**Abstract**—Millimeter-wave (mmWave) multiple-input-multiple-output (MIMO) systems relying on lens antenna arrays are capable of achieving a high antenna-gain at a considerably reduced number of radio frequency (RF) chains via beam selection. However, the traditional beam selection network suffers from significant performance loss in wideband systems due to the effect of beam squint. In this paper, we propose a phase shifter-aided beam selection network, which enables a single RF chain to support multiple focused-energy beams, for mitigating the beam squint in wideband mmWave MIMO systems. Based on this architecture, we additionally design an efficient transmit precoder (TPC) for maximizing the achievable sum-rate, which is composed of beam selection and beamspace precoding. Specifically, we decouple the design problems of beamspace precoding and beam selection by exploiting the fact that the beam selection matrix has a limited number of candidates. For the beamspace precoding design, we propose a successive interference cancellation (SIC)-based method, which decomposes the associated optimization problem into a series of subproblems and solves them successively. For the beam selection design, we propose an energy-max beam selection method for avoiding the high complexity of exhaustive search, and derive the number of required beams for striking an attractive trade-off between the hardware cost and system performance. Our simulation results show that the proposed beamspace precoding and beam selection methods achieve both a higher sum-rate and a higher energy efficiency than its conventional counterparts.

**Index Terms**—Wideband mmWave MIMO, beam squint, lens antenna array, beamspace precoding, beam selection.

**I. INTRODUCTION**

**M**illimeter-wave (mmWave) communication has become a key technique for next-generation wireless communication systems owing to its substantial bandwidth [1]–[3], but unfortunately it has a high free-space path loss [4]. A promising technique for mitigating this problem is to involve massive multiple-input-multiple-output (MIMO) techniques [5]–[7]. Fortunately the mm-scale wavelengths allow for example 256 antenna elements to be packed in a relatively small physical size [4], which is capable of compensating for the high path loss with the aid of transmit precoder (TPC) [5]. Despite this potential, a range of practical challenges hamper the implementation of mmWave massive MIMO systems. Traditionally, MIMO systems tend to rely on the fully digital precoding, where each antenna is supported by a dedicated radio frequency (RF) chain. This leads to high hardware cost and high power consumption for mmWave massive MIMO systems relying on large antenna arrays [8].

Hence, hybrid TPC solutions have been proposed for circumventing this problem [9]–[13], where the TPC is decomposed into the digital precoding requiring a reduced number of RF chains and the large analog precoding, realized by analog phase shifters. Hybrid precoding combined with beam selection has been discussed in [14] and [15], when the phase shifters are used for analog precoding. A switching network is incorporated between the RF chains and phase shifters in [14] for reducing the hardware cost, while retaining good performance. The authors of [15] propose to add a fully-/sub-connected switching network between the phase shifters and the antennas for reducing the number of phase shifters without degrading the spectral efficiency. A particularly promising way of implementing the analog precoding is using the lens antenna array [16]–[20], which includes a lens and an antenna array located on the focal surface of the lens. Lenses focus the incident mmWave beams (signals) on different antennas. In this way, the traditional spatial channel is transformed into the so-called “beamspace channel.” Due to the limited scattering experienced at mmWave frequencies, the number of focused-energy beams of the beamspace channel is small [21]. Thus the transmitter can select a subset of focused-energy beams by switches for ensuring that the number of RF chains is reduced without significant performance loss [22], [23]. Then, the digital precoding is performed on the reduced-dimensional beamspace channel, which is termed as “beamspace precoding” in this paper.

However, designing the TPC consisting of beam selection and beamspace precoding is not a trivial task for wideband mmWave MIMO systems. Due to the effect of beam squint in wideband systems [24], the focused-energy beams of beamspace channels become frequency-dependent. However, the traditional beam selection network is frequency-independent [11], which will lead to considerable performance loss due to the power leakage of beamspace channel at certain frequencies. To overcome this problem, an intuitive technique is to select more beams to cover the entire channel bandwidth. However, this will unfortunately increase the number of RF chains, hence the hardware cost and power consumption as well. In conclusion, it is challenging to design the beam selection and beamspace precoding schemes for wideband systems to maintain satisfactory performance across the entire channel bandwidth without increasing the number of RF chains.

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A. Prior work

For the family of wideband mmWave MIMO systems relying on lens antenna arrays, the authors of [25], [26] adopted a single-carrier transmission scheme relying on a path delay-compensation technique, where the frequency-selective multi-user MIMO channels are transformed to several low-dimensional parallel frequency-flat MIMO channels for different users. Then, a joint antenna selection and beamforming scheme was proposed for eliminating the inter-user interference. The authors of [27] analyzed the effect of beam squint on the system’s performance loss and quantified the number of dominant beams required for maintaining a satisfactory performance. The results were obtained for single-input multiple-output systems communicating over line-of-sight channels. Gao et al. [28] proposed an adaptive beam selection network, which consists of a small number of 1-bit phase shifters. This architecture is used adaptively as a random combiner during the channel estimation while during data transmission as a traditional beam selection network. However, the performance erosion caused by the effect of beam squint was still not alleviated.

B. Contributions

In this paper, we propose a phase shifter-aided selection network combined with an efficient TPC design for wideband mmWave MIMO systems relying on lens antenna arrays. The main contributions of this paper are summarized as follows:

- We propose a phase shifter-aided selection network for mitigating the effect of beam squint. The wideband beamspace MIMO channel is frequency-dependent due to the beam squint, while the beam selection network is frequency-independent, which will lead to the power leakage of beamspace channel at certain frequencies. In the proposed selection network, we capture most of the channel’s output energy over the entire bandwidth without increasing the number of RF chains, where each RF chain is designed for supporting multiple focused-energy beams via a sub-array connected phase shifter network. Upon relying on a carefully designed TPC composed of beamspace precoding and beam selection, the proposed architecture achieves a near-unimpaired sum-rate despite relying on a reduced number of RF chains.

- To design an efficient TPC maximizing the sum-rate, we first decouple the design problems of beamspace precoding and beam selection by exploiting the fact that the number of candidate beam selection matrices is limited. For a given beam selection matrix, we propose a successive interference cancellation (SIC)-based beamspace precoding scheme, which is capable of achieving the maximum mutual information (MI) of wideband mmWave MIMO channels. Specifically, the beamspace precoding is realized by intrinsically amalgamating our baseband precoding and the phase shifter network. Given the sub-array connected structure of the phase shifter network, the optimization problem of MI maximization may be readily decomposed into several subproblems. Then, by appropriately adapting the classic concept of SIC signal detection [12], we propose a SIC-based beamspace precoding design, where each subproblem is solved after removing the contributions of the previously solved subproblems.

- Once the beamspace precoding method has been designed for a given beam selection matrix, the TPC design problem is reduced to the beam selection design problem, which can be solved by the optimal exhaustive search, but its complexity may still be excessive. To avoid the high complexity of exhaustive search, we develop an energy-max beam selection method. Specifically, we first prove that the energy-max beam selection design is capable of approaching the maximum MI, which means that the selection matrix should be designed to select the focused-energy beams of the wideband beamspace channel. Then, we derive the number of beams required for striking an attractive hardware cost/power consumption vs sum-rate trade-off. Extensive simulations verify the superior performance of the proposed beamspace precoding and beam selection methods.

The rest of the paper is organized as follows. In Section II, we present both our system and channel models. In Section III, we first propose a phase shifter-aided selection network for mmWave MIMO systems relying on lens antenna arrays. Then the SIC-based beamspace precoding and energy-max beam selection methods are developed. Our simulation results are provided in Section IV. Finally, our conclusions are drawn in Section V.

Notation: Lower-case and upper-case boldface letters denote vectors and matrices, respectively. $(\cdot)^T$, $(\cdot)^H$, $(\cdot)^*$, and $(\cdot)^{-1}$ denote the transpose, conjugate transpose, conjugate, and inverse of a matrix, respectively. $|\cdot|$ denotes the determinant of a matrix. $\lceil \cdot \rceil$ denotes rounding toward its nearest higher integer. $\text{tr}(\cdot)$ is the trace of a matrix. $||\cdot||_F$ is the Frobenius norm of a matrix. Finally, $I_N$ denotes the identity matrix of size $N \times N$.

II. System and Channel Models

In this section, we describe our wideband mmWave MIMO system relying on a lens antenna array and wideband beamspace channel model. The effect of beam squint over different frequencies is also highlighted.

A. System model

We commence by briefly introducing the wideband mmWave MIMO system model relying on a lens antenna array. We consider a MIMO aided orthogonal-frequency-division-multiplexing (OFDM) system using $K$ subcarriers. There are $N_t$ transmit antennas (TAs) and $N_{tRF}$ RF chains at the transmitter ($N_t \geq N_{tRF}$). At the receiver side, there are $N_r$ receive antennas (RAs) and $N_{rRF}$ RF chains ($N_r \geq N_{rRF}$). The transmitter conveys $N_s$ data streams at each subcarrier to the receiver, so that we have $N_s \leq N_{sRF}$ and $N_s \leq N_{sRF}$.

The transmitter is illustrated in Fig. 1 while the receiver relies on the inverse architecture. For the $k$-th subcarrier, the transmit data $s[k] \in \mathbb{C}^{N_r \times 1}$ ($k = 1, 2, \cdots, K$) is first precoded for reducing the interference between the different data
streams. The baseband precoding matrix $\mathbf{F}_{\text{BB}}[k] \in \mathbb{C}^{N_{\text{RF}} \times N_s}$ can adjust both the amplitude and phase of each data stream. Then the precoded data $\mathbf{F}_{\text{BB}}[k]s[k]$ is transformed into the time domain using the $K$-point inverse fast Fourier transform (IFFT). After adding CP before each OFDM symbol, the time-domain signals are transmitted through a subset of antennas selected by a selection network.

where $\mathbf{H}_k[k] \in \mathbb{C}^{N_r \times N_t}$ is the beamspace channel at the $k$-th subcarrier generated by the lens, which will be discussed in the following Subsection II-B. Finally, $\mathbf{n}[k] \sim \mathcal{N}(0, \sigma^2 \mathbf{I}_{N_r})$ is the additive white Gaussian noise (AWGN), where $\sigma^2$ is the noise power.

B. Beamspace channel model

To incorporate the multi-path structure of wideband mmWave MIMO channels, we adopt a ray-based channel model having $L$ clusters of scatterers \cite{31, 32}. Each cluster has a limited angle-of-arrival/departure (AoD/AoA) spread $\psi_t^s \in \psi_t^r$, and a complex path gain $\alpha_{s,k} (s=1,2,\ldots,S_{\ell})$. The spatial AoD and AoA at the $k$-th subcarrier are defined as $\phi_t^{s,k} = (d/\lambda_k) \sin \theta_t^s$ and $\phi_r^{s,k} = (d/\lambda_k) \sin \theta_r^s$, where $d$ is the antenna spacing and $\lambda_k$ is the wavelength of the $k$-th subcarrier. Under this model, the delay-$\ell$ tap $\mathbf{H}_d[k]$ of the wideband mmWave MIMO channel can be expressed as \cite{11, 32}.

$$\mathbf{H}_d = \sum_{\ell=1}^{L} \sum_{s=1}^{S_{\ell}} \alpha_{s,k} \mathbf{F}_{\text{BB}}[k] \mathbf{F}_{\text{BB}}[k]^H \mathbf{a}_r(\phi_r^{s,k}) \mathbf{a}_t(\phi_t^{s,k})^H,$$
that in narrowband systems. Specifically, by exploiting the fact that a lens computes a spatial Fourier transformation of incident signals, the spatial channel \( \mathbf{H}[k] \) is transformed into its beamspace representation \( \mathbf{H}_b[k] \) as [28, 30]

\[
\mathbf{H}_b[k] = \mathbf{U}^H \mathbf{H}[k] \mathbf{U}_t,
\]

where \( \mathbf{U}_t = \left[ \mathbf{a}_t \left( \phi_1^t \right), \mathbf{a}_t \left( \phi_2^t \right), \ldots, \mathbf{a}_t \left( \phi_{N_r}^t \right) \right] \) and \( \mathbf{U}_t = \left[ \mathbf{a}_t \left( \phi_1^t \right), \mathbf{a}_t \left( \phi_2^t \right), \ldots, \mathbf{a}_t \left( \phi_{N_t}^t \right) \right] \) are discrete Fourier transform (DFT) matrices, and \( \phi_r^t = \frac{\pi}{N_r} \left( n - \frac{N_r + 1}{2} \right) \) for \( n = 1, 2, \ldots, N_r \) and \( \phi_s^t = \frac{\pi}{N_t} \left( n - \frac{N_t + 1}{2} \right) \) for \( n = 1, 2, \ldots, N_t \). By substituting (6) into (7), we have

\[
\mathbf{H}_b[k] = \sum_{\ell=1}^L \sum_{s=1}^{S_\ell} \beta_{s_\ell,k} \mathbf{U}_r^H \mathbf{a}_r \left( \phi_{s_\ell,k}^r \right) \mathbf{a}_t^H \left( \phi_{s_\ell,k}^t \right) \mathbf{U}_t
\]

\[
= \sum_{\ell=1}^L \sum_{s=1}^{S_\ell} \beta_{s_\ell,k} \overline{\mathbf{a}}_r \left( \phi_{s_\ell,k}^r \right) \overline{\mathbf{a}}_t^H \left( \phi_{s_\ell,k}^t \right), \quad (8)
\]

where the antenna array response \( \overline{\mathbf{a}}_r \left( \phi_{s_\ell,k}^r \right) \in \mathbb{C}^{N_r \times 1} \) of the beamspace channel at the transmitter is given by

\[
\overline{\mathbf{a}}_r \left( \phi_{s_\ell,k}^r \right) = \mathbf{U}_r^H \mathbf{a}_r \left( \phi_{s_\ell,k}^r \right)
\]

\[
= \left[ \Xi_{N_r} \left( \phi_{s_\ell,k}^r - \bar{\phi}_1^r \right), \ldots, \Xi_{N_r} \left( \phi_{s_\ell,k}^r - \bar{\phi}_{N_r}^r \right) \right]^T,
\]

where \( \Xi_N(x) = \sum_{n=0}^{N-1} \frac{1}{N} e^{j2\pi nx} = \frac{\sin N\pi x}{N\sin \pi x} e^{j\pi(N-1)x} \). Similarly, the antenna array response \( \overline{\mathbf{a}}_t \left( \phi_{s_\ell,k}^t \right) \in \mathbb{C}^{N_t \times 1} \) of the beamspace channel at the receiver is given by

\[
\overline{\mathbf{a}}_t \left( \phi_{s_\ell,k}^t \right) = \mathbf{U}_t^H \mathbf{a}_t \left( \phi_{s_\ell,k}^t \right)
\]

\[
= \left[ \Xi_{N_t} \left( \phi_{s_\ell,k}^t - \bar{\phi}_1^t \right), \ldots, \Xi_{N_t} \left( \phi_{s_\ell,k}^t - \bar{\phi}_{N_t}^t \right) \right]^T.
\]

Observe that \( \Xi_N(x) \) has the following characteristics: \( \left| \Xi_N(x) \right| \approx 0 \) when \( |x| > 1/N \) [23]. Thus \( \overline{\mathbf{a}}_r \left( \phi_{s_\ell,k}^r \right) \) and \( \overline{\mathbf{a}}_t \left( \phi_{s_\ell,k}^t \right) \) can be regarded as sparse vectors. As a result, the beamspace channel \( \mathbf{H}_b[k] \) in (8) is a sparse matrix when the number of clusters \( L \) is limited and the AoD/AoA spread of each cluster is small, which are commonly assumed for mmWave channels [4]. In other words, the number of focused-energy beams in the beamspace channel is limited. Thus the selection network can select a subset of focused-energy beams, so that the number of RF chains and effective MIMO dimension is reduced without any substantial performance loss. However, as we have mentioned above, the spatial AoD \( \phi_{s_\ell,k}^r \) and spatial AoA \( \phi_{s_\ell,k}^t \) are frequency-dependent. Therefore, the beamspace channel \( \mathbf{H}_b[k] \) is also subcarrier-dependent, where both the values and positions of the non-zero elements of \( \mathbf{H}_b[k] \) vary with the subcarrier index. This effect of the frequency-dependent beamspace channel is termed ‘beam squint’ [27, 28, 34]. This is an important feature of the wideband beamspace channel that differentiates it from the traditional narrowband beamspace channel. Fig. 2 shows the power distribution of a single-path beamspace channel at different subcarriers with the physical AoD of \( \pi/4 \). The central carrier frequency is 28 GHz and the system bandwidth is 4 GHz. We observe that the beamspace channel is frequency-dependent.

III. PROPOSED BEAMSPACE PRECODING AND BEAM SELECTION

In this section, we first propose a phase shifter-aided selection network for coping with the effect of beam squint in wideband mmWave MIMO systems. Based on this architecture, a SIC-based beamspace precoding technique is presented for a given beam selection design. Then, a low-complexity energy-max beam selection method is proposed.

A. Phase shifter-aided selection network

Note that the beamspace MIMO channel is sparse associated with frequency-dependent non-zero elements (both the locations and values vary over frequencies). However, the selection network realized in the time-domain is frequency-independent, which will lead to power leakage at certain frequencies. To address this problem, we propose a phase shifter-aided selection network for mmWave MIMO systems relying on a lens antenna array, as shown in Fig. 3.

Fig. 3 illustrates the transmitter, while the receiver obeys the inverse architecture. The main difference between the proposed phase shifter-aided selection network and the traditional selection network seen in Fig. 1 is that \( N_t^{RF} \) RF chains are connected to \( N_t^B \) switches through a sub-array connected phase shifter network. Each switch is associated with two phase shifters and each RF chain is associated with \( [N_t^B/N_t^{RF}] \) switches. The two-phase-shifter structure is adopted for facilitating the amplitude variations of beamspace precoding [35], which will be explained in detail later. Within this architecture, the beamspace precoding is realized by the baseband precoding and the sub-array connected phase.
Through careful design of beam selection and beamspace precoding, the channel’s output energy across the entire bandwidth can be captured without increasing the number of RF chains. The design of beam selection and beamspace precoding methods will be discussed in the following Subsection III-B and Subsection III-C.

The model of the proposed phase shifter-aided mmWave MIMO system can be formulated as a modification of (2). Specifically, the received discrete-time complex baseband signal \( y[k] \) at the \( k \)-th subcarrier is given by

\[
y[k] = \sqrt{\rho/N_b} W_{BB}[k] \tilde{W}_{BB}[k] S[k] F_{PS} F_{BB}[k] s[k] + \tilde{W}_{BB}[k] \tilde{W}_{PS} S[k] n[k],
\]

(11)

where \( F_{PS} \in C^{N_r \times N_{RF}} \) is the precoding matrix realized by phase shifters, while \( N_t \) is the number of beams selected at the transmitter. Similarly, \( W_{PS} \in C^{N_r \times N_{RF}} \) is the combining matrix realized by phase shifters at the receiver, while \( N_r \) is the number of beams selected at the receiver. Note that \( F_{PS} \) and \( W_{PS} \) are realized relying on the sub-array connected phase shifters, i.e., each RF chain is associated with \( 2[N_B/N_t] \) phase shifters as shown in Fig. 3. Thus, \( F_{PS} \) and \( W_{PS} \) are block-diagonal matrices. Still referring to (11), \( S_t \in C^{N_t \times N_B} \) and \( S_r \in C^{N_r \times N_B} \) are selection matrices at the transmitter and the receiver, respectively. The baseband precoding matrix \( F_{BB}[k] \) and the phase shifter-aided precoding matrix \( F_{PS} \) satisfy the transmit power constraint of \( \text{tr} (F_{BB}[k] \tilde{F}_{BB}[k] F_{PS} F_{BB}[k]) = N_s \). The baseband combining matrix \( \tilde{W}_{BB}[k] \) and the phase shifter-aided combining matrix \( \tilde{W}_{PS} \) satisfy the similar power constraint of \( \text{tr} (W_{BB}[k] W_{PS} W_{BB}[k]) = N_s \). In the following subsections, we will focus our attention on the TPC design, which is composed of beamspace precoding and beam selection.

### B. SIC-based beamspace precoding

To obtain an efficient TPC maximizing the sum-rate, we first decouple the design problems of beam selection and beamspace precoding, based on the fact that the selection matrix is taken from a limited number of candidates. We consider the throughput optimization problem that captures the fact that the beam selection matrices \( S_t \) and \( S_r \) are taken from the sets of candidates \( S_t \) and \( S_r \).

\[
R^* = \max_{S_t, S_r, W_{PS}, W_{BB}[k], F_{PS}, F_{BB}[k]} R \quad \text{s.t.} \quad S_r \in S_r, \quad S_t \in S_t.
\]

(12)

The throughput \( R \) of the wideband channels is given by

\[
R = \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( I_{N_B} + \frac{\rho}{\sigma^2 N_s} \tilde{H}_b[k] F_{PS} F_{BB}[k] \tilde{W}_{BB}[k] F_{PS} \tilde{H}_b[k] \right),
\]

(13)

where \( C[k] = W_{PS} W_{BB}[k] (W_{BB}[k] W_{PS} W_{BB}[k])^{-1} W_{BB}[k] \). The reduced-dimensional beamspace channel \( \tilde{H}_b[k] \in C^{N_r \times N_B} \) after beam selection is defined as

\[
\tilde{H}_b[k] = S_t^H H_b[k] S_t.
\]

(14)

Since the selection matrices are taken from a limited number of candidates, the throughput optimization problem in (12) can be equivalently expressed in the following form

\[
R^* = \max_{S_t, S_r} \left\{ \max_{W_{PS}, W_{BB}[k]} \left\{ (F_{PS} F_{BB}[k]) R \right\} \right\}. \quad (15)
\]

The outer maximization is over the legitimate selection matrix candidates \( (S_t, S_r) \), while the inner maximization is obtained by finding the beamspace precoding \( F_{PS}, F_{BB}[k] \) and beamspace combining \( (W_{PS}, W_{BB}[k]) \) methods, given the selection matrices \( S_t, S_r \). In this way, we decouple the design problems of beam selection and beamspace precoding/combining.

However, unfortunately the joint optimization of the throughput over the TPC and receiver combiner is generally intractable [9], [32]. Hence, we follow the common assumption that the receiver combiner is optimal and focus our attention on the TPC design. The design ideas of TPC given in this paper can be directly used for the receiver combiner design [10]. Without considering the combining, the throughput tends towards the MI, and we design both \( F_{PS} \) as well as \( F_{BB}[k] \) to maximize the MI \( I \left( F_{PS}, F_{BB}[k], \tilde{H}_b[k] \right) \), which can be presented as [9], [10]

\[
I \left( F_{PS}, F_{BB}[k], \tilde{H}_b[k] \right) = \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( I_{N_B} + \frac{\rho}{\sigma^2 N_s} \tilde{H}_b[k] F_{PS} F_{BB}[k] F_{PS} \tilde{H}_b[k] \right).
\]

(16)

The overall problem of MI maximization is modeled as the following outer-inner problem form based on (15)

\[
I^* = \max_{S_t, S_r} \left\{ \max_{F_{PS}, F_{BB}[k]} \left\{ I \left( F_{PS}, F_{BB}[k], \tilde{H}_b[k] \right) \right\} \right\}. \quad (17)
\]

This subsection focuses on the inner problem formulated in (17). Our goal is to obtain the beamspace precoding scheme including the phase shifter-aided precoding matrix \( F_{PS} \) and the baseband precoding matrix \( F_{BB}[k] \), that maximizes \( I \left( F_{PS}, F_{BB}[k], \tilde{H}_b[k] \right) \) for a given \( S_t \) and \( S_r \). This problem is solved in two steps. The first step is to design the baseband precoding matrix \( F_{BB}[k] \) for a fixed \( F_{PS} \). The optimal baseband precoding matrix is given as \( F_{BB}[k] = (F_{PS} F_{PS})^{-1/2} V_{eff}[k] \Gamma_{eff}[k] \). Here \( V_{eff}[k] \in C^{N_{RF} \times N_s} \) gathers the right singular vectors corresponding to the \( N_s \) largest singular values of \( \tilde{H}_b[k] F_{PS} F_{PS} \tilde{H}_b[k] \) and \( \Gamma_{eff}[k] \in C^{N_s \times N_s} \) is a diagonal matrix of the power allocated to data streams according to the water filling solution. In the high signal-to-noise ratio (SNR) regime, we have \( \Gamma_{eff}[k] \approx I_{N_s} \). Note that \( F_{PS} \) is a block-diagonal matrix due to the sub-array connected structure of the proposed phase shifter-aided selection network. Therefore, by normalizing the precoding vector of each sub-array of phase shifters\(^2\), we have \( F_{PS}^H F_{PS} = \delta I_{N_{RF}} \), where \( \delta \) is a normalization coefficient ensuring the transmit power constraint of \( \text{tr} (F_{BB}[k] F_{PS} F_{BB}[k]) = N_s \). Substituting

\(^2\)As we will discuss later in this subsection, the precoding vector of each sub-array of phase shifters, i.e., the non-zero part of the column vectors of \( F_{PS} \) is a singular vector.
$F_{BB} [k] = (F_{BB}^H F_{PS}^{-1/2} V_{eff} [k] \Gamma_{eff} [k])$ into this constraint, we have $\delta = 1$. This property of $F_{PS}$ ensures that the optimal baseband precoding matrix for $N_t^{RF} = N_s$ typically satisfies $F_{BB} [k] F_{BB}^H [k] = I_{N_t^{RF}}$ in the high-SNR regime.

By substituting $F_{BB} [k] F_{BB}^H [k] = I_{N_t^{RF}}$ into (16), the problem of MI maximization only depends on $F_{PS}$. Thus, the second step of solving the maximization problem of (16) is to design $F_{PS}$ for maximizing the MI expressed as

$$I \left( F_{PS}; \hat{H}_b^H [k] \right) \leq \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H F_{PS} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right).$$

Employing Jensen’s inequality, (13) can be upper-bounded as

$$I \left( F_{PS}; \hat{H}_b^H [k] \right) \leq \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H F_{PS} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right).$$

Upon defining $R = \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H F_{PS}$, the desired phase shifter-aided precoding matrix $F_{PS}$ maximizing the upper bound of $I \left( F_{PS}; \hat{H}_b^H [k] \right)$ in (19) is given by

$$F_{PS}^* = \arg \max_{F_{PS}} \log_2 \left( I_{N_s} + \frac{\rho}{\sigma^2} F_{PS}^H R F_{PS} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right).$$

Once $F_{PS}^*$ has been obtained, the baseband precoding matrix is correspondingly given by $F_{BB} [k] = F_{PS}^* F_{PS}^{-1/2} V_{eff} [k] \Gamma_{eff} [k]$. Next, we present a SIC-based algorithm to solve (20).

Since $R$ is a Hermitian positive definite matrix, it can be decomposed as $R = Q^H Q$. Then, the optimization target of (20) can be presented as

$$\log_2 \left( I_{N_s} + \frac{\rho}{\sigma^2} F_{PS}^H Q F_{PS} \right) = \log_2 \left( I_{N_s} + \frac{\rho}{\sigma^2} F_{PS}^H Q F_{PS} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right).$$

By expressing $F_{PS} = [F_{PS,N_s-1}, f_{PS,N_s}]$, where $f_{PS,N_s}$ is the $N_s$-th column vector of $F_{PS}$ and $F_{PS,N_s-1}$ is the submatrix of $F_{PS}$ by removing $f_{PS,N_s}$, (21) is rewritten as [12]

$$\log_2 \left( I_{N_s} + \frac{\rho}{\sigma^2} Q F_{PS,N_s-1} F_{PS,N_s-1}^H \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right) \leq \log_2 \left( I_{N_s} + \frac{\rho}{\sigma^2} Q F_{PS} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right).$$

Algorithm 1 Proposed SIC-based beamspace precoding

1: Input: $T_1, G_1, F_{PS,0} = \emptyset$
2: for $1 \leq n \leq N_s$ do
3: Compute $f_{PS,n}$ by solving (23)
4: $F_{PS,n} = [F_{PS,n-1}, f_{PS,n}]$
5: Update $T_{n+1}$ and $G_{n+1}$
6: end for
7: Output: $F_{PS} = F_{PS,N_s}$
9: $F_{BB} [k] = (F_{PS}^H F_{PS})^{-1/2} V_{eff} [k] \Gamma_{eff} [k]$

where $T_{N_s} = I_{N_s} + \frac{\rho}{\sigma^2} Q F_{PS,N_s-1} F_{PS,N_s-1}^H \Gamma_{eff} [k] \Gamma_{eff} [k]$ and $T_1 = I_{N_s}$. Thus, the optimization problem of (23) can be decomposed into $N_s$ subproblems, where the $n$-th subproblem is

$$f_{PS,n}^* = \arg \max_{f_{PS,n}} \log_2 \left( 1 + \frac{\rho}{\sigma^2} F_{PS,n} Q F_{PS,n}^H \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right).$$

where $G_n = Q^H T_{n-1}^H$. Inspired by the SIC concept [12], we propose to successively solve the above $N_s$ subproblems. The proposed SIC-based beamspace precoding scheme is summarized in Algorithm 1 and explained as follows. The algorithm starts by solving the first subproblem (24). The first column of $F_{PS}$ is obtained as $f_{PS,1}^*$. Then, it is used to update the matrices $T_2 = I_{N_s} + \frac{\rho}{\sigma^2} Q F_{PS,2-1} F_{PS,2-1}^H Q^H$ and $G_2 = Q^H T_2^H Q$. Where $F_{PS,2-1} = f_{PS,1}$. The second column $f_{PS,2}$ of $F_{PS}$ is obtained by solving the second subproblem. Then we repeat this procedure until the desired precoding matrix $F_{PS}$ is obtained. Then, the baseband precoding matrix is designed as $F_{BB} [k] = (F_{PS}^H F_{PS})^{-1/2} V_{eff} [k] \Gamma_{eff} [k]$. Finally, we discuss how to solve (24) for obtaining $f_{PS,n}^*$. Defining a vector $\tilde{f}_{PS,n}$, which gathers the non-zero elements of $f_{PS,n}$, (24) can be rewritten as

$$\tilde{f}_{PS,n} = \arg \max_{\tilde{f}_{PS,n}} \log_2 \left( 1 + \frac{\rho}{\sigma^2} F_{PS,n} \tilde{G}_n \tilde{f}_{PS,n} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right) \left( \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( \frac{1}{\rho} \frac{\sigma^2}{\sigma^2 + \rho} F_{PS}^H \hat{H}_b^H [k] \right) \frac{1}{K} \sum_{k=1}^{K} \frac{1}{\rho} F_{PS}^H H_{b} [k] F_{PS} \right).$$

where $G_n = \tilde{G}_n \tilde{f}_{PS,n}$ and $\tilde{f}_{PS,n}$ corresponding to the largest singular value of $G_n$. [12].
Note the phase shifter-aided precoding matrix is realized by a sub-array connected phase shifter network and each element of \(v_n\) is realized by a pair of phase shifters, as shown in Fig. 3. Given that the amplitude of each element of \(v_n\) is not larger than 1, since \(v_n^H v_n = 1\), we can assume that each element of \(v_n\) is expressed as \(\alpha e^{j\beta}\) (\(\alpha \leq 1\)). Our target is to design two phase shifters with the phases of \(\beta_1\) and \(\beta_2\) to satisfy \(e^{j\beta_1} + e^{j\beta_2} = \alpha e^{j\beta}\), which is equivalently expressed as

\[
\begin{cases}
\cos (\beta_1 - \beta) + \cos (\beta_2 - \beta) = \alpha \\
\sin (\beta_1 - \beta) + \sin (\beta_2 - \beta) = 0
\end{cases}
\tag{26}
\]

Solving (26), we have \(\beta_1 = \cos^{-1} (\alpha/2) + \beta\) and \(\beta_2 = \beta - \cos^{-1} (\alpha/2)\).

We also estimate the computational complexity of the proposed Algorithm 1. Specifically, to obtain the phase shifter-aided precoding matrix \(F_{PS}\), we have to iteratively compute \(G_n = Q^H T_n^{-1} Q\) and the SVD of \(G_n\) \((n = 1, 2, \ldots, N_s)\). The complexity of computing \(G_n\) is \(O \left( 4N_t^3 + 2nN_t^2 \right) \) and that of computing the SVD of \(G_n\) is \(O \left( (N_t^B/N_t^RF)^3 \right) \). Therefore, the complexity of computing \(F_{PS}\) is \(O \left( 4N_sN_t^B + N_t(N_s + 1)N_t^B^2 + N_s(N_t^B/N_t^RF)^3 \right) \). The baseband precoding matrix \(F_{BB}[k]\) is computed by step 9, which requires the computational complexity of \(O \left( N_t^B/N_t^RF^2 \right) \). To sum up, the complexity of our proposed SIC-based beamspace precoding is \(O \left( 4N_sN_t^B^3 + N_t(N_s + 1)N_t^B^2 + N_s(N_t^B/N_t^RF)^3 + KN_t^B/N_t^RF^2 \right) \).

### C. Energy-max beam selection

This subsection focuses on the design problem of beam selection. The inner maximization problem in (17) is solved by the proposed SIC-based beamspace precoding scheme given the beam selection matrices \(S_t\) and \(S_s\). Then, the overall solution to the MI maximization problem in (17) can be obtained by solving the outer maximization problem through an exhaustive search over \(S_t\) and \(S_s\). Since the exhaustive search is of high complexity, in this subsection, we develop a low-complexity beam selection method.

The beamspace precoding scheme \((F_{PS}^*, F_{BB}^*[k])\) obtained through the inner problem maximization in (17) only depends on the reduced-dimensional beamspace channel \(\tilde{H}_b^H[k]\). Thus, the MI optimization problem (17) is rewritten as

\[
\mathcal{I} = \max_{S_t \in S_t, S_s \in S_s} \mathcal{I} \left( \tilde{H}_b^H[k] \right).
\tag{27}
\]

By using (18), we have

\[
\mathcal{I} \left( \tilde{H}_b^H[k] \right) = \frac{1}{K} \sum_{k=1}^{K} \log_2 \left| I_{N_s} + \frac{\rho}{\sigma^2 N_s} F_{PS}^* H_{H}^H[k] H_b[k] F_{PS}^* \right|.
\tag{28}
\]

Similar to the decoupling of the beamspace precoding and combining design, we focus our attention on the beam selection design at the transmitter and temporarily assume that \(S_t^H S_t^H = I_{N_t}\). The design method of the transmit beam selection can be directly used for the receive beam selection. Thus, the MI in (28) is rewritten as

\[
\mathcal{I} (S_t) = \frac{1}{K} \sum_{k=1}^{K} \log_2 \left| I_{N_s} + \frac{\rho}{\sigma^2 N_s} F_{PS}^* H_{H}^H[\alpha_e \tilde{H}_b[k] H_b[k] F_{PS}^*] \right| = \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( 1 + \frac{\rho}{\sigma^2 N_s} \lambda_s[k] \right),
\]

where (a) is obtained by the singular value decomposition (SVD) of the matrix \(H_{H}^H[k] S_t^H F_{PS}^*\) and \(\{\lambda_s[k]\}_{s=1}^{N_s}\) are the singular values.

By employing Jensen’s inequality, (29) is upper-bounded as

\[
I (S_t) \leq \frac{1}{K} \sum_{k=1}^{K} \log_2 \left( 1 + \frac{\rho}{\sigma^2 N_s} \sum_{s=1}^{N_s} \lambda_s^2[k] \right) \leq N_s \log_2 \left( 1 + \frac{\rho}{\sigma^2 N_s} \sum_{k=1}^{N_s} \lambda_s^2[k] \right)
\]

The transmit beam selection matrix \(S_t^*\) is designed for maximizing the upper bound of the MI (30). Due to the fact that \(\sum_{s=1}^{N_s} \lambda_s^2[k] = ||H_b[k] S_t F_{PS}^*||_F^2\), the selection matrix of the transmitter is designed as

\[
S_t^* = \arg \max_{S_t} \sum_{k=1}^{K} ||H_b[k] S_t F_{PS}^*||_F^2
\tag{31}
\]

We observe from (31) that to maximize the upper bound of the MI, the selection matrix \(S_t^*\) should select the specific transmit beams (i.e., columns of the beamspace channel \(H_b[k]\)), which are then combined by \(F_{PS}^*\) to capture as much of the channel’s output energy as possible. To solve (31), we first define

\[
\tilde{S}_t = S_t F_{PS}^*.
\tag{32}
\]

Since \(F_{PS}^*\) is a block-diagonal matrix, \(\tilde{S}_t \in C^{N_t \times N_s}\) is a matrix composed of \(N_s\) columns, each of which has \(\lfloor N_t^B/N_t^RF \rfloor\) non-zero elements. This means that each column of \(\tilde{S}_t\) can select \(\lfloor N_t^B/N_t^RF \rfloor\) transmit beams and combine them with different weights, which are determined by the diagonal block of \(F_{PS}^*\). Then we rewrite the optimization target in (31) as

\[
\sum_{k=1}^{K} ||H_b[k] S_t F_{PS}^*||_F^2 = \text{tr} \left( S_t^H \left( \sum_{k=1}^{K} H_b^H[k] H_b[k] \right) S_t \right).
\tag{33}
\]

Observe that \(\sum_{k=1}^{K} H_b^H[k] H_b[k]\) is a diagonal-dominant matrix \([33]\). The \(i\)-th diagonal element of \(\frac{1}{K} \sum_{k=1}^{K} H_b^H[k] H_b[k]\) represents the energy of the \(i\)-th transmit beam of the beamspace channel averaged over the entire bandwidth. Thus, if \(\tilde{S}_t\) is going to select \(N_t^B\) transmit beams and combine them as \(N_s\) beams to transmit signals, \(S_t^*\) should be designed to select the \(N_t^B\) largest diagonal elements\(^4\) of \(\frac{1}{K} \sum_{k=1}^{K} H_b^H[k] H_b[k]\).

\(^4\)The transmit beams associated with the \(N_t^B\) largest diagonal elements of \(\frac{1}{K} \sum_{k=1}^{K} H_b^H[k] H_b[k]\) are selected as candidate beams. Then, the indices of the candidate beams are sorted in an ascending order. Thus, the locations of non-zero elements in the columns of \(S_t^*\) are in an ascending order.
Similar to the transmitter, the selection matrix \( S_r^* \) at the receiver is designed to select the \( N_r^B \) largest diagonal elements of \( \frac{1}{K} \sum_{k=1}^{K} \mathbf{H}_b[k] \mathbf{S}_r^r \mathbf{S}_r^H \mathbf{H}_b^H[k] \) to capture as much of the energy of the receive beams of the beamspace channel as possible. The procedure of the energy-max beam selection method is shown in Fig. 4. The transmitter first selects \( N_t^B \) focused-energy transmit beams corresponding to the \( N_t^B \) largest diagonal elements of \( \frac{1}{K} \sum_{k=1}^{K} \mathbf{H}_b[k] \mathbf{S}_t^r \mathbf{S}_t^H \mathbf{H}_b^H[k] \). The selection matrix at the transmitter is denoted by \( \mathbf{S}_t^r \). Then the receiver selects \( N_r^B \) focused-energy receive beams corresponding to the \( N_r^B \) largest diagonal elements of \( \frac{1}{K} \sum_{k=1}^{K} \mathbf{H}_b[k] \mathbf{S}_r^r \mathbf{S}_r^H \mathbf{H}_b^H[k] \). The selection matrix at the receiver is denoted by \( \mathbf{S}_r^r \). Thus, the reduced-dimensional beamspace channel is obtained by \( \mathbf{H}_b[k] = \mathbf{S}_r^r \mathbf{H}_b[k] \mathbf{S}_t^r \).

Naturally, having more beams selected captures more energy of the beamspace channel. Ideally, setting \( N_t^B = N_t \) and \( N_r^B = N_r \) enables all the beamspace channel energy to be captured. However, increasing \( N_t \) and \( N_r \) leads to higher hardware cost and higher power consumption. Given the sparsity of the beamspace channel, a small number of transmit beams and receive beams are capable of capturing most of the channel’s output energy. It is expected that a better trade-off between the hardware cost/power consumption and the sum-rate performance can be achieved by appropriately choosing the values of \( N_t^B \) and \( N_r^B \). It is proved in Appendix that \( N_t^B \) and \( N_r^B \) should be set as

\[
N_t^B = \left\lceil \frac{L N_t B}{2 f_c} \right\rceil, \quad N_r^B = \left\lceil \frac{L N_r B}{2 f_c} \right\rceil.
\]

We observe that the number of required beams increases with the system bandwidth. This is because the effect of beam squint leads to heavier power leakage for a high bandwidth.

### IV. Simulation Results

In this section, we evaluate the performance of the proposed phase shifter-aided selection network as well as the associated TPC design through numerical simulations. The key system parameters are summarized in Table I. The delays of the dominant paths are assumed to follow the uniform distribution within \([0, 20](\text{ns})\) \([67]\). Each dominant path is contributed by 20 subpaths, whose delay offsets are within \([-0.1, 0.1](\text{ns})\).

The path gain \( \alpha_{kn} \) follows the complex Gaussian distribution \( CN(0, 1) \). The AoAs/AoDs of dominant paths are assumed to follow the uniform distribution within \([-1/2\pi, 1/2\pi]\). The maximum AoA/AoD spread of dominant paths is \(5^\circ\) \([38]\). We simulate the MI as defined in (17), which is achieved via the proposed SIC-based beamspace precoding and energy-max beam selection. We also present the energy efficiency (EE) defined as \( \eta = \mathcal{T} / P_{\text{total}} \), where \( P_{\text{total}} \) is the total energy consumption, defined as \( P_{\text{total}} = P_{RF} + P_{RF}^T + 2 N_t^B P_{PS} + N_r^B P_{\text{Switch}} \). Here, \( P_{RF} \), \( P_{PS} \) and \( P_{\text{Switch}} \) are the energy consumed by each RF chain, phase shifter and switch, respectively. We adopt the typical values of \( P_{RF} = 250 \text{ mW} \), \( P_{PS} = 10 \text{ mW} \), and \( P_{\text{Switch}} = 5 \text{ mW} \) \([29]\). \( P_c \) is the energy consumption per \(10^6\) elementary operations (MOps) of the digital signal processor (DSP) and \( N_c \) being the computational complexity as discussed in Section III.

For comparison, we also simulate the MI of the traditional wideband mmWave MIMO systems relying on the SVD-based TPC, which is defined as

\[
\mathcal{I}_{\text{SVD}} = \frac{1}{K} \sum_{k=1}^{K} \log_2 \left| I_{N_k^B} + \frac{\rho}{\sigma^2 N_k} \mathbf{H}_b[k] \mathbf{F}_B[k] \mathbf{F}_B^H[k] \mathbf{H}_b^H[k] \right|,
\]

where the SVD-based baseband precoding matrix \( \mathbf{F}_B[k] \) is composed of the right singular vectors of \( \mathbf{H}_b[k] \) corresponding to the \( N_r^B \) largest singular values \([56]\). The EE of the traditional SVD-based wideband mmWave MIMO systems is defined as \( \eta_{\text{SVD}} = \mathcal{I}_{\text{SVD}} / (\rho + p_c N_{\text{SV}} + N_t^B P_{RF} + N_r^B P_{\text{Switch}}) \), where \( N_{\text{SV}} \) is the computational complexity of the SVD-based TPC. For fair comparison, we adopt the low-complexity energy-max beam selection for both systems.

In Fig. 6 we present the MI vs the SNR, which is defined as \( \rho / \sigma^2 \). The system’s bandwidth is \( B = 2 \text{ GHz} \). The number of RF chains is \( N_r^B = 8 \text{ for the proposed SIC-based method, which is denoted by “SIC 8×8”} \). The number of data streams is set to \( N_t = N_r^B = 8 \text{ RF}. \) For the given system parameters, the number of selected beams should be set as \( N_t^B = N_r^B = 24 \) according to (34). The traditional SVD-based methods using 8 RF chains and 24 RF chains are denoted by “SVD 8×8” and “SVD 24×24”, respectively. We also present the fully digital precoding method using 64 RF chains at the transmitter and receiver as the benchmark. We observe that the proposed SIC-based method outperforms the traditional SVD-based methods, when the same number of RF chains is

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**Table I: System parameters for simulation**

| Parameter | Values |
|-----------|--------|
| Central carrier frequency \( f_c \) (GHz) | 28 |
| System’s bandwidth \( B \) (GHz) | 1 ~ 2 |
| \# of dominant channel paths \( L \) | 10 |
| \# of subpaths of each cluster \( N_l \) | 20 |
| \# of subcarriers \( K \) | 1024 |
| \# of IAs and RAs \( N_c = N_r \) | 64 |
| \# of data streams \( N_b \) | 4 ~ 10 |
| \# of RF chains \( N_r^B = N_r^T \) | 4 ~ 10 |
| Transmit power \( \rho \) (W) | 0.2 ~ 2 |
considered. As expected, the “SIC $8 \times 8$” method has worse performance than the “SVD $24 \times 24$” method, since the SIC-based method combines the selected 24 beams into 8 beams through a normalized precoding matrix $F_{PS}$. Thus, the array beamforming gain in the most desired direction is lower than that of the traditional “SVD $24 \times 24$” method. However, the traditional “SVD $24 \times 24$” method requires more RF chains, which will lead to higher hardware cost and higher energy consumption, as it will be verified in the following Fig. 6.

In Fig. 6, we present the EE against the transmit power $\rho$. We adopt the same simulation parameters as in Fig. 5. The noise power $\sigma^2$ is set as 0.01. We observe that the proposed SIC-based method has higher EE than that of both the traditional SVD-based methods and the fully digital method.

In Fig. 7 and Fig. 8, we present the MI and the EE against the bandwidth. We adopt the same simulation parameters as in Fig. 5 except for the bandwidth. The bandwidth is set to $1 \sim 4$ GHz, while the SNR is set to 20 dB. The number of selected beams is set to $N_B^D = N_B^R = 24$. We observe that both the proposed “SIC $8 \times 8$” method and the traditional “SVD $8 \times 8$” method have reduced MI and EE, when the bandwidth increases. This is because the effect of beam squint becomes serious, when the bandwidth increases. This means that the leaked energy of wideband channels increases with the bandwidth.
In Fig. 9 and Fig. 10, we present the MI and the EE against the number of data streams when SNR is set to 20 dB. We adopt the same simulation parameters as in Fig. 5 except for the number of data streams and RF chains. The number of data streams is set to $4 \sim 10$. The number of RF chains is $N_{RF}^t = N_{RF}^r = N_s$ for the proposed SIC-based method and for the traditional SVD-based method denoted by “SVD (one RF per stream)”. We denote the traditional SVD-based method using $3N_s$ RF chains at the transmitter and receiver as “SVD (three RF per stream)”. We also characterize the fully digital precoding method using $64$ RF chains. Fig. 9 shows that both the proposed SIC-based method, the traditional SVD-based methods, and fully digital method have increased MI with a large number of data streams. Furthermore, the proposed SIC-based method has higher MI than the traditional SVD-based method when the same number of RF chains is considered. The proposed SIC-based method has higher EE than the traditional SVD-based methods and the fully digital method.

In Fig. 11 and Fig. 12, we present the MI and EE against the number of selected beams when SNR is set to 20 dB. We adopt the same simulation parameters as in Fig. 5 except for the number of selected beams and RF chains. The number of selected beams is set to $16 \sim 48$. The number of RF chains is 8 for the proposed SIC-based method and $16 \sim 48$ for the traditional SVD-based method. We observe that the MI performance of the proposed SIC-based method becomes worse when the number of selected beams exceeds 32. This is because of the sub-optimality of the proposed beam selection and beamspace precoding methods, when the number of selected beams becomes very large. More explicitly, this is caused by the low-complexity sub-optimal energy-max beam selection design and the decoupling design of $F_{PS}$ and $F_{BB}$. Fig. 12 shows that $N_{B}^t = N_{B}^r = 24$ achieves a good trade-off between the MI performance and the hardware cost/power consumption, which is consistent with our analytical results.

**V. Conclusions**

In this paper, we proposed a phase shifter-aided selection network for wideband mmWave MIMO systems relying on lens antenna arrays for coping with the beam squint, where each RF chain can support multiple focused-energy beams through a sub-array connected phase shifter network. We also proposed an associated TPC design, which is composed of the SIC-based beamspace precoding scheme and the low-complexity energy-max beam selection scheme. Our simulations verified that the proposed beamspace precoding and beam selection design achieves higher sum-rate and EE than its conventional counterparts. For our future research, the proposed schemes may be extended to multi-user scenarios.

**Appendix**

To capture most of the channel’s output energy, the number of beams selected at the transmitter and receiver can be set as

$$N_{t}^B = \left\lceil \frac{LN_sB}{2f_c} \right\rceil, \quad N_{r}^B = \left\lceil \frac{LN_sB}{2f_c} \right\rceil.$$  \hspace{1cm} (36)

**Proof:** We first consider the case of single-path channels, i.e., $L = 1$. Then we will expand the results to the case
of multi-path channels having \( L \geq 2 \). For the single-path channels, the beamspace channel \( \mathbf{H}_b[k] \) can be simplified as
\[
\mathbf{H}_b[k] = \beta \hat{\mathbf{a}}_t(\phi^c_k) \mathbf{a}_t^H(\phi^s_k),
\]
where \( \beta \) is the path gain, \( \phi^c_k \) and \( \phi^s_k \) are the spatial AoD and AoA. The antenna array response vectors \( \hat{\mathbf{a}}_t(\phi^c_1) \) and \( \hat{\mathbf{a}}_t(\phi^c_2) \) are given by
\[
\hat{\mathbf{a}}_t(\phi^c_k) = \left[ \Xi_{N_t} \left( \phi^c_k - \phi^c_1 \right), \ldots, \Xi_{N_t} \left( \phi^c_k - \phi^c_{N_t} \right) \right]^T,
\]
\[
\hat{\mathbf{a}}_t(\phi^c_k) = \left[ \Xi_{N_t} \left( \phi^c_k - \phi^c_1 \right), \ldots, \Xi_{N_t} \left( \phi^c_k - \phi^c_{N_t} \right) \right]^T.
\]

The selection matrix at the transmitter is obtained by choosing the largest \( N^B_t \) diagonal element of \( \frac{1}{K} \sum_{k=1}^{K} \mathbf{H}_b[k] \mathbf{H}_b[k] \), which is given by
\[
\frac{\beta^2}{K} \sum_{k=1}^{K} \left[ \Xi_{N_t}^2 \left( \phi^c_k - \phi^c_1 \right) \right] \ldots \left[ \Xi_{N_t}^2 \left( \phi^c_k - \phi^c_{N_t} \right) \right]^{-1}. \tag{40}
\]
Observe that \( \Xi_N(x) \) has the following characteristics: \( \Xi_N(x) \approx 0 \) when \( |x| \gg 1/N \). Thus the diagonal element \( \Xi_{N_t}^2 \) is a non-negligible value only when \( \phi^c_k \approx \phi^c_0 \). Without loss of generality, we assume that \( \phi^c_k = \phi^c_0 > 0 \), where \( \phi^c_0 \) is the spatial AoD. Note that \( f_{k'} = f_k \left( 1 + \frac{\phi^c_k - \phi^c_{k'} - \phi^c_0}{\phi^c_0} \right) \) for \( k = 1, 2, \ldots, K \). Therefore, the spatial AoD \( \phi^c_k \) at the k-th subcarrier can be bounded as
\[
\phi^c_k \leq \phi^c_0, \tag{41}
\]
where \( \alpha \) is rewritten as
\[
\left( n - \frac{N_t + 1}{2} \right) \left( 1 - \alpha \right) \leq n \leq \left( n - \frac{N_t + 1}{2} \right) \left( 1 + \alpha \right), \tag{42}
\]
which can be simplified as
\[
n \leq n + \alpha \left( n - \frac{N_t + 1}{2} \right). \tag{43}
\]
This implies that \( \phi^c_k = \phi^c_0 \) may happen when \( n \) satisfies (43). Hence, the number of non-negligible diagonal elements of \( \frac{1}{K} \sum_{k=1}^{K} \mathbf{H}_b[k] \mathbf{H}_b[k] \) is smaller than \( 2 \alpha \left( n - \frac{N_t + 1}{2} \right) \). Since \( \frac{N_t + 1}{2} \leq n \leq N_t \), the number of non-negligible diagonal elements is smaller than \( \alpha \left( N_t - 1 \right) = \frac{B(K-1)(N_t-1)}{2K f_c} \). Therefore, the number of beams \( N^B_t \) selected at the transmitter can be approximated by \( \alpha \left( N_t - 1 \right) \).

For multi-path channels, the beamspace channel \( \mathbf{H}_b[k] \) is modeled as
\[
\mathbf{H}_b[k] = \sum_{l=1}^{L} \sum_{s=1}^{S_l} \beta_{s,l,k} \hat{\mathbf{a}}_t(\phi^s_{e,l,k}) \mathbf{a}_t^H(\phi^s_{e,l,k}). \tag{44}
\]
The expression of the diagonal elements of \( \frac{1}{K} \sum_{k=1}^{K} \mathbf{H}_b[k] \mathbf{H}_b[k] \) in (40) becomes very complicated.
\[
\text{diag}\left(\frac{1}{K} \sum_{k=1}^{K} H_b^H[k] H_b[k]\right) = \frac{1}{K} \sum_{k=1}^{K} \left[ \sum_{l=1}^{L} \sum_{s_l=1}^{S_l} \beta_{s_l,k} \Xi_{N_l}^2 \left( \phi_{t, \ell,k} - \bar{\phi}_{t, \ell, k} \right) \right].
\]
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