Rate-Splitting Multiple Access for 6G—Part II: Interplay With Integrated Sensing and Communications

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Abstract—This letter is the second part of a three-part tutorial focusing on rate-splitting multiple access (RSMA) for 6G. As Part II of the tutorial, this letter addresses the interplay between RSMA and integrated radar sensing and communications (ISAC). In particular, we introduce a general RSMA-assisted ISAC architecture, where the ISAC platform has a dual capability to simultaneously communicate with downlink users and probe detection signals to a moving target. Then, the metrics of radar sensing and communications are respectively introduced, followed by a RSMA-assisted ISAC waveform design example which jointly minimizes the Cramér-Rao bound (CRB) of target estimation and maximizes the minimum fairness rate (MFR) amongst communication users subject to the per-antenna power constraint. The superiority of RSMA-assisted ISAC is verified through simulation results in both terrestrial and satellite scenarios. RSMA is demonstrated to be a powerful multiple access and interference management strategy for ISAC, and provides a better communication-sensing trade-off compared with the conventional benchmark strategies. Consequently, RSMA is a promising technology for next generation multiple access (NGMA) and future networks such as 6G and beyond.

Index Terms—Rate-splitting multiple access (RSMA), 6G, integrated sensing and communications (ISAC), next generation multiple access (NGMA).

I. INTRODUCTION

INTEGRATED sensing and communications (ISAC) has been envisioned as a key technique for future 6G wireless networks to fulfill the increasing demands on high-quality wireless connectivity as well as accurate and robust sensing capability [1]. ISAC merges wireless communications and remote sensing into a single system, where both functionalities are combined via shared use of the spectrum, the hardware platform, and a joint signal processing framework. ISAC systems are typically categorized into three types: radar-centric design, communication-centric design, and joint waveform design [2]. This letter focuses on the joint waveform design of ISAC rather than relying on existing radar or communication waveforms [3], [4]. In the literature, a novel waveform design was proposed in [5], where the ISAC transmit beamforming is designed to formulate an appropriate desired radar beampattern, while guaranteeing the signal-to-interference-plus-noise ratio (SINR) requirements of the communication users. The authors in [6] proposed the joint waveform design such that the multi-user interference (MUI) is minimized while formulating a desired radar beampattern. [2] investigated the joint waveform design with emphasis on optimizing the target estimation performance, measured by Cramér-Rao bound (CRB) considering both point and extended target scenarios.

As introduced in Part I of this tutorial, rate-splitting multiple access (RSMA) is a flexible and robust interference management strategy for multi-antenna systems, which relies on linearly precoded RS at the transmitter and successive interference cancellation (SIC) at the receivers. Inspired by the advantages of RSMA in spectral efficiency (SE), energy efficiency (EE), user fairness, reliability, and quality of service (QoS), performance enhancements in a wide range of network loads (underloaded and overloaded) and channel conditions, etc, the interplay between RSMA and ISAC systems was proposed in [7]. As a step further, RSMA-assisted ISAC was studied in [8] considering partial CSIT and mobility of communication users as a practical application. RSMA was shown to better manage the interference and improve the trade-off between weighted sum rate (WSR) and mean square error (MSE) of beampattern approximation compared with other commonly used strategies such as space division multiple access (SDMA) and non-orthogonal multiple access (NOMA). [9] investigated the RSMA-assisted ISAC satellite system, which facilitates the integration of communications and moving target sensing in space networks. The design of RSMA-assisted ISAC with low resolution digital-to-analog converter (DAC) units was introduced in [10].

In this letter, an overview of the interplay between RSMA and ISAC is provided. RSMA-assisted ISAC waveform optimization is for the first time studied to jointly minimize the CRB of the target estimation and maximize the minimum fairness rate (MFR) amongst communication users subject to transmit power constraints. RSMA is demonstrated to be very effective for both terrestrial and satellite ISAC systems to manage the multi-user/inter-beam interference as well as performing the radar functionality.

II. RSMA-ASSISTED ISAC

Consider a general RSMA-assisted ISAC, where the antenna array is shared by a co-located monostatic multiple-input multiple-output (MIMO) radar system and a communication system as depicted in Fig. 1. The ISAC platform equipped with $N_t$ transmit antennas and $N_r$ receive antennas simultaneously...
tracks a radar target and serves $K$ downlink single-antenna users indexed by the set $K = \{1, \ldots, K\}$. Since RSMA\(^1\) is adopted, the messages $W_1, \ldots, W_K$ intended for the communication users are split into common parts and private parts. All common part messages $\{W_{c,1}, \ldots, W_{c,K}\}$ are jointly encoded into a common stream $s_c$, while all private part messages $\{W_{p,1}, \ldots, W_{p,K}\}$ are respectively encoded into private streams $s_{p,1}, \ldots, s_{p,K}$. Thus, we can denote $s[l] = [s_c[l], s_{p,1}[l], \ldots, s_{p,K}[l]]^T$ as a $(K + 1) \times 1$ vector of unit-power signal streams, where $l \in L = \{1, \ldots, L\}$ is the discrete-time index within one coherent processing interval ( CPI), and the transmit signal at time index $l$ writes as

$$x[l] = Ps[l] = p_c s_c[l] + \sum_{k \in K} p_{p,k} s_{p,k}[l],$$

where $P = [p_c, p_{p,1}, \ldots, p_{p,K}] \in \mathbb{C}^{N_t \times (K+1)}$ is the precoding matrix, which is fixed within one CPI. If $L$ is sufficiently large, and the data streams are assumed to be independent from each other, satisfying $\frac{1}{L} \sum_{l=1}^{L} s[l] s[l]^H = I_K$. The covariance matrix of the transmit signal is written as

$$R_X = \frac{1}{L} \sum_{l=1}^{L} x[l] x[l]^H = PP^H.$$

### A. Radar Sensing Model and Metrics

The $N_t \times 1$ reflected echo signal at the radar receiver writes as

$$y[l] = H_l x[l] + m[l] = a e^{j2\pi f_D T} b(\theta) a^H(\theta) x[l] + m[l],$$

where $H_l \in \mathbb{C}^{N_r \times N_t}$ is the effective radar sensing channel. $\alpha$ stands for the complex reflection coefficient which is related to the radar cross-section (RCS) of the target. $F_D = \frac{2v_f}{c}$ denotes the Doppler frequency, with $f_c$ and $c$ respectively denoting the carrier frequency and the speed of the light. $v$ is the relative radar target velocity. $T$ denotes the symbol period. Note that for a monostatic radar, the direction of arrival (DoA) and the direction of departure (DoD) are the same, and can be denoted by $\theta$. $a(\theta) \in \mathbb{C}^{N_t \times 1}$ and $b(\theta) \in \mathbb{C}^{N_r \times 1}$ are the transmit and receive steering vectors, respectively. $m[l]$ is the additive white Gaussian noise (AWGN) distributed by $m[l] \sim \mathcal{CN}(0, \sigma^2_m I_{N_t})$, with $\sigma^2_m$ denoting the variance of each entry. The commonly used radar sensing metrics are as follows.

1) **Cramér-Rao Bound**: Cramér-Rao bound (CRB) is the radar sensing metric for target estimation, which represents a theoretical lower bound on the variance of unbiased estimators [2]. The CRB matrix can be calculated as $\text{CRB} = F^{-1}$, where $F$ is the Fisher information matrix (FIM) for estimating the real-valued target parameters $\xi = [\theta, \alpha, \sigma^2, F_D]^T$ given by

$$F = \begin{bmatrix} F_{\theta\theta} & F_{\theta\alpha} & F_{\theta\sigma^2} & F_{\theta F_D} \\ F_{\alpha\theta} & F_{\alpha\alpha} & F_{\alpha\sigma^2} & F_{\alpha F_D} \\ F_{\sigma^2\theta} & F_{\sigma^2\alpha} & F_{\sigma^2\sigma^2} & F_{\sigma^2 F_D} \\ F_{F_D\theta} & F_{F_D\alpha} & F_{F_D\sigma^2} & F_{F_D F_D} \end{bmatrix}. \quad (4)$$

By denoting $\mu[l] = y[l] - m[l]$, the elements of FIM can be calculated as

$$[F]_{i,j} = \frac{2}{\sigma^2_m} \text{Re} \left( \sum_{l=1}^{L} \frac{\partial \mu[l]^H \partial \mu[l]}{\partial \xi_i \partial \xi_j} \right), \quad i, j \in \{1, \ldots, 4\}, \quad (5)$$

where $\xi_i, \xi_j$ are the elements of $\xi$, and $[F]_{i,j}$ are dependent on $R_X$. As shown in [12], $R_X$ can be designed appropriately to improve the estimation capability of a MIMO radar by minimizing the trace, determinant or largest eigenvalue of the CRB matrix.

2) **Beampattern Approximation**: As widely studied in the literature, such as in [5], [7], etc, the transmit beampattern can be designed to approximate a highly directional transmit beampattern $P_d$. The radar sensing metric, MSE of beampattern approximation, is defined as

$$\text{MSE} = \sum_{m=1}^{M} \left| P_d(\theta_m) - a^H(\theta_m) R_X a(\theta_m) \right|^2, \quad (6)$$

where $\theta_m$ is the $m$-th direction grid among all $M$ grids. $P_d(\theta_m)$ is the desired beampattern at $\theta_m$. The smaller the MSE of beampattern approximation is, the higher SNR is obtained at the radar receiver, which leads to a higher probability of target detection and accuracy of target estimation.

3) **Radar Mutual Information**: The radar mutual information (RMI) represents the mutual information between the received radar signal and channel parameters. It contains all parameters of the target during the propagation and can be calculated as

$$\text{RIM} = I(y_r; H_r|s) = \log_2 \left( \frac{1}{|H_r + \frac{\sigma^2_m}{\sigma^2_m} R_X H_r^H|} \right)$$

$$= \log_2 \left( \frac{1}{|H_r + \alpha^2 b(\theta) a^H(\theta) R_X a(\theta) b^H(\theta)|} \right). \quad (7)$$

From (7), maximizing the RMI is equivalent to maximizing the transmit beampattern gain [8].

**Remark 1**: The CRB metric focuses on explicit optimization of the estimation performance characterized at the radar receiver, while the beampattern approximation MSE metric captures only the transmit design. Target estimation performance is guaranteed implicitly by approaching some well-designed radar beampattern featured with good estimation capability under communication constraints [2].

### B. Multi-User Communication Model and Metrics

At each user side, the received signal is written as

$$y_k[l] = h_k^T x[l] + z_k[l]$$

$$= h_k^T p_c s_c[l] + h_k^T \sum_{i \in K} p_{p,i} s_{p,i}[l] + z_k[l], \quad (8)$$

where $h_k \in \mathbb{C}^{N_r \times 1}$ denotes the channel between the ISAC transmitter and user-$k$, and $z_k[l] \sim \mathcal{CN}(0, \sigma^2_z)$ is the AWGN.

\(^1\)Layer RS is considered in this work for brevity and ease of illustration. We refer interested readers to [11] for a comprehensive study on different forms of RSMA.
Following the decoding order of RSMA, each user first decodes the common stream by treating all private streams as noise. After the common stream is reconstructed and subtracted from the received signal through SIC, each user then decodes its desired private stream. As detailed in Part I of the tutorial, the achievable rate of user-$k$, assuming Gaussian signalling, is $R_k = C_k + R_{p,k}$, where $C_k$ denotes the portion of the common stream rate carrying $W_c$, and $R_{p,k}$ denotes the private stream rate of user-$k$. To mitigate MUI, precoders can be optimized based on different metrics, e.g., WSR, MFR, EE, etc.

For the baseline strategies, SDMA-assisted ISAC is enabled by turning off the common stream in (1) [11]. NOMA-assisted ISAC relies on superposition coding at the transmitter and SIC at each user. The precoders and decoding orders are typically jointly optimized. By taking a two-user system as an example, and considering the specific decoding order, where the message of user-1 is decoded before the message of user-2, user-2 is able to decode messages of both users, while user-1 only decodes its desired message. Therefore, RSMA boils down to NOMA by encoding $W_1$ into the common stream $s_c$, encoding $W_2$ into $s_{p,2}$ and turning off $s_{p,1}$.

C. A Design Example

The ISAC waveform can be designed by investigating the trade-off between different communication and radar sensing metrics. We select the MFR and CRB as the communication and radar metrics, respectively. We consider a RSMA-assisted ISAC waveform optimization problem to investigate the trade-off by maximizing the communication MFR while minimizing the largest eigenvalue of the CRB matrix, which is equivalent to maximizing the smallest eigenvalue of FIM. Assuming perfect CSIT, the optimization problem is formulated as

$$\text{max} \left[ \min_{c \in \mathcal{C}} \left( C_k + R_{p,k} \right) \right] + \lambda t$$

s.t.

$$F \succeq \mathbf{I}$$

$$\text{diag}(\mathbf{PP}^H) = \frac{P_l N_t \times 1}{N_t}$$

$$R_{c,k} \geq \sum_{i=1}^{K} C_i, \quad \forall k \in \mathcal{K}$$

$$C_k \geq 0, \quad \forall k \in \mathcal{K},$$

where $c = [C_1, \ldots, C_K]^T$ is vector of common rate portions, $t$ is an auxiliary variable, and $\mathbf{I}$ is an identity matrix (here of the same dimension as $F$). $\lambda$ is the regularization parameter to prioritize either communications or radar sensing. $P$ denotes the sum transmit power budget. The constraint (11) ensures the transmit power of each antenna to be the same, which is commonly used for MIMO radar to avoid saturation of transmit power amplifiers in practical systems. The constraint (12) ensures that the common stream can be successfully decoded by all communication users, and (13) guarantees the non-negativity of all common rate portions. Note that the formulated problem is non-convex due to the non-convex rate expressions. It can be solved iteratively by a sequential convex approximation (SCA)-based algorithm [9]. The approximated convex problem involves only second-order cone (SOC) and linear matrix inequality (LMI) constraints, and can be solved by interior-point methods with the worst-case computational complexity $O\left( \log(e^{-1}) [N_t^2 (K + 1)]^{3.5} \right)$.

III. NUMERICAL RESULTS

In this section, simulation results are presented. The performance of RSMA-assisted ISAC is evaluated in terms of the trade-off between MFR and Root CRB (RCRB).

First, we consider a terrestrial radar-communication system where the ISAC base station (BS) is equipped with $N_t = 8$ antennas and $N_r = 9$ antennas. The system employs uniform linear array (ULA) with half-wavelength adjacent antenna spacing. The sum transmit power budget is $P = 20$ dBm, and the noise power at each user is $\sigma_n^2 = 0$ dBm. The communication channel is set as Rayleigh fading with each entry drawn from $CN \sim (0, 1)$. We assume $K = 4$ communication users, and the target is located at $\theta = 45^\circ$. The relative target velocity is $v = 8$ m/s. The number of transmit symbols within one CPI is $L = 1024$. In Fig. 2, the curves of the trade-off between MFR and RCRB of different target parameters are plotted. All results are obtained by solving the optimal waveform problem (9) - (13), and averaged over 100 channel realizations. The radar SNR is defined as $\text{SNR}_{\text{radar}} = |\alpha|^2 P/\sigma_n^2 = -20$ dB. For the baseline strategies, SDMA-assisted ISAC can be simulated as a special case of RSA by turning off the common stream. The decoding order of NOMA-assisted ISAC is the ascending order of channel strengths. No user grouping is considered. We can observe that when the priority is the communication functionality, both RSMA-assisted and SDMA-assisted ISAC achieve similar MFR. As the priority is shifted to sensing, the RSMA-assisted ISAC achieves a considerably better trade-off compared with SDMA. Similar trade-off performance can be observed in [8] where the ISAC waveform was designed by optimizing the communication and radar metric, namely, WSR and beampattern MSE. From Fig. 2, the NOMA-assisted ISAC achieves the poorest trade-off due to the degrees of freedom (DoF) loss in multi-antenna NOMA [13]. At the leftmost points which correspond to prioritizing the radar functionality, the optimized precoders are linearly dependent[2] on each other. Thus, the SDMA-assisted ISAC can no longer exploit spatial DoF provided by multiple antennas and leads to lower MFR compared with the RSMA-assisted and NOMA-assisted ISAC which employ SIC at user sides to manage the MUI.

Sensing capability at the radar receiver is evaluated in Fig. 3 in terms of the target estimation root mean square error (RMSE). Subspace-based estimation algorithms, e.g., multiple signal classification (MUSIC) can be used to estimate the Doppler frequency, direction of the target and reflection coefficient from the radar received signal. Throughout the simulations, communication symbols $s[l]$ in (1) are generated as random quadrature-phase-shift-keying (QPSK) modulated sequences, and the precoders are obtained by solving the optimal waveform problem (9) - (13). Fig. 3 depicts the RMSE and RCRB with the increase of radar SNR while fixing the MFR of RSMA-assisted and SDMA-assisted ISAC to be 6 bps/Hz. NOMA-assisted ISAC is not evaluated due to its poor MFR performance, and 6 bps/Hz cannot be satisfied. We can observe that the RMSEs of different target parameters are lower-bounded by the corresponding RCRBs.

[2]From the simulation results, we can observe that the optimized precoders are linearly dependent on each other. Intuitively, when mostly prioritizing the radar functionality, the optimized precoders are designed to radiate the highest power towards the target angle.
and are expected to approach the RCRBs at high radar SNR regime. As expected, the RSMA-assisted ISAC always outperforms SDMA-assisted ISAC in terms of the target estimation performance.

Second, a satellite radar-communication system is considered, where the ISAC satellite could be a multi-beam LEO satellite simultaneously serving single-antenna satellite users and sensing a moving target within the satellite coverage area. Considering a single feed per beam (SFPB) architecture, which is popular in modern satellites such as Eutelsat Ka-Sat, where one antenna feed is required to generate one beam. We can simply assume \( \rho \) uniformly distributed users in each beam, and the satellite channel model is given in \[14\]. Satellite users follow multibeam multicast transmission. Fig. 4 shows the trade-off curves between MFR and RCRB in a multi-beam satellite ISAC system. Form Fig. 4, the trade-off performance gain provided by RSMA-assisted design is more obvious than the terrestrial scenario given in Fig. 2. The gap between RSMA-assisted and SDMA-assisted ISAC starts from the rightmost points which correspond to prioritizing the communication functionality. This is due to the superiority of using RSMA in an overloaded communication system [14]. Since NOMA leads to extremely high receiver complexity when the number of user is large and also a waste of spatial resources in multi-antenna settings, we do not compare with NOMA-assisted ISAC in this scenario. Above all, we can conclude that RSMA is a very effective and powerful strategy for both terrestrial and satellite ISAC systems to manage the multi-user/inter-beam interference as well as performing the radar functionality.

IV. FUTURE PERSPECTIVES

In this section, we illustrate the potential research directions and applications of RSMA-assisted ISAC in emerging technologies that are of great importance in 6G.

A. RSMA-Assisted ISAC With mmWave Massive MIMO

Millimeter wave (mmWave) has been considered as a key technique for the future 6G wireless networks. A number of ISAC scenarios involve millimeter wave (mmWave) frequencies [15]. It shows potentials to achieve high data rates for communication and high resolution for radar operation due to the huge available bandwidth in the mmWave frequency bands and multiplexing gains achievable with massive antenna arrays. To reduce the transceiver hardware complexity and power consumption, HAD precoding design for ISAC systems at the mmWave band was investigated in [15], [16]. Inspired by the appealing advantages of RSMA in multi-antenna systems, the benefits of introducing RSMA and HAD to tackle the MUI in the context of mmWave communications were demonstrated in [17], [18]. As a consequence, the interplay between RSMA-assisted ISAC and HAD for mmWave is becoming another interesting research topic [10].

B. RSMA-Assisted ISAC With SAGIN

The space-air-ground integrated network (SAGIN), which integrates spaceborne, airborne and terrestrial/marine networks has been envisioned to provide heterogeneous services and seamless network coverage. ISAC is expected to benefit the SAGIN by improving the utilization of spectral and hardware resources. In [19], the authors investigated the application of ISAC in a massive MIMO low earth orbit (LEO) satellite system and proposed a beam squint-aware scheme to operate...
communications and target sensing simultaneously. Compared with satellites and terrestrial BSs, UAVs enjoy much higher mobility, ease of deployment, and may act as aerial ISAC platforms, relays or sensing targets, which present great compatibility with the SAGIN to enhance ISAC services [1]. Note that RSMA has been demonstrated to be very promising for satellite-terrestrial/aerial integrated networks [20] and satellite ISAC systems [9]. The interplay between RSMA-assisted ISAC and SAGIN is worth being investigated.

C. RSMA-Assisted ISAC With V2X

For the coming generation of vehicle-to-everything (V2X) networks, ISAC serves as a particularly suitable technology aiming to jointly provide high throughput vehicular communication service and remote sensing service for vehicle localization and anti-collision detection [21]. The characteristics of vehicular networks include high mobility, rigorous requirements on transmission latency and reliability, etc. Since RSMA has been demonstrated to be robust against CSIT imperfections resulting from user mobility and feedback delay in multi-user (Massive) MIMO [22], RSMA-assisted ISAC has great potential to become a promising research topic for the future V2X networks.

D. RSMA-Assisted ISAC With OFDM

Orthogonal frequency-division multiplexing (OFDM) has been widely investigated as one of the key techniques in wireless networks. It was also found to be useful in radar sensing [4]. Due to the promising application to radar sensing, and the key role in 4G and 5G wireless communication standards, OFDM waveforms for ISAC systems have been explored in [23], [24]. The benefits of implementing RSMA in a multicarrier communication system have been demonstrated in [25], [26]. Thus, implementing RSMA in an OFDM-based ISAC systems is worth being investigated.

V. CONCLUSION

In this letter, we provide an overview on the interplay between two promising techniques for future 6G, namely, RSMA and ISAC. We start from a general RSMA-assisted ISAC model and summarize the commonly used performance metrics for both radar sensing and communications. Then, we introduce a design example which jointly minimizes the CRB of target estimation and maximizes MFR amongst communication users subject to the per-antenna power constraint. Simulation results demonstrate that RSMA is a very effective and powerful strategy for both terrestrial and satellite ISAC systems to manage the multi-user/inter-beam interference as well as performing the radar functionality. Finally, we illustrate the future perspectives and potential applications including, but are not limited to, RSMA-assisted ISAC with mmWave, SAGIN, V2X and OFDM-based systems. We conclude that RSMA is a promising technology for NGMA and future networks such as 6G and beyond.

REFERENCES

[1] F. Liu et al., “Integrated sensing and communications: Toward dual-functional wireless networks for 6G and beyond,” IEEE J. Sel. Areas Commun., vol. 40, no. 6, pp. 1728–1767, Jun. 2022.
[2] F. Liu et al., “Cramér-rao bound optimization for joint radar-communication beamforming,” IEEE Trans. Signal Process., vol. 70, pp. 240–253, 2021.
[3] A. Hassanien et al., “Dual-function radar-communications: Information embedding using sidelobe control and waveform diversity,” IEEE Trans. Signal Process., vol. 64, no. 8, pp. 2168–2181, Aug. 2015.
[4] C. Sturm and W. Wiesbeck, “Waveform design and signal processing aspects for fusion of wireless communications and radar sensing,” Proc. IEEE, vol. 99, no. 7, pp. 1236–1259, Jul. 2011.
[5] F. Liu et al., “MU-MIMO communications with MIMO radar: From co-existence to joint transmission,” IEEE Trans. Wireless Commun., vol. 17, no. 4, pp. 2755–2770, Apr. 2018.
[6] F. Liu et al., “Toward dual-functional radar-communication systems: Optimal waveform design,” IEEE Trans. Signal Process., vol. 66, no. 16, pp. 4264–4279, Aug. 2018.
[7] C. Xu et al., “Rate-splitting multiple access for multi-antenna joint radar and communications,” IEEE J. Sel. Topics Signal Process., vol. 15, no. 6, pp. 1332–1347, Nov. 2021.
[8] R. C. Loli et al., “Rate-splitting multiple access for multi-antenna joint radar and communications with partial CSIT: Precoder optimization and link-level simulations,” 2022, arXiv:2201.10621.
[9] L. Yin and B. Clerckx, “Rate-splitting multiple access for dual-functional radar-communication satellite systems,” in Proc. IEEE Wireless Commun. Netw. Conf. (WCNC), Apr. 2022, pp. 1–6.
[10] O. Dizdar et al., “Energy efficient dual-functional radar-communication: Rate-splitting multiple access, low-resolution DACs, and RF chain selection,” 2022, arXiv:2202.09128.
[11] Y. Mao et al., “Rate-splitting multiple access: Fundamentals, survey, and future research trends,” 2022, arXiv:2001.03192.
[12] J. Li et al., “Range compression and waveform optimization for MIMO radar: A Cramér–Rao bound based study,” IEEE Trans. Signal Process., vol. 56, no. 1, pp. 218–232, Jan. 2007.
[13] B. Clerckx et al., “Is NOMA efficient in multi-antenna networks? A critical look at next generation multiple access techniques,” IEEE Open J. Commun. Soc., vol. 2, pp. 1310–1343, 2021.
[14] L. Yin and B. Clerckx, “Rate-splitting multiple access for multigroup multicast and multibeam satellite systems,” IEEE Trans. Commun., vol. 69, no. 2, pp. 976–990, Feb. 2020.
[15] A. Kaushik et al., “Hardware efficient joint radar-communications with hybrid precoding and RF chain optimization,” in Proc. IEEE Int. Conf. Commun. (ICC), Jun. 2021, pp. 1–6.
[16] F. Liu and C. Masouros, “Hybrid beamforming with sub-arrayed MIMO radar: Enabling joint sensing and communication at mmWave band,” in Proc. IEEE Int. Conf. Acoust., Speech Signal Process. (ICASSP), May 2019, pp. 7770–7774.
[17] M. Dai and B. Clerckx, “Multiuser millimeter wave beamforming strategies with quantized and statistical CSIT,” IEEE Trans. Wireless Commun., vol. 16, no. 11, pp. 7025–7038, Nov. 2017.
[18] Z. Li et al., “A general rate splitting scheme for hybrid precoding in mmWave systems,” in Proc. IEEE Int. Conf. Commun. (ICC), May 2018, pp. 1–6.
[19] L. You et al., “Beam squint-aware integrated sensing and communications for hybrid massive MIMO LEO satellite systems,” 2022, arXiv:2203.00235.
[20] L. Yin and B. Clerckx, “Rate-splitting multiple access for satellite-terrestrial integrated networks: Benefits of coordination and cooperation,” 2021, arXiv:2111.14074.
[21] F. Liu and C. Masouros, “A tutorial on joint radar and communication transmission for vehicular networks—Part I: Background and fundamentals,” IEEE Commun. Lett., vol. 25, no. 2, pp. 322–326, Feb. 2021.
[22] O. Dizdar et al., “Rate-splitting multiple access to mitigate the curse of mobility in (Massive) MIMO networks,” IEEE Trans. Commun., vol. 69, no. 10, pp. 6765–6780, Oct. 2021.
[23] M. F. Keskin et al., “MIMO-OFDM joint radar-communications: Is ICI friend or foe?” IEEE J. Sel. Topics Signal Process., vol. 15, no. 6, pp. 1393–1408, Nov. 2021.
[24] J. A. Zhang et al., “An overview of signal processing techniques for joint communications and radar sensing,” IEEE J. Sel. Topics Signal Process., vol. 15, no. 6, pp. 1295–1315, Nov. 2021.
[25] H. Chen et al., “Rate-splitting for multicarrier multigroup multicast: Precoder design and error performance,” IEEE Trans. Broadcast., vol. 67, no. 3, pp. 619–630, Sep. 2021.
[26] O. Dizdar and B. Clerckx, “Rate-splitting multiple access for communications and jamming in multi-antenna multi-carrier cognitive radio systems,” IEEE Trans. Inf. Forensics Security, vol. 17, pp. 628–643, 2022.