Reconfigurable Intelligent Surface-Empowered MIMO Systems

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Abstract—Reconfigurable intelligent surface (RIS)-assisted communications appear as a promising candidate for future wireless systems due to its attractive advantages in terms of implementation cost and end-to-end system performance. In this article, two new multiple-input multiple-output (MIMO) system designs using RISs are presented to enhance the performance and boost the spectral efficiency of state-of-the-art MIMO communication systems. Vertical Bell Labs layered space-time (VBLAST) and Alamouti’s schemes have been considered in this article and RIS-based simple transceiver architectures are proposed. For the VBLAST-based new system, an RIS is used to enhance the performance of the nulling and canceling-based suboptimal detection procedure as well as to noticeably boost the spectral efficiency by conveying extra bits through the adjustment of the phases of the RIS elements. In addition, RIS elements have been utilized in order to redesign Alamouti’s scheme with a single radio frequency signal generator at the transmitter side and to enhance its bit error rate (BER) performance. Monte Carlo simulations are provided to show the effectiveness of our system designs and it has been shown that they outperform the reference schemes in terms of BER performance and spectral efficiency.

Index Terms—Alamouti’s scheme, error probability analysis, multiple-input multiple-output (MIMO) systems, reconfigurable intelligent surface (RIS), Vertical Bell Labs layered space-time (VBLAST).

I. INTRODUCTION

RECONFIGURABLE intelligent surfaces (RISs) have received significant attention from the wireless communications community as effective, cheap, reconfigurable, easy to deploy, and passive system modules that can be used to control the wireless propagation environment by re-engineering the electromagnetic waves [1], [2]. Manipulating the propagation environment using RISs has been regarded as a promising candidate for the next-generation wireless technologies such as terahertz communications, nonorthogonal multiple access, and low-cost massive multiple-input multiple-output (MIMO) systems. Without loss of generality, mitigating the fading channel impairments and compensating for the propagation losses in order to enhance the signal quality at the receiver side are the main objectives behind the development of this technology. Nevertheless, RISs can also be utilized to minimize the transmitted signal power, to boost the system transmission capacity, and to enhance the physical layer security [3], [4].

In the preliminary study of [5], RISs are employed for two different purposes. First, an RIS is used to realize an ultra-reliable communication scheme that operates at considerably low signal-to-noise ratio (SNR) values. While in the second scenario, an RIS is used as an access point to create virtual phase-shift keying (PSK) symbols at the receiver side. The latter concept is also used to perform index modulation (IM) at the receiver side [6]. Considering a dual-hop communication scenario, Canbilen et al. [7] proposed an RIS-based space-shift keying system where the RIS is used as a reflector, which is positioned between a transmitter with multiple antennas and a receiver with a single antenna. In [8], the indoor multiple-user network-sharing capacity is enhanced by optimally adjusting the phases of a passive reconfigurable reflect array to cancel the interference and enhance the users’ signal quality. An energy-efficient multiple-input single-output (MISO) system is proposed in [9], by jointly optimizing the transmit powers of the users and the phases of RIS elements. In [10], Wu and Zhang proposed an RIS-assisted simultaneous wireless information and power transfer system, where an information decoding set and energy harvesting set of single-antenna receivers are served by a multiple-antenna access point. In [11], Guan et al. investigated the use of artificial noise in order to increase the secrecy rate in an RIS-based communication system, where a single-antenna user is served by a multiple-antenna transmitter in the presence of multiantenna eavesdroppers. In [12] and [13], the received signal power for a MISO user is maximized by optimizing the active beamforming at the transmitter jointly with the passive beamforming at the RIS by adjusting its phase shifters. The latter concept is also used in [14], where the beamformer at the access point and RIS are jointly optimized in order to increase the spectral efficiency for an RIS-assisted multiuser MISO system. In [15], an RIS-assisted multiuser MISO system is considered with different channel types where RIS phase optimization is utilized in order to maximize the minimum signal-to-noise-and-interference ratio. RIS is used in [16] to improve the channel rank for MIMO systems by adding additional multipaths with distinctively different spatial angles in addition to the low-rank direct channel path. RIS reflection coefficients and the transmit covariance matrix are jointly optimized in [17] in order to maximize the capacity of a point-to-point MIMO system. In [18], Chen et al. considered the channel estimation...
problem in multiuser MIMO systems and proposed an uplink channel estimation protocol to estimate the cascaded channel from the base station to the RIS and from the RIS to the user. The use of passive intelligent mirrors (PIMs) with a multiuser MISO downlink system is investigated in [19], where the transmit powers and the PIM reflection coefficients are designed to maximize the sum rate considering the individual quality of service for mobile users. An overview of the holographic MIMO surface is presented in [20], where the authors investigated their hardware architectures, functionalities, and characteristics. In [21], Huang et al. exploited the deep learning reinforcement learning algorithm to jointly optimize the beamforming matrix and the RIS phase shifts. Since the introduced algorithm learns directly from the environment and updates the beamforming matrix and RIS phase shifts accordingly. Based on the cosine similarity theorem, a low-complexity RIS phase-shift design algorithm is proposed in [22], where the RIS is used to assist a MIMO communication system. In [23] and [24], Basar et al. investigated physical channel modeling for mmWave bands considering indoor and outdoor environments. Furthermore, the authors provided an open-source comprehensive channel simulator that can be used to examine the different channel models discussed in their work. However, the use of RISs to boost the spectral efficiency and/or reliability of existing MIMO systems along with applications of IM is not well explored in the open literature. Against this background, two new RIS-assisted communication schemes are presented in this article by focusing on the integration of RISs into the existing MIMO systems in a simple and effective way. VBLAST [25] and Alamouti’s schemes [26] are considered in this article as the most common and practical MIMO schemes while a generalization to other advanced MIMO signaling schemes might be possible using our concept. We summarize the main contributions of this article as follows.

1) We propose an RIS-assisted Alamouti’s scheme in which we redesign the classical Alamouti’s scheme with a single RF signal generator at the transmitter side instead of two RF chains.

2) We show that our RIS-assisted Alamouti’s scheme preserves the diversity order of the classical Alamouti’s scheme and provides a significant BER performance enhancement.

3) We propose an RIS-assisted and IM-based VBLAST scheme using nulling and canceling-based suboptimal detection with a zero forcing (ZF) technique. Compared to the classical VBLAST scheme, we show that our proposed RIS-assisted and IM-based VBLAST scheme provides superior performance in terms of the spectral efficiency and the BER performance.

4) For RIS-assisted and IM-based VBLAST scheme, we propose two novel nulling-based optimal and suboptimal detectors to detect the indices of the antennas targeted by the IM.

In the RIS-assisted Alamouti’s scheme, at the receiver side, the classical Alamouti’s detector is used to recover the transmitted symbols, assuming that the channel state information (CSI) is available at this unit. It is worth noting that, for our RIS-assisted scheme, no CSI is needed at the RIS side, which reduces the overhead for CSI acquisition and simplifies the RIS design. Compared to the blind RIS-AP scheme in [5] and the classical Alamouti’s scheme, our results show that the proposed scheme provides a significant improvement in the BER performance. Furthermore, theoretical analysis and simulation results of the RIS-assisted Alamouti’s scheme show that the same concept can be generalized to space–time block code (STBC) systems. In other words, in a large-scale MIMO setup, our scheme can replace a large number of RF chains at the transmitter side by a single RF signal generator. Furthermore, our RIS-assisted scheme can provide a significant bit error rate (BER) performance enhancement while preserving the diversity order of the STBC system. In the RIS-assisted and IM-based VBLAST scheme, the RIS can be operated in multiple modes with and without IM. With IM, the RIS eliminates the channel phases between a specific transmit–receive antenna pair, which is selected according to the additional information bits in an IM fashion. On the other hand, without IM, the RIS eliminates the channel phases between a fixed and predefined antenna pair in order to provide the maximum BER performance enhancement. In this scheme, the CSI between each transmit–receive antenna pair through the RIS is required at both the RIS and the receiver side. At the receiver side, the IM bits are obtained using our novel nulling-based detectors while the transmitted symbols will be detected as in the plain VBLAST scheme using the ZF-based nulling and canceling algorithm without requiring additional signal processing steps [25]. The proposed schemes are simple in design and do not require major modifications for the state-of-the-art systems. Furthermore, comprehensive computer simulations are provided in this article under realistic environment setups to assess their practical feasibility.

The rest of this article is organized as follows. In Section II, we introduce the system model of the RIS-assisted Alamouti’s scheme and evaluate its symbol error probability (SEP). Section III introduces the RIS-assisted and IM-based VBLAST scheme, the nulling-based detectors, and the analysis of computational complexity of the receiver. In Section IV, we provide our computer simulation results along with comparisons along with path loss models considered in these simulations. Finally, conclusions are given in Section V.

II. RIS-ASSISTED ALAMOUTI’S SCHEME: SIGNAL MODEL AND ERROR PERFORMANCE ANALYSIS

In the proposed RIS-assisted Alamouti’s scheme, an unmodulated carrier signal is being transmitted from a low-cost RF signal generator close to the source (S) unit. The RF signal generator contains an RF digital-to-analog converter with an internal memory and a power amplifier as discussed in [5]. Fig. 1 shows the block diagram of the proposed scheme where $r_s$ and $r_d$ are the distances (in meters) of S-RIS and RIS-D, respectively. $r_s$ is selected in a way that the channel between S and RIS is assumed to be line-of-sight (LOS) dominated. In our setup, the RIS is divided into two parts each having $N/2$ elements adjusted to a common reflection phase value. Each part employs two different common reflection phase values over two time slots. The proposed system emulates the Alamouti’s scheme by...
adjusting the phases of the RIS elements to modify the RF carrier signal and invoke the phases of the two data symbols. In this way, the Alamouti’s scheme can be redesigned with a single RF signal generator instead of two full RF chains at the transmitter side. The RIS is a blind one with respect to the CSI, whereas its intelligence stems from the fact that it adjusts the unmodulated RF signal to mimic the PSK symbol phases. In this article, due to the high path loss experienced by the signals reflected from the RIS, the power of the signals reflected from the RIS for two or more times is ignored and only the first reflection is considered in our signal model [12].

Let \( h_i \) denotes the small-scale fading channel coefficient between the destination (D) and the \( i \)th element of the RIS, we have \( h_i \sim CN(0, 1) \) under Rayleigh fading assumption, where \( CN(0, \sigma^2) \) stands for complex Gaussian distribution with zero mean and \( \sigma^2 \) variance, \( \theta_0 \) and \( \theta_1 \) stand for the phases of two \( M \)-PSK symbols to be transmitted according to \( 2 \log_2(M) \) bits. Assuming quasi-static fading channels, where the channels will remain constant over the two time slots, the received signal at the first time slot can be written as

\[
    r_0 = \sqrt{P_L} \left[ \sqrt{E_s e^{j\theta_0}} \sum_{i=1}^{N/2} h_i + \sqrt{E_s e^{j\theta_1}} \sum_{i=N/2+1}^{N} h_i \right] + n_0
\]

(1)

where \( n_0 \) is the additive white Gaussian noise (AWGN) sample at the first time slot, i.e., \( n_0 \sim CN(0, N_0) \). \( E_s \) is the transmitted RF signal energy and \( \theta_0 \) and \( \theta_1 \) are the common RIS reflection phases for the first and second parts, respectively, for the first time slot. \( P_L \) is the total path gain (loss), and more details regarding the considered path loss model and environmental setups will be given in Section IV. According to the Alamouti’s transmission scheme, in the second time slot, we obtain the following received signal by carefully adjusting the common RIS phase terms as \(-(\theta_1 + \pi)\) and \(-\theta_0\) for the first and second parts, respectively:

\[
    r_1 = \sqrt{P_L} \left[ \sqrt{E_s e^{-j(\theta_1 + \pi)}} \sum_{i=1}^{N/2} h_i + \sqrt{E_s e^{-j\theta_0}} \sum_{j=N/2+1}^{N} h_j \right] + n_1
\]

(2)

where \( n_1 \sim CN(0, N_0) \). Defining \( s_0 = \sqrt{E_s} e^{j\theta_0}, s_1 = \sqrt{E_s} e^{j\theta_1}, A_0 = \sqrt{P_L} \sum_{i=1}^{N/2} h_i \), and \( A_1 = \sqrt{P_L} \sum_{i=N/2+1}^{N} h_i \), (1) and (2) can be re-expressed as

\[
    r_0 = s_0 A_0 + s_1 A_1 + n_0
\]

(3)

\[
    r_1 = -s_1^* A_0 + s_0^* A_1 + n_1
\]

(4)

where \( s_0 \) and \( s_1 \) stand for two virtual \( M \)-PSK symbols to be delivered to the receiver and \( M \) is the modulation order. As in the classical Alamouti’s scheme, the combiner will construct the combined signals as follows:

\[
    s_0 = r_0 A_0^* + r_1 A_1^* = (|A_0|^2 + |A_1|^2) s_0 + A_0^* n_0 + A_1 n_1^*
\]

(5)

\[
    s_1 = r_1 A_1^* - r_0 A_0^* = (|A_0|^2 + |A_1|^2) s_1 - A_0 n_1^* + A_1^* n_0.
\]

(6)

Then, \( s_0 \) and \( s_1 \) will be passed to the maximum-likelihood detector to estimate \( s_0 \) and \( s_1 \). Considering the symmetry of \( s_0 \) and \( s_1 \), the instantaneous received SNR per symbol can be obtained as

\[
    \gamma = \frac{(|A_0|^2 + |A_1|^2) E_s}{N_0}
\]

(7)

Considering \( h_i \sim CN(0, 1) \), we obtain \( A_0 \) and \( A_1 \sim CN(0, P_L E_s \frac{\pi}{2}) \). Consequently, \( \gamma \) becomes a central chi-square distributed random variable (RV) with four degrees of freedom with the following MGF [27]:

\[
    M_{\gamma}(s) = \left( \frac{1}{1 - \frac{\pi^2 P_L N E_s}{2 N_0}} \right)^2.
\]

(8)

From (8), the average SEP for \( M \)-PSK signaling can be obtained as [28]

\[
    P_e = \frac{1}{\pi} \int_{0}^{(M-1)\pi/M} \left( \frac{1}{1 + \frac{\sin((\pi/M)^2) P_L N E_s}{2 N_0 \sin(\eta)^4}} \right)^2 d\eta
\]

(9)

which can be simplified for BPSK as

\[
    P_e = \frac{1}{\pi} \int_{0}^{\pi/2} \left( \frac{1}{1 + \frac{P_L N E_s}{2 N_0 \sin(\eta)^4}} \right)^2 d\eta.
\]

(10)

From (9), we observe that a transmit diversity of order two is still achieved by this scheme due to the fact that the RIS mimics a similar transmission methodology to the Alamouti’s scheme over two time slots and preserves its orthogonality. Furthermore, the transmitted symbols have an SNR amplification achieved by the combination of the signals reflected from the RIS, where the SNR is enhanced by a factor of \( N \), as seen from (10).

III. RIS-ASSISTED AND IM-BASED VBLAST SYSTEM: THE SIGNAL MODEL

The proposed RIS-assisted and IM-based VBLAST scheme is assumed to utilize the RIS through a feedback link to eliminate the channel phases between a transmit–receive antenna pair. Instead of randomly selecting it, this transmit–receive antenna pair is selected according to the bits incoming to the RIS from S through the feedback link, in IM fashion. This means that the proposed scheme benefits from the RIS in an effective way.

Fig. 1. RIS-assisted Alamouti’s scheme with the RIS as the transmitter.
Consequently, with the help of the RIS, the proposed scheme provides a significant BER performance enhancement and an effective boost in the spectral efficiency compared to the classical VBLAST scheme. The proposed scheme is shown in Fig. 2, where an $N_t \times N_r$ VBLAST system is being operated along with an RIS with $N$ reflecting elements. The RIS-assisted and IM-based VBLAST scheme can be operated in the following three different modes:

1) full IM mode;
2) partial-IM mode;
3) enhancing mode.

Furthermore, we describe these three operating modes.

In the full-IM mode, at the transmitter side, the IM-based mapping and transmission can be described as follows. The incoming bits will be divided into two groups. The first group of $N_t \log_2 M$ bits will be used to select $N_t$ independent $M$-QAM symbols to be transmitted from the available $N_t$ transmitting antennas at S, as in classical VBLAST scheme.

The second group of $\log_2 (N_t N_r)$ bits, where $N_t N_r$ is assumed to be an integer power of two and corresponds to the all possible transmit–receive antenna combinations, is sent through a feedback link from S to the RIS. The RIS uses these incoming bits to select the indices of the transmit and receive antennas, shown by $l^*$ and $m^*$, corresponding to a $T_{l^*} - R_{m^*}$ pair of antennas, respectively. According to the CSI associated with the selected antenna pair, the phase shifts of the RIS elements will be adjusted to make the $T_{l^*}$-RIS-$R_{m^*}$ equivalent channel phases equal to zero. Consequently, the signals transmitted from $T_{l^*}$ and reflected from the RIS will be constructively combined to provide SNR amplification at $R_{m^*}$. The overall spectral efficiency of the system becomes $N_t \log_2 M + \log_2 (N_t N_r)$ bits per channel use (bpcu).

In the partial-IM mode, the transmission procedure is the same as in the full-IM mode except that a smaller set of the transmit–receive antenna combinations is used to convey the IM bits. That is, the index of the targeted transmit antenna is determined first and the same index is used for the targeted receive antenna, i.e., $T_{l^*} - R_{l^*}$. Hence, there are only $\log_2 N_t$ possible combinations that can be used to convey the IM bits. The resulting spectral efficiency under this mode will be $N_t \log_2 M + \log_2 N_t$ bpcu, where the motivation here is to sacrifice spectral efficiency gained by IM in order to further enhance the BER performance through increasing the reliability of IM bits.

Finally, in the enhancing mode, IM is not performed at all and, therefore, there is no need for a feedback link between S and D, instead, a fixed and predefined antenna pair will always be targeted by the RIS for the elimination of channel phases. Consequently, the best BER performance is achieved while preserving the same spectral efficiency for classical VBLAST, $N_t \log_2 M$. For all operating modes, we assume that perfect CSI is available at the both RIS and receiver side. The acquisition of CSI in RIS-based systems is discussed in [18], [29], and [30]. We describe our scheme with an example.

**Example:** A $4 \times 4$ RIS-assisted VBLAST system operated in the full-IM mode with QPSK modulation transmits the bit-stream of $[00 \ 01 \ 10 \ 11 \ 00 \ 01]$ as follows. The first eight bits are modulated to four QPSK symbols and transmitted, in parallel, from the available four transmit antennas. The remaining four bits are used by the RIS to select the pair $T_1 - R_2$, since we implement the following mapping rule: $00 \rightarrow 1$ and $01 \rightarrow 2$. The RIS adjusts the reflection phases to eliminate the channel phases between $T_1$ and $R_2$. Here, we assume that the adjustment of the phases of the RIS elements and the $M$-QAM symbols’ transmission from S are being performed simultaneously. On the other hand, the receiver tries to first estimate the indices of the selected antennas, and then successively detects the $N_t$ independent $M$-QAM symbols.

For an $N_t \times N_r$ VBLAST system assisted by an RIS with $N$ reflectors, the vector of the received signals $r \in \mathbb{C}^{N_r \times 1}$ can be written as [31]

$$
\mathbf{r} = \left[ \sqrt{P_{t,1}} \mathbf{G}_1^T \mathbf{H}_1 + \sqrt{P_{t,2}} \mathbf{H}_2 \right] \mathbf{x} + \mathbf{n} = \mathbf{Vx} + \mathbf{n}
$$

where $\mathbf{H}_1 \in \mathbb{C}^{N_r \times N_t}$ and $\mathbf{G}_1 \in \mathbb{C}^{N_r \times N_t}$ are the S-RIS and RIS-D uncorrelated Rician fading channel matrices, respectively. The $S$–$D$ channel matrix $\mathbf{H}_2 \in \mathbb{C}^{N_r \times N_t}$ is a random matrix where its elements are independent and identically distributed (i.i.d.) complex Gaussian RVs with zero mean and unit variance, i.e., $\sim \mathcal{C}\mathcal{N}(0, 1)$. $K$ is the Rician factor standing for the ratio of LOS and non-LOS power and $\mathbf{G} = \text{diag}(\exp(j \Phi_1), \ldots, \exp(j \Phi_n), \ldots, \exp(j \Phi_N))$ is the matrix of RIS reflection phases. In theory, each RIS element can be adjusted to any arbitrary phase shift, $\Phi_i \in [0, 2\pi]$; however, this is challenging to implement in practice. Therefore, in implementation, the RIS elements are being designed so that they can be adjusted to a finite number of discrete phase shifts. Hence, assuming uniform quantization for the interval $[0, 2\pi)$, the phase shift associated with each RIS element can be controlled by $b$ bits that correspond to $Z = 2^b$ possible different phase shifts belong to the finite set $\mathcal{F} = \{0, \Delta \Phi, \ldots, \Delta \Phi(Z-1)\}$, where $\Delta \Phi = \frac{2\pi}{2^b}$ [32]. $\mathbf{x} \in \mathbb{C}^{N_t \times 1}$ is the vector of transmitted $M$-QAM symbols, $\mathbf{V} \in \mathbb{C}^{N_r \times N_t}$ is the S–RIS–D and S–D equivalent channel matrix, and $\mathbf{n} \in \mathbb{C}^{N_r \times 1}$ is the vector of AWGN noise samples. $P_{t,1}$ and $P_{t,2}$ are the total path losses for S–RIS–D and S–D transmission paths, respectively, and more details are given in Section IV. The received signal by the $n$th receiving antenna ($R_m$) can be represented as

$$
R_m = \sqrt{P_{t,1} g_{m}^T \mathbf{H}_1} h_2 + n_m
$$

where $h_2 = \sqrt{P_{t,2}} h_2 \in \mathbb{C}^{1 \times N_t}$ is the $S$–$D$ channel vector for the $n$th receiving antenna. $g_m \in \mathbb{C}^{N_r \times 1}$ is the RIS-D channel vector for the $n$th receiving antenna, and $n_m$ is the AWGN sample with $n_m \sim \mathcal{C}\mathcal{N}(0, N_0)$. Expanding (12), we
have

\[ r_m = \sqrt{P_L I} \sum_{i=1}^{N_i} \sum_{l=1}^{N} h_{i(l)}^{(l)} e^{j\Phi_i} g_i^{(m)} x^{(l)} + h_2 x + n_m \]  

(13)

where \( h_{i(l)}^{(l)} = \alpha_i^{(l)} e^{-j\psi_{i(l)}} \) is the \( S \) (\( i \)th transmitting antenna)–RIS (\( l \)th element) channel coefficient and \( g_i^{(m)} = \beta_i^{(m)} e^{-j\psi_i^{(m)}} \) is RIS (\( i \)th element)–\( D \) (\( m \)th receiving antenna) channel coefficient. In this way, in order to eliminate the phases of the S–RIS–D channel between the antenna pair \( T_l - R_m \), elements’ phases of the RIS are adjusted as \( \Phi_i = \theta_i^{(l)} + \psi_i^{(m)} \), and (13) can be re-expressed for the \((m^*)\)th receive antenna as

\[
r_{m^*} = \sqrt{P_L I} \left[ \sum_{i=1}^{N_i} \sum_{l=1}^{N} \alpha_i^{(l)} \beta_i^{m^*} x^{(l)} \right] + \sum_{l=1,l \neq l^*}^{N_i-1} \sum_{i=1}^{N} h_{i(l)}^{(l)} e^{j\Phi_i} g_i^{(m^*)} x^{(l)} + h_2 x + n_{m^*}.
\]

(14)

Equation (14) can be interpreted as follows. The first term illustrates the amplification gained by the constructive combining of the signals reflected from the RIS and belongs to the symbol \( x^{(l^*)} \). This constructive combining will result in an SNR gain of \( N^2 \) for this symbol as in [5]. Hence, the BER performance of this symbol will be boosted up and, consequently, it will be the strongest symbol where the nulling and canceling algorithm starts with. This also means that the error propagation from the first symbol to the remaining ones will be significantly mitigated and the overall BER performance will be improved. The second term shows the destructive interference of the signals reflected from the RIS, which belongs to the other symbols. Finally, the third term corresponds to the interference received by the \((m^*)\)th receiving antenna through the S–D transmission path for all the transmitted symbols. Hence, compared to the classical VBLAST, an RIS will introduce \( N \) times extra interference for each received symbol. Nevertheless, assuming that the CSI over S–RIS–D is available at the receiver side, this interference can be handled readily.

A. Detection Algorithms

At the receiver side, we introduce two novel nulling-based detectors to detect the transmit–receive antenna indices targeted by the RIS for channel phases elimination. According to Algorithm 1, the optimal detector performs an exhaustive search for \( l^* \) and \( m^* \) jointly, as follows. For each iteration, the detector determines \( \Theta^{(l,m)} \) and then constructs \( \tilde{V}^{(l,m)} \). Next, the nulling-based procedure will be used to detect the first symbol, which has the highest SNR, assumed to be transmitted and received by the antenna pair \( T_l - R_m \). Finally, the pair \( l \) and \( m \) associated with the symbol that has the minimum squared Euclidean distance will be picked as the pair \( l^* \) and \( m^* \).

In Algorithm 1, \( \tilde{V}^{(l,m)} \) is the index-estimated S–RIS–D and S–D equivalent channel matrix assuming the antenna pair \( T_l - R_m \) was targeted by the RIS for channel phases elimination. \( \Theta^{(l,m)} \) is the diagonal RIS phases matrix where its \( i \)th element \( \Phi_i = \theta_i^{(l)} + \psi_i^{(m)} \) corresponds to the phase elimination for the S–RIS–D channel between \( T_l \) and \( R_m \). \( \cdot^+ \) is a Moore–Penrose pseudoinverse operator, and \( (\tilde{V}^{(l,m)})^+ = (\tilde{V}^{(l,m)\dagger}) (\tilde{V}^{(l,m)\dagger} \tilde{V}^{(l,m)})^{-1} \), where \( \dagger \) is the Hermitian operator. \( (\tilde{W}^{(l,m)})_{jk} \) is the \( j \)th row of \( \tilde{W}^{(l,m)} \), \( k_l \) is the index of the row, which has the minimum squared Euclidean norm, of the matrix \( \tilde{W}^{(l,m)} \), corresponding to the symbol with the highest SNR. \( (\tilde{W}^{(l,m)})_{ki} \) is the \( k \)th row of \( \tilde{W}^{(l,m)} \), and \( y_{ki} \) is the \( k \)th symbol after nulling the interference of the other symbols. \( Q(\cdot) \) returns the squared Euclidean distance \( D_{l,m} \) of the closest \( M\)–QAM symbol associated with \( y_{ki} \). Hence, \( \tilde{V}^{(l,m)} \) associated with the minimum distance \( D_{l,m} \) is the most likely equivalent channel matrix that corresponds to the current adjustment of the RIS phases. By estimating \( l^* \) and \( m^* \), the IM bits conveyed by the adjustment of the RIS phases will be obtained. In order to reduce the complexity of the detector proposed in Algorithm 1, the receiving antenna index \( m^* \) can be detected using a greedy detector instead of the joint exhaustive search for \( l^* \) and \( m^* \). In this way, \( m^* \) can be detected by finding the receiving antenna with the highest instantaneous energy

\[ \hat{m} = \arg \max_{m \in \{1,2,...,N_r\}} |r_m|^2. \]

(15)

Next, a nulling-based procedure is used to search for \( l^* \) while fixing \( \hat{m} \) found from (15). This detector is represented as a suboptimal one and Algorithm 2 illustrates its detection steps.

After obtaining the S–RIS–D and S–D equivalent channel matrix \( \tilde{V} \), it will be used by Algorithm 3 to detect the \( N_t \) independent symbols as in the case of classical VBLAST [25]. In Algorithm 3, \( Q(\cdot) \) is the slicing function, \( (\tilde{V})_{ki} \) is the \( k \)th column of \( \tilde{V} \) and \( (\tilde{V})^\top \) is pseudoinverse of the matrix obtained by zeroing the columns of \( \tilde{V} \) with indices \( k_1, k_2, ..., k_i \). Finally, \( r_{i+1} \) is the signal vector after subtracting the interference contribution of the previously detected symbol \( \tilde{z}_{k_i} \).
TABLE I

COMPUTATIONAL COMPLEXITY DERIVATION STEPS OF NULLING-BASED DETECTION ALGORITHMS

| Operation                                                                 | CMs |
|---------------------------------------------------------------------------|-----|
| Searching for $\hat{m}$                                                  | $N_r$ |
| Constructing $\tilde{V}(l,\hat{m})$                                      | $N_r + N_r N_t + N_r N_t$ |
| Pseudo inverse of $\tilde{V}(l,\hat{m})$                                | $2 N_r^2 N_t + N_r^3$ |
| Ordering $N_t N_r$                                                        | $N_t N_r$ |
| Nulling $N_t^2 N_r$                                                       | $N_t^2 N_r$ |
| Getting $D_{l,m}$                                                        | $2^M$ |

Algorithm 2: Sub-optimal Detector: Detecting the Antenna Indices $l^*$ and $m^*$ Sequentially.

Require: $H_1$, $H_2$, $G_1$, $r$, $\sqrt{P_L}$, $\sqrt{P_L}$

1: $\hat{m} = \arg\max_{m\in\{1,2,\ldots,N_r\}} |r_m|^2$
2: for $l = 1 : N_t$ do
3: $\tilde{V}(l,\hat{m}) = \sqrt{P_L} G_1^T \Theta(l,\hat{m}) H_1 + \sqrt{P_L} H_2$
4: $W(l,\hat{m}) = (\tilde{V}(l,\hat{m}))^+$
5: $k_l = \arg\min_{j\in\{1,2,\ldots,N_t\}} ||W(l,\hat{m})_j||^2$ Ordering
6: $s_{k_l} = (W(l,\hat{m}))_{k_l}$
7: $y_{k_l} = s_{k_l}^T r$ Nulling
8: $D_l = \hat{Q}(y_{k_l})$
9: end for
10: $\hat{l} = \arg\min_{l\in\{1,2,\ldots,N_t\}} D_l$
11: $\hat{V} = \sqrt{P_L} G_1^T \Theta(l,\hat{m}) H_1 + \sqrt{P_L} H_2$
12: return $\hat{V}$

Algorithm 3: ZF-Based Successive Nulling and Canceling to Detect the Transmitted $M$-QAM Symbols.

Require: $\tilde{V}$, $r$

1: $r_1 = r$
2: $W_1 = \tilde{V}^+$
3: $k_1 = \arg\min_{j\in\{1,2,\ldots,N_t\}} ||(W_1)_j||^2$
4: for $i = 1 : N_t - 1$ do
5: $s_{k_i} = (W_{i-1})_{k_i}$
6: $y_{k_i} = s_{k_i}^T r_i$
7: $\hat{x}_{k_i} = \tilde{Q}(y_{k_i})$ Slicing
8: $r_{i+1} = r_i - \hat{x}_{k_i}(\tilde{V})_{k_i}$ Canceling
9: $W_{i+1} = (\tilde{V}_{i+1})^+$
10: $k_{i+1} = \arg\min_{j\in\{1,2,\ldots,N_t\}\setminus\{k_1,\ldots,k_i\}} ||(W_{i+1})_j||^2$
11: end for
12: $s_{k_{i+1}} = (W_{i+1})_{k_{i+1}}$
13: $y_{k_{i+1}} = s_{k_{i+1}}^T r_{i+1}$
14: $\hat{x}_{k_{i+1}} = \tilde{Q}(y_{k_{i+1}})$
15: return $\hat{x}$

$C_3 = 4 N_r^3 + 3 N_r^2 + N_r 2^M + 3 N_r^4 + N_r + 2^M$. (21)

From (16) and (17), it can be seen that the number of antennas and the number of RIS reflecting elements both have a significant contribution to the computational complexity of detecting the antenna indices $l^*$ and $m^*$. This is still valid even if we consider the fact that $N_t$ is practically, much larger than $N_r$, since the latter still has a higher exponent. From (19) and (20), the overall computational complexity levels can be obtained as $O(N(N_r^2 + N_r^3))$ and $O(N(N_r^2 + N_r^3) + N_r^2)$, for Algorithms 1 and 2, respectively. Comparing both algorithms, we observe that the overall computational complexity is dominated by that of constructing $\tilde{V}(l,\hat{m})$. Compared to the classical VBLAST scheme, where the receiver computational complexity will be equivalent to that of Algorithm 3 only, i.e., $O(N_r^4)$, the RIS-assisted IM-based VBLAST scheme has an extra complexity of $O(N(N_r^2 + N_r^3) + N_r^2)$ and $O(N(N_r^2 + N_r^3))$ for Algorithms 1 and 2, respectively. The cost of this additional complexity stems from the construction of $\tilde{V}(l,\hat{m})$, which is required to detect the transmitted symbols and the indices of the targeted transmit–receiving antennas. In a brief, $C_1$ or $C_2$ is the additional computational complexity that is required to operate a classical VBLAST system as an RIS-assisted system.

IV. SIMULATION RESULTS

In this section, exhaustive computer simulations are provided for the proposed schemes against their counterparts. We consider realistic setups and path loss models where both transmitter and receiver located in an indoor environment and Fig. 3 shows the block diagrams of the benchmark schemes. In all simulations, the SNR is defined to be $E_s/N_0$. Furthermore, perfect CSI is assumed to be available for the proposed and benchmark schemes at the receiver side only, except RIS-assisted and IM-based VBLAST scheme where the CSI need also to be available at the RIS side.
Fig. 4. BER performance of the RIS-assisted Alamouti’s scheme versus classical Alamouti’s scheme and RIS-AP (blind) scheme, with different $N$ values and BPSK.

Fig. 5. BER performance of the RIS-assisted and IM-based VBLAST scheme versus classical VBLAST with a $2 \times 2$ MIMO setup, BPSK, and $K = -\infty$ dB.

In Figs. 5 and 6, we compare RIS-assisted and IM-based VBLAST with a classical VBLAST for a $2 \times 2$ MIMO setup. Algorithms 1 and 2 correspond to the optimal and suboptimal detectors for the transmit–receive antenna indices and they are
denoted in Figs. 5 and 6 by “opt.” and “sub-opt.,” respectively. Here, the spectral efficiency of classical VBLAST is 2 b/s/Hz, whereas it is 3 and 4 b/s/Hz for the RIS-assisted and IM-based VBLAST, in partial and full-IM modes, respectively, with BPSK. Thus, the spectral efficiency for the RIS-assisted IM-based VBLAST is significantly boosted compared to the classical VBLAST due to the extra IM bits conveyed by our antenna pair selection methodology.

In Fig. 5, we compare the BER performance of the RIS-assisted and IM-based VBLAST scheme, under the Rayleigh fading assumption for the S–RIS–D transmission path, with the classical VBLAST scheme. We observe that the proposed scheme operated in full-IM mode provides an improved BER at low to mid SNR values while saturating to the BER performance of the classical VBLAST at high SNR. Nevertheless, compared to the classical VBLAST scheme, the spectral efficiency is doubled for the proposed scheme. In addition, the suboptimal detector represented by Algorithm 2 is shown to achieve the same performance as for the optimal one, which is represented by Algorithm 1. This means that the detection process of the antenna indices can be simpler in terms of the complexity level. It can also be seen that the performance of the continuous phases adjustment can be almost achieved with only four discrete phase shifts [32], [36]. In this way, the RIS design can be simplified to control the phase shift of each element using 2 b only.

Furthermore, the partial-IM and enhancing modes show a significant BER performance gain of 16 and 28 dB, respectively, compared to the classical VBLAST scheme. It is worth noting that, the partial-IM scheme increases the spectral efficiency by 50%.

In Fig. 6, the BER performance curves are shown for the RIS-assisted and IM-based VBLAST scheme considering the existence of LOS component (K = 5 dB) between S and the RIS and between the RIS and D. Compared to the classical VBLAST scheme, the impact of the LOS component on the RIS performance can be clearly seen for the proposed scheme. The full-IM and partial-IM modes need an SNR increase of 5–10 dB in order to increase the spectral efficiency by 2 and 1.5 fold, respectively. This can be explained by the lack of diversity due to the LOS component in the S-RIS and RIS-D channels. Therefore, the receiver makes more errors while detecting the targeted antenna indices l∗ and m∗. Since the construction of V depends on l∗ and m∗, the erroneous detection of them reflects on the detection of the M-QAM symbols. Furthermore, it can be noted that the suboptimal (greedy) detector provides an error floor even with an RIS of 512 elements. This is, also, due to the LOS component, which makes the difference in the instantaneous energy, received by all the receiving antennas, trivial. Nevertheless, the enhancing mode still provides a BER performance gain of 4 dB compared to the classical VBLAST scheme.

Comparing the three operating modes, Figs. 5 and 6 show that the enhancing mode, where there is no IM, achieves the best BER performance. This is due to the fact that, in enhancing mode, the amplification is directed toward a fixed and predefined antenna pair. Furthermore, the proposed scheme has the flexibility to operate in one of these three modes according to the K value, hence, increase the spectral efficiency and/or enhance the BER performance accordingly.

Finally, it is worth noting that, from (22) and (23), comparing the S–RIS–D and S–D communication links, we obtain 7.86-dB additional path loss, under the given separation distances, for RIS-AP and RIS-assisted Alamouti schemes compared to the classical Alamouti’s scheme. In the same way, for the RIS-assisted and IM-based VBLAST scheme, where there are two communication links, there is an additional path loss of 12.29 dB for the S–RIS–D compared to the S–D communication link. This shows that the transmission over the RIS has considerably higher path loss compared to the direct transmission without RIS. Nevertheless, the additional path loss can be compensated by choosing the proper RIS size [34], [37], as we showed in our proposed schemes.

**V. Conclusion**

In this article, we have proposed novel designs for MIMO systems with the assistance of RISs. Although only VBLAST and Alamouti’s schemes have been considered in this article, our concepts can be applied for other MIMO schemes as well. Applying the considered concept to space–time codes in large-scale MIMO setups with a large number of antennas may show the remarkable advantage for this scheme by utilizing a single RF signal generator instead of multiple RF chains. Furthermore, an RIS-assisted Alamouti’s scheme is capable of achieving an N times SNR enhancement in addition to a transmit diversity order of two and the RIS-assisted IM-based VBLAST scheme is able to provide a significant BER performance gain in addition to the noticeable increasing in the spectral efficiency by the smart methodology of channel phases elimination. Considering the simplicity of implementation and deployment, both schemes do not require a significant reconfiguration for the existing MIMO setups, particularly in their receiver architectures, which makes them practical and feasible alternatives for future wireless systems. The design of a practical phase shift model, which captures the phase-dependent amplitude variation, for the proposed schemes appears as an interesting problem, which we will consider in our future research.
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