
**QoS-Based Resource
Allocation and
Transceiver Optimization**

QoS-Based Resource Allocation and Transceiver Optimization

Martin Schubert

*Fraunhofer German-Sino Lab for
Mobile Communications MCI*

Holger Boche

*Technical University of Berlin
Fraunhofer Institute for Telecommunications
Heinrich-Hertz-Institut HHI
Fraunhofer German-Sino Lab for
Mobile Communications MCI*

now

the essence of **know**ledge

Boston – Delft

Foundations and Trends[®] in Communications and Information Theory

Published, sold and distributed by:

now Publishers Inc.
PO Box 1024
Hanover, MA 02339
USA
Tel. +1-781-985-4510
www.nowpublishers.com
sales@nowpublishers.com

Outside North America:

now Publishers Inc.
PO Box 179
2600 AD Delft
The Netherlands
Tel. +31-6-51115274

A Cataloging-in-Publication record is available from the Library of Congress

The preferred citation for this publication is M. Schubert and H. Boche, QoS-Based Resource Allocation and Transceiver Optimization, *Foundations and Trends[®] in Communications and Information Theory*, vol 2, no 6, pp 383–529, 2005

Printed on acid-free paper

ISBN: 1-933019-73-5
© 2006 M. Schubert and H. Boche

All rights reserved. No part of this publication may be reproduced, stored in a retrieval system, or transmitted in any form or by any means, mechanical, photocopying, recording or otherwise, without prior written permission of the publishers.

Photocopying. In the USA: This journal is registered at the Copyright Clearance Center, Inc., 222 Rosewood Drive, Danvers, MA 01923. Authorization to photocopy items for internal or personal use, or the internal or personal use of specific clients, is granted by now Publishers Inc for users registered with the Copyright Clearance Center (CCC). The 'services' for users can be found on the internet at: www.copyright.com

For those organizations that have been granted a photocopy license, a separate system of payment has been arranged. Authorization does not extend to other kinds of copying, such as that for general distribution, for advertising or promotional purposes, for creating new collective works, or for resale. In the rest of the world: Permission to photocopy must be obtained from the copyright owner. Please apply to now Publishers Inc., PO Box 1024, Hanover, MA 02339, USA; Tel. +1 781 871 0245; www.nowpublishers.com; sales@nowpublishers.com

now Publishers Inc. has an exclusive license to publish this material worldwide. Permission to use this content must be obtained from the copyright license holder. Please apply to now Publishers, PO Box 179, 2600 AD Delft, The Netherlands, www.nowpublishers.com; e-mail: sales@nowpublishers.com

**Foundations and Trends[®] in
Communications and Information Theory**
Volume 2 Issue 6, 2005
Editorial Board

Editor-in-Chief:

Sergio Verdú

*Department of Electrical Engineering
Princeton University
Princeton, New Jersey 08544,
USA
verdu@princeton.edu*

Editors

Venkat Anantharam (UC. Berkeley)	Amos Lapidoth (ETH Zurich)
Ezio Biglieri (U. Torino)	Bob McEliece (Caltech)
Giuseppe Caire (Eurecom)	Neri Merhav (Technion)
Roger Cheng (U. Hong Kong)	David Neuhoff (U. Michigan)
K.C. Chen (Taipei)	Alon Orlitsky (UC. San Diego)
Daniel Costello (U. Notre Dame)	Vincent Poor (Princeton)
Thomas Cover (Stanford)	Kannan Ramchandran (Berkeley)
Anthony Ephremides (U. Maryland)	Bixio Rimoldi (EPFL)
Andrea Goldsmith (Stanford)	Shlomo Shamai (Technion)
Dave Forney (MIT)	Amin Shokrollahi (EPFL)
Georgios Giannakis (U. Minnesota)	Gadiel Seroussi (HP-Palo Alto)
Joachim Hagenauer (TU Munich)	Wojciech Szpankowski (Purdue)
Te Sun Han (Tokyo)	Vahid Tarokh (Harvard)
Babak Hassibi (Caltech)	David Tse (UC. Berkeley)
Michael Honig (Northwestern)	Ruediger Urbanke (EPFL)
Johannes Huber (Erlangen)	Steve Wicker (Georgia Tech)
Hideki Imai (Tokyo)	Raymond Yeung (Hong Kong)
Rodney Kennedy (Canberra)	Bin Yu (UC. Berkeley)
Sanjeev Kulkarni (Princeton)	

Editorial Scope

Foundations and Trends[®] in Communications and Information Theory will publish survey and tutorial articles in the following topics:

- Coded modulation
- Coding theory and practice
- Communication complexity
- Communication system design
- Cryptology and data security
- Data compression
- Data networks
- Demodulation and Equalization
- Denoising
- Detection and estimation
- Information theory and statistics
- Information theory and computer science
- Joint source/channel coding
- Modulation and signal design
- Multiuser detection
- Multiuser information theory
- Optical communication channels
- Pattern recognition and learning
- Quantization
- Quantum information processing
- Rate-distortion theory
- Shannon theory
- Signal processing for communications
- Source coding
- Storage and recording codes
- Speech and Image Compression
- Wireless Communications

Information for Librarians

Foundations and Trends[®] in Communications and Information Theory, 2005, Volume 2, 6 issues. ISSN paper version 1567-2190. ISSN online version 1567-2328. Also available as a combined paper and online subscription.

QoS-Based Resource Allocation and Transceiver Optimization

Martin Schubert¹ and Holger Boche²

¹ *Fraunhofer German-Sino Lab for Mobile Communications MCI,
Einsteinufer 37, 10587 Berlin, Germany, schubert@hhi.fhg.de*

² *Technical University of Berlin, Dept. of Electrical Engineering,
Heinrich-Hertz Chair for Mobile Communications, HFT-6,
Einsteinufer 25, 10587 Berlin, Germany; Fraunhofer German-Sino Lab for
Mobile Communications MCI, Einsteinufer 37, 10587 Berlin, Germany;
Fraunhofer Institute for Telecommunications, Heinrich-Hertz-Institut
(HHI), Einsteinufer 37, 10587 Berlin, Germany, boche@hhi.fhg.de*

Abstract

The control and reduction of multiuser interference is a fundamental problem in wireless communications. In order to increase the spectral efficiency and to provide individual quality-of-service (QoS), it is required to jointly optimize the power allocation together with possible receive and transmit strategies. This often leads to complex and difficult-to-handle problem formulations. There are many examples in the literature, where the special structure of the problem is exploited in order to solve special cases of this problem (e.g. multiuser beamforming or CDMA). So it is desirable to have a general theory, which can be applied to many practical QoS measures, like rates, delay, BER, etc. These measures can all be related to the signal-to-interference ratio (SIR) or the signal-to-interference-plus-noise ratio (SINR). This leads to the problem of SIR and SINR balancing, which is fundamental for many problems in communication theory.

In this text we derive a comprehensive theoretical framework for SIR balancing, with and without noise. The theory considers the possible use of receive strategies (e.g. interference filtering or channel assignment), which can be included in the model in an abstract way. Power allocation and receiver design are mutually interdependent, thus joint optimization strategies are derived. The main purpose of this text is to provide a better understanding of interference balancing and the characterization of the QoS feasible region. We also provide a generic algorithmic framework, which may serve as a basis for the development of new resource allocation algorithms.

We study different interference models, which are general enough to be valid for a wide range of system designs, but which are also specific enough to facilitate efficient algorithmic solutions. One important class of interference functions is based on axioms, which characterize the impact of the power allocation of the interference received by the individual users. Another class of interference functions is based on non-negative coupling matrices, which may be parameter-dependent in order to model the possible impact of receive strategies. Both models are studied with and without noise. We analyze the resulting QoS feasible region (the set of jointly achievable QoS) and discuss different allocation strategies for min-max fairness and sum-power minimization. Finally we study geometrical properties of the QoS region, which can be shown to be convex for log-convex interference functions.

Contents

1	Introduction	1
1.1	QoS-based power and resource allocation	3
1.2	Related results in wireless communications	6
1.3	Outline	9
2	Axiomatic SIR-Balancing Theory	13
2.1	Axiom-based interference model	14
2.2	Existence of a min-max optimal power allocation	21
2.3	Achievable balanced SIR margin	29
2.4	Generalized achievability of SIR targets	32
2.5	Special monotonicity properties	34
2.6	Comparison of min-max and max-min optimization	36
2.7	Summary	37
3	Matrix-Based SIR Balancing	39
3.1	Min-max balancing and Perron root minimization	39
3.2	Characterization of boundary points	49
3.3	Achievability under an adaptive receive strategy	56
3.4	Uniqueness of the power allocation	63
3.5	Irreducible coupling matrices	73
3.6	Min-max and max-min balancing	80

3.7	Duality	84
3.8	Summary	86
4	General SINR Balancing Theory	89
4.1	Axiomatic interference model	89
4.2	Continuity of interference functions	91
4.3	Feasibility	91
4.4	Sum power minimization and fixed-point iteration	94
4.5	Relation with SINR balancing	96
4.6	Summary	99
5	Matrix-Based SINR Balancing and Algorithmic Solutions	101
5.1	Matrix-based interference function	101
5.2	Sum-power minimization	102
5.3	Fixed-point iteration	104
5.4	Matrix-based iteration	104
5.5	Convergence and comparison with the fixed-point iteration	106
5.6	Relationship with spectral radius optimization	110
5.7	Application example: Beamforming	116
5.8	Summary	121
6	Geometrical Properties for Log-Convex Interference Functions	123
6.1	Log-convexity of linear interference functions	124
6.2	Worst-case interference functions	126
6.3	Convexity of the QoS feasible region	127
6.4	Resource allocation by weighted QoS optimization	132
6.5	Summary	133
	Acknowledgements	135
	Appendix	137

A.1	Some definitions and results	137
A.2	Proof of Theorem 2.9	138
A.3	Proof of Theorem 2.22	139
A.4	Proof of Theorem 3.2	140
A.5	Proof of Theorem 4.3	141
	References	143

1

Introduction

The wireless channel is a broadcast medium, so each communication link is possibly interfered by other users transmitting at the same resource. The traditional way of handling interference is to assign all links orthogonal resources, in time (TDMA), frequency (FDMA), or code space (CDMA). This considerably simplifies the system design since the links are no longer coupled by interference. However, reserving each link a fixed resource often comes at the cost of sacrificing spectral efficiency. The available bandwidth is generally best exploited by letting transmitted signals interfere with each other in a controlled way (see e.g. [76, 75]). Also, orthogonality may be lost because of system imperfections and the effects of the time-varying multipath channel. It can be said that interference and power constraints are the main hurdle in achieving a high per-user throughput in heavily loaded multiuser networks, as will be required in the future.

Since interference plays an important role in the optimal exploitation of the given bandwidth, it is generally not sufficient to regard the system as a collection of point-to-point communication links. The quality-of-service (QoS) of each link depends on its own transmission power, but also on the power levels of the other links, which

2 Introduction

are experienced as interference. This results in a competitive situation, where all users try to compensate interference by increasing its own transmission power, which in turn increases the overall interference in the system. A transmission strategy which neglects these interdependencies is likely to cause uncontrollable and exceeding interference, which means a waste of the overall system efficiency. Thus, it is desirable to find a suitable equilibrium that optimally exploits the available resources. This requires a joint optimization of all communication links.

Optimization can be performed with respect to various design goals, like the overall efficiency, max-min fairness, proportional fairness, network utility maximization, etc. There is no such thing as “the” optimal communication strategy. There exists a great deal of literature on resource allocation from various points of view. For example, there are network-centric strategies, which aim at finding a stable performance trade-off by bidding strategies, accounting for traffic, channel quality, and revenues. User-centric strategies, which are closely related to power control, aim at fulfilling user-specific QoS requirements. Both strategies have in common that they are determined and limited by the QoS feasible region (the set of jointly achievable QoS).

The purpose of this text is not to give a comprehensive overview on allocation strategies, but rather to provide a fundamental theoretical framework which helps to understand the underlying effects of interference coupling, and to characterize the QoS feasible region. A fundamental question in this context is: what is the region of jointly achievable QoS, and how can certain points be achieved in a spectrally efficient way? This question is closely related to the classical power allocation problem, but in this text we will go one step further in assuming that interference not only depends on the power allocation, but also on adaptive *receive strategies*, like interference filtering or channel assignment. The additional optimization of the receive strategy adds new degrees of freedom to the problem of resource allocation. Thus, new concepts and algorithms are required.

Since power allocation and receive strategies are intricately intertwined, our approach is to use abstract models, which provide a better understanding of the underlying structure of the problem at hand.

In this respect, the work can be seen as a theoretical basis, which can be applied to solve existing problems in wireless communications.

1.1 QoS-based power and resource allocation

In this section we give an overview on some aspects of QoS-Based power allocation. We start by introducing the basic model used throughout this text, which will be refined later on.

1.1.1 Interference functions

Consider a network with K communication links, whose transmission powers are collected in a power allocation vector

$$\mathbf{p} = [p_1, \dots, p_K]^T > 0.$$

as illustrated in Fig. 1.1.

The interference power experienced by the k th user can be modeled by a function $\mathcal{I}_k(\mathbf{p})$. The functions $\mathcal{I}_1, \dots, \mathcal{I}_K$ describe how the links are affected by mutual cross-talk. Different definitions of $\mathcal{I}_k(\mathbf{p})$ and the resulting QoS region will be analyzed in this text.

It should be noted that the mapping $\mathcal{I}_k : \mathbb{R}_+^K \mapsto \mathbb{R}_+$ can be linear or non-linear, and it can also model the impact of adaptive receiver designs, like MMSE or interference cancellation, as well as other system aspects. A few examples are listed in the following.

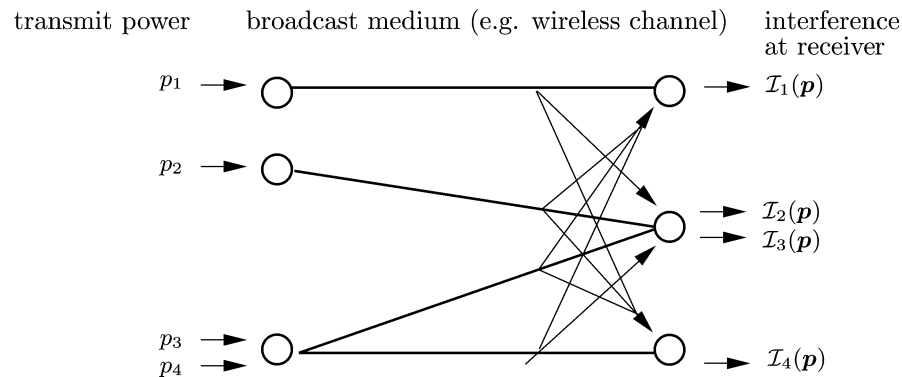


Fig. 1.1 Example of an interference-coupled multiuser system with four transmitter-receiver pairs.

4 Introduction

- $\mathcal{I}_k(\mathbf{p}) = [\Psi \mathbf{p}]_k$, where the positive coupling matrix $\Psi > 0$ contains interference coefficients, which determine in which way the users are affected by cross-talk (interference). This is a common model in power control theory (see e.g. [96]).
- $\mathcal{I}_k(\mathbf{p}) = \min_{z \in \mathcal{Z}} [\Psi(z) \mathbf{p}]_k$, where the adjustable receive strategy z (from a compact set of possible strategies \mathcal{Z} , as discussed later in Section 3.1.1) has impact on the interference structure. This specific model, which holds e.g. for multi-antenna beamforming or CDMA designs, and many more, will be studied in Sections 3 and 5.
- $\mathcal{I}_k(\mathbf{p}) = \max_c f_k(\mathbf{p}, c)$, where $f_k(\mathbf{p}, c)$ is the interference for a given power allocation \mathbf{p} under some interference uncertainty c . This definition can be used, e.g. to model worst-case interference under imperfect channel knowledge. This model will be discussed in Section 6.

But instead of focusing on a particular model, this text aims at characterizing basic properties, which are a common for a wide range of interference functions. To this end, we introduce an axiomatic characterization of interference functions in Section 2. This generic model contains the above examples as special cases. The axiomatic framework will be gradually refined in the following sections. By introducing additional properties, more results can be shown.

1.1.2 The QoS feasible region

The signal-to-interference ratio (SIR) of the k th user is

$$\text{SIR}_k(\mathbf{p}) = \frac{p_k}{\mathcal{I}_k(\mathbf{p})}, \quad 1 \leq k \leq K, \quad (1.1)$$

where p_k is the desired transmission power of the k th user. Note, that the function $\mathcal{I}_k(\mathbf{p})$ can include receiver noise. If noise is part of the assumed model (as in Sections 4 and 5), then we will emphasize this by using “SINR” instead of “SIR”. If we use SIR, then we discuss the general case where noise can be included or not. In this case, we need $\mathcal{I}_k(\mathbf{p}) > 0$ to ensure that (1.1) is well defined.

The term “QoS” is commonly used to describe the performance and reliability of a communication link. In order to keep the results

as general as possible, we do not make any specific assumption on QoS, except that it is related to the SIR by a monotonic and bijective function ϕ :

$$\text{QoS}_k(\mathbf{p}) = \phi(\text{SIR}_k(\mathbf{p})), \quad 1 \leq k \leq K. \quad (1.2)$$

Some examples are BER: $\phi(x) = Q(\sqrt{x})$, MMSE: $\phi(x) = 1/(1+x)$, BER-slope for α -fold diversity: $\phi(x) = x^{-\alpha}$, or capacity: $\phi(x) = \log(1+x)$.

Let γ be the inverse function of ϕ , then

$$\gamma_k = \gamma(Q_k), \quad 1 \leq k \leq K, \quad (1.3)$$

is the minimum SIR level needed by the k th user to satisfy the QoS target Q_k . Thus, the problem of achieving certain QoS requirements, carries over to the problem of achieving SIR targets $\gamma_k > 0, \forall k$. In the following we will also summarize the targets in a diagonal matrix

$$\mathbf{\Gamma} = \text{diag}\{\gamma_1, \dots, \gamma_K\}. \quad (1.4)$$

It is desirable to find a power allocation $\mathbf{p} > 0$ such that $\text{SIR}_k(\mathbf{p}) \geq \gamma_k$, for all $k = 1, \dots, K$. This can be rewritten as $\min_k \text{SIR}_k(\mathbf{p})/\gamma_k \geq 1$ or equivalently as $\max_k \gamma_k \mathcal{I}_k(\mathbf{p})/p_k \leq 1$. We say that the target $\mathbf{\Gamma}$ is *feasible* if and only if $C(\mathbf{\Gamma}) \leq 1$, where

$$C(\mathbf{\Gamma}) = \inf_{\mathbf{p} > 0} \left(\max_{1 \leq k \leq K} \frac{\gamma_k \mathcal{I}_k(\mathbf{p})}{p_k} \right). \quad (1.5)$$

In the following we will refer to (1.5) as the “min-max balancing problem”.

The optimum $C(\mathbf{\Gamma})$ provides a single measure for the joint feasibility of the targets $\mathbf{\Gamma}$. Note that the optimization is over $\mathbf{p} > 0$ to ensure that $\mathcal{I}_k(\mathbf{p})/p_k$ is always defined (see Section 2.1.1). However, this does not restrict the generality of the results since \mathbf{p} can be made arbitrarily small.

The min-max optimum $C(\mathbf{\Gamma})$ can be used to characterize the QoS feasible region:

$$\mathcal{Q} = \{[\phi(\gamma_1), \dots, \phi(\gamma_K)] : C(\mathbf{\Gamma}) \leq 1\}. \quad (1.6)$$

6 Introduction

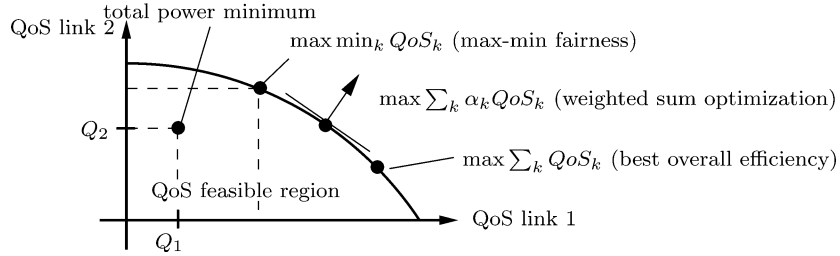


Fig. 1.2 QoS-based resource allocation strategies, illustrated for two users with QoS requirements Q_1 and Q_2

A fundamental problem in resource allocation theory is to find a feasible point $[Q_1, \dots, Q_K] \in \mathcal{Q}$ according to certain design criteria, like network efficiency, stability, or fairness. The optimization strategy can depend on many parameters, like operator revenue, user requests, queuing lengths, individual link priorities, etc. Examples for different points of interest are depicted in Fig. 1.2. But there exists no joint optimization framework. Their actual problem structure strongly depends on the geometry of \mathcal{Q} and on the definition of the underlying interference function.

So the purpose of this text is not to give a comprehensive overview on allocation strategies, but rather to provide a theoretical framework which helps to understand underlying principles. Most of the optimization problems illustrated in Fig. 1.2 are directly connected with the min-max balancing problem (1.5) and the associated QoS feasible region \mathcal{Q} .

In the following we will study the QoS (resp. SIR) feasible region for different interference functions $\mathcal{I}_k(\mathbf{p})$, including adaptive receive strategies and worst-case designs. But before we start with the most basic (axiomatic) interference model in Section 2, we provide some additional motivation by discussing the relationship of the generic interference model with problems in wireless communications.

1.2 Related results in wireless communications

A few examples for possible definitions of the interference function $\mathcal{I}_k(\mathbf{p})$ have already been given in Section 1.1.2. We will now discuss the SIR balancing problem in the context of previous work.

The linear function $\mathcal{I}_k(\mathbf{p}) = [\mathbf{\Psi} \mathbf{p}]_k$ is a classical model, which is used, e.g. in power control theory. The square matrix $\mathbf{\Psi} \geq 0$ models the link gains between all receiver/transmitter pairs. The min-max problem (1.5) for this case was already studied in [1] in the context of power balancing for satellite communication systems employing frequency reuse. Under the assumption that $\mathbf{\Psi}$ is non-negative and irreducible (see Section 3.1.4 for a definition), it was shown that the min-max-optimal power allocation is given as the principal eigenvector of $\mathbf{\Psi}$, and the optimum is the maximal eigenvalue (Perron root). This work was later extended by [42, 2, 46, 94, 95, 93, 33, 34]. An overview is given in [96, 38].

The above model can be extended to include AWG receiver noise, i.e., $\mathcal{I}_k(\mathbf{p}) = [\mathbf{\Psi} \mathbf{p}]_k + \sigma^2$. The presence of noise results in a situation where possible constraints on the transmission power do matter. Thus, the power allocation problem can be formulated so as to minimize the total power while maintaining certain SINR levels at the receiver. The optimal power allocation is obtained as the solution of a system of linear equations. Iterative solutions were proposed in [26, 43, 31, 4, 3, 7, 87].

The same power minimization problem was considered in [91, 39], where $\mathcal{I}_k(\mathbf{p})$ was not defined by a coupling matrix, but by using an axiomatic framework, equivalent to the one used in Section 4.1.

Since the mid-nineties, there has been a series of publications on *multiuser beamforming* for the downlink channel. In analogy to the power control problem, it was first proposed in [28, 29], to maximize the minimum SIR, assuming that the SIR not only depends on the power allocation, but also on a set of transmit beamformers $\mathbf{u}_1, \dots, \mathbf{u}_K \in \mathbb{C}^M$, which can be seen as a bank of linear unity-norm filters, which distribute all K signals across the M elements of an antenna array. Given $M \times M$ array covariance matrices $\mathbf{R}_1, \dots, \mathbf{R}_K$, the interference experienced by the k th receiver is $\sum_{l \neq k} p_l \mathbf{u}_l^* \mathbf{R}_k \mathbf{u}_l$. This is illustrated in Fig. 1.3.

The resulting min-max balancing problem is

$$\inf_{\mathbf{p} > 0, \mathbf{u}_1, \dots, \mathbf{u}_K} \left(\max_{1 \leq k \leq K} \frac{\sum_{l \neq k} p_l \mathbf{u}_l^* \mathbf{R}_k \mathbf{u}_l}{p_k \mathbf{u}_k^* \mathbf{R}_k \mathbf{u}_k} \right) \quad \text{s.t.} \quad \|\mathbf{u}_k\|_2 = 1. \quad (1.7)$$

8 Introduction

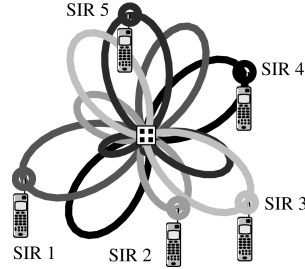


Fig. 1.3 Crosstalk is caused by non-orthogonal beams in a cellular system with multiuser beamforming, where a base station (BS) is simultaneously connected with K mobiles.

It can be observed that the interference in the numerator is not only affected by the powers, but also by the beamformers, thus beamforming adds an additional degree of freedom to the optimization. Problem (1.7) is difficult to handle in its direct form, since all the interference terms are coupled by the transmit beamformers $\mathbf{u}_1, \dots, \mathbf{u}_K$. The k th beamformer \mathbf{u}_k can be adjusted such that the desired power $\mathbf{u}_k^* \mathbf{R}_k \mathbf{u}_k$ becomes maximal. However, this strategy is generally not optimal for the other users, which are affected by the interference caused by \mathbf{u}_k . There is no obvious way how to obtain a good tradeoff between desired power and interference.

It was recognized in [44] that problem (1.7) can be reformulated as an eigenvalue optimization problem, which can be solved by an iterative algorithm. This work was further extended by [10, 13], where it was shown that this algorithm is closely connected with an equivalent uplink channel (see also the discussion in Sections 5.6.4, 3.5.4 and 5.7). By optimizing the uplink interference functions

$$\mathcal{I}_k(\mathbf{p}) = \min_{\mathbf{u}_k} \frac{\mathbf{u}_k^* (\sum_{l \neq k} p_l \mathbf{R}_l) \mathbf{u}_k}{\mathbf{u}_k^* \mathbf{R}_k \mathbf{u}_k} \quad \text{s.t.} \quad \|\mathbf{u}_k\|_2 = 1, \quad (1.8)$$

the optimal downlink beamformers can be found. Note that the beamformer \mathbf{u}_k in (1.8) is adaptively adjusted for each power allocation \mathbf{p} . This results in a non-linear dependency between the powers and the experienced interference. Nevertheless, the min-max SIR balancing problem (1.5) can be solved efficiently for the special choice of interference functions (1.8).

Downlink beamforming was also studied under the assumption of additional receiver noise [97, 21, 22, 70, 30, 89, 50, 78, 73, 6, 57, 58, 85]. Similar to the noiseless case, the uplink/downlink duality can be exploited in order to develop iterative algorithmic solutions. In [50, 78], an optimal algorithm was proposed, which consists of an iterative optimization of powers and beamformers for a “virtual uplink” problem. In retrospective, this algorithm can also be understood as a special case of the axiomatic interference model proposed in [91]. An equivalent axiomatic model will be studied in detail in Section 4. Another iterative solution was proposed in [57, 58, 9], where techniques from the theory of non-negative matrices were used to prove monotonicity and convergence. This was extended in [52], where it was shown that additional constraints on the beamformers can be added without affecting the convergence. This already points to the existence of a more general framework for interference balancing which will be introduced in Sections 4 and 5. Many of the results in [57, 58, 9, 52] can also be understood in the context of this general theory.

Besides beamforming, there are other examples for joint power allocation and receiver/transmitter optimization. This includes results on CDMA equalization [76, 71, 79, 72], multi-antenna MMSE filtering [88, 53, 51, 54, 20, 36, 37, 61], as well as recent progress on transceiver optimization for point-to-point MIMO systems [47]. Information-theoretical aspects of MIMO communication have been studied, e.g. in [25, 69, 86, 74, 77, 80, 82, 92]

All these results all have in common that they aim at a better understanding of the joint optimization of interference-coupled links in a network. While the discussed examples are focused on particular scenarios, it is desirable to have a general theory for resource allocation over the QoS region, where QoS can stand for different performance measures, like SINR, MMSE, or capacity. So the motivation behind this text is to find general principles behind interference balancing, which include some of the discussed results as special cases.

1.3 Outline

The sections of this text build on each other. Starting with the most general case, we successively add specifying assumptions, which

10 *Introduction*

sometimes restrict the generality, but also allow to show more specific properties. We will conclude each section with a short summary of the main results.

We start in Section 2 with an axiomatic interference model, which describes an interference situation in a most abstract and general way. The properties shown here can be regarded as the most common basis for interference balancing.

Section 3 focuses on the practically relevant case where interference can be modeled with a non-negative coupling matrix. This is known in the literature as the “SIR Balancing Problem”. But unlike classical power control theory, we assume that the powers are optimized jointly with an adaptive receiver design. This generalizes known results and algorithms from the aforementioned beamforming example [28, 29, 44, 10, 13]. The impact of the receiver design on the interference is modeled by a parameter-dependent coupling matrix. This adds an additional degree of freedom, so classical results and concepts need to be reconsidered.

From Section 4 on, we study the impact of an additional noise component, which leads to the problem of SINR balancing. Section 4 starts with an axiomatic model, which extends the model of Section 2 by an additional axiom which requires that the interference function is strictly monotone with respect to noise. This constant power level can also be regarded as a fixed interferer, thus the model can be seen as a special case of the more general model used in Section 2, where all interferers are assumed to be varying.

Section 5 further specifies the interference functions. As for the SIR balancing case, we use a parameter-dependent coupling matrix in order to model the impact of interference and noise. The assumption of a fixed noise component leads to additional properties. We study the problems of SINR-constrained power minimization and power-constrained SINR balancing.

Section 6 investigates the QoS feasible region under the assumption of log-convex interference functions. In this case, it can be shown that the resulting QoS region is convex. This useful property is the basis for the development of fast-convergent algorithms for resource allocation and scheduling.

Notation

Some general notational conventions are: matrices and vectors are set in boldface. Let \mathbf{y} be a vector, then $\mathbf{y}_l := [\mathbf{y}]_l$ is the l th component. We use $:=$ for definitions. Finally, $\mathbf{y} \geq 0$ means component-wise inequality, i.e., $\mathbf{y}_l \geq 0$ for all indices l . The set \mathbb{R}_+ does include the zero element, while \mathbb{R}_{++} only contains strictly positive elements.

References

- [1] J. M. Aein, "Power balancing in systems employing frequency reuse," *COMSAT Tech. Rev.*, vol. 3, no. 2, pp. 277–300, 2002.
- [2] H. Alavi and R. Nettleton, "Downstream power control for a spread spectrum cellular mobile radio system," in *Proc. IEEE Globecom*, pp. 84–88, 1982.
- [3] M. Andersin, Z. Rosberg, and J. Zander, "Gradual removals in cellular PCS with constrained power control and noise," *ACM/Baltzer Wireless Networks*, vol. 2, no. 1, pp. 27–43, 1996.
- [4] N. Bambos, S. C. Chen, and G. J. Pottie, "Radio link admission algorithms for wireless networks with power control and active link quality protection," in *INFOCOM (1)*, pp. 97–104, 1995.
- [5] M. Bengtsson, "Jointly optimal downlink beamforming and base station assignment," in *Proc. IEEE Internat. Conf. on Acoustics, Speech, and Signal Proc. (ICASSP)*, May 2001.
- [6] M. Bengtsson and B. Ottersten, *Handbook of Antennas in Wireless Communications*. CRC press, August 2001. ch. 18: Optimal and Suboptimal Transmit Beamforming.
- [7] F. Berggren, R. Jäntti, and S.-L. Kim, "A generalized algorithm for constrained power control with capability of temporary removal," *IEEE Trans. on Vehicular Technology*, vol. 50, no. 6, pp. 1604–1612, November 2001.
- [8] D. Bertsekas and R. Gallager, *Data Networks*. Englewood Cliffs, NJ: Prentice-Hall, 1992.
- [9] H. Boche and M. Schubert, "Duality theory for uplink downlink multiuser beamforming," in *Smart Antennas – State-of-the-Art*. EURASIP, Hindawi Publishing Corp., 2006. [Online]. Available: www.hindawi.com/books/spc/volume-3/index.html.

- [10] H. Boche and M. Schubert, "A unifying theory for uplink and downlink multi-user beamforming," in *Proc. IEEE Intern.*, (Zurich Seminar, Switzerland), February 2002.
- [11] H. Boche and M. Schubert, "A semialgebraic approach to multiuser beamforming and power control," in *Proc. IEEE Int. Symp. on Inf. Theory (ISIT)*, (Chicago, USA), June 2004.
- [12] H. Boche and M. Schubert, "Resource allocation in multi-antenna systems – achieving max-min fairness by optimizing a sum of inverse SIR," *IEEE Trans. Signal Processing*, vol. 54, no. 6, June 2006.
- [13] H. Boche and M. Schubert, "Optimal multi-user interference balancing using transmit beamforming," *Wireless Personal Communications*, vol. 26, no. 4, pp. 305–324, September 2003.
- [14] H. Boche, M. Schubert, and S. Stańczak, "A unifying approach to multiuser receiver design under QoS constraints," in *Proc. IEEE Vehicular Techn. Conf. (VTC)*, (Stockholm, Sweden), May 2005.
- [15] H. Boche, M. Schubert, S. Stańczak, and M. Wiczanowski, "An axiomatic approach to resource allocation and interference balancing," in *Proc. IEEE Internat. Conf. on Acoustics, Speech, and Signal Proc. (ICASSP)*, Philadelphia, USA, March 2005.
- [16] H. Boche and S. Stańczak, "Convexity of some feasible QoS regions and asymptotic behavior of the minimum total power in CDMA systems," *IEEE Trans. Commun.*, vol. 52, no. 12, pp. 2190–2197, December 2004.
- [17] H. Boche and S. Stańczak, "Log-convexity of the minimum total power in CDMA systems with certain quality-of-service guaranteed," *IEEE Trans. Inform. Theory*, vol. 51, no. 1, pp. 374–381, January 2005.
- [18] H. Boche, M. Wiczanowski, and S. Stańczak, "Characterization of optimal resource allocation in cellular networks," in *Proc. IEEE Workshop SPAWC*, (Lisboa, Portugal), 2004.
- [19] D. Catrein, L. Imhof, and R. Mathar, "Power control, capacity, and duality of up- and downlink in cellular CDMA systems," *IEEE Trans. Commun.*, vol. 52, no. 10, pp. 1777–1785, 2004.
- [20] R. Choi and R. Murch, "MIMO transmit optimization for wireless communication systems," in *DELTA'02*, 2002.
- [21] C. Farsakh and J. A. Nossek, "Channel allocation and downlink beamforming in an SDMA mobile radio system," in *Proc. Int. Symp. on Personal, Indoor and Mobile Radio Comm. (PIMRC)*, (Toronto, Canada), pp. 687–691, 1995.
- [22] C. Farsakh and J. A. Nossek, "Spatial covariance based downlink beamforming in an SDMA mobile radio system," *IEEE Trans. Commun.*, vol. 46, no. 11, pp. 1497–1506, November 1998.
- [23] G. Fettweis, M. Löhning, D. Petrovic, M. Windisch, P. Zillmann, and W. Rave, "Dirty RF: A new paradigm," in *Proc. 16th IEEE Intern. Symposium on Personal, Indoor and Mobile Radio Commun. (PIMRC)*, (Berlin, Germany), September 2005.
- [24] G. Foschini, G. Golden, R. Valenzuela, and P. Wolniansky, "Simplified processing for wireless communication at high spectral efficiency," *IEEEjsac*, vol. 17, no. 11, pp. 1841–1852, November 1999.

- [25] G. J. Foschini and M. J. Gans, "On limits of wireless communications in a fading environment when using multiple antennas," *Signal Processing (Elsevier Science)*, vol. 6, pp. 311–335, 1999.
- [26] G. J. Foschini and Z. Miljanic, "A simple distributed autonomous power control algorithm and its convergence," *IEEE Trans. on Vehicular Technology*, vol. 42, no. 4, pp. 541–646, November 1993.
- [27] F. R. Gantmacher, *The Theory of Matrices*. Vol. 2, New York: Chelsea Publishing Comp., 1959.
- [28] D. Gerlach and A. Paulraj, "Adaptive transmitting antenna methods for multipath environments," in *Proc. IEEE Globecom*, (San Francisco, CA), pp. 425–492, November 1994.
- [29] D. Gerlach and A. Paulraj, "Base station transmitting antenna arrays for multipath environments," *Signal Processing (Elsevier Science)*, vol. 54, pp. 59–73, 1996.
- [30] J. Goldberg and J. R. Fonollosa, "Downlink beamforming for cellular mobile communications," in *Proc. IEEE Vehicular Techn. Conf. (VTC)*, pp. 632–636, 1997.
- [31] S. A. Grandhi, R. Vijayan, and D. J. Goodman, "Distributed power control in cellular radio systems," *IEEE Trans. Commun.*, vol. 42, pp. 226–228, 1994.
- [32] S. Hanly, "An algorithm for combined cell-site selection and power control to maximize cellular spread spectrum capacity," *IEEE Journal on Selected Areas in Communications*, vol. 13, no. 7, pp. 1332–1340, September 1995.
- [33] B. He, M. Wang, and E. Li, "A new distributed power balancing algorithm for CDMA cellular systems," in *Proc. IEEE International Symposium on Circuits and Systems (ISCAS)*, pp. 1768–1771, 1997.
- [34] W. Hongyu, H. Aiging, H. Rong, and G. Weikang, "Balanced distributed power control," in *IEEE International Symposium on Personal, Indoor and Mobile Radio Communications (PIMRC)*, pp. 1415–1419, 2000.
- [35] R. A. Horn and C. R. Johnson, *Matrix Analysis*. MA: Cambridge University Press, 1985.
- [36] M. Joham, K. Kusume, M. H. Gzara, W. Utschick, and J. A. Nossek, "Transmit Wiener filter for the downlink of TDD DS-CDMA systems," in *Symp. on Spread-Spectrum Tech. and Appli.*, (Prague, Czech Republic), September 2002.
- [37] M. Joham, W. Utschick, and J. A. Nossek, "Linear transmit processing in MIMO communications systems," *IEEE Trans. Signal Processing*, vol. 53, no. 8, pp. 2700–2712, August 2005.
- [38] S. Koskie and Z. Gajic, "Sir-based power control algorithms wireless CDMA networks: An overview," in *International Conference on Dynamics of Continuous, Discrete and Impulsive Systems*, (Guelph, Ontario), May 2003.
- [39] K. K. Leung, C. W. Sung, W. S. Wong, and T. Lok, "Convergence theorem for a general class of power-control algorithms," *IEEE Trans. Commun.*, vol. 52, no. 9, pp. 1566–1574, September 2004.
- [40] J. C. Liberti and T. S. Rappaport, *Smart Antennas for Wireless Communications*. Upper Saddle River, NJ: Prentice Hall, 1999.
- [41] C. D. Meyer, *Matrix Analysis and Applied Linear Algebra*. SIAM, 2000.

- [42] H. J. Meyerhoff, "Method for computing the optimum power balance in multi-beam satellites," *COMSAT Tech. Rev.*, vol. 4, no. 1, pp. 139–146, 1974.
- [43] D. Mitra, "An asynchronous distributed algorithm for power control in cellular radio systems," in *Wireless and Mobile Communications*, pp. 177–186, Kluwer, 1994.
- [44] G. Montalbano, I. Ghauri, and D. T. M. Slock, "Spatio-temporal array processing for DS-CDMA downlink transmission," in *Proc. Asilomar Conf. on Signals, Systems and Computers*, (Monterey), 2000.
- [45] G. Montalbano and D. T. M. Slock, "Matched filter bound optimization for multiuser downlink transmit beamforming," in *Proc. IEEE Internat. Conf. on Universal Personal Communications (ICUPC)*, (Florence, Italy), October 1998.
- [46] R. Nettleton and H. Alavi, "Power control for a spread spectrum cellular mobile radio system," in *Proc. IEEE Vehicular Techn. Conf. (VTC)*, pp. 242–246, 1983.
- [47] D. P. Palomar, J. M. Cioffi, and M. A. Lagunas, "Joint Tx-Rx beamforming design for multicarrier MIMO channels: A unified framework for convex optimization," *IEEE Trans. Signal Processing*, vol. 51, no. 9, pp. 2381–2401, September 2003.
- [48] S. U. Pillai, *Array Signal Processing*. NY: Springer, 1989.
- [49] J. G. Proakis, *Digital Communications*. McGraw Hill, 1989.
- [50] F. Rashid-Farrokhi, L. Tassiulas, and K. J. Liu, "Joint optimal power control and beamforming in wireless networks using antenna arrays," *IEEE Trans. Commun.*, vol. 46, no. 10, pp. 1313–1323, October 1998.
- [51] H. Sampath, P. Stoica, and A. Paulraj, "Generalized linear precoder and decoder design for MIMO channels using the weighted MMSE criterion," *IEEE Trans. Signal Processing*, vol. 49, no. 12, pp. 2198–2206, December 2001.
- [52] D. Samuelsson, M. Bengtsson, and B. Ottersten, "An efficient algorithm for solving the downlink beamforming problem with indefinite constraints," in *Proc. IEEE Internat. Conf. on Acoustics, Speech, and Signal Proc. (ICASSP)*, Philadelphia, USA, March 2005.
- [53] A. Scaglione, G. B. Giannakis, and S. Barbarossa, "Redundant filterbank precoders and equalizers, parts i and ii," *IEEE Trans. Signal Processing*, vol. 47, pp. 1988–2002, July 1999.
- [54] A. Scaglione, P. Stoica, S. Barbarossa, G. B. Giannakis, and H. Sampath, "Optimal designs for space-time linear precoders and decoders," *IEEE Trans. Signal Processing*, vol. 50, no. 5, pp. 1051–1064, May 2002.
- [55] M. Schubert and H. Boche *Advanced Interference Function Calculus*.
- [56] M. Schubert and H. Boche, "Comparison of infinity-norm and 1-norm optimization criteria for SIR-balanced beamforming," *Signal Processing (Elsevier Science)*, vol. 84, no. 2, pp. 367–368, February 2004.
- [57] M. Schubert and H. Boche, "Solution of the multi-user downlink beamforming problem with individual SINR constraints," *IEEE Trans. Veh. Technol.*, vol. 53, no. 1, pp. 18–28, January 2004.
- [58] M. Schubert and H. Boche, "Iterative multiuser uplink and downlink beamforming under SINR constraints," *IEEE Trans. Signal Processing*, vol. 53, no. 7, pp. 2324–2334, July 2005.

- [59] M. Schubert, S. Shi, E. A. Jorswieck, and H. Boche, "Downlink sum-MSE transceiver optimization for linear multi-user MIMO systems," in *Proc. Asilomar Conf. on Signals, Systems and Computers*, (Monterey, CA), September 2005.
- [60] E. Seneta, *Non-Negative Matrices and Markov Chains*. Springer, 1981.
- [61] S. Serbetli and A. Yener, "Transceiver optimization for multiuser MIMO systems," *IEEE Trans. Signal Processing*, vol. 52, no. 1, pp. 214–226, January 2004.
- [62] S. Shi and M. Schubert, "MMSE transmit optimization for multiuser multi-antenna systems," in *Proc. IEEE Internat. Conf. on Acoustics, Speech, and Signal Proc. (ICASSP)*, (Philadelphia, USA), March 2005.
- [63] S. Stanczak and H. Boche, "Information theoretic approach to the Perron root of nonnegative irreducible matrices," in *Proc. of 2004 Information Theory Workshop (ITW)*, (San Antonio, Texas, USA), pp. 24–29, October 2004.
- [64] S. Stanczak and H. Boche, "The infeasible SIR region is not a convex set," *IEEE Trans. Commun.*, Accepted, November 2005.
- [65] S. Stanczak and M. Wiczanowski, "Distributed fair power control for wireless networks: Objectives and algorithms," in *Proc. the 43rd Annual Allerton Conference on Communications, Control, and Computing*, pp. 28–30, September 2005. Invited.
- [66] S. Stanczak, M. Wiczanowski, and H. Boche, "Distributed power control for optimizing a weighted sum of QoS parameter values," in *Proceedings of the IEEE Global Telecommunications Conference (GLOBECOM)*, (St. Louis, MO, USA), November 25–December 2 2005.
- [67] S. Stanczak, M. Wiczanowski, and H. Boche, *Theory and Algorithms for Resource Allocation in Wireless Networks*. Springer-Verlag, 2006. ser. Lecture Notes in Computer Science (LNCS).
- [68] C. W. Sung, "Log-convexity property of the feasible SIR region in power-controlled cellular systems," *IEEE Communications Letters*, vol. 6, no. 6, pp. 248–249, June 2002.
- [69] E. Telatar, "Capacity of multiple-antenna Gaussian channels," *European Trans. Telecommun.*, vol. 10, pp. 585–595, November 1999.
- [70] M. Torlak, G. Xu, B. L. Evans, and H. Liu, "Estimation of optimal weight vectors for spatial broadcast channels," in *Proc. IEEE Internat. Conf. on Acoustics, Speech, and Signal Proc. (ICASSP)*, p. 4009, 1997.
- [71] S. Ulukus and R. Yates, "Adaptive power control and MMSE interference suppression," *ACM Wireless Networks*, vol. 4, no. 6, pp. 489–496, 1998.
- [72] S. Ulukus and A. Yener, "Iterative transmitter and receiver optimization for CDMA networks," *IEEE Trans. Wireless Commun.*, vol. 3, no. 6, pp. 1879–1884, November 2004.
- [73] W. Utschick and J. A. Nossek, "Downlink beamforming for FDD mobile radio systems based on spatial covariances," in *Proc. European Wireless 99*, (Munich, Germany), pp. 65–68, October 1999.
- [74] M. K. Varanasi and T. Guess, "Achieving vertices of the capacity region of the Gaussian correlated-waveform multiple-access channel with decision feedback

148 *References*

- receivers,” in *Proc. IEEE Int. Symp. on Inf. Theory (ISIT)*, (Ulm, Germany), p. 270, June 1997.
- [75] M. K. Varanasi and T. Guess, “Optimum decision feedback multiuser equalization with successive decoding achieves the total capacity of the Gaussian multiple-access channel,” in *Proc. Asilomar Conf. on Signals, Systems and Computers, Monterey*, pp. 1405–1409, November 1997.
- [76] S. Verdú, *Multiuser Detection*. Cambridge, UK: Cambridge University Press, 1998.
- [77] S. Vishwanath, N. Jindal, and A. Goldsmith, “Duality, achievable rates, and sum-rate capacity of Gaussian MIMO broadcast channels,” *IEEE Trans. Inform. Theory*, vol. 49, no. 10, pp. 2658–2668, October 2003.
- [78] E. Visotsky and U. Madhow, “Optimum beamforming using transmit antenna array,” in *Proc. IEEE Vehicular Techn. Conf. (VTC) spring*, (Houston, Texas), pp. 851–856, May 1999.
- [79] P. Viswanath, V. Anantharam, and D. Tse, “Optimal sequences, power control and capacity of synchronous CDMA systems with linear MMSE multiuser receivers,” *IEEE Trans. Inform. Theory*, vol. 45, no. 6, pp. 1968–1993, September 1999.
- [80] P. Viswanath and D. Tse, “Sum capacity of the vector Gaussian broadcast channel and uplink-downlink duality,” *IEEE Trans. Inform. Theory*, vol. 49, no. 8, pp. 1912–1991, August 2003.
- [81] S. Vorobyov, A. Gershman, and Z.-Q. Luo, “Robust adaptive beamforming using worst-case performance optimization: a solution to the signal mismatch problem,” *IEEE Trans. Signal Processing*, vol. 51, no. 2, pp. 313–324, February 2003.
- [82] H. Weingarten, Y. Steinberg, and S. Shamai (Shitz), “The capacity region of the Gaussian MIMO broadcast channel,” *IEEE Trans. Inform. Theory*, Submitted, 2004.
- [83] M. Wiczanowski, H. Boche, and S. Stańczak, “Characterization of optimal resource allocation in cellular networks – optimization theoretic view and algorithmic solutions,” in *Internat. Teletraffic Congress (ITC)*, (Beijing, China), September 2005.
- [84] H. Wielandt, “Unzerlegbare, nicht negative Matrizen,” in *Math. Z.*, pp. 642–648, 1950. and *Mathematische Werke/Mathematical Works*, vol. 2, 100–106 de Gruyter, Berlin, 1996.
- [85] A. Wiesel, Y. C. Eldar, and S. Shamai, “Linear precoding via conic optimization for fixed MIMO receivers,” *IEEE Trans. Signal Processing*, vol. 54, no. 1, pp. 161–176, 2006.
- [86] P. Wolniansky, G. Foschini, G. Golden, and R. Valenzuela, “V-BLAST: an architecture for realizing very high data rates over the rich-scattering wireless channel,” in *Proc. URSI Intern. Symp. on Signals, Systems, and Electronics.*, (New York, NY, USA), pp. 295–300, 1998.
- [87] C. Wu and D. Bertsekas, “Distributed power control algorithms for wireless networks,” *IEEE Trans. on Vehicular Technology*, vol. 50, no. 2, pp. 504–514, March 2001.

- [88] J. Yang and S. Roy, "On joint transmitter and receiver optimization for multiple-input-multiple-output (MIMO) transmission systems," *IEEE Trans. Commun.*, vol. 42, no. 12, pp. 3221–3231, December 1994.
- [89] W. Yang and G. Xu, "Optimal downlink power assignment for smart antenna systems," in *Proc. IEEE Internat. Conf. on Acoustics, Speech, and Signal Proc. (ICASSP)*, May 1998.
- [90] R. Yates and H. Ching-Yao, "Integrated power control and base station assignment," *IEEE Trans. on Vehicular Technology*, vol. 44, no. 3, pp. 638–644, August 1995.
- [91] R. D. Yates, "A framework for uplink power control in cellular radio systems," *IEEE J. Select. Areas Commun.*, vol. 13, no. 7, pp. 1341–1348, September 1995.
- [92] W. Yu and J. M. Cioffi, "Sum capacity of Gaussian vector broadcast channels," *IEEE Trans. Inform. Theory*, vol. 50, no. 9, pp. 1875–1892, 2004.
- [93] J. Zander, "Distributed cochannel interference control in cellular radio systems," *IEEE Trans. on Vehicular Technology*, vol. 41, no. 3, pp. 305–311, August 1992.
- [94] J. Zander, "Performance of optimum transmitter power control in cellular radio systems," *IEEE Trans. on Vehicular Technology*, vol. 41, no. 1, pp. 57–62, February 1992.
- [95] J. Zander and M. Frodigh, "Comment on performance of optimum transmitter power control in cellular radio systems," *IEEE Trans. on Vehicular Technology*, vol. 43, no. 3, p. 636, August 1994.
- [96] J. Zander and S.-L. Kim, *Radio Resource Management for Wireless Networks*. Boston, London: Artech House, 2001.
- [97] P. Zetterberg and B. Ottersten, "The spectrum efficiency of a base station antenna array system for spatially selective transmission," in *Proc. IEEE Vehicular Techn. Conf. (VTC) fall*, 1995.

