHG	116	7	PSHRTM	179	5	INCREMG	314
11	ROOTP	118	8	SMNSPM	180	6	ARTIFG	316
12	DELTREP	119	9	PRIMAL	187	7	PATHPG	323
13	ADDTREP	120	10	SELECT	189	8	PGAINS	332
14	TREINT	121	11	INCREMENT	199	U	TOAINS	332
15	FLOCHG	122	12	SHORT2	200	OT.	A DADED 44	
16	MFLO	122				-	APTER 11	241
			СН	APTER 8		1	CONVEX	361
CH	APTER 5		1	READOK	208	2	HFUN	361
1	DSHORT	129	2	KILTER	213	-	SMNCNVX	362
2	<b>PSHORT</b>	133	3	ADOKA	215	4	ORGSCNV	363
3	STARTD	134	4	FIND	216			
4	SMNSP	134	5	FLOW	216		APTER 12	
5	Table Control Control Control		3	LLOW	21/	1	READCC	372
,	SHORT	137	6	EDATU2	210			
6	SHORT NBSHORT	137 139	6	FPATH3	218	2	ENUMER	381
		139	7	OKAMFLO	219	2	ENUMER SETUP	381 383
6	NBSHORT		7			2 3 4	SETUP SET	1.8 8.5
6	NBSHORT DUALSP	139	7	OKAMFLO POTENT	219	2	SETUP SET RELAX	383
6 7 <b>CH</b>	NBSHORT DUALSP APTER 6	139 141	7 8 CH	OKAMFLO POTENT APTER 9	219 222	2 3 4	SETUP SET	383 383
6 7 <b>CH</b> 1	NBSHORT DUALSP APTER 6 MAXFLO	139 141	7 8 CHL 1	OKAMFLO POTENT  APTER 9 TRAIL	219 222 246	2 3 4 5	SETUP SET RELAX ROUND SEPRAT	383 383 383
6 7 <b>CH</b> 1 2	NBSHORT DUALSP APTER 6 MAXFLO LABEL1	139 141 156 157	7 8 CHL 1 2	OKAMFLO POTENT  APTER 9 TRAIL ROOTG	219 222 246 247	2 3 4 5 6	SETUP SET RELAX ROUND	383 383 383 384
6 7 <b>CH</b> 1	NBSHORT DUALSP APTER 6 MAXFLO	139 141	7 8 CHL 1	OKAMFLO POTENT  APTER 9 TRAIL	219 222 246	2 3 4 5 6 7	SETUP SET RELAX ROUND SEPRAT	383 383 383 384 384

# CHAPTER 1 NETWORK FLOW MODELS

#### 1.1 INTRODUCTION

As illustrated in Figure 1.1, a *network* is a collection of *nodes* and *arcs*. This representation is useful for modeling a wide range of physical and conceptual situations. Networks have been used in innumerable applications to represent such things as inventory systems, river systems, distribution systems, precedence ordering of events, flowcharts, and organization charts. The network representation is such a valuable visual and conceptual aid to the understanding of the relationships between events and objects that it is used in virtually every field of scientific, social, and economic endeavor.

Some practical situations that can be represented by a network also have the characteristic of flow; that is, water may flow in a pipe network, traffic may flow in a street network and products may flow in a distribution network. Models of such situations are called *network flow models*. In this book, we restrict our attention to models of this type and, as we will see, many problems not obviously in this class can be represented by network flow models.

Further, we consider network flow models in which the amount of flow in each arc is controllable and the objective is to choose values for the arc flows that optimize some measure of effectiveness. To illustrate, suppose Figure 1.2 defines a network flow model. Each arc in the network has flow directed as specified by the arrow head of the arc. In this model, each arc has been assigned three parameters: a *lower bound*, which is the minimal amount that can flow over the arc; a *capacity*, which is the maximum amount of flow that the arc can carry; and a *cost* for each unit of flow that passes through the arc.

Since a time period is implied in most network formulations, flows and capacities are usually stated in terms of "flow-per-unit-time." If no lower bound parameter is present on an arc, it is assumed that the lower bound is zero. In

#### 2 NETWORK FLOW MODELS

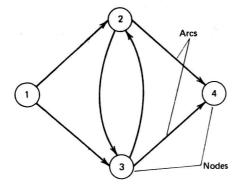


Figure 1.1 Network

Definition of terms [External flow] (Flow, lower bound, eapacity, cost)

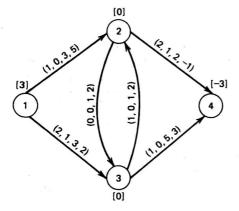


Figure 1.2
Network Flow Problem with Solution

addition lower bounds present no practical problems since a very simple transformation, discussed in Chapter 3, may be used to remove nonzero lower bounds from any network. The required quantities of flow entering or leaving the network at each node are also specified. These parameters of the nodes are called *external flows*. A positive external flow enters the network at a node and a negative external flow leaves the network. Flow is conserved at each node. Thus, the flow entering a node from the arcs of the network plus the external flow at the node must equal the flow leaving the node on the arcs of the network.

The flows on the arcs are controllable within the limits, or constraints, set by arc capacities, conservation of flow, and node external flows. Clearly, these arc flows are the decision variables of an optimization problem. The optimization

problem is to choose the arc flows, within the above restrictions, to minimize the total cost of the flow. As the reader may easily verify, the flows shown in Figure 1.2 are optimal for the given parameter values.

## 1.2 RELATIONSHIPS BETWEEN NETWORK FLOW PROGRAMMING PROBLEMS

The problem of optimizing some objective subject to constraints is called a mathematical programming problem. Because all the problems considered in this text are defined by a network that carries flow, we use the term network flow programming problem. The problem of Figure 1.2 is a specific example of the pure, linear, minimum cost flow problem. The schematic relationships joining this basic problem to other network flow programming problems that are considered in this text are shown in Figure 1.3. The central point in this figure is the pure, linear, minimum cost flow problem. The problems listed to the left are less general in the sense that they are specializations of this basic problem. Problems listed to the right are more general in that this basic problem is in some way a specialization of each of these problems. The general linear programming problem is also shown in this figure to indicate its relationship to the network programming problems. Algorithms have been identified to solve each of the problem classes in Figure 1.3. Algorithms for the less general problems are more efficient, in computational time and memory requirements, than those for the more general problems. Algorithms defined for a more general problem can solve a less general form of that problem, while the converse is rarely true.

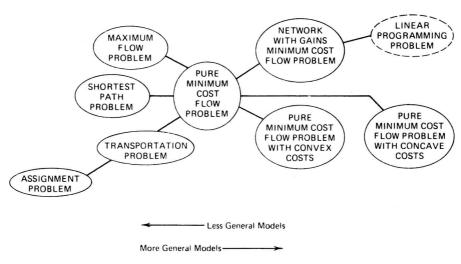


Figure 1.3
Network Flow Programming Problems Relationships

#### 4 NETWORK FLOW MODELS

## 1.3 SPECIALIZATIONS OF THE PURE, LINEAR, MINIMUM COST FLOW PROBLEM

#### Transportation Problem

Of frequent use in practice, the transportation problem is a special case of the minimum cost flow problem where the network representation has a distinct form: the nodes can be partitioned into two sets  $N_1$  and  $N_2$  such that all arcs originate in  $N_1$  and terminate in  $N_2$ . Three other special properties are:

- All arcs have infinite capacities, which allows omission of the capacity parameter from the arc parameter list.
- 2. All nodes have nonzero fixed external flows.
- 3. The sum of the external flows over all nodes is zero.

An example transportation problem network model, with its associated optimal flows, is presented in Figure 1.4.

#### **Assignment Problem**

An important specialization of the transportation problem is the assignment problem, in which both  $|N_1| = |N_2|$  and all demands and supplies are unity. Given the cost associated with pairing any object  $i \in N_1$  with any object  $j \in N_2$ , the problem is to find an exhaustive one-to-one pairing of the two sets' elements that minimizes the sum of the pairing costs. This problem is illustrated by the network of Figure 1.5. Again the flows present in Figure 1.5 are optimal and

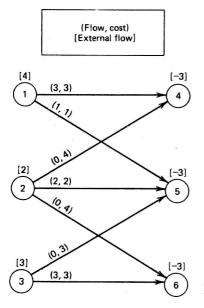


Figure 1.4
Example Transportation Problem

(Flow, cost) [External flow]

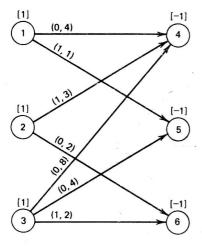


Figure 1.5
Example Assignment Problem

imply the following optimal pairings: node 1 and node 5, node 2 and node 4, and node 3 and node 6.

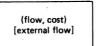
#### Shortest Path Problem

In the shortest path problem, two nodes are designated as the source and the sink. The arc cost is commonly given the physical interpretation of arc length. The optimal path is that sequence of arcs connecting the source to the sink such that the sum of the arc costs on the path is minimized. An example shortest path problem appears in Figure 1.6. The optimal flow pattern in Figure 1.6 implies the shortest path consists of arcs (1,3), (3,4), and (4,5).

Since nodes that are neither source nor sink will always have a fixed external flow of zero, no external flow designation is given for nodes 2, 3, and 4. This convention will be followed consistently throughout the book; that is, the absence of fixed external flow parameters will be interpreted as zero external flow.

#### **Maximum Flow Problem**

For the maximum flow problem, are capacity is the only relevant parameter. Once a source node and a sink node are identified, the problem is to maximize the flow passing from source to sink. An example maximum flow problem is shown in Figure 1.7, where node 1 is the source and node 6 is the sink. One of the several alternate optimal flow patterns is also given in Figure 1.7.



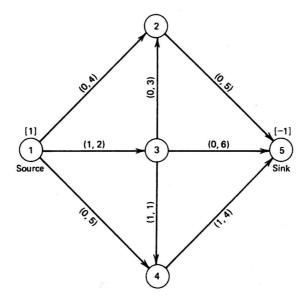


Figure 1.6
Example Shortest Path Problem

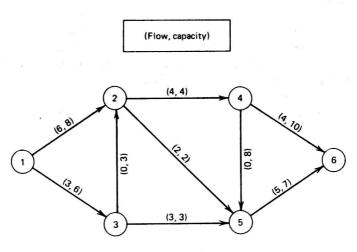


Figure 1.7 Example Maximum Flow Problem