Optimal OFDM Pilot Sequences for Time-Delay and Channel Estimation Based on the Expected CRB for a Large Number of Subcarriers

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Abstract—A key aspect to design an OFDM system for combined positioning and high-data-rate communications is to find optimal data and pilot power allocations. Previous work has investigated the capacity maximizing design taking into account the effects of channel and time-delay estimation for finite number of subcarriers and channel taps. In this paper, we propose a method based on the asymptotic expected Cramér-Rao bound of joint time-delay and channel coefficients that reduces the complexity of the bounds by increasing the number of subcarriers or channel taps for data and pilot power allocations design. Specifically, for long channels a general form of matrix inversion, which is computationally complex, is converted to only the inversion at strong eigenvalues or pilots. Numerical results show that as the number of subcarriers increases, the non-asymptotic bounds converge to the asymptotic bounds at a fast speed. Moreover, even for a finite number of subcarriers or channel taps the difference between joint data and pilot power allocations is negligible compared to the non-asymptotic expected Cramér-Rao bounds.

Index Terms—Combined positioning and communications, OFDM, channel and time-delay estimation, asymptotic expected Cramér-Rao bound, joint design of data and pilot power allocations.

I. INTRODUCTION

The design of combined positioning and communications systems that can perform well in terms of high-data-rate transmission and estimation accuracy is a challenging problem. In general, the signals used for one application perform poorly in the other case. To design a signal which can be applied for both purposes, one needs to consider the system specifications for time-delay estimation accuracy and data-rate communications.

To date, different approaches have been adopted to design pilot symbols that improve the performance of channel estimators [1]–[3]. The results show that equi-spaced, equi-powered pilots

are optimal in terms of mean squared error. Pilot designs for carrier frequency offset (CFO) estimation [4], or joint channel and CFO estimation [5]-[7] are considered by others. However, pilot design based on time-delay estimation has received little attention. A pilot design based on joint CRB of channel and time delay is proposed in [8], [9]. However, since CRBs in [8], [9] are functions of specific channel realizations, the resulting pilots cannot be guaranteed to be optimal for all instances of random channels. The problem has been solved by designing based on averaging the CRB over a set of channel realizations known as Expected CRB (ECRB) [10]. Furthermore, joint data and pilot power allocations for the case of limited number of subcarriers N and channel taps Lis investigated [10]. A method based on the effect of increasing the number of subcarriers N and channel taps L on joint channel coefficients and clock offset estimation is proposed in [11]. However, the bounds are limited to a specific type of channel.

In this paper, we consider the effect of increasing the number of subcarriers N and channel taps L on the joint expected Cramér-Rao bound of time-delay and channel coefficients. Using asymptotic bounds reduce the computational complexity. Specifically, applying matrix inversion algorithms such as Gaussian elimination requires the computational complexity of the order $O(L^3)$ for non-asymptotic bounds that makes non-asymptotic bounds in [10] complex and close to singular. However, we reduce the complexity to O(L) by doing the inversion only at strong pilots. Asymptotic expected Cramér-Rao bound for joint time-delay and channel coefficients is investigated for two cases. First, we assume the number of subcarriers N is sufficiently large but the number of channel taps L is finite, second we consider the case where L and Nboth grows large, with their relative ratio L/N approaching zero. Furthermore, we compare joint design of data and pilot allocations based on asymptotic bounds with non-asymptotic bounds. Our results show that the difference for joint design of data and pilot allocations based on asymptotic bounds is negligible.

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II. SYSTEM MODEL AND PRELIMINARIES

In this section, first we present a model of the OFDM system, then we propose the ECRB of the joint timing offset and channel coefficients estimation as a performance metric.

A. OFDM Signal Model

Using the same notation as [11], the continuous-time received signal from a standard OFDM symbol passed through a frequency selective channel, after removing the guard interval, is

$$y_N^{(a)}(t) = \sum_{k \in \mathbb{Z}} d_{N,k} g^{(a)}(t - kT) + v^{(a)}(t), \qquad (1)$$

where T is the sampling period at the transmitter such that $T_0 = NT$ is the observation window, N is the total number of subcarriers, $d_{N,k}$ represents the output from the inverse FFT (IFFT) block at the transmitter, and $v^{(a)}(t)$ is additive zeromean complex Gaussian noise. Unlike [11], but without loss of generality, we assume that the impulse response $g^{(a)}(t)$ is a delta function with the time limit of [0, LT) where L is the number of channel taps

$$g^{(a)}(t) = \sum_{l=0}^{L-1} h_l \delta(t - lT - \tau), \qquad (2)$$

where h_l is the channel coefficient of *l*th path and τ is the timing offset or equivalently the time delay of first path. Assuming the transmitter's and receiver's clocks are synchronized, the discrete-time received signal $y_N[n] = y_N^{(a)}(nT)$ is

$$y_N[n] = \frac{1}{\sqrt{N}} \sum_{n'=0}^{N-1} \sum_{l=0}^{L-1} D_{N,n'} h_l e^{j\frac{2\pi}{N}n'(n-l-\tau_d)} + v[n], \quad (3)$$

where $D_{N,n'}$ represents pilot subcarrier at the n'th frequency, and $\tau = \tau_d T$. In vector form we have

$$\boldsymbol{y}_N = \boldsymbol{R}_N(\tau_d)\boldsymbol{h} + \boldsymbol{v}_N, \qquad (4)$$

where $\boldsymbol{y}_N = [y_N[0], \dots, y_N[N-1]]^T$, $\boldsymbol{v}_N = [v_N[0], \dots, v_N[N-1]]^T$, $\boldsymbol{h} = [h_0, \dots, h_{L-1}]^T$, and the (n,l) element of $\boldsymbol{R}_N(\tau_d)$ is $[\boldsymbol{R}_N(\tau_d)]_{n,l} = \frac{1}{\sqrt{N}} \sum_{n'=0}^{N-1} D_{N,n'} e^{j\frac{2\pi}{N}n'(n-l-\tau_d)}$. Finally, taking the FFT of the output, we find

$$\boldsymbol{Y}_{N} = \boldsymbol{F}_{N,N} \boldsymbol{y}_{N} = \boldsymbol{F}_{N,N} \boldsymbol{R}_{N}(\tau_{d}) \boldsymbol{h} + \boldsymbol{V}_{N}, \qquad (5)$$

which describes the output of the OFDM system at the training phase. Also, one can obtain a similar model as in [8] by taking the DFT of (3) and writing the result in a vector notation as

$$\boldsymbol{Y}_{N} = \boldsymbol{D}_{N}\boldsymbol{\Gamma}(\tau_{d})\boldsymbol{F}_{N,L}\boldsymbol{h} + \boldsymbol{V}_{N}, \qquad (6)$$

where D_N represents an $N \times N$ diagonal matrix of the input with the *k*th diagonal element $D_{N,k}$ representing the input at the *k*th subcarrier, Γ is an $N \times N$ diagonal matrix with the *k*th diagonal element $\exp(-j2\frac{\pi}{T}k\tau_d)$, and $F_{N,L}$ contains the first *L* columns of a Discrete Fourier Transform (DFT) matrix. Note that the difference between our problem and the problems presented in [5], [11] is that here the timing offset is only multiplied by n', while in [11], sampling clock frequency offset is multiplied by both n' and n. Also, the problem differs from the case when frequency offset occurs [5] since in that case frequency offset is multiplied by n not n'.

B. Expected Cramér-Rao Bound

In this section, we present a closed-form expression for the expected CRB for the channel coefficients h and the timing offset τ_d . Note that the related results can be found in [5], [8], [10], [11]. Defining the parameter vector by $\boldsymbol{\theta} = [\boldsymbol{h}_R^T, \boldsymbol{h}_I^T, \tau_d]^T$, the corresponding Fisher information matrix (FIM) can be written as

$$\boldsymbol{J}_{F} = \frac{2}{\sigma^{2}} \Re \left[\frac{\partial \boldsymbol{\mu}^{H}}{\partial \boldsymbol{\theta}} \frac{\partial \boldsymbol{\mu}}{\partial \boldsymbol{\theta}^{T}} \right], \tag{7}$$

where $\boldsymbol{\mu} = \boldsymbol{F}_{N,N} \boldsymbol{R}_N(\tau_d) \boldsymbol{h}$. One can easily find the FIM

$$\boldsymbol{J}_{F} = \frac{2}{\sigma^{2}} \begin{bmatrix} N \Re[\boldsymbol{U}_{N}] & -N \Im[\boldsymbol{U}_{N}] & N^{2} \Re[\boldsymbol{V}_{N}\boldsymbol{h}] \\ N \Im[\boldsymbol{U}_{N}] & N \Re[\boldsymbol{U}_{N}] & N^{2} \Im[\boldsymbol{V}_{N}\boldsymbol{h}] \\ N^{2} \Re[\boldsymbol{h}^{H} \boldsymbol{V}_{N}^{H}] & -N^{2} \Im[\boldsymbol{h}^{H} \boldsymbol{V}_{N}^{H}] & N^{3} \boldsymbol{h}^{H} \boldsymbol{W}_{N} \boldsymbol{h} \end{bmatrix},$$
(8)
where $\boldsymbol{U}_{N} = \frac{1}{N} \boldsymbol{R}_{N}^{H}(\tau_{d}) \boldsymbol{R}_{N}(\tau_{d}), \boldsymbol{V}_{N} = \frac{1}{N^{2}} \boldsymbol{R}_{N}^{H}(\tau_{d}) \boldsymbol{Q}_{N}(\tau_{d}),$

where $\mathbf{U}_N = \frac{1}{N} \mathbf{A}_N(\tau_d) \mathbf{R}_N(\tau_d)$, $\mathbf{V}_N = \frac{1}{N^2} \mathbf{R}_N(\tau_d) \mathbf{Q}_N(\tau_d)$, $\mathbf{W}_N = \frac{1}{N^3} \mathbf{Q}_N^H(\tau_d) \mathbf{Q}_N(\tau_d)$, and $\mathbf{Q}_N(\tau_d) = d\mathbf{R}_N(\tau_d)/d\tau_d$. Using the well known block inversion matrix lemma [12] and defining a new estimation parameter $\tilde{\boldsymbol{\theta}} = [\boldsymbol{h}^T, \tau_d]^T$, we find

$$\mathbb{E}\left[||\hat{\boldsymbol{h}}_{N} - \boldsymbol{h}||^{2}\right] \geq \frac{\sigma^{2}}{2N} \left(2 \operatorname{tr}(\boldsymbol{U}_{N}^{-1}) + \gamma_{N}^{-1} ||\boldsymbol{\beta}_{N}||^{2}\right), \quad (9)$$

$$\mathbb{E}\left[(\hat{\tau}_d^N - \tau_d)^2\right] \ge \frac{\sigma^2}{2N^3\gamma_N},\tag{10}$$

where

$$\boldsymbol{\beta}_N = \boldsymbol{U}_N^{-1} \boldsymbol{V}_N \boldsymbol{h}, \qquad (11)$$

$$\gamma_N = \boldsymbol{h}^H (\boldsymbol{W}_N - \boldsymbol{V}_N^H \boldsymbol{U}_N^{-1} \boldsymbol{V}_N) \boldsymbol{h}.$$
(12)

Finally, taking the expectation with respect to channel coefficients and using Jensen's inequality, the approximate expression for the ECRB of the timing offset and the channel coefficients would be

$$\mathrm{ECRB}_{h} \approx \frac{\sigma^{2}}{2N} \left(2\mathrm{tr}(\boldsymbol{U}_{N}^{-1}) + \overline{\gamma}_{N}^{-1} || \overline{\boldsymbol{\beta}}_{N} ||^{2} \right), \qquad (13)$$

$$\mathrm{ECRB}_{\tau_d} \approx \frac{\sigma^2}{2N^3 \overline{\gamma}_N},\tag{14}$$

where

$$||\overline{\boldsymbol{\beta}}_{N}||^{2} = \operatorname{tr}\left((\boldsymbol{V}_{N}^{H}\boldsymbol{U}_{N}^{-H}\boldsymbol{U}_{N}^{-1}\boldsymbol{V}_{N})\boldsymbol{R}_{h}\right), \qquad (15)$$

$$\overline{\gamma}_N = \operatorname{tr}\left((\boldsymbol{W}_N - \boldsymbol{V}_N^H \boldsymbol{U}_N^{-1} \boldsymbol{V}_N) \boldsymbol{R}_h \right), \qquad (16)$$

where \mathbf{R}_h is the channel covariance matrix. Note that the actual expected CRB is tighter than the above expressions due to Jensen's inequality. In the next section, we obtain the asymptotic ECRB for channel coefficients and timing offset where the number of subcarriers N is assumed to be sufficiently large.

III. ASYMPTOTIC EXPECTED CRAMÉR-RAO BOUND

In this section, we obtain an asymptotic expression for U_N , and using a similar procedure as in [11], we conclude the expressions for V_N and W_N . Unlike the expressions proposed in [11], we express the asymptotic values of U_N , V_N , and W_N in discrete frequency as a function of pilot powers since the final goal would be to design a system with optimal power allocation for joint communication and navigation.

The (p,q) element of U_N is

$$[\boldsymbol{U}_{N}]_{p,q} = \frac{1}{N} \sum_{l_{1}, l_{2}=0}^{N-1} \boldsymbol{D}_{N, l_{1}}^{*} \boldsymbol{D}_{N, l_{2}} e^{j \frac{2\pi}{N} (l_{1}-l_{2})\tau_{d}} [\boldsymbol{M}^{H}(l_{1}) \boldsymbol{M}(l_{2})]_{p,q},$$
(17)

where

$$[\boldsymbol{M}^{H}(l_{1})\boldsymbol{M}(l_{2})]_{p,q} = e^{-j\frac{2\pi}{N}(ql_{2}-pl_{1})}\psi_{N}^{(0)}((l_{1}-l_{2})), \quad (18)$$

being

$$\psi_N^{(0)}((l_1 - l_2)) = \frac{1}{N} \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}n(l_1 - l_2)}.$$
 (19)

We propose the following lemma [11]:

Lemma 3.1: Let $\alpha > 0$ and let $\phi_N : \mathbb{N} \to \mathbb{C}$ be a function such for every integer k with $1 < |k| < \lfloor \alpha N \rfloor$, $|\phi_N(k)| < C/k$, where C is a constant that does not depend on N. Then, for every real number r

$$\psi_N^{(0)}((l_1 - l_2)) = \frac{1}{N} \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}n(l_1 - l_2)},$$
 (20)

converges almost surely to 0 as $N \to \infty$.

Replacing (18) and (19) in (17), and using the above lemma with $r = q - p + \tau_d$, $\alpha = 1/2$ and $\phi_N(k) = e^{-j\frac{2\pi}{N}pk}\psi_N^{(0)}(k)$, one can conclude that the terms for $l_1 \neq l_2$ in (17) almost surely converge to zero. Therefore, we obtain

$$[\boldsymbol{U}]_{p,q} = [\boldsymbol{U}_{N \to \infty}]_{p,q} = \frac{1}{N} \sum_{l=0}^{N-1} |D_{N,l}|^2 e^{-j\frac{2\pi}{N}(q-p)l}.$$
 (21)

Similarly, the asymptotic expressions for V_N and W_N can be found as

$$[\mathbf{V}]_{p,q} = [\mathbf{V}_{N \to \infty}]_{p,q} = \frac{j\pi}{N} \sum_{l=0}^{N-1} \frac{l}{N} |D_{N,l}|^2 e^{-j\frac{2\pi}{N}(q-p)l},$$
(22)

and

$$[\mathbf{W}]_{p,q} = [\mathbf{W}_{N \to \infty}]_{p,q} = \frac{4\pi^2}{3N} \sum_{l=0}^{N-1} (\frac{l}{N})^2 |D_{N,l}|^2 e^{-j\frac{2\pi}{N}(q-p)l}.$$
(23)

In the following, we find a more compact form for the asymptotic expressions of channel coefficients and time delay and also we consider the case when the number of channel taps Lis sufficiently large such that $L/N \to 0$. Replacing $|D_{N,l}|^2$ by P_l as the pilot power at the *l*th subcarrier, and using (21), (22), and (23) we obtain $U = U_{N\to\infty} = \frac{1}{N} \sum_{l=0}^{N-1} P_l e(l) e^H(l)$, $V = V_{N\to\infty} = \frac{j\pi}{N} \sum_{l=0}^{N-1} \frac{l}{N} P_l e(l) e^H(l)$, and W = $W_{N\to\infty} = \frac{4\pi^2}{3N} \sum_{l=0}^{N-1} (\frac{l}{N})^2 P_l e(l) e^H(l)$ where e(l) = $[e^{j\frac{2\pi}{N}l(0)}, \ldots, e^{j\frac{2\pi}{N}l(L-1)}]^T$. This can be interpreted as the sum of N, $L \times L$ matrices with the eigenvalues P_l and eigenvectors e(l) for $l = 0, \ldots, N - 1$. In matrix form, we have $U = F_{L,N} P F_{L,N}^H$, $V = j\pi F_{L,N} D P F_{L,N}^H$, and $W = \frac{4\pi^2}{3} F_{L,N} D^2 P F_{L,N}^H$ where $F_{L,N}$ is the first L rows of the discrete Fourier transform matrix, and Pand D are pilot power and derivative matrices defined as $P = \text{diag}\{P_0, \ldots, P_{N-1}\}$ and $D = \text{diag}\{\frac{0}{N}, \ldots, \frac{N-1}{N}\}$ respectively. Consequently, asymptotic ECRB of channel coefficients and time delay can be found by replacing U, V, and W in (13) and (14),

$$NECRB_{h}^{as} \approx \frac{\sigma^{2}}{2} \left(2tr(\boldsymbol{U}^{-1}) + \overline{\gamma}^{-1} || \overline{\boldsymbol{\beta}} ||^{2} \right), \qquad (24)$$

$$N^{3} \text{ECRB}^{as}_{\tau_{d}} \approx \frac{\sigma^{2}}{2\overline{\gamma}},$$
 (25)

where

$$||\overline{\beta}||^{2} = \operatorname{tr}\left((\boldsymbol{V}^{H}\boldsymbol{U}^{-H}\boldsymbol{U}^{-1}\boldsymbol{V})\boldsymbol{R}_{h}\right), \qquad (26)$$

and

$$\overline{\gamma} = \operatorname{tr}\left((\boldsymbol{W} - \boldsymbol{V}^{H} \boldsymbol{U}^{-1} \boldsymbol{V}) \boldsymbol{R}_{h} \right).$$
 (27)

Finally, we analyze a special case where the number of channel taps L is sufficiently large such that $L/N \rightarrow 0$. In this case the Fisher information matrix (FIM) is close to singular or badly conditioned. For this type of problem, it is proved in [13] that instead of using the inverse of the FIM, we should apply the pseudo-inverse. This means that the inverse terms in (24) and (25), i.e. U^{-1} and U^{-2} , should be replaced by their pseudo-inverse. Equivalently, the corresponding eigenvectors to weak eigenvalues are set to zero. Applying eigenvalue decomposition to (24) and (25), we obtain

$$N\overline{\mathrm{ECRB}}_{h}^{as} \approx \sigma^{2} \left[\mathrm{tr}(\boldsymbol{\Lambda}^{-1}) + \frac{3\mathrm{tr}\left((\boldsymbol{E}\tilde{\boldsymbol{D}}^{2}\boldsymbol{E}^{H})\boldsymbol{R}_{h}\right)}{2\mathrm{tr}\left((\boldsymbol{E}\tilde{\boldsymbol{D}}^{2}\boldsymbol{\Lambda}\boldsymbol{E}^{H})\boldsymbol{R}_{h}\right)} \right], (28)$$

and

$$N^{3}\overline{\mathrm{ECRB}}_{\tau_{d}}^{as} \approx \frac{3\sigma^{2}}{2\pi^{2}} \frac{1}{\mathrm{tr}\left((\boldsymbol{E}\tilde{\boldsymbol{D}}^{2}\boldsymbol{\Lambda}\boldsymbol{E}^{H})\boldsymbol{R}_{h}\right)}, \qquad (29)$$

where E represents the eigenvectors corresponding to the strong eigenvalues or pilots. Note that the terms eigenvalues and pilots have been used interchangeably since, by increasing the number of channel taps, the eigenvalues and pilots are asymptotically the same. \tilde{D} is the derivative matrix at the subcarriers with strong pilots, and Λ represents strong eigenvalues or pilots. Note that (28) and (29) have been computed by replacing U^{-1} , V, and W by $E\Lambda^{-1}E^{H}$, $E\tilde{D}\Lambda E^{H}$, and $E\tilde{D}^{2}\Lambda E^{H}$ respectively, and using the fact that E is unitary matrix (i.e., $E^{H}E = I$). This is pretty similar to the expressions proposed in [5] for the case of Asymptotic CRB of joint frequency offset and channel coefficients.

IV. CHANNEL CAPACITY

In this section we present a pilot design for joint communication and navigation based on asymptotic expected CRB for two cases. First, a more general form when the number of channel taps can be any limited number is investigated. Second, we assume sufficiently large number of channel taps L such that the ratio between channel taps and the number of subcarriers goes to zero $L/N \rightarrow 0$. We can rewrite signal model (6) as

$$\boldsymbol{Y}_N = \boldsymbol{H} \boldsymbol{\dot{D}}_N + \boldsymbol{V}_N, \qquad (30)$$

where $\dot{\mathbf{D}}_N = [D_{N,0}, \dots, D_{N,N-1}]^T$, $\mathbf{H} = \text{diag}\{H(0), \dots, H(N-1)\}, H(k) = \gamma_{\tau}(k)\overline{H}(k), \gamma_{\tau}(k) = \exp(-j\frac{2\pi}{T_0}k\tau), \overline{H}(k) = \mathbf{F}_{k,L}\mathbf{h}, \text{ and } \mathbf{F}_{k,L} \text{ is the }k\text{th row of } \mathbf{F}_{N,L}.$ Replacing \mathbf{H} by $\hat{\mathbf{H}} + \tilde{\mathbf{H}}$ in (30) we obtain

$$\boldsymbol{y} = \hat{\boldsymbol{H}} \boldsymbol{\dot{\boldsymbol{D}}}_N + \hat{\boldsymbol{H}} \boldsymbol{\dot{\boldsymbol{D}}}_N + \boldsymbol{V}_N, \qquad (31)$$

where \hat{H} and \hat{H} represent the estimated value and the error, with the *k*th diagonal elements $\hat{H}(k) = \gamma_{\hat{\tau}}(k) F_{k,L} \hat{h}$ and $\tilde{H}(k) = \gamma_{\hat{\tau}}(k) F_{k,L} \tilde{h}$ respectively. Note that channel coefficients h are estimated in the receiver through known pilot symbols inserted at the transmitter. The receiver feeds the estimated channel coefficients back to the transmitter (Adaptive Modulation). However, the estimated coefficients at the transmitter are not error-free due to estimation error. Consequently, it can be shown that the lower bound of ergodic capacity \bar{C}_{lb} for the so called partially known channel at the transmitter is of the form of [14]

$$\bar{C}_{lb}(\boldsymbol{p}_p, \boldsymbol{p}_d) = \frac{1}{N} \mathbb{E} \left[\log_2 \det(\boldsymbol{I} + \boldsymbol{P}_d \boldsymbol{R}_{y_e}^{-1} \hat{\boldsymbol{H}} \hat{\boldsymbol{H}}^H) \right], \quad (32)$$

where P_d is an $N \times N$ diagonal matrix of data power with $k \in D$ diagonal entry $p_{d,k}$ and zero elsewhere, D being the set of subcarriers used for data transmission, and

$$\boldsymbol{R}_{y_e} = \boldsymbol{P}_d \mathbb{E}\left[\tilde{\boldsymbol{H}} \tilde{\boldsymbol{H}}^H\right] + \sigma^2 \boldsymbol{I}.$$
 (33)

So, we obtain

$$\mathbb{E}\left[\tilde{\boldsymbol{H}}\tilde{\boldsymbol{H}}^{H}\right] = \operatorname{diag}\left\{\boldsymbol{F}_{k,L}\mathbb{E}\{\tilde{\boldsymbol{h}}\tilde{\boldsymbol{h}}^{H}\}\boldsymbol{F}_{k,L}^{H}\right\}_{k\in\mathcal{D}}.$$
 (34)

Using (33) and (34), and replacing $\mathbb{E}\{\tilde{h}\tilde{h}^H\}$ by the matrix form of (24) or (28), we find the ergodic capacity as

$$\bar{C}_{lb}(\boldsymbol{p}_p, \boldsymbol{p}_d) = \mathbb{E}\left[C_{lb}(\boldsymbol{p}_p, \boldsymbol{p}_d)\right],\tag{35}$$

where $C_{lb}(\boldsymbol{p}_p, \boldsymbol{p}_d)$ is the instantaneous capacity defined as

$$C_{lb}(\boldsymbol{p}_p, \boldsymbol{p}_d) = \frac{1}{N} \sum_{k \in \mathcal{D}} \log_2(1 + \frac{p_{d,k} \zeta_k}{p_{d,k} \widetilde{\zeta}_k(\boldsymbol{p}_p) + \sigma^2}), \quad (36)$$

where $\zeta_k = \mathbf{F}_{k,L} \hat{\mathbf{h}} \hat{\mathbf{h}}^H \mathbf{F}_{k,L}^H$, and $\tilde{\zeta}_k(\mathbf{p}_p) = \mathbf{F}_{k,L} \mathbf{J}_h(\mathbf{p}_p) \mathbf{F}_{k,L}^H$ with $\mathbf{J}_h(\mathbf{p}_p)$ being the matrix form of (24) or (28), and \mathbf{p}_p represents pilot vector or subcarriers used for estimation, i.e. $\tilde{\zeta}_k(\mathbf{p}_p)$ is a function of pilot vector \mathbf{p}_p .

V. POWER ALLOCATION OPTIMIZATION

In this section, we formulate the optimization problem used for pilot design for joint communications and positioning. To maximize the cost function which is the lower bound of ergodic capacity (35), one needs to solve the following optimization problem

$$\bar{C}_{CL} = \mathbb{E}\left[\max_{\boldsymbol{p}_p, \boldsymbol{p}_d \in \mathcal{K}} C_{lb}(\boldsymbol{p}_p, \boldsymbol{p}_d)\right],$$
(37)

where

$$\mathcal{K} = \{ \boldsymbol{p}_p, \boldsymbol{p}_d | J_{\tau_d}(\boldsymbol{p}_p) \leq \epsilon, \mathbf{1}^T \boldsymbol{p}_d + \mathbf{1}^T \boldsymbol{p}_p \leq P_t, \\ \boldsymbol{p}_p^T \boldsymbol{p}_d = 0, \boldsymbol{p}_p \succeq 0, \boldsymbol{p}_d \succeq 0 \},$$
(38)

where \bar{C}_{CL} is the channel capacity for the closed-loop (CL) system using partially known CSI at the transmitter side, \mathcal{K} is a set such that \boldsymbol{p}_p and \boldsymbol{p}_d satisfy some constraints, \boldsymbol{p}_p and p_d represent pilot and data power vectors respectively, P_t is the total power, $J_{\tau_d}(\boldsymbol{p}_p)$ represents the asymptotic bound for time-delay in (25) or (29) and ϵ is the minimum accuracy in the estimation, constraint $\mathbf{1}^T \boldsymbol{p}_d + \mathbf{1}^T \boldsymbol{p}_p \leq P_t$ limits the total power in the design of data and pilots to be smaller than P_t , $p_p^T p_d = 0$ means that subcarriers used for data transmission cannot be used as pilots which leads to a combinatorial optimization that is not convex. To solve this issue, we use the relaxation approach [10] by omitting the constraint $\boldsymbol{p}_p^T \boldsymbol{p}_d = 0$ and solving the relaxed problem. Solving the relaxed problem, one needs only a few subcarriers as pilots with higher amplitudes while the rest are set for data transmission. Finally, $p_p \succeq 0$ and $p_d \succeq 0$ emphasizes that pilots and data are non-negative values where \succeq is an elementwise operator.

VI. SIMULATION RESULTS

In this section we present the simulation results based on the asymptotic bounds. We use the simulation parameters as follows. The number of channel taps L = 4, number of subcarriers N = 40, noise power $\sigma^2 = 0.01$, minimum accuracy in time-delay estimation using asymptotic bound is set to $\epsilon = 0.001$ and $\epsilon = 0.0001$, and a diagonal channel covariance matrix has been used with $(R_h =$ diag([1.8949; 1.6222; 1.4209; 1.9405])).

Fig. 1 and Table. I show the asymptotic behavior of the bounds for channel coefficients and time delay estimation respectively. Fig. 1 shows that after 40 subcarriers the asymptotic bound for channel coefficients converges to the non-asymptotic ECRB of channel coefficients. Note that convergence behavior of the asymptotic ECRB of time delay is faster than channel coefficients due to the fact that it decreases by a factor of N^3 while the asymptotic ECRB of channel coefficients decreases by a factor of N. Therefore, to make the difference visible we used the Root Mean Square Error (RMSE) of the difference between asymptotic and non-asymptotic ECRB of time delay.

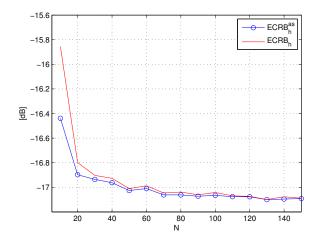


Fig. 1. Asymptotic behavior of ECRB $_h^{as}$ versus number of subcarriers N.

TABLE I RMSE of the difference between asymptotic and non-asymptotic ECRB of time-delay

N	10	50	100	150
RMSE [dB]	-56.3961	-86.8914	-98.9957	-106.1279

As it is clear from Fig. I the difference is of the order of -56dB to -106dB for N from 10 to 150 respectively.

Fig. 2 shows ergodic capacity for closed-loop system \bar{C}_{CL} obtain based on the optimization problem in (37) and (38) versus SNR. From Fig. 2, it is clear that by improving the accuracy in the estimation of time-delay \bar{C}_{CL} is reduced considerably for SNR below 30dB while by increasing SNR the effect of time-delay estimation accuracy on \bar{C}_{CL} is reduced.

VII. CONCLUSION

Applying asymptotic bounds one can use the simpler expressions of the bounds to the optimization problem specially by increasing the number of subcarriers N and channel taps L. In this paper, the performance of near-optimal pilot and data power allocations for the case of asymptotic bounds has been compared with the traditional non-asymptotic bounds. Results show that after a certain number of subcarriers which can be as low as N = 40, asymptotic bounds converge to the non-asymptotic bounds. Further, the performance of pilot power allocations is only affected negligibly even for the limited number of subcarriers and channel taps.

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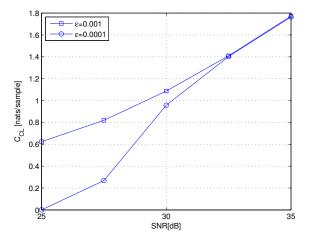


Fig. 2. Ergodic channel capacity for the closed-loop (CL) system using partially known CSI at the transmitter.

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