When Cell-Free Massive MIMO Meets OTFS Modulation: The Downlink Case

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I. INTRODUCTION

Future beyond-5G (B5G) wireless communication networks will provide ultra-reliable services as well as ubiquitous connectivity for a range of emerging mobile applications, including vehicle-to-vehicle (V2V) communications, high-speed railways, and unmanned aerial vehicles. However, wireless channels in high mobility environments are inherently linear time-variant fading channels, which are so called doubly-selective channels [1]. In practice, communications over these channels suffer from severe Doppler spread. Therefore, the work of [11], [12] cannot accommodate high Doppler spread applications. In order to serve high-mobility users with time-variant channels, the integration of cell-free massive MIMO and OFDM modulation is expected to further improve the network performance. To the authors’ best knowledge, the consolidation of OTFS modulation with cell-free massive MIMO has not been reported before. Thus, this paper will focus on the downlink achievable rate analysis of OTFS in cell-free massive MIMO. The main contributions of our work are as follows:

- We apply the embedded pilot-aided channel estimation with reduced guard interval and derive the minimum mean-square error (MMSE) estimate of the channel gains at the APs.
- We investigate the achievable downlink rate of OTFS modulation in cell-free massive MIMO with conjugate beamforming. We derive closed-form expression for the individual user downlink throughput for finite numbers of APs and users, taking into account the effects of channel estimation errors.
- Our findings demonstrate that a significant performance improvement can be achieved by the OTFS over the OFDM modulation in cell-free massive MIMO over high-mobility channels.

*Notation:* We use bold upper case letters to denote matrices, and bold lower case letters to denote vectors; the superscripts $(\cdot)^*\quad$ and $(\cdot)^\dagger\quad$ stand for the conjugate and
conjugate-transpose, respectively; $\text{tr}(A)$ denotes the trace of $A$; $\text{diag}(a_1, a_2, \ldots, a_n)$ denotes a square $n \times n$ diagonal matrix whose element in the $i$th row and $i$th column is $a_i$, $i = 1, \ldots, n$; the matrix $F_X = \begin{pmatrix} e^{-j2\pi k_0} & e^{-j2\pi k_1} & \cdots & e^{-j2\pi k_{N-1}} \end{pmatrix}$ denotes the unitary DFT matrix of dimension $X \times X$; the operator $\otimes$ denotes the Kronecker product of two matrices; $(\cdot)\%N$ denotes the modulo $N$ operation; $\mathbb{R}[\cdot]$ denotes the real part of the quantity within the brackets; $\mathbb{N}[a,b]$ represents the set of natural numbers ranging from $a$ to $b$; finally, $\mathbb{E}\{-\}$ denotes the statistical expectation.

II. SYSTEM MODEL

We consider a cell-free massive MIMO system consisting of $M_a$ APs and $K_u$ users. The APs and users are all equipped with single antenna, and they are randomly located in a large area. APs are connected to a central processing unit (CPU) via a backhaul network. The users are assumed to move at different speeds, thus the channels between the AP and users experience doubly-selective fading. An OTFS frame is divided into two phases: uplink payload transmission with channel estimation, and downlink payload transmission. Uplink and downlink transmissions are underpinned by time-division duplex (TDD) operation.

OTFS Modulation and Channel Model: Consider an OTFS system with $M$ sub-carriers having $\Delta f$ bandwidth each, and $N$ symbols having $T$ symbol duration. Therefore, the total bandwidth of the system is $M\Delta f$ and $NT$ is the duration of an OTFS block. Modulated data symbols of the $q$th user $\{x_q[k,\ell], k \in \mathbb{N}[0, N-1], \ell \in \mathbb{N}[0, M-1]\}$ are arranged over DD lattice $\Lambda = \{k/T, \ell/N\}$, where $k$ and $\ell$ represent the Doppler shift and delay index, respectively. Data symbols $x_q[k,\ell]$ are firstly converted to $X_q[n,m]$ in the TF domain through an ISFFT according to

$$X_q[n,m] = \frac{1}{\sqrt{MN}} \sum_{k=0}^{N-1} \sum_{\ell=0}^{M-1} x_q[k,\ell] e^{j2\pi (\frac{nT}{N} - \frac{m\ell}{M})},$$

where $n \in \mathbb{N}[0, N-1]$ and $m \in \mathbb{N}[0, M-1]$. Accordingly, by using Heisenberg transform, $X_q[n,m]$ is converted to a time domain signal as

$$s_q(t) = \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} X_q[n,m] g_{tx}(t-nT) e^{-j2\pi \Delta f (t-nT)},$$

where $g_{tx}(t)$ is the transmitter pulse of duration $T$.

The uplink channel between the $q$th user and the $p$th AP is given by [4]

$$h_{pq}(\tau, \nu) = \sum_{i=1}^{L_{pq}} \gamma_{pq,i} \delta(\tau - \tau_{pq,i}) \delta(\nu - \nu_{pq,i}),$$

where $L_{pq}$ denotes the number of paths from the $q$th user to the $p$th AP, $\tau_{pq,i}$, $\nu_{pq,i}$, and $\gamma_{pq,i}$ denote the delay, Doppler shift, and the channel gain, respectively, of the $i$th path of the $q$th user to the $p$th AP. The complex channel gains $\gamma_{pq,i}$ for different $(pq,i)$ are independent and identically distributed (i.i.d.) RVs with $\gamma_{pq,i} \sim \mathcal{CN}(0, \sigma_{pq,i}^2)$. The delay and Doppler shift for the $i$th path are given by $\tau_{pq,i} = \frac{k_{pq,i}}{\Delta f}$ and $\nu_{pq,i} = \frac{\ell_{pq,i}}{M} + \kappa_{pq,i}$, respectively, where $k_{pq,i} \in \mathbb{N}[0, M-1]$ and $\ell_{pq,i} \in \mathbb{N}[0, N-1]$ are the delay index and Doppler index of the $i$th path, and $\kappa_{pq,i}$ is $(-0.5, 0.5)$ is a fractional Doppler associated with the $i$th path. Let $k_{pq}$ and $\ell_{pq}$ denote the delay and Doppler taps corresponding to the largest delay and Doppler between the $q$th user and the $p$th AP. We note that the typical value of the sampling time $1/(M \Delta f)$ is usually sufficiently small in the delay domain. Hence, the impact of fractional delays in typical wideband systems can be neglected [4].

A. Uplink Payload Data Transmission

In the uplink, all $K_u$ users simultaneously send their data to the APs. The received signal at the $p$th AP is expressed as

$$r_p(t) = \sum_{q=1}^{K_u} h_{pq}(\tau, \nu)s_q(t-\tau)e^{j2\pi \nu(t-\tau)} dt \nu + w_p(t),$$

where $w_p(t)$ represents the noise signal in the time domain following a stationary Gaussian random process and we have $w_p(t) \sim \mathcal{CN}(0, \sigma_p^2)$ with $\sigma_p^2$ denoting the noise variance. The received signal is processed via a Wigner transform, implementing a receiver filter with an impulse response $g_{rx}(t)$ followed by a sampler, to obtain the received samples $\{y_p[n,m], n \in \mathbb{N}[0, N-1], m \in \mathbb{N}[0, M-1]\}$ in the TF domain

$$y_p[n,m] = \int r_p(t)g_{rx}(t-nT)e^{-j2\pi \Delta f (t-nT)} dt.$$  

Finally by applying a SFFT to $y_p[n,m]$, assuming that practical non-ideal rectangular waveforms are used for transmit and receive pulse shaping filters [4], the received signal at the $p$th AP in the DD domain can be written as

$$y_p[k,\ell] = \sqrt{\rho_u} \sum_{q=1}^{K_u} \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} b[k,\ell] \sum_{c=0}^{N-1/2} h_{pq}[k,\ell] \alpha[k,l,c] x_q[n-k+c,\ell-n] + w_p[k,\ell],$$

where $\rho_u$ is the normalized uplink signal-to-noise ratio (SNR); $b[k,\ell]$ is 0 for i.i.d. the path index, i.e., $b[k,\ell] = 1$ indicates there is a path with Doppler tap $k$ and delay tap $\ell$, otherwise $b[k,\ell] = 0$ (i.e., $\sum_{k=0}^{K_u-1} \sum_{\ell=0}^{M-1} b[k,\ell] = L_{pq}$). Moreover $\alpha[k,l,c]$ is given by [4]

$$\alpha[k,l,c] = \begin{cases} \frac{1}{N} \beta(c) e^{-j2\pi (k+c)\ell} e^{-j2\pi \nu \ell} & \ell \leq M \\ \frac{1}{N} \beta(c) e^{-j2\pi (k+c)\ell} e^{-j2\pi \nu \ell} & 0 < \ell < \ell' \end{cases}$$

where $\beta(c) = e^{-j2\pi \nu (c-1)\ell} - e^{-j2\pi \nu (-c-1)\ell}$. Finally by applying a SFFT to $y_p[k,\ell]$ following a stationary Gaussian random process, we have given $y_p[k,\ell] \sim \mathcal{CN}(0, \sigma_p^2)$

For the sake of simplicity of presentation and analysis, we consider the vector form representation of the input-output relationship of OTFS system in the DD domain. Let $v \in \{x_q, y_p, w_p\} \subset C^{MN \times 1}$ where $x_q$, $y_p$, and $w_p$ denote the vector of transmitted symbols from the $q$th user, the received signal vector at the $p$th AP, and the corresponding noise vector, respectively. Hence, the input-output relationship in (6) can be expressed in vector form as

$$y_p = \sum_{q=1}^{K_u} \sqrt{\rho_u} h_{pq} x_q + w_p$$

where $h_{pq} \in C^{MN \times MN}$ is the effective DD domain channel between the $q$th user and $p$th AP, given by [13]

$$h_{pq} = \sum_{i=1}^{L_{pq}} h_{pq,i} \mathbf{F}_N^\dagger \mathbf{f}_{pq,i} \mathbf{f}_{pq,i}^\dagger,$$

where $\mathbf{F}_N^\dagger = \mathbf{I}_M \mathbf{P}^\dagger_{diag} \mathbf{D}^\dagger_{k_{pq}+\kappa_{pq,i}} (\mathbf{F}_N \otimes \mathbf{I}_M)$ and $\mathbf{P}$ denotes a $MN \times MN$ permutation matrix, given in [13], and $\mathbf{D} = \text{diag} \{z^0, z^1, \ldots, z^{MN-1}\}$ is a diagonal matrix with $z = e^{j2\pi /N}$. 

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Channel estimation: By considering TDD operation, we rely on channel reciprocity to acquire channel state information (CSI). An intuitive method to estimate CSI is to transmit an impulse in the DD domain as the training pilot and then estimate the DD channel impulse response using the least square (LS) or MMSE estimator [5]. Transmitted impulse pilots are spread by the channel and interfere with data symbols in the DD domain. Therefore, insertion of guard symbols to avoid the interference between the pilot and data symbols is required [5], [14]. This technique, however, incurs a huge pilot overhead in cell-free massive MIMO. For instance, in the considered system, $K_u$ impulses are required to be transmitted. Assume that the DD channel responses of $K_u$ users have a finite support $[0, \ell_{max}]$ along the delay dimension and $[-k_{max}, k_{max}]$ along the Doppler dimension $\ell_{max} = \max q \epsilon q \text{pp}$ and $k_{max} = \max q \epsilon q \text{pp}$. Then, the guard intervals between two adjacent impulse along the Doppler and delay dimension should not be smaller than $2k_{max}$ and $\ell_{max}$, respectively. Moreover, users cannot use dedicated pilot and guard grids of each other for data transmission. As a result, the pilot length of pilot overhead should be $2K_uN\ell_{max}$. Therefore, with a large number of users, the pilot overhead would be prohibitively high.

As an alternative solution, we deploy embedded-pilot channel estimation method with reduced guard symbols, while users are allowed to use dedicated pilot and guard DD grids of each other for data transmission. Consider $\varphi_q$, with $\mathbb{E}\{\varphi_q^2\} = P_p$ denoting a known pilot symbol for the $q$th user at a specific DD grid location $[k_q, \ell_q]$. Let $x_{q}[k, \ell]$ denote the data symbol at grid point $[k, \ell]$, and assume that each pilot is surrounded by a guard region of zero symbols. Therefore, for the $q$th user the pilot, guard, and data symbols in the DD grid are arranged as

$$x_{q}[k, \ell] = \begin{cases} \varphi_q & k = k_q, \ell = \ell_q \\ 0 & k \in K, k \neq k_q, \ell \in \mathcal{L}, \ell \neq \ell_q \end{cases}$$

(9)

where $K = \{k_q - 2k_{max} - 2k; \ldots; k_q + 2k_{max} + 2k\}$, $\mathcal{L} = \{\ell - \ell_{max} \leq \ell \leq \ell_q + \ell_{max}\}$ and $k$ denotes the additional guard to mitigate the spread due to fractional Doppler shift [5]. In this case, the total overhead is $N_{\text{guard}} = (24k_{max} + 4k + 1)$ per each user. At the receiver, the received symbols $y_{p}[k, \ell]$, $k_q - k_{max} - k < k < k_q + k_{max} + k$, $\ell_q \leq \ell \leq \ell_q - \ell_{max}$ are used for channel estimation. Therefore, from (6), we have

$$y_{p}[k, \ell] = \sqrt{P_p} b[\ell - \ell_q] h_{pq}[k-k_q] N, (\ell - \ell_q)_M \varphi_q + I_q(k, \ell, \ell_q) + I_q(k, \ell, \ell_q) + w[k, \ell],$$

(10)

where

$$b[\ell - \ell_q] = \begin{cases} 1, & \sum_{k' = 0}^{k_{pq}} b[k', \ell - \ell_q] \geq 1 \\ 0, & \text{otherwise} \end{cases}$$

is the path indicator and

$$\bar{h}_{pq}[k-k_q] N, (\ell - \ell_q)_M = \sum_{k' = 0}^{k_{pq}} b[k', \ell - \ell_q] h_{pq}[k', \ell - \ell_q] \alpha(k, \ell, c).$$

In (10), $I_q(k, \ell)$ denotes the interference spread from the $q$th user’s data symbols due to fractional Doppler

$$I_q(k, \ell) = \sqrt{P_u} \sum_{k' = 0}^{k_{pq}} \sum_{\ell' = 0}^{\ell_{pq}} b[k', \ell'] \sum_{c \in K} h_{pq}[k-k_q] N, (\ell - \ell_q)_M \times \alpha(k, \ell, c) x_{dq}[k-k' + c] N, (\ell - \ell'_q)_M,$$

(11)

and $I_q(k, \ell)$ denotes the interference spread from the data symbols of other users, which can be expressed as

$$\mathbb{E}\{\|\bar{x}_{d,q}\|^2\} \leq \rho_d.$$
where \( w_{d,q} \in \mathbb{C}^{MN \times 1} \) is the AWGN vector at the user \( q \).

### III. Performance Analysis

In this section, we derive a new closed-form expression for the downlink achievable rate, using the bounding technique from [10]. We further assume that the proposed low-complexity DD domain detector in [15] is utilized for detection of information symbols at each user.

We start by presenting some characteristics of the DD domain channel representation in (8), which will facilitate the ensuing achievable rate analysis.

**Lemma 1:** For the matrix \( \mathbf{T}_{pq} \), we have \( \mathbf{T}_{pq} \mathbf{T}_{pq}^\dagger = I_{MN} \).

**Proof:** The proof is relegated to the journal version.

**Theorem 1:** For any two different paths with different delay indices, we have \( \mathbb{E} \left\{ \mathbf{T}_{pq} \mathbf{T}_{pq}^\dagger \right\} = 0 \) for \( (\ell_{pq,i} - \ell_{pq,j}) \neq 0 \).

**Proof:** The proof is relegated to the journal version.

**Lemma 2:** For any two matrices of \( \mathbf{T}_{pq} \) and \( \mathbf{T}_{pq}^\dagger \) in (8), we have

\[
\sum_{r=1}^{MN} \mathbb{E} \left\{ \mathbf{T}_{pq} \mathbf{T}_{pq}^\dagger \right\} = 1, \quad \forall q, q', i, j.
\]

**Proof:** The proof is relegated to the journal version.

**Lemma 4:** The power constraint in (19), is expressed as

\[
K_q \sum_{i=1}^{L_q} \eta_{pq,i}^2 \leq 1,
\]

where \( \gamma_{pq,i} \triangleq \mathbb{E} \left\{ \hat{h}_{pq,i}^2 \right\} = \sqrt{\mathbb{E} \left\{ \hat{h}_{pq,i}^2 \right\}} c_{pq,i} \).

**Proof:** By invoking (18), we have

\[
\mathbb{E} \left\{ \| \mathbf{x}_{d,q} \|^2 \right\} = \mathbb{E} \left\{ \| \sqrt{\rho_d \sum_{q=1}^{K_q} \eta_{pq,i}^2 \mathbf{H}_i^\dagger \mathbf{h}_{pq} s_i \} \|^2 \right\}
\]

\[
= \rho_d \mathbb{E} \left\{ \left( \sum_{q=1}^{K_q} \eta_{pq,i}^2 \mathbf{H}_i^\dagger \mathbf{h}_{pq} s_i \right) \left( \sum_{q=1}^{K_q} \eta_{pq,i}^2 \mathbf{H}^\dagger \mathbf{h}_{pq} \right) \right\}
\]

\[
= \rho_d \mathbb{E} \left\{ \left( \sum_{q=1}^{K_q} \eta_{pq,i}^2 \mathbf{H}_i^\dagger \mathbf{h}_{pq} s_i \right) \mathbf{E} \left\{ \mathbf{H}_i^\dagger \mathbf{H}_i \right\} \right\},
\]

where the equality holds since \( s_{q,r} \) are assumed to be i.i.d. RVs. By using (8), and noticing that channel gains of different path are zero-mean i.i.d RVs, i.e., \( \mathbb{E} \left\{ \mathbf{h}_{pq,i} \right\} = 0 \), \( \forall i \neq j \), and then applying Lemma 1 we get \( \mathbb{E} \left\{ \| \mathbf{x}_{d,q} \|^2 \right\} = \rho_d \sum_{q=1}^{K_q} \sum_{i=1}^{L_q} \eta_{pq,i}^2 \mathbb{E} \left\{ \mathbf{h}_{pq,i} \right\} \) which completes the proof.

We assume that each user has knowledge of the channel statistics in DD domain but not of the channel realizations [10]. The signal received at the qth user (20) can be re-arranged to be suitable for detection of the rth entry of the received signal in DD domain with only statistical channel knowledge at users, given by (23) at the top of the next page.

The sum of second, third, forth and last term in (23) are

\[
\sum_{r=1}^{MN} \mathbb{E} \left\{ \mathbf{T}_{pq} \mathbf{T}_{pq}^\dagger \right\} = 1, \quad \forall q, q', i, j.
\]

**Proof:** Since the channel model in (8) consists of simply \( MN \) parallel channels, an achievable rate per channel input by the qth user is given by

\[
R_{d,q} = \frac{\sum_{r=1}^{MN} I_{q,r} (\text{sinr}_{d,r})}{MN},
\]

with \( I_{q,r} (\text{sinr}_{d,r}) = \log_2 (1 + \text{sinr}_{d,r}) \), and \( \text{sinr}_{d,r} \), i.e., the signal-to-interference-plus-noise ratio (SINR) at the qth user is

\[
\text{SINR}_{d,r} = \frac{\mathbb{E} \left\{ |B_{pq} |^2 \right\}}{\mathbb{E} \left\{ |l_{q} |^2 \right\} + \mathbb{E} \left\{ |l_{r} |^2 \right\} + 1/p_d}.
\]

We now proceed to derive the achievable rate \( R_{d,q} \). Noticing that \( \mathbf{h}_{pq} = \sum_{i=1}^{L_q} \mathbf{h}_{pq,i} \mathbf{T}_{pq}^i \) and \( \mathbf{h}_{pq} = \sum_{i=1}^{L_q} \mathbf{h}_{pq,i} \mathbf{T}_{pq}^i \), the term in the numerator of (26) can be derived as

\[
\mathbb{E} \left\{ |B_{pq} |^2 \right\} = \sum_{i=1}^{L_q} \mathbb{E} \left\{ |\mathbf{h}_{pq,i} |^2 \right\} \mathbb{E} \left\{ |\mathbf{h}_{pq,i} |^2 \right\}.
\]

Before proceeding to derive \( \mathbb{V}_1 \), we define \( X \triangleq \sum_{i=1}^{L_q} \mathbf{h}_{pq,i} \mathbf{h}_{pq,i}^\dagger \) and \( Y \triangleq \sum_{i=1}^{L_q} \mathbf{h}_{pq,i} \mathbf{h}_{pq,j}^\dagger \). We note that \( \mathbb{E} \left\{ |X + Y |^2 \right\} = \mathbb{E} \left\{ |X |^2 \right\} + \mathbb{E} \left\{ |Y |^2 \right\} \) if \( X \) and \( Y \) are independent RVs and \( \mathbb{E} \{ Y \} = 0 \). It can be readily checked that \( Y \) is a zero mean RV which is independent from \( X \). Hence, \( \mathbb{V}_1 \) can be evaluated as

\[
\mathbb{V}_1 = \sum_{i=1}^{L_q} \mathbb{E} \left\{ |\mathbf{h}_{pq,i} |^2 \right\} \mathbb{E} \left\{ |\mathbf{h}_{pq,i} |^2 \right\}.
\]

where

\[
\Psi = \left\{ \left( \mathbf{h}_{pq} \mathbf{T}_{pq} \right)^i \right\} = \mathbb{E} \left\{ (\mathbf{h}_{pq,i} |^2) \mathbb{E} \left\{ (\mathbf{h}_{pq,i} |^2 \right\}.
\]

1We note that the channel overhead is ignored as the uplink achievable rate is not considered in this paper. We emphasize that this assumption does not affect the insights obtained in our work.
interference term can be derived as

\[ z_{d,q,r} = \sqrt{\mathbb{E}} \left\{ \sum_{p=1}^{M_q} |\eta_{pq}^{1/2} \mathbf{H}_{pq}(\cdot,\cdot) \mathbf{H}_p^\dagger(\cdot,\cdot) s_{qr}\right\} + \sqrt{\mathbb{E}} \left\{ \sum_{p=1}^{M_q} |\eta_{pq}^{1/2} \mathbf{H}_{pq}(\cdot,\cdot) \mathbf{H}_p^\dagger(\cdot,\cdot) r_{qr}\right\} \]

desired signal \( \oplus \) noise

\[ + \sqrt{\mathbb{E}} \left\{ \sum_{p=1}^{M_N} \eta_{pq}^{1/2} \mathbf{H}_{pq}(\cdot,\cdot) \mathbf{H}_p^\dagger(\cdot,\cdot) s_{qr'}\right\} + \sqrt{\mathbb{E}} \left\{ \sum_{p=1}^{M_N} \eta_{pq}^{1/2} \mathbf{H}_{pq}(\cdot,\cdot) \mathbf{H}_p^\dagger(\cdot,\cdot) r_{qr'}\right\} \]

inter-symbol interference \( \oplus \) noise

\[ + \sqrt{\mathbb{E}} \left\{ \sum_{p=1}^{M_N} K_u \sum_{q'=q'}^{M_N} \eta_{pq}^{1/2} \mathbf{H}_{pq}(\cdot,\cdot) \mathbf{H}_p^\dagger(\cdot,\cdot) s_{qr'}\right\} + \sqrt{\mathbb{E}} \left\{ \sum_{p=1}^{M_N} K_u \sum_{q'=q'}^{M_N} \eta_{pq}^{1/2} \mathbf{H}_{pq}(\cdot,\cdot) \mathbf{H}_p^\dagger(\cdot,\cdot) r_{qr'}\right\} + w_{d,q,r} \text{ AWGN} \]

inter-user interference \( \oplus \) noise

\[ R_{d,q} = \frac{1}{MN} \sum_{r=1}^{M_N} \log_2 \left( 1 + \frac{\rho_d \left( \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \eta_{pq}^{1/2} \gamma_{pq,i} \right)^2}{\rho_d \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \beta_{pq,i} + \sum_{q''=q''}^{K_u} \sum_{j=1}^{L_{pq}'} \eta_{pq,j} \gamma_{pq,j} + 1} \right) \]

(23)

\[ R_{d,q} = \log_2 \left( 1 + \frac{\rho_d \left( \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \eta_{pq}^{1/2} \gamma_{pq,i} \right)^2}{\rho_d \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \beta_{pq,i} + \sum_{q''=q''}^{K_u} \sum_{j=1}^{L_{pq}'} \eta_{pq,j} \gamma_{pq,j} + 1} \right) \]  

(36)

To this end, by using the facts that \( \mathbb{E} \{ |\varepsilon_{pq,i}|^2 \} = \beta_{pq,i} - \gamma_{pq,i} \)

and \( \mathbb{E} \{ \Re(\Psi) \} = 0 \), (30) is reduced to

\[ V_1 = \sum_{i=1}^{L_{pq}} \left( \gamma_{pq,i}(\beta_{pq,i} - \gamma_{pq,i}) + 2 \chi_{pq,i} \right) \]

\[ + \sum_{i=1}^{L_{pq}} \sum_{j \neq i} \left( \gamma_{pq,i} \gamma_{pq,j} + \chi_{pq,i} \beta_{pq,j} \gamma_{pq,j} \right) \]  

(31)

Therefore, by substituting (31) into (28) we obtain

\[ \mathbb{E} \{ |\mathbf{H}_u|^2 \} = \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \eta_{pq}^{1/2} \beta_{pq,i} + \sum_{j \neq i} \chi_{pq,i} \gamma_{pq,j} \]  

(32)

The inter-symbol interference term can be obtained as

\[ \mathbb{E} \{ |\mathbf{H}_q,1|^2 \} = \mathbb{E} \left\{ \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \eta_{pq}^{1/2} \mathbf{H}_{pq}(\cdot,\cdot) \mathbf{H}_p^\dagger(\cdot,\cdot) \right\} \]

\[ = \mathbb{E} \left\{ \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \eta_{pq}^{1/2} \left( \sum_{i=1}^{L_{pq}} \eta_{pq}^{1/2} \mathbf{H}_{pq,j} \mathbf{H}_j^\dagger(\cdot,\cdot) \right) \right\} \]

\[ \left( 35 \right) \]

To this end, by using Lemma 1, i.e., \( \left( T_{ij} \right)_{pq} \) \( \mathbf{H}_{pq}(\cdot,\cdot) \) \( \mathbf{H}_p^\dagger(\cdot,\cdot) \) \( \mathbf{H}_q,1 \) \( 0 \), and then by substituting \( h_{pq,i} = \varepsilon_{pq,i} + \hat{h}_{pq,i} \), we get

\[ \mathbb{E} \{ |\mathbf{H}_q,1|^2 \} = \sum_{p=1}^{M_q} \sum_{i=1}^{L_{pq}} \sum_{j \neq i} \mathbb{E} \{ |\varepsilon_{pq,i}|^2 \} \mathbb{E} \{ |\hat{h}_{pq,j}|^2 \} \]

\[ + \mathbb{E} \{ |\hat{h}_{pq,i}|^2 \} |\hat{h}_{pq,j}|^2 \left| j \neq i \right) \]  

(34)

where the first equality holds as the variance of a sum of independent RVs is equal to the sum of the variances and the second equality holds since \( \varepsilon_{pq,i} \) has zero mean and is independent of \( \hat{h}_{pq,i} \).

Noticing that the channel gains of different users are independent zero mean RVs, and applying Lemma 3, the inter-user interference term can be derived as

\[ \mathbb{E} \{ |\mathbf{H}_q,2|^2 \} = \sum_{p=1}^{M_q} \sum_{q'=q'}^{K_u} \sum_{i=1}^{L_{pq}} \eta_{pq}^{1/2} \beta_{pq,i} \gamma_{pq,q'} \]  

(35)

To this end, by substituting (27), (32), (34), and (35) into (26), after some algebraic manipulations the desired result in (24) is obtained.

**Corollary 1:** In the special case that different paths corresponding to the channel between the \( q \)th user and \( p \)th AP experience different delays (i.e., \( \varepsilon_{pq,i} \neq \varepsilon_{pq,j} \)), the achievable downlink rate in (25) reduces to (36) at the top of the page.

**Proof 6:** By invoking Lemma 2 and Lemma 3, it can be checked that when \( \varepsilon_{pq,i} \neq \varepsilon_{pq,j} \), we have \( \chi_{pq,i} = 0 \) and \( \kappa_{pq,j} = 1 \). Thus, the received SINR becomes independent of \( r \), which completes the proof.

**Remark 1:** By inspecting (36), we see that with perfect knowledge of delay and Doppler indices and using conjugate beamforming, the impact of delay shift and Doppler spread can be efficiently mitigated in the special case of \( \varepsilon_{pq,i} \neq \varepsilon_{pq,j} \).

**IV. Numerical Results and Discussions**

We consider an OTFS system with \( N = 20 \) and \( M = 30 \). The carrier frequency is \( f_c = 4 \text{ GHz} \) and the sub-carrier spacing is \( \Delta f = 15 \text{ kHz} \). The maximum moving speed in the scenario is 500 km/h, thus, the maximum Doppler index is \( k_{max} = 3 \). We set maximum delay index as \( \ell_{max} = 2 \) and the number of channel paths as \( L_{pq} = 5 \). The associated delay and Doppler indices for each channel path are randomly chosen from \([0, \ell_{max}] \) and \([0, k_{max}] \), respectively.

We assume that \( M_q \) APs and \( K_u \) users are uniformly distributed at random within a square of size \( D \times D \) km² whose edges are wrapped around to avoid the boundary effects. The large-scale fading coefficient \( \beta_{pq} \) models the path loss and shadow fading, according to \( \beta_{pq,i} = P_{L_{pq},i} 10^{\frac{s_{pq,i} - \text{SNR}}{10}} \), where \( P_{L_{pq},i} \) represents the path loss, and \( 10^{\frac{s_{pq,i} - \text{SNR}}{10}} \) represents the shadow fading with the standard deviation \( \sigma_{sh} \), and \( s_{pq,i} \sim \mathcal{CN}(0, 1) \). We use the three-slope model for the path-loss \( P_{L_{pq},i} \) (in dB) as

\[ P_{L_{pq},i} = \begin{cases} \left(-L - 35 \log_{10}(d_{pq})\right) & d_{pq} > d_1, \\ \left(-L - 15 \log_{10}(d_{pq})\right) & d_1 < d_{pq} \leq d_1 \\ \left(-L - 15 \log_{10}(d_{pq})\right) & d_1 < d_{pq} \leq d_1 \end{cases} \]

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where $L$ is a constant depending on the carrier frequency, the user and AP heights, given in [10]. We further use the correlated shadowing model for $d_{pq} > d_{q}$ as described in [10]. Here, we choose $\sigma_{sh} = 8$ dB, $D = 1$ km, $d_{1} = 50$ m, and $d_{0} = 10$ m. We further set the noise figure $F = 9$ dB, and thus the noise power $\sigma_{n}^{2} = -108$ dBm ($\sigma_{n}^{2} = k_{B}T_{0}(M_{\Delta f})F$ W, where $k_{B} = 1.381 \times 10^{-23}$ Joules/K is the Boltzmann constant, while $T_{0} = 290^{\circ}$K is the noise temperature). Let $\rho_{d} = 1$ W, $\rho_{u} = 0.2$ W and $\rho_{p} = 1$ W be the maximum transmit power of the APs, users and uplink training pilots, respectively. The normalized maximum transmit power $\rho_{d}$, $\rho_{u}$, and $\rho_{p}$ are calculated by dividing these powers by the noise power $\sigma_{n}^{2}$. We assume that all APs transmit with full power, and at the $p$th AP, the power control coefficients $\eta_{pq}$, $q = 1, \ldots, K_{u}$, are the same, i.e., $\eta_{pq} = \left(\sum_{q'=1}^{K_{u}} \sum_{l=1}^{L_{pq}} \gamma_{pq'} \right)^{-1}$, $\forall q = 1, \ldots, K_{u}$.

In Fig. 1, we compare the performance of OTFS with OFDM modulation system for $M_{a} = 40$ and $K_{u} = 20$ with and without shadow fading correlation. It is clear that OTFS significantly outperforms OFDM in both median and in 95%-likely performance. The performance improvement of the OTFS over the OFDM is more pronounced in the uncorrelated shadowing case. More specifically, the 95%-likely throughput gain of the OTFS system over the OFDM counterpart is nearly 4-fold greater in uncorrelated shadowing as compared to correlated shadowing.

Fig. 2 shows the average downlink throughput of the system versus number of APs and for two different number of users, with and without shadow fading correlation. We evaluate the SE of the system over 200 random realizations of the locations of APs, users and fading channels. The analytical results are based on Theorem 1. It is observed that by increasing $K_{u}$, the per user rate decreases. This is because the same amount of resources are shared between more users. Moreover, the interference from other users increases, which degrades both the MMSE estimate of the channel gains and the received per-user SINR. This motivates AP selection and user scheduling algorithm as well as pilot assignment algorithm design to manage the radio resources more efficiently.

V. CONCLUSION

We analyzed the downlink performance of cell-free massive MIMO systems with OTFS modulation, taking into account the effects of channel estimation. A new closed-form expression for the average downlink throughput was presented. Our results confirmed the superiority of OTFS against OFDM in improving the throughput of the cell-free massive MIMO systems in high-mobility scenarios. Moreover, the improvement in 95%-likely per-user throughput in uncorrelated shadowing is almost 4 times higher than in the correlated fading scenarios.

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