

Robust MIMO Beamforming for Cellular and Radar Coexistence

Fan Liu, *Student Member, IEEE*, Christos Masouros, *Senior Member, IEEE*, Ang Li, *Student Member, IEEE*,
and Tharmalingam Ratnarajah, *Senior Member, IEEE*

Abstract—In this letter, we consider the coexistence and spectrum sharing between downlink multi-user multiple-input-multiple-output (MU-MIMO) communication and an MIMO radar. For a given performance requirement of the downlink communication system, we design the transmit beamforming such that the detection probability of the radar is maximized. While the original optimization problem is non-convex, we exploit the monotonically increasing relationship of the detection probability with the non-centrality parameter of the resulting probability distribution to obtain a convex lower-bound optimization. The proposed beamformer is designed to be robust to imperfect channel state information (CSI). Simulation results verify that the proposed approach facilitates the coexistence between radar and communication links, and illustrates a scalable tradeoff between the two systems' performance.

Index Terms—MU-MIMO downlink, radar-communication coexistence, spectrum sharing, robust beamforming.

I. INTRODUCTION

TO ADDRESS the explosive growth of wireless communication devices and services, a broadband plan has been agreed to free additional spectrum that is currently exclusive for military and governmental operations [1]. Typically, this spectrum is occupied by air surveillance and weather radar systems, and henceforth spectrum sharing between radar and communication has drawn much attention as an enabling solution [2]. While policy and regulations may delay the practical application of such solutions, research efforts are well under way to address the practical implementation of radar and communication coexistence. In [3], Opportunistic Spectrum Sharing (OSS) between cellular system and rotating radar has been considered, where the communication system are allowed to transmit signals when the space and frequency spectra are not occupied by the radar. Although the OSS method is straightforward, it does not allow radar and communication

to work simultaneously. Additionally, traditional rotating radar will soon be replaced by MIMO radar in the near future due to the advantages of waveform diversity and higher detection capability [4]. In recent years, several methods that consider the coexistence between MIMO radar and MIMO communication have been proposed, among which the Null Space Projection (NSP) method has been widely discussed [5], [6]. More relevant to this letter, optimization techniques have also been proposed to solve the problem. Li and Petropulu [7] optimize the Signal-to-Interference-plus-Noise-Ratio (SINR) of radar subject to power and capacity constraints. Related work discusses the coexistence between MIMO-Matrix Completion (MIMO-MC) radar and MIMO communication system, where the radar beamforming matrix and communication covariance matrix are jointly optimized [8]. Similar work has been done in [9], where the transmit beamforming design for the base station (BS) based on Linearly Constrained Minimum Variance (LCMV) optimization is proposed. Nevertheless, all of these works assume that CSI is perfectly known by radar or BS, which is not possible in practical scenarios. While robust beamformers exist in the broader area of cognitive radio networks for unicast and multicast transmission [10], [11], robust radar-specific coexistence solutions are yet to be explored in the related literature.

In this letter, we consider the transmit beamforming for spectrum sharing between downlink MU-MIMO communication and colocated MIMO radar. Focusing on a radar-specific optimization, we maximize the detection probability of radar while guaranteeing the transmit power budget of the BS and the received SINR of each downlink user. The beamforming design is initially formulated as an optimization problem under the perfect CSI assumption. Since the objective function is non-concave, we then optimize its lower bound instead. We further consider two optimization approaches where both the communication channel and interference channel are subject to CSI quantization errors. The proposed problems can be transformed into Semidefinite Programs (SDP) and solved by Semidefinite Relaxation (SDR) techniques. Simulation results validate the effectiveness of the proposed beamforming approach under the coexistence scenario for both perfect and imperfect CSI cases.

II. SYSTEM MODEL

We consider a Time Division Duplex (TDD) downlink MU-MIMO communication system that coexists with a MIMO radar on the same frequency band. As shown in Fig. 1, an N -antenna BS transmits signals to K single-antenna users. Meanwhile, a MIMO radar with M_t TX and M_r RX antennas

Manuscript received March 13, 2017; accepted April 9, 2017. Date of publication April 12, 2017; date of current version June 19, 2017. This work was supported in part by the Engineering and Physical Sciences Research Council (EPSRC) under Project EP/M014150/1, in part by the China Scholarship Council (CSC), and in part by the Seventh Framework Programme for Research of the European Commission under Project ADEL619647. The associate editor coordinating the review of this paper and approving it for publication was M. Nafie. (*Corresponding author: Fan Liu.*)

F. Liu is with the School of Information and Electronics, Beijing Institute of Technology, Beijing 100081, China, and also with the Department of Electronic and Electrical Engineering, University College London, London WC1E 7JE, U.K. (e-mail: liufan92@bit.edu.cn).

C. Masouros and A. Li are with the Department of Electronic and Electrical Engineering, University College London, London WC1E 7JE, U.K. (e-mail: chris.masouros@ieee.org; ang.li.14@ucl.ac.uk).

T. Ratnarajah is with the Institute for Digital Communications, School of Engineering, University of Edinburgh, Edinburgh EH9 3JL, U.K. (e-mail: t.ratnarajah@ed.ac.uk).

Digital Object Identifier 10.1109/LWC.2017.2693985

is detecting a point-like target in the far-field. The received signal at the i -th user is

$$y_i^C[l] = \mathbf{h}_i^T \sum_{k=1}^K \mathbf{t}_k d_k[l] + \sqrt{P_R} \mathbf{f}_i^T \mathbf{s}_l + n_i[l], i = 1, 2, \dots, K, \quad (1)$$

where $\mathbf{h}_i \in \mathbb{C}^{N \times 1}$, $\mathbf{f}_i \in \mathbb{C}^{M_r \times 1}$, $\mathbf{t}_i \in \mathbb{C}^{N \times 1}$, $d_i[l]$ and $n_i[l] \sim \mathcal{CN}(0, \sigma_C^2)$ denote the communication channel vector, the interference channel vector from radar, the beamforming vector, the communication symbol and the received noise for the i -th user respectively. $l = 1, 2, \dots, L$ is the symbol duration index, and L is the length of the communication frame. $\mathbf{S} = [\mathbf{s}_1, \mathbf{s}_2, \dots, \mathbf{s}_L] \in \mathbb{C}^{M_r \times L}$ denotes the radar transmit waveforms, with \mathbf{s}_l being the l -th snapshot across the transmit antennas. P_R is the power of the radar signals. Without loss of generality, we assume that the communication symbol has unit power, i.e., $\mathbb{E}[|d_k[l]|^2] = 1$, where \mathbb{E} denotes the ensemble average. It is also assumed that MIMO radar uses orthogonal waveforms, i.e., $\mathbb{E}[\mathbf{s}_l \mathbf{s}_l^H] = \frac{1}{L} \sum_{l=1}^L \mathbf{s}_l \mathbf{s}_l^H = \mathbf{I}$. The received SINR at the i -th user is thus given as

$$\gamma_i = \frac{|\mathbf{h}_i^T \mathbf{t}_i|^2}{\sum_{k=1, k \neq i}^K |\mathbf{h}_i^T \mathbf{t}_k|^2 + P_R \|\mathbf{f}_i\|^2 + \sigma_C^2}, \quad \forall i. \quad (2)$$

Considering the echo wave in a single range-Doppler bin of the radar detector, at the l -th snapshot, the discrete signal vector \mathbf{y}_l^R received by radar is given as

$$\mathbf{y}_l^R = \alpha \sqrt{P_R} \mathbf{A}(\theta) \mathbf{s}_l + \mathbf{G}^T \sum_{k=1}^K \mathbf{t}_k d_k[l] + \mathbf{w}_l, \quad (3)$$

where $\mathbf{G} = [\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_{M_r}] \in \mathbb{C}^{N \times M_r}$ is the interference channel matrix between BS and radar RX, θ is the azimuth angle of the target, α is the complex path loss of the radar-target-radar path, $\mathbf{w}_l = [w_1[l], w_2[l], \dots, w_{M_r}[l]]^T \in \mathbb{C}^{M_r \times 1}$ is the received noise vector at the l -th snapshot with $w_m[l] \sim \mathcal{CN}(0, \sigma_R^2)$, $\forall m$, $\mathbf{A}(\theta) = \mathbf{a}_R(\theta) \mathbf{a}_T^T(\theta)$, in which $\mathbf{a}_T(\theta) \in \mathbb{C}^{M_t \times 1}$ and $\mathbf{a}_R(\theta) \in \mathbb{C}^{M_r \times 1}$ are transmit and receive steering vectors of radar antenna array. In this letter, the model in [12] is used, for which

$$\begin{aligned} M_r &= M_t = M, \\ \mathbf{a}_R(\theta) &= \mathbf{a}_T(\theta) = \mathbf{a}(\theta), \\ \mathbf{A}_{im}(\theta) &= \mathbf{a}_i(\theta) \mathbf{a}_m(\theta) = \exp(-j\omega\tau_{im}(\theta)) \\ &= \exp\left(-j\frac{2\pi}{\lambda} [\sin(\theta); \cos(\theta)]^T (\mathbf{x}_i + \mathbf{x}_m)\right), \end{aligned} \quad (4)$$

where ω and λ denote the frequency and the wavelength of the carrier, $\mathbf{A}_{im}(\theta)$ is the i -th element at the m -th column of the matrix \mathbf{A} , which is the total phase delay of the signal, transmitted by the i -th element and received by the m -th element of the antenna array, $\mathbf{x}_i = [x_i^1; x_i^2]$ is the location of the i -th element of the antenna array. In the above model, we assume that $\mathbf{H} = [\mathbf{h}_1, \mathbf{h}_2, \dots, \mathbf{h}_K]$, $\mathbf{F} = [\mathbf{f}_1, \mathbf{f}_2, \dots, \mathbf{f}_K]$ and \mathbf{G} are flat Rayleigh fading and independent with each other and can be estimated by the BS through the pilot symbols. Note that for

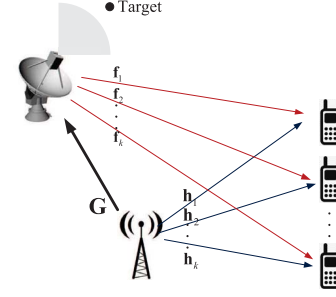


Fig. 1. Spectrum sharing scenario.

a typical TDD downlink, users will remain silence when BS is transmitting signals, so the radar only receives interference from the BS. For convenience, the index l is omitted in the rest of the letter.

The interference from BS to radar RX will affect the detection probability of radar. Under the Neyman-Pearson criterion, by using the Generalized Likelihood Ratio Test (GLRT), the asymptotic radar detection probability P_D is given as¹ [12]

$$P_D = 1 - \mathfrak{F}_{\chi_2^2(\rho)}\left(\mathfrak{F}_{\chi_2^2}^{-1}(1 - P_{FA})\right), \quad (5)$$

where P_{FA} is radar's probability of false alarm, $\mathfrak{F}_{\chi_2^2(\rho)}$ is the non-central chi-square Cumulative Distribution Function (CDF) with 2 Degrees of Freedom (DoF), $\mathfrak{F}_{\chi_2^2}^{-1}$ is the inverse

function of chi-square CDF with 2 DoFs. Let $\tilde{\mathbf{T}} = \sum_{k=1}^K \mathbf{t}_k \mathbf{t}_k^H$, the non-centrality parameter ρ for $\chi_2^2(\rho)$ is given by [13]

$$\rho = |\alpha|^2 L P_R \text{tr}\left(\mathbf{A} \mathbf{A}^H (\mathbf{G}^T \tilde{\mathbf{T}} \mathbf{G}^* + \sigma_R^2 \mathbf{I})^{-1}\right). \quad (6)$$

III. PROPOSED BEAMFORMING OPTIMIZATION

The average transmit power per frame of the BS is

$$P_C = \sum_{k=1}^K \|\mathbf{t}_k\|^2 = \sum_{k=1}^K \text{tr}(\mathbf{t}_k \mathbf{t}_k^H) = \sum_{k=1}^K \text{tr}(\mathbf{T}_k), \quad (7)$$

where $\mathbf{T}_k = \mathbf{t}_k \mathbf{t}_k^H$. The goal is to maximize the detection performance of radar while guaranteeing the received SINR per user and the power budget for the BS. We first consider the optimization with perfect CSI, followed by two optimization approaches, the upper bound minimization and the weighted minimization with norm-bounded CSI errors.

A. Beamforming for Perfect CSI

The optimization problem can be formulated as

$$\begin{aligned} \mathcal{P}_0 : \quad & \max_{\mathbf{t}_k} P_D \\ \text{s.t.} \quad & \gamma_i \geq \Gamma_i, \quad \forall i, \\ & P_C \leq P_0, \end{aligned} \quad (8)$$

¹It should be highlighted that the derivation in [12] is for the scenario with white Gaussian noise only while the proposed model in (3) includes both interference and noise. However, it can be shown that the resultant interference-plus-noise is still i.i.d. Gaussian distributed, but with a non-identity covariance matrix. We therefore apply a whitening-filter to normalize the interference-plus-noise, such that the derivation in [12] is still valid.

where Γ_i is the required SINR of the i -th communication user, P_0 is the power budget of the BS, γ_i , P_D and P_C are defined as (2), (5) and (7) respectively. It is well-known that P_D is a monotonically increasing function with respect to ρ [13], thus problem \mathcal{P}_0 can be equivalently formulated as

$$\begin{aligned} \mathcal{P}_1 : & \max_{\mathbf{t}_k} \text{tr}(\mathbf{A}\mathbf{A}^H(\mathbf{G}^T\tilde{\mathbf{T}}\mathbf{G}^* + \sigma_R^2\mathbf{I})^{-1}) \\ \text{s.t. } & \gamma_i \geq \Gamma_i, \forall i, \\ & P_C \leq P_0. \end{aligned} \quad (9)$$

As the objective function is non-concave, we consider a relaxation of optimizing its lower bound. Let $\mathbf{J} = \mathbf{G}^T\tilde{\mathbf{T}}\mathbf{G}^* + \sigma_R^2\mathbf{I}$. Noting that both \mathbf{J} and $\mathbf{A}\mathbf{A}^H$ are positive-definite, we have

$$\begin{aligned} \text{tr}(\mathbf{A}\mathbf{A}^H\mathbf{J}^{-1}\mathbf{J}) & \leq \text{tr}(\mathbf{A}\mathbf{A}^H\mathbf{J}^{-1})\text{tr}(\mathbf{J}) \\ \Rightarrow \text{tr}(\mathbf{A}\mathbf{A}^H\mathbf{J}^{-1}) & \geq \frac{\text{tr}(\mathbf{A}\mathbf{A}^H)}{\text{tr}(\mathbf{J})} = \frac{M^2}{\text{tr}(\mathbf{G}^T\tilde{\mathbf{T}}\mathbf{G}^*) + M\sigma_R^2}. \end{aligned} \quad (10)$$

$$(11)$$

Based on (11), \mathcal{P}_1 can be relaxed as

$$\begin{aligned} \mathcal{P}_2 : & \min_{\mathbf{t}_k} \text{tr}(\mathbf{G}^T\tilde{\mathbf{T}}\mathbf{G}^*) \\ \text{s.t. } & \gamma_i \geq \Gamma_i, \forall i, \\ & P_C \leq P_0. \end{aligned} \quad (12)$$

\mathcal{P}_2 is non-convex, and is equivalent to minimizing the total interference power from BS to radar. Fortunately, it can be efficiently solved by SDR technique. We refer readers to [14] for details on this topic.

B. Upper Bound Minimization for Imperfect CSI

Let us first model the channel vectors as

$$\begin{aligned} \mathbf{h}_i &= \hat{\mathbf{h}}_i + \mathbf{e}_{hi}, \mathbf{f}_i = \hat{\mathbf{f}}_i + \mathbf{e}_{fi}, \forall i, \\ \mathbf{g}_m &= \hat{\mathbf{g}}_m + \mathbf{e}_{gm}, \forall m, \end{aligned} \quad (13)$$

where $\hat{\mathbf{h}}_i$, $\hat{\mathbf{g}}_m$ and $\hat{\mathbf{f}}_i$ denote the estimated channel vectors known to the BS, \mathbf{e}_{hi} , \mathbf{e}_{gm} and \mathbf{e}_{fi} denote the CSI uncertainty within the spherical sets $\mathcal{U}_{hi} = \{\mathbf{e}_{hi} | \|\mathbf{e}_{hi}\|^2 \leq \delta_{hi}^2\}$, $\mathcal{U}_{gm} = \{\mathbf{e}_{gm} | \|\mathbf{e}_{gm}\|^2 \leq \delta_{gm}^2\}$ and $\mathcal{U}_{fi} = \{\mathbf{e}_{fi} | \|\mathbf{e}_{fi}\|^2 \leq \delta_{fi}^2\}$. This model is reasonable for scenarios where the CSI is quantized at the receiver and fed back to the BS. Particularly, if the quantizer is uniform, the quantization error region can be covered by spheres of given sizes [15].

It is assumed that BS has no knowledge about the error vectors except for the bounds of their norms. Given the partially known \mathbf{G} , following the process of [10], the upper bound of the interference power for the m -th radar antenna is given by

$$\left| \mathbf{g}_m^T \sum_{k=1}^K \mathbf{t}_k d_k \right|^2 \leq \sum_{k=1}^K \text{tr}(\hat{\mathbf{g}}_m^* \hat{\mathbf{g}}_m^T \mathbf{T}_k) + \zeta_{gm} \sum_{k=1}^K \text{tr}(\mathbf{T}_k), \quad (14)$$

where $\zeta_{gm} = 2\delta_{gm}\|\hat{\mathbf{g}}_m\| + \delta_{gm}^2$. We optimize the upper bound of the total interference power, which is obtained as

$$\begin{aligned} \text{tr}(\mathbf{G}^T\tilde{\mathbf{T}}\mathbf{G}^*) &= \sum_{k=1}^K \text{tr}(\mathbf{G}^* \mathbf{G}^T \mathbf{T}_k) = \sum_{m=1}^M \left| \mathbf{g}_m^T \sum_{k=1}^K \mathbf{t}_k d_k \right|^2 \\ &\leq \sum_{k=1}^K \text{tr}(\hat{\mathbf{G}}^* \hat{\mathbf{G}}^T \mathbf{T}_k) + \sum_{m=1}^M \zeta_{gm} \sum_{k=1}^K \text{tr}(\mathbf{T}_k). \end{aligned} \quad (15)$$

For the SINR constraint with partially known channel \mathbf{H} and \mathbf{F} , a worst-case approach is considered to guarantee that the solution is robust to all the uncertainties. Based on the triangle inequality, the maximum interference power from radar to the i th user is given as

$$\begin{aligned} P_R \|\mathbf{f}_i\|^2 &= P_R \|\hat{\mathbf{f}}_i + \mathbf{e}_{fi}\|^2 \\ &\leq P_R (\|\hat{\mathbf{f}}_i\| + \|\mathbf{e}_{fi}\|)^2 \leq P_R (\|\hat{\mathbf{f}}_i\| + \delta_{fi})^2. \end{aligned} \quad (16)$$

Following the well known S-procedure [16], the upper bound minimization with worst-case constraints is given as

$$\begin{aligned} \mathcal{P}_3 : & \min_{\mathbf{T}_i, s_i} \sum_{i=1}^K \text{tr}(\hat{\mathbf{G}}^* \hat{\mathbf{G}}^T \mathbf{T}_i) + \sum_{m=1}^M \zeta_{gm} \sum_{i=1}^K \text{tr}(\mathbf{T}_i) \\ \text{s.t. } & \begin{bmatrix} \hat{\mathbf{h}}_i^T \mathbf{Q}_i \hat{\mathbf{h}}_i^* - \Gamma_i \beta_i - s_i \delta_{hi}^2 & \hat{\mathbf{h}}_i^T \mathbf{Q}_i \\ \mathbf{Q}_i \hat{\mathbf{h}}_i^* & \mathbf{Q}_i + s_i \mathbf{I} \end{bmatrix} \geq 0, \\ & \mathbf{T}_i \geq 0, \mathbf{T}_i = \mathbf{T}_i^*, \text{rank}(\mathbf{T}_i) = 1, s_i \geq 0, \forall i, \\ & \sum_{i=1}^K \text{tr}(\mathbf{T}_i) \leq P_0, \end{aligned} \quad (17)$$

where $\mathbf{Q}_i = \mathbf{T}_i - \Gamma_i \sum_{n=1, n \neq i}^K \mathbf{T}_n$, $\beta_i = P_R (\|\hat{\mathbf{f}}_i\| + \delta_{fi})^2 + \sigma_C^2$.

Similar to problem \mathcal{P}_2 , by dropping the rank constraint, the non-convex problem \mathcal{P}_3 becomes a standard SDP and can be solved by SDR.

C. Weighted Minimization for Imperfect CSI

It is important to note that the upper bound minimization can only guarantee that the obtained beamformer does not generate strong interference, and it may not perform well for all the realizations of the interference channel \mathbf{G} . Here we use a weighted minimization for the case. Consider that \mathbf{G} is perfectly known to the BS, thus the actual power of the interference can be minimized. On the contrary, if BS has no knowledge of \mathbf{G} , the best strategy is to minimize the transmit power since a large power may cause higher interference. Obviously, the case with the partially known \mathbf{G} falls in between these two extreme cases. In other word, BS knows the estimated form of the interference power $\sum_{i=1}^K \text{tr}(\hat{\mathbf{G}}^* \hat{\mathbf{G}}^T \mathbf{T}_i)$, and the uncertainty about it, which is decided by the norm bound δ_{gm} . If δ_{gm} is large, we are more uncertain about the estimated interference, so we put more weight on minimizing the transmit power. Based on this, we rewrite \mathcal{P}_3 as

$$\begin{aligned} \mathcal{P}_4 : & \min_{\mathbf{T}_i, s_i} \sum_{i=1}^K \text{tr}(\hat{\mathbf{G}}^* \hat{\mathbf{G}}^T \mathbf{T}_i) + \phi(\delta_{g1}, \dots, \delta_{gm}) \sum_{i=1}^K \text{tr}(\mathbf{T}_i) \\ \text{s.t. } & \text{The same constraints with } \mathcal{P}_3, \end{aligned} \quad (18)$$

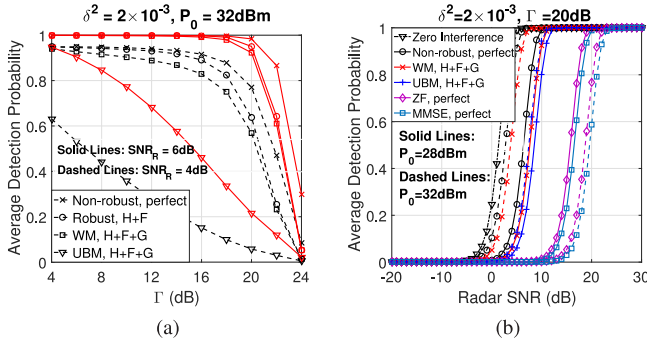


Fig. 2. Numerical results. (a) Average detection probability vs. SINR level for different cases, $\delta^2 = 2 \times 10^{-3}$, $P_0 = 32\text{dBm}$, $P_{FA} = 10^{-5}$; (b) Average detection probability vs. radar SNR for different cases, $\delta^2 = 2 \times 10^{-3}$, $\Gamma = 20\text{dB}$, $P_{FA} = 10^{-5}$.

where $\phi(\delta_{g1}, \dots, \delta_{gm})$ is an increasing function of error bounds. It can be observed that the upper bound minimization is a particular case of the weighted minimization with its weight function equalling to $\sum_{m=1}^M \zeta_{gm}$. Our results below show that this weight function is in general large, which means that it puts too much weight on minimizing the BS power while the uncertainty about the estimated channel $\hat{\mathbf{G}}$ is small.

IV. NUMERICAL RESULTS

In this section, numerical results based on Monte Carlo simulations have been provided to validate the effectiveness of the proposed method. Without loss of generality, each entry of the channel matrices follows the standard complex Gaussian distribution. We assume that radar uses a Uniform Linear Array (ULA) and has unit power. For convenience, we set $\Gamma_i = \Gamma$, $\delta_{hi} = \delta_{fi} = \delta_{gm} = \delta$, $\forall i, m$. While it is plausible that the benefits of the proposed scheme extend to various scenarios, here we assume $N = 8$, $K = 4$, $M = 4$, $\delta^2 = 2 \times 10^{-3}$ and $P_{FA} = 10^{-5}$ in all simulations. The power of all the noise vectors are set as $\sigma_R^2 = \sigma_C^2 = 0\text{dBm}$. The radar SNR is defined as $\text{SNR}_R = \frac{L|\alpha|^2 P_R}{\sigma_R^2}$ [12]. And we use a weight function $\phi(\delta) = \frac{\delta}{M}$ in the weighted minimization.

In Fig. 2 (a), the average detection probability P_D with increasing SINR level is shown for $P_0 = 32\text{dBm}$. The legends denote the beamformer used and the CSI state assumed, where ‘Non-robust’, ‘Robust’, ‘UBM’ and ‘WM’ denote beamforming optimization \mathcal{P}_2 , \mathcal{P}_2 using robust constraints, upper bound minimization \mathcal{P}_3 and weighted minimization \mathcal{P}_4 respectively. ‘Perfect’, ‘H+F’ and ‘H+F+G’ denote the case of perfect CSI, the case that \mathbf{H} and \mathbf{F} suffer from CSI errors while \mathbf{G} is perfectly known and the case that all channels are with errors. It can be seen that P_D decreases with the growth of Γ , which formulates the trade-off between the performance of radar and downlink communications. Note that the case ‘Robust, H+F’ and the weighted minimization show detection performance close to that for the perfect CSI case. Nevertheless, a significant performance loss occurs for upper bound minimization since it puts too much weight on minimizing the transmit power when δ is small. Similar results have been shown in Fig. 2 (b), where P_D with increased radar SNR for different power budget has been given with $\Gamma = 20\text{dB}$.

The idealistic case that BS causes no interference to radar has been provided as a reference. Once again, the weighted minimization outperforms the upper bound minimization. To further prove the optimality of the proposed methods, we also show the performance of Zero-forcing (ZF) and Minimum Mean Squared Error (MMSE) beamforming methods with perfect CSI. Unsurprisingly, even the upper bound minimization achieves a far better performance than the conventional methods. It can be also observed that larger P_0 leads to higher P_D for the proposed methods due to the extension of the feasible domain.

V. CONCLUSION

A beamforming approach has been introduced to facilitate the coexistence between downlink MU-MIMO communication system and MIMO radar. Given a target communication link SINR and the transmit power budget of BS, the proposed beamformer optimizes the radar performance. The proposed optimization has also been made robust to CSI errors by the upper bound minimization and the weighted minimization. The trade-off between radar and communication performance as well as the effectiveness of the proposed approach have been revealed by numerical simulations.

REFERENCES

- [1] FCC. (Mar. 2010). *The National Broadband Plan*. [Online]. Available: <http://www.broadband.gov/plan/5-spectrum>
- [2] B. Paul, A. R. Chiriyath, and D. W. Bliss, “Survey of RF communications and sensing convergence research,” *IEEE Access*, vol. 5, pp. 252–270, 2017.
- [3] R. Saruthirathanaworakun, J. M. Peha, and L. M. Correia, “Opportunistic sharing between rotating radar and cellular,” *IEEE J. Sel. Areas Commun.*, vol. 30, no. 10, pp. 1900–1910, Nov. 2012.
- [4] J. Li and P. Stoica, “MIMO radar with colocated antennas,” *IEEE Signal Process. Mag.*, vol. 24, no. 5, pp. 106–114, Sep. 2007.
- [5] A. Babaei, W. H. Tranter, and T. B. B. Bose, “A nullspace-based precoder with subspace expansion for radar/communications coexistence,” in *Proc. IEEE GLOBECOM*, Atlanta, GA, USA, Dec. 2013, pp. 3487–3492.
- [6] A. Khawar, A. Abdelhadi, and C. Clancy, “Target detection performance of spectrum sharing MIMO radars,” *IEEE Sensors J.*, vol. 15, no. 9, pp. 4928–4940, Sep. 2015.
- [7] B. Li and A. Petropulu, “MIMO radar and communication spectrum sharing with clutter mitigation,” in *Proc. IEEE Radar Conf. (RadarConf)*, Philadelphia, PA, USA, May 2016, pp. 1–6.
- [8] B. Li, A. P. Petropulu, and W. Trappe, “Optimum co-design for spectrum sharing between matrix completion based MIMO radars and a MIMO communication system,” *IEEE Trans. Signal Process.*, vol. 64, no. 17, pp. 4562–4575, Sep. 2016.
- [9] E. H. G. Yousif, M. C. Filippou, F. Khan, T. Ratnarajah, and M. Sellathurai, “A new LSA-based approach for spectral coexistence of MIMO radar and wireless communications systems,” in *Proc. IEEE Int. Conf. Commun. (ICC)*, Kuala Lumpur, Malaysia, May 2016, pp. 1–6.
- [10] K. T. Phan, S. A. Vorobyov, N. D. Sidiropoulos, and C. Tellambura, “Spectrum sharing in wireless networks via QoS-aware secondary multicast beamforming,” *IEEE Trans. Signal Process.*, vol. 57, no. 6, pp. 2323–2335, Jun. 2009.
- [11] H. Du and T. Ratnarajah, “Robust utility maximization and admission control for a MIMO cognitive radio network,” *IEEE Trans. Veh. Technol.*, vol. 62, no. 4, pp. 1707–1718, May 2013.
- [12] I. Bekkerman and J. Tabrikian, “Target detection and localization using MIMO radars and sonars,” *IEEE Trans. Signal Process.*, vol. 54, no. 10, pp. 3873–3883, Oct. 2006.
- [13] S. M. Kay, *Fundamentals of Statistical Signal Processing: Detection Theory*, vol. 2. Upper Saddle River, NJ, USA: Prentice-Hall, 1998.
- [14] Z.-Q. Luo, W.-K. Ma, A. M.-C. So, Y. Ye, and S. Zhang, “Semidefinite relaxation of quadratic optimization problems,” *IEEE Signal Process. Mag.*, vol. 27, no. 3, pp. 20–34, May 2010.
- [15] M. B. Shenuoda and T. N. Davidson, “On the design of linear transceivers for multiuser systems with channel uncertainty,” *IEEE J. Sel. Areas Commun.*, vol. 26, no. 6, pp. 1015–1024, Aug. 2008.
- [16] S. P. Boyd and L. Vandenberghe, *Convex Optimization*. Cambridge, U.K.: Cambridge Univ. Press, 2004.