An Uplink-Downlink Duality for Cloud Radio Access Network

Liang Liu, Pratik Patil, and Wei Yu Department of Electrical and Computer Engineering University of Toronto, Toronto, ON, M5S 3G4, Canada Emails: lianguot.liu@utoronto.ca, {ppatil,weiyu}@comm.utoronto.ca

Abstract-Uplink-downlink duality refers to the fact that the Gaussian broadcast channel has the same capacity region as the dual Gaussian multiple-access channel under the same sumpower constraint. This paper investigates a similar duality relationship between the uplink and downlink of a cloud radio access network (C-RAN), where a central processor (CP) cooperatively serves multiple mobile users through multiple remote radio heads (RRHs) connected to the CP with finite-capacity fronthaul links. The uplink of such a C-RAN model corresponds to a multipleaccess relay channel; the downlink corresponds to a broadcast relay channel. This paper considers compression-based relay strategies in both uplink and downlink C-RAN, where the quantization noise levels are functions of the fronthaul link capacities. If the fronthaul capacities are infinite, the conventional uplinkdownlink duality applies. The main result of this paper is that even when the fronthaul capacities are finite, duality continues to hold for the case where independent compression is applied across each RRH in the sense that when the transmission and compression designs are jointly optimized, the achievable rate regions of the uplink and downlink remain identical under the same sum-power and individual fronthaul capacity constraints. As an application of the duality result, the power minimization problem in downlink C-RAN can be efficiently solved based on its uplink counterpart.

I. INTRODUCTION

As a promising candidate for the future 5G standard, cloud radio access network (C-RAN) enables a centralized processing architecture, using multiple relay-like base stations (BSs), named remote radio heads (RRHs), to serve mobile users cooperatively under the coordination of a central processor (CP). The practically achievable throughput of C-RAN is largely constrained by the finite-capacity fronthaul links between the RRHs and the CP. In the literature, a considerable amount of effort has been dedicated to the study of efficient fronthaul techniques [1]. In the uplink, the compression-based strategy can be used, where each RRH samples, quantizes and forwards its received signals to the CP over its fronthaul link such that each user's messages can be jointly decoded [2]. Likewise, in the downlink, the CP can pre-form the beamforming vectors, then compress and transmit the beamformed signals via the fronthaul links to RRHs for coherent transmission [3]. This paper focuses on these compression-based strategies and reveals an uplink-downlink duality for C-RAN.

Information theoretically, the uplink of C-RAN model corresponds to a multiple-access relay channel, while the downlink corresponds to a broadcast relay channel. When the fronthaul links have infinite capacities, the C-RAN model reduces to a multiple-access channel (MAC) in the uplink, and a broadcast channel (BC) in the downlink, for which there is a well-known uplink-downlink duality, i.e., the achievable rate regions with linear beamforming or the entire capacity regions of the MAC and the dual BC are identical under the same sum-power constraint [4]-[7]. In this paper, we extend such an uplinkdownlink duality relationship to the C-RAN model where the fronthaul capacities are finite. We show that if independent compression is applied across each RRH, the achievable rate regions of the uplink and downlink C-RAN are identical under the same sum-power and the same individual fronthaul capacity constraints when all the terminals are equipped with a single antenna. Specifically, given any downlink (uplink) power assignment, transmit (receive) beamforming and quantization noise levels, we construct the corresponding respective uplink (downlink) power assignment, receive (transmit) beamforming and quantization noise levels such that each user's access rate, each RRH's fronthaul rate, as well as the sum-power are all preserved. Based on this duality result, we further show that the downlink sum-power minimization problem can be efficiently solved based on its uplink counterpart.

Although this paper focuses on the compression-based strategies for C-RAN, we remark that other coding strategies can potentially outperform the compression strategy for certain channel parameters. For example, the so-called "compute-andforward" and "reverse compute-and-forward" strategies have been applied for the uplink and downlink C-RAN, respectively, where linear functions of transmitted codewords are computed and then sent between the RRHs and CP [8]. Moreover, in the downlink, the CP may opt to share user data directly with the RRHs [9]. However, the capacity of the uplink and downlink C-RAN model is still an open problem.

Notation: Scalars are denoted by lower-case letters, vectors denoted by bold-face lower-case letters, and matrices denoted by bold-face upper-case letters. I and 1 denote an identity matrix and an all-one vector, respectively, with appropriate dimensions. For a square matrix S, S^{-1} denotes its inverse (if S is full-rank). For a matrix M of arbitrary size, $[M]_{i,j}$ denotes the entry on the *i*th row and *j*th column of M. Diag (x_1, \dots, x_K) denotes a diagonal matrix with the diagonal elements given by x_1, \dots, x_K . $\mathbb{C}^{x \times y}$ denotes the space of $x \times y$ complex matrices.

II. SYSTEM MODEL

The uplink and downlink models for a C-RAN consist of one CP, M RRHs, and K users. It is assumed that all the RRHs



Fig. 1. System model of the uplink and downlink C-RAN.

and users are equipped with a single antenna. In the uplink, the channel from user k to RRH m is denoted by $h_{m,k}$; in the dual downlink, the channel from RRH m to user k is then given by $h_{m,k}^H$. The sum-power constraint is denoted by P for both the uplink and downlink. It is assumed that RRH m is connected to the CP via a noiseless digital fronthaul link with capacity C_m . In this paper, we focus on compression-based strategies to relay the information between the CP and RRHs via fronthaul links.

A. Uplink C-RAN

The uplink C-RAN model is as shown in Fig. 1(a). The discrete-time baseband channel between the users and RRHs can be modelled as

$$\begin{bmatrix} y_1^{\mathrm{ul}} \\ \vdots \\ y_M^{\mathrm{ul}} \end{bmatrix} = \begin{bmatrix} h_{1,1} & \cdots & h_{1,K} \\ \vdots & \ddots & \vdots \\ h_{M,1} & \cdots & h_{M,K} \end{bmatrix} \begin{bmatrix} x_1^{\mathrm{ul}} \\ \vdots \\ x_K^{\mathrm{ul}} \end{bmatrix} + \begin{bmatrix} z_1^{\mathrm{ul}} \\ \vdots \\ z_M^{\mathrm{ul}} \end{bmatrix}, \quad (1)$$

where $x_k^{\rm ul}$ denotes the transmit signal of user k, and $z_m^{\rm ul} \sim \mathcal{CN}(0, \sigma^2)$ denotes the additive white Gaussian noise (AWGN) at RRH m. In this paper, we assume that each of the users transmits using a Gaussian codebook, i.e., $x_k^{\rm ul} = \sqrt{p_k^{\rm ul} s_k^{\rm ul}}, \forall k$, where $s_k^{\rm ul} \sim \mathcal{CN}(0, 1)$ denotes the message of user k, and $p_k^{\rm ul}$ denotes the transmit power of user k.

After receiving the wireless signals from users, RRH m compresses y_m^{ul} and sends the compressed signals to the CP, $\forall m$. The quantization noise can be modeled as an independent Gaussian random variable, i.e,

$$\tilde{y}_{m}^{\text{ul}} = y_{m}^{\text{ul}} + e_{m}^{\text{ul}} = \sum_{k=1}^{K} h_{m,k} x_{k}^{\text{ul}} + z_{m}^{\text{ul}} + e_{m}^{\text{ul}}, \ \forall m, \quad (2)$$

where $e_m^{\rm ul} \sim \mathcal{CN}(0, q_m^{\rm ul})$, and $q_m^{\rm ul}$ denotes the variance of the compression noise at RRH m. After receiving the compressed signals, the CP first decodes the compression codewords and then applies linear beamforming to decode each user's message, i.e.,

$$\tilde{s}_k^{\mathrm{ul}} = \boldsymbol{w}_k^H \tilde{\boldsymbol{y}}^{\mathrm{ul}}, \quad \forall k,$$
 (3)

where $\boldsymbol{w}_k = [w_{k,1}, \cdots, w_{k,M}]^T$ with $\|\boldsymbol{w}_k\|^2 = 1$ denotes the decoding beamforming vector for user k's message, and $\tilde{\boldsymbol{y}}^{\text{ul}} = [\tilde{y}_1^{\text{ul}}, \cdots, \tilde{y}_M^{\text{ul}}]^T$ denotes the collective compressed signals from all RRHs.

The total transmit power of all the users is expressed as

$$P^{\rm ul}(\{p_i^{\rm ul}\}) = \sum_{i=1}^{K} \mathbb{E}[|x_i^{\rm ul}|^2] = \sum_{i=1}^{K} p_i^{\rm ul}.$$
 (4)

Moreover, in this paper we assume that the compression process is done independently across RRHs. Based on ratedistortion theory, the fronthaul rate for transmitting $\tilde{y}_m^{\rm ul}$ is expressed as

$$C_{m}^{\text{ul}}(\{p_{i}^{\text{ul}}\}, q_{m}^{\text{ul}}) = I(y_{m}^{\text{ul}}; \tilde{y}_{m}^{\text{ul}})$$

= $\log_{2} \frac{\sum_{i=1}^{K} p_{i}^{\text{ul}} |h_{m,i}|^{2} + q_{m}^{\text{ul}} + \sigma^{2}}{q_{m}^{\text{ul}}}, \forall m.$ (5)

Finally, with interference treated as noise, the achievable rate of user k is expressed as

$$R_{k}^{\text{ul}}(\{p_{i}^{\text{ul}}, \boldsymbol{w}_{i}\}, \{q_{m}^{\text{ul}}\}) = I(s_{k}^{\text{ul}}; \tilde{s}_{k}^{\text{ul}})$$

$$= \log_{2} \frac{\sum_{i=1}^{K} p_{i}^{\text{ul}} |\boldsymbol{w}_{k}^{H} \boldsymbol{h}_{i}|^{2} + \sum_{m=1}^{M} q_{m}^{\text{ul}} |\boldsymbol{w}_{k,m}|^{2} + \sigma^{2}}{\sum_{j \neq k} p_{j}^{\text{ul}} |\boldsymbol{w}_{k}^{H} \boldsymbol{h}_{j}|^{2} + \sum_{m=1}^{M} q_{m}^{\text{ul}} |\boldsymbol{w}_{k,m}|^{2} + \sigma^{2}}, \quad \forall k, \quad (6)$$

where $h_k = [h_{1,k}, \cdots, h_{M,k}]^T$ denotes the collective channel from user k to all RRHs.

Given the individual fronthaul capacity constraints C_m 's and sum-power constraint P, define the set of feasible transmit power, compression noise levels, and receive beamforming vectors as

$$\mathcal{T}^{\mathrm{ul}}(\{C_m\}, P) = \left\{ (\{p_i^{\mathrm{ul}}, \boldsymbol{w}_i\}, \{q_m^{\mathrm{ul}}\}) : P^{\mathrm{ul}}(\{p_i^{\mathrm{ul}}\}) \le P, \\ C_m^{\mathrm{ul}}(\{p_i^{\mathrm{ul}}\}, q_m^{\mathrm{ul}}) \le C_m, \forall m, \|\boldsymbol{w}_i\|^2 = 1, \forall i \right\}.$$
(7)

With independent compression and linear decoding, the achievable rate region in the uplink C-RAN is thus given by

$$\mathcal{R}^{\mathrm{ul}}(\{C_m\}, P) \triangleq \bigcup_{(\{p_i^{\mathrm{ul}}, \boldsymbol{w}_i\}, \{q_m^{\mathrm{ul}}\}) \in \mathcal{T}^{\mathrm{ul}}(\{C_m\}, P)} \{(r_1^{\mathrm{ul}}, \cdots, r_K^{\mathrm{ul}}) : r_k^{\mathrm{ul}} \le R_k^{\mathrm{ul}}(\{p_i^{\mathrm{ul}}, \boldsymbol{w}_i\}, \{q_m^{\mathrm{ul}}\}), \forall k\}.$$
(8)

B. Downlink C-RAN

The downlink C-RAN model is as shown in Fig. 1(b). The discrete-time baseband channel model between the RRHs and the users is the dual of the uplink channel given by

$$\begin{bmatrix} y_1^{\text{dl}} \\ \vdots \\ y_K^{\text{dl}} \end{bmatrix} = \begin{bmatrix} h_{1,1}^H & \cdots & h_{M,1}^H \\ \vdots & \ddots & \vdots \\ h_{1,K}^H & \cdots & h_{M,K}^H \end{bmatrix} \begin{bmatrix} x_1^{\text{dl}} \\ \vdots \\ x_M^{\text{dl}} \end{bmatrix} + \begin{bmatrix} z_1^{\text{dl}} \\ \vdots \\ z_K^{\text{dl}} \end{bmatrix}, \quad (9)$$

where $x_m^{\rm dl}$ denotes the transmit signal of RRH m, and $z_k^{\rm dl} \sim \mathcal{CN}(0, \sigma^2)$ denotes the AWGN at receiver k.

The transmit signals x_m^{dl} 's are compressed versions of the beamformed signals \tilde{x}_m^{dl} 's. Similar to (2), the compression noise is modelled as a Gaussian random variable, i.e.,

$$x_m^{\rm dl} = \tilde{x}_m^{\rm dl} + e_m^{\rm dl}, \quad \forall m, \tag{10}$$

where $e_m^{\rm dl} \sim \mathcal{CN}(0, q_m^{\rm dl})$, and $q_m^{\rm dl}$ denotes the variance of the compression noise at RRH m.

Similar to the uplink, a Gaussian codebook for each user is used for downlink transmission. Define the beamformed signal intended for user k to be transmitted across all the RRHs as $v_k \sqrt{p_k^{\text{dl}} s_k^{\text{dl}}}$, $\forall k$, where $s_k^{\text{dl}} \sim \mathcal{CN}(0,1)$ denotes the message for user k, p_k^{dl} denotes the transmit power, and $v_k = [v_{k,1}, \cdots, v_{k,M}]^T$ with $||v_k||^2 = 1$ denotes the transmit beamforming vector. The aggregate signal intended for all the RRHs is thus given by $\sum_{i=1}^{K} v_i \sqrt{p_i^{\text{dl}} s_i^{\text{dl}}}$, which is compressed and then sent to the RRHs via fronthaul links. The transmit signal across the RRHs is therefore expressed as

$$\begin{bmatrix} x_1^{\text{dl}} \\ \vdots \\ x_M^{\text{dl}} \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^K v_{i,1} \sqrt{p_i^{\text{dl}}} s_i^{\text{dl}} \\ \vdots \\ \sum_{i=1}^K v_{i,M} \sqrt{p_i^{\text{dl}}} s_i^{\text{dl}} \end{bmatrix} + \begin{bmatrix} e_1^{\text{dl}} \\ \vdots \\ e_M^{\text{dl}} \end{bmatrix}.$$
(11)

Then, the transmit power of all the RRHs is expressed as

$$P^{\rm dl}(\{p_i^{\rm dl}\}, \{q_m^{\rm dl}\}) = \sum_{m=1}^M \mathbb{E}[|x_m^{\rm dl}|^2] = \sum_{i=1}^K p_i^{\rm dl} + \sum_{m=1}^M q_m^{\rm dl}.$$
 (12)

Again, the compression is done independently across RRHs. As a result, the fronthaul rate for transmitting x_m^{dl} is expressed as

$$C_{m}^{\text{dl}}(\{p_{i}^{\text{dl}}, \boldsymbol{v}_{i}\}, q_{m}^{\text{dl}}) = I(\tilde{x}_{m}^{\text{dl}}; x_{m}^{\text{dl}})$$
$$= \log_{2} \frac{\sum_{i=1}^{K} p_{i}^{\text{dl}} |v_{i,m}|^{2} + q_{m}^{\text{dl}}}{q_{m}^{\text{dl}}}, \quad \forall m. \quad (13)$$

Finally, with linear encoding, the achievable rate of user k can be expressed as

$$R_{k}^{dl}(\{p_{i}^{dl}, \boldsymbol{v}_{i}\}, \{q_{m}^{dl}\}) = I(s_{k}^{dl}; y_{k}^{dl})$$

= $\log_{2} \frac{\sum_{i=1}^{K} p_{i}^{dl} |\boldsymbol{v}_{i}^{H} \boldsymbol{h}_{k}|^{2} + \sum_{m=1}^{M} q_{m}^{dl} |h_{m,k}|^{2} + \sigma^{2}}{\sum_{j \neq k} p_{j}^{dl} |\boldsymbol{v}_{j}^{H} \boldsymbol{h}_{k}|^{2} + \sum_{m=1}^{M} q_{m}^{dl} |h_{m,k}|^{2} + \sigma^{2}}, \quad \forall k.$ (14)

Given the individual fronthaul capacity constraints C_m 's and sum-power constraint P, define the set of feasible transmit power and beamforming vectors as well as compression noise levels as

$$\mathcal{T}^{\rm dl}(\{C_m\}, P) = \{(\{p_i^{\rm dl}, \boldsymbol{v}_i\}, \{q_m^{\rm dl}\}) : P^{\rm dl}(\{p_i^{\rm dl}\}, \{q_m^{\rm dl}\}) \le P, \\ C_m^{\rm dl}(\{p_i^{\rm dl}, \boldsymbol{v}_i\}, q_m^{\rm dl}) \le C_m, \forall m, \|\boldsymbol{v}_i\|^2 = 1, \forall i\}.$$
(15)

With independent compression and linear encoding, the achievable rate region in the downlink C-RAN is thus given

by

$$\mathcal{R}^{\mathrm{dl}}(\{C_m\}, P) \triangleq \bigcup_{(\{p_i^{\mathrm{dl}}, \boldsymbol{v}_i\}, \{q_m^{\mathrm{dl}}\}) \in \mathcal{T}^{\mathrm{dl}}(\{C_m\}, P)} \{(r_1^{\mathrm{dl}}, \cdots, r_K^{\mathrm{dl}}) : r_k^{\mathrm{dl}} \leq R_k^{\mathrm{dl}}(\{p_i^{\mathrm{dl}}, \boldsymbol{v}_i\}, \{q_m^{\mathrm{dl}}\}), \forall k\}.$$
(16)

III. UPLINK-DOWNLINK DUALITY FOR C-RAN

In this section, we establish a duality relationship for C-RAN under the compression strategy by comparing the rate regions $\mathcal{R}^{\mathrm{ul}}(\{C_m\}, P)$ and $\mathcal{R}^{\mathrm{dl}}(\{C_m\}, P)$. As mentioned before, if the fronthaul links between the RRHs and the CP have infinite capacities, i.e., $C_m \to \infty$, $\forall m$, then the uplink and downlink C-RAN models reduce to the MAC and BC, respectively, thus the conventional uplink-downlink duality applies, i.e., $\mathcal{R}^{\mathrm{ul}}(\{C_m \to \infty\}, P) = \mathcal{R}^{\mathrm{dl}}(\{C_m \to \infty\}, P)$. With finite values of C_m 's, however, it is not obvious if duality still holds. On one hand, the compression noise contributes to the transmit power in the downlink, but has no contribution to the transmit power in the uplink. On the other hand, background Gaussian noise contributes to the fronthaul rate in the uplink, but it does not affect the fronthaul transmission in the downlink. Interestingly, the following theorem shows that these two effects counteract with each other and the uplinkdownlink duality continues to exist in C-RAN even with finite fronthaul capacities.

Theorem 1: Consider a C-RAN model implementing compression-based strategies in both the uplink and the down-link, where all the users and RRHs are equipped with one single antenna. Then, any rate tuple that is achievable in the uplink is also achievable with the same sum-power and individual fronthaul capacity constraints in the downlink, and vice versa, i.e., $\mathcal{R}^{ul}(\{C_m\}, P) = \mathcal{R}^{dl}(\{C_m\}, P)$. *Proof:* First, we show that given any feasible uplink

Proof: First, we show that given any feasible uplink solution $(\{\bar{p}_i^{\text{ul}}, \bar{\boldsymbol{w}}_i\}, \{\bar{q}_m^{\text{ul}}\}) \in \mathcal{T}^{\text{ul}}(\{C_m\}, P)$, the following downlink problem is always feasible.

find
$$\{p_i^{\mathrm{dl}}, \boldsymbol{v}_i\}, \{q_m^{\mathrm{dl}}\}$$
 (17)

s.t. $R_k^{\text{dl}}(\{p_i^{\text{dl}}, \boldsymbol{v}_i\}, \{q_m^{\text{dl}}\}) = R_k^{\text{ul}}(\{\bar{p}_i^{\text{ul}}, \bar{\boldsymbol{w}}_i\}, \{\bar{q}_m^{\text{ul}}\}), \ \forall k, \ (18)$

$$C_m^{\rm dl}(\{p_i^{\rm dl}, \boldsymbol{v}_i\}, q_m^{\rm dl}) = C_m^{\rm ul}(\{\bar{p}_i^{\rm ul}\}, \bar{q}_m^{\rm ul}), \ \forall m,$$
(19)

$$P^{\rm dl}(\{p_i^{\rm dl}\}, \{q_m^{\rm dl}\}) = P^{\rm ul}(\{\bar{p}_i^{\rm ul}\}).$$
⁽²⁰⁾

According to (5), we have

$$\bar{q}_{m}^{\text{ul}} = \frac{\sum_{i=1}^{K} \bar{p}_{i}^{\text{ul}} |h_{m,i}|^{2} + \sigma^{2}}{\eta_{m}}, \quad \forall m,$$
(21)

where $\eta_m = 2^{C_m^{\text{ul}}(\{\bar{p}_i^{\text{ul}}\},\bar{q}_m^{\text{ul}})} - 1$. By substituting the quantization noise level from (21) in (6), it follows that

 $\gamma_k =$

$$\frac{\bar{p}_{k}^{\mathrm{ul}}|\bar{\boldsymbol{w}}_{k}^{H}\boldsymbol{h}_{k}|^{2}}{\sum_{j\neq k}\bar{p}_{j}^{\mathrm{ul}}|\bar{\boldsymbol{w}}_{k}^{H}\boldsymbol{h}_{j}|^{2} + \sum_{m=1}^{M}\frac{(\sum_{i=1}^{K}\bar{p}_{i}^{\mathrm{ul}}|h_{m,i}|^{2} + \sigma^{2})|\bar{\boldsymbol{w}}_{k,m}|^{2}}{\eta_{m}} + \sigma^{2}}, \quad \forall k,$$
(22)

where $\gamma_k = 2^{R_k^{\text{ul}}(\{\bar{p}_i^{\text{ul}}, \bar{\boldsymbol{w}}_i\}, \{\bar{q}_m^{\text{ul}}\})} - 1.$

Define $\bar{p}^{\mathrm{ul}}, \bar{\tau}^{\mathrm{ul}} \in \mathbb{C}^{K \times 1}$ with the *k*th elements denoted as \bar{p}_k^{ul} and $\sum_{m=1}^M |\bar{w}_{k,m}|^2 / \eta_m$, respectively, $\bar{D}^{\mathrm{ul}} = \mathrm{diag}(\gamma_1 / |\bar{w}_1^H h_1|^2, \cdots, \gamma_K / |\bar{w}_K^H h_K|^2)$, and $\bar{A}^{\mathrm{ul}} \in \mathbb{C}^{K \times K}$ with

$$\begin{bmatrix} \bar{\boldsymbol{A}}^{\mathrm{ul}} \end{bmatrix}_{i,j} = \begin{cases} \sum_{m=1}^{M} \frac{|h_{m,i}|^2 |\bar{\boldsymbol{w}}_{i,m}|^2}{\eta_m}, & \text{if } i = j, \\ |\bar{\boldsymbol{w}}_i^H \boldsymbol{h}_j|^2 + \sum_{m=1}^{M} \frac{|h_{m,j}|^2 |\bar{\boldsymbol{w}}_{i,m}|^2}{\eta_m}, & \text{otherwise.} \end{cases}$$
(23)

Then, we can rewrite (22) in the following matrix form:

$$\left(\boldsymbol{I}-\bar{\boldsymbol{D}}^{\mathrm{ul}}\bar{\boldsymbol{A}}^{\mathrm{ul}}\right)\bar{\boldsymbol{p}}^{\mathrm{ul}}=\sigma^{2}\bar{\boldsymbol{D}}^{\mathrm{ul}}\boldsymbol{1}+\sigma^{2}\bar{\boldsymbol{D}}^{\mathrm{ul}}\bar{\boldsymbol{\tau}}^{\mathrm{ul}}.$$
 (24)

Since $\bar{p}^{\text{ul}} > 0$ is a feasible solution to (24) by choice, according to [10, Theorem 2.1], it follows that $\rho(\bar{D}^{\text{ul}}\bar{A}^{\text{ul}}) < 1$, where $\rho(\cdot)$ denotes the spectral radius of the argument matrix.

Next, consider problem (17). It follows from constraint (19) that

$$q_m^{\rm dl} = \frac{\sum_{i=1}^{K} p_i^{\rm dl} |v_{i,m}|^2}{\eta_m}, \quad \forall m.$$
 (25)

By substituting the quantization noise level from (25) in (18), we have

$$\gamma_{k} = \frac{p_{k}^{dl} |\boldsymbol{v}_{k}^{H} \boldsymbol{h}_{k}|^{2}}{\sum_{j \neq k} p_{j}^{dl} |\boldsymbol{v}_{j}^{H} \boldsymbol{h}_{k}|^{2} + \sum_{m=1}^{M} \frac{\sum_{i=1}^{K} p_{i}^{dl} |v_{i,m}|^{2} |h_{m,k}|^{2}}{\eta_{m}} + \sigma^{2}}, \quad \forall k. \quad (26)$$

We can rewrite (26) in the following matrix form:

$$\left(\boldsymbol{I} - \boldsymbol{D}^{\mathrm{dl}}\boldsymbol{A}^{\mathrm{dl}}\right)\boldsymbol{p}^{\mathrm{dl}} = \sigma^{2}\boldsymbol{D}^{\mathrm{dl}}\boldsymbol{1},$$
 (27)

where $\boldsymbol{p}^{\mathrm{dl}} \in \mathbb{C}^{K \times 1}$ with the *k*th element denoted as p_k^{dl} , $\boldsymbol{D}^{\mathrm{dl}} = \mathrm{diag}(\gamma_1/|\boldsymbol{v}_1^H \boldsymbol{h}_1|^2, \cdots, \gamma_K/|\boldsymbol{v}_K^H \boldsymbol{h}_K|^2)$, and $\boldsymbol{A}^{\mathrm{dl}} \in \mathbb{C}^{K \times K}$ with

$$\left[\boldsymbol{A}^{\mathrm{dl}}\right]_{i,j} = \begin{cases} \sum_{m=1}^{M} \frac{|h_{m,i}|^2 |v_{i,m}|^2}{\eta_m}, & \text{if } i = j, \\ |\boldsymbol{v}_j^H \boldsymbol{h}_i|^2 + \sum_{m=1}^{M} \frac{|h_{m,i}|^2 |v_{j,m}|^2}{\eta_m}, & \text{otherwise.} \end{cases}$$
(28)

Next, we choose the downlink beamforming solution as

$$\boldsymbol{v}_k = \bar{\boldsymbol{w}}_k, \quad \forall k.$$
 (29)

Then we have $D^{dl} = \bar{D}^{ul}$ and $A^{dl} = (\bar{A}^{ul})^T$. As a result, (27) reduces to

$$\left(\boldsymbol{I} - \bar{\boldsymbol{D}}^{\mathrm{ul}} (\bar{\boldsymbol{A}}^{\mathrm{ul}})^T \right) \boldsymbol{p}^{\mathrm{dl}} = \sigma^2 \bar{\boldsymbol{D}}^{\mathrm{ul}} \boldsymbol{1}.$$
 (30)

Since $\bar{\boldsymbol{D}}^{\mathrm{ul}} \bar{\boldsymbol{A}}^{\mathrm{ul}}$ and $\bar{\boldsymbol{D}}^{\mathrm{ul}} (\bar{\boldsymbol{A}}^{\mathrm{ul}})^T$ possess the same eigenvalues, we have $\rho(\bar{\boldsymbol{D}}^{\mathrm{ul}} (\bar{\boldsymbol{A}}^{\mathrm{ul}})^T) = \rho(\bar{\boldsymbol{D}}^{\mathrm{ul}} \bar{\boldsymbol{A}}^{\mathrm{ul}}) < 1$. According to [10, Theorem 2.1], this implies the existence of a feasible power solution $\boldsymbol{p}^{\mathrm{dl}} > \boldsymbol{0}$ to equation (27), which is given by

$$\boldsymbol{p}^{\text{dl}} = \left(\boldsymbol{I} - \bar{\boldsymbol{D}}^{\text{ul}} (\bar{\boldsymbol{A}}^{\text{ul}})^T \right)^{-1} \sigma^2 \bar{\boldsymbol{D}}^{\text{ul}} \boldsymbol{1}.$$
 (31)

By substituting the above downlink power solution into (25), the corresponding quantization noise power levels can be obtained.

Given any feasible uplink solution $(\{\bar{p}_i^{\text{ul}}, \bar{w}_i\}, \{\bar{q}_m^{\text{ul}}\}) \in \mathcal{T}^{\text{ul}}(\{C_m\}, \bar{P})$, we have constructed a downlink solution as given by (29), (31) and (25) which satisfies constraints (18) and (19) in problem (17). Next, we show that this solution also satisfies the sum-power constraint (20):

$$P^{dl}(\{p_{i}^{dl}\},\{q_{m}^{dl}\}) = \sum_{i=1}^{K} p_{i}^{dl} + \sum_{m=1}^{M} q_{m}^{dl}$$

$$= \sum_{i=1}^{K} p_{i}^{dl} + \sum_{m=1}^{M} \sum_{i=1}^{K} \frac{p_{i}^{dl} |v_{i,m}|^{2}}{\eta_{m}} = \mathbf{1}^{T} \mathbf{p}^{dl} + (\bar{\boldsymbol{\tau}}^{ul})^{T} \mathbf{p}^{dl}$$

$$= \sigma^{2} (\mathbf{1} + \bar{\boldsymbol{\tau}}^{ul})^{T} \left(\mathbf{I} - \bar{\mathbf{D}}^{ul} (\bar{\mathbf{A}}^{ul})^{T} \right)^{-1} \bar{\mathbf{D}}^{ul} \mathbf{1}$$

$$= \sigma^{2} \mathbf{1}^{T} \left(\mathbf{I} - \bar{\mathbf{D}}^{ul} \bar{\mathbf{A}}^{ul} \right)^{-1} \bar{\mathbf{D}}^{ul} (\mathbf{1} + \bar{\boldsymbol{\tau}}^{ul})$$

$$= \mathbf{1}^{T} \bar{\mathbf{p}}^{ul} = \sum_{i=1}^{K} \bar{p}_{i}^{ul} = P^{ul}(\{\bar{p}_{i}^{ul}\}). \tag{32}$$

As result, given any feasible uplink solution $(\{\bar{p}_i^{\mathrm{ul}}, \bar{\boldsymbol{w}}_i\}, \{\bar{q}_m^{\mathrm{ul}}\}) \in \mathcal{T}^{\mathrm{ul}}(\{C_m\}, P)$, the constructed downlink solution given in (29), (31) and (25) satisfies all the constraints in problem (17). In other words, any rate tuple that is achievable in the uplink is also achievable in the downlink.

Similarly, we can show that given any feasible downlink solution $(\{\bar{p}_i^{\text{dl}}, \bar{v}_i\}, \{\bar{q}_m^{\text{dl}}\}) \in \mathcal{T}^{\text{dl}}(\{C_m\}, P)$, the following uplink problem is always feasible.

find
$$\{p_i^{\text{ul}}, \boldsymbol{w}_i\}, \{q_m^{\text{ul}}\}$$
 (33)

s.t.
$$R_k^{\text{ul}}(\{p_i^{\text{ul}}, \boldsymbol{w}_i\}, \{q_m^{\text{ul}}\}) = R_k^{\text{dl}}(\{\bar{p}_i^{\text{dl}}, \bar{\boldsymbol{v}}_i\}, \{\bar{q}_m^{\text{dl}}\}), \forall k,$$
(34)

$$C_m^{\mathrm{ul}}(\{p_i^{\mathrm{ul}}\}, q_m^{\mathrm{ul}}) = C_m^{\mathrm{dl}}(\{\bar{p}_i^{\mathrm{dl}}, \bar{\boldsymbol{v}}_i\}, \bar{q}_m^{\mathrm{dl}}), \ \forall m,$$
(35)

$$P^{\rm ul}(\{p_i^{\rm ul}\}) = P^{\rm dl}(\{\bar{p}_i^{\rm dl}\}, \{\bar{q}_m^{\rm dl}\}).$$
(36)

In other words, any rate tuple that is achievable in the downlink is also achievable in the uplink. Theorem 1 is thus proved.

Remark 1: Besides linear encoding and decoding, to achieve larger rate regions, dirty-paper coding can be used in the downlink and successive interference cancellation can be used in the uplink at the CP. Consider again the compression-based strategy C-RAN with single-antenna terminals. Using similar proof technique as for Theorem 1, it can be shown that by reversing the encoding and decoding order, any rate tuple that is achievable in the uplink is achievable with the same sum-power and individual fronthaul capacity constraints in the downlink, and vice versa. As a result, uplink-downlink duality also applies with non-linear encoding and decoding techniques.

Remark 2: In [7], it is shown based on the minimax optimization technique that the capacity region of the BC with perantenna power constraints is the same as that of its dual MAC with an uncertain noise at the receiver constrained by a certain positive semidefinite covariance matrix. Following a similar approach, it can be shown that the achievable rate region of the downlink C-RAN with per-RRH power constraints is the same as that of its dual uplink C-RAN with uncertain noises across all the RRHs.

IV. APPLICATION OF UPLINK-DOWNLINK DUALITY

This section illustrates an application of uplink-downlink duality. We first show that the sum-power minimization problem in the uplink C-RAN can be globally solved by the celebrated fixed-point method [11]. Then, the optimal solution to the downlink sum-power minimization problem is obtained based on the uplink solution. Specifically, in the uplink, the sumpower minimization problem is formulated as

$$\begin{array}{ll}
\begin{array}{l} \underset{\{p_{i}^{\mathrm{ul}}, \boldsymbol{w}_{i}\}, \{q_{m}^{\mathrm{ul}}\}}{\text{subject to}} & P^{\mathrm{ul}}(\{p_{i}^{\mathrm{ul}}\}) & (37) \\ \\ \underset{k}{\mathrm{subject to}} & R_{k}^{\mathrm{ul}}(\{p_{i}^{\mathrm{ul}}, \boldsymbol{w}_{i}\}, \{q_{m}^{\mathrm{ul}}\}) \geq R_{k}, & \forall k, \\ & C_{m}^{\mathrm{ul}}(\{p_{i}^{\mathrm{ul}}, q_{m}^{\mathrm{ul}}\}) \leq C_{m}, & \forall m, \end{array}$$

where R_k denotes the rate requirement of user k. It can be shown that with the optimal solution, the fronthaul capacities should be fully used, i.e., (21) holds with $\eta_m = 2^{C_m} - 1$, $\forall m$. Moreover, the optimal beamforming solution is the wellknown minimum-mean-square-error (MMSE) based receiver, i.e.,

$$\boldsymbol{w}_{k} = \left(\sum_{j \neq k} p_{j}^{\mathrm{ul}} \boldsymbol{h}_{j} \boldsymbol{h}_{j}^{H} + \boldsymbol{Q}^{\mathrm{ul}}(\boldsymbol{p}^{\mathrm{ul}}) + \sigma^{2} \boldsymbol{I}\right)^{-1} \boldsymbol{h}_{k}, \ \forall k, \ (38)$$

where $\boldsymbol{p}^{\mathrm{ul}} \in \mathbb{C}^{K \times 1}$ with the *k*th element denoted by p_k^{ul} , and $\boldsymbol{Q}^{\mathrm{ul}}(\boldsymbol{p}^{\mathrm{ul}}) = \mathrm{diag}(q_1^{\mathrm{ul}}, \cdots, q_M^{\mathrm{ul}})$ is a diagonal matrix with q_m^{ul} 's given by (21). By substituting the quantization noise levels and receive beamforming vectors with (21) and (38), problem (37) reduces to the following power control problem

$$\begin{array}{ll} \underset{\boldsymbol{p}^{\mathrm{ul}}}{\min \mathrm{inimize}} & P^{\mathrm{ul}}(\{p_i^{\mathrm{ul}}\}) & (39) \\ & \text{subject to} & \boldsymbol{p}^{\mathrm{ul}} \geq \boldsymbol{\Gamma}(\boldsymbol{p}^{\mathrm{ul}}), \end{array}$$

where $\Gamma(p^{\mathrm{ul}}) \in \mathbb{C}^{K \times 1}$ with the *k*th element as

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$$[\boldsymbol{\Gamma}(\boldsymbol{p}^{\mathrm{ul}})]_{k} = \frac{2^{R_{k}} - 1}{\boldsymbol{h}_{k}^{H} \left(\sum_{j \neq k} p_{j}^{\mathrm{ul}} \boldsymbol{h}_{j} \boldsymbol{h}_{j}^{H} + \boldsymbol{Q}^{\mathrm{ul}}(\boldsymbol{p}^{\mathrm{ul}}) + \sigma^{2} \boldsymbol{I}\right)^{-1} \boldsymbol{h}_{k}}, \quad (40)$$

It can be shown that $\Gamma(p^{\text{ul}})$ is a standard interference function [11]. According to [11, Theorem 2], the following fixed-point algorithm can converge to the globally optimal power solution to problem (39) given any initial point:

$$\boldsymbol{p}^{\mathrm{ul},(n+1)} = \boldsymbol{\Gamma}(\boldsymbol{p}^{\mathrm{ul},(n)}), \tag{41}$$

where $p^{\text{ul},(n)}$ denotes the power allocation solution obtained in the *n*th iteration of the above method. After the optimal power allocation is obtained, the optimal quantization noise levels and receive beamforming vectors to problem (37) can be obtained according to (21) and (38), respectively. Next, consider the sum-power minimization problem in the downlink.

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$$\begin{array}{ll} \underset{i}{\text{minimize}}{\text{minimize}} & P^{\text{dl}}(\{p_i^{\text{dl}}\}, \{q_m^{\text{dl}}\}) & (42) \\ \text{ubject to} & R_k^{\text{dl}}(\{p_i^{\text{dl}}, \boldsymbol{v}_i\}, \{q_m^{\text{dl}}\}) \geq R_k, \quad \forall k, \\ & C_m^{\text{dl}}(\{p_i^{\text{dl}}, \boldsymbol{v}_i\}, q_m^{\text{dl}}) \leq C_m, \quad \forall m. \end{array}$$

In general, the downlink sum-power minimization problem is more involved since unlike the uplink, the optimal transmit beamforming vectors are not easy to obtain. However, Theorem 1 indicates that the optimal values of problems (37) and (42) are identical. As a result, we can find the optimal solution to problem (37) first, based on which we can then construct the optimal downlink solution according to (29), (31), and (25).

Remark 3: Similar to the power minimization problem, uplink-downlink duality can be applied to solve the downlink weighted sum-rate maximization problem with per-RRH power constraints based on the dual uplink, which is typically easier to deal with (see Remark 2).

V. CONCLUSION

An uplink-downlink duality relationship is established for C-RAN. Specifically, we show that if transmission and compression designs are jointly optimized, the achievable rate regions of the uplink and downlink C-RAN are identical under the same sum-power and individual fronthaul capacity constraints when independent compression is performed across RRHs. Furthermore, this duality result proves useful for solving the downlink power minimization problem based on its dual problem in the uplink.

References

- O. Simeone, A. Maeder, M. Peng, O. Sahin, and W. Yu, "Cloud radio access network: virtualizing wireless access for dense heterogeneous systems," to appear in *J. Commun. Networks*, Apr. 2016. [Online]. Available: http://arxiv.org/abs/1512.07743
- [2] Y. Zhou and W. Yu, "Optimized backhaul compression for uplink cloud radio access network," *IEEE J. Sel. Areas Commun.*, vol. 32, no. 6, pp. 1295-1307, Jun. 2014.
- [3] S. H. Park, O. Simeone, O. Sahin and S. Shamai, "Joint precoding and multivariate backhaul compression for the downlink of cloud radio access networks," *IEEE Trans. Signal Process.*, vol. 61, no. 22, pp. 5646-5658, Nov. 2013.
- [4] F. Rashid-Farrokhi, K. J. R. Liu, and L. Tassiulas, "Transmit beamforming and power control for cellular wireless systems," *IEEE J. Sel. Areas Commun.*, vol. 16, no. 8, pp. 1437-1450, Oct. 1998.
- [5] P. Viswanath and D. N. C. Tse, "Sum capacity of the multiple antenna Gaussian broadcast channel and uplink-downlink duality," *IEEE Trans. Inf. Theory*, vol. 49, no. 8, pp. 1912-1921, Jul. 2003.
- [6] S. Vishwanath, N. Jindal, and A. Goldsmith, "Duality, achievable rates and sum rate capacity of Gaussian MIMO broadcast channels," *IEEE Trans. Inf. Theory*, vol. 49, pp. 2658-2668, Oct. 2003.
- [7] W. Yu, T. Lan, "Transmitter optimization for the multi-antenna downlink with per-antenna power constraints," *IEEE Trans. Signal Process.*, vol. 55, no. 6, pp. 2646-2660, Jun. 2007.
- [8] S. N. Hong and G. Caire, "Compute-and-forward strategies for cooperative distributed antenna systems," *IEEE Trans. Inf. Theory*, vol. 59, no. 9, pp. 5227-5243, Sep. 2013.
- [9] B. Dai and W. Yu, "Sparse beamforming and user-centric clustering for downlink cloud radio access network," *IEEE Access*, vol. 2, pp. 1326-1339, 2014.
- [10] E. Seneta, Non-Negative Matrices and Markov Chains: Springer, 1981.
- [11] R. D. Yates, "A framework for uplink power control in cellular radio systems," *IEEE J. Select. Areas Commun.*, vol. 13, pp. 1341-1347, Sep. 1995.