

Reviewers Comments & Authors Replies

Manuscript No.	Paper T-SP-18051-2014, submitted to “ <i>IEEE Transactions on Signal Processing</i> ”
Title	“Traffic Aware Resource Allocation Schemes for Multi-Cell MIMO-OFDM Systems”
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The authors are grateful to the associate editor for being considerate enough to provide us with this additional opportunity to revise and correct our convergence analysis. We thank all the reviewers for their valuable comments on the manuscript, which have been greatly helpful to improve the paper quality. Based on the comments, we have revised our manuscript to ensure that all the reviewers comments are addressed properly and also verified the technical correctness of the convergence proof. We have also modified some parts of text to improve the readability.

1. We have rewritten Appendix A-D to discuss the stationarity of limit points of the sequence of beam-former iterates generated by the centralized algorithms.
2. We have updated the convergence of distributed algorithm to reflect the same.

In what follows, the comments are listed, each followed immediately by the corresponding reply from the authors. The reviewers questions in the revised manuscript are highlighted in blue color and the authors responses are presented in black. Unless otherwise stated, all the numbered items (figures, equations, references, citations, etc) in this response letter refer to the revised manuscript. All revisions in the manuscript are highlighted in blue color.

List of Changes:

- Page 6, paragraph after (28)
- Page 8, last paragraph in Section IV-B
- Page 12, paragraph following itemized requirements in Appendix A
- Last sentence after (51) on page 14
- Last paragraph in Appendix A-C
- Appendix A-D is rewritten using subsequence convergence
- Last sentence after (60) on page 15

Response to first reviewer comments

The reviewer's concerns have been addressed and the manuscript can be published as it is.

Reply: We thank the reviewer for all constructive comments and recommending the revised manuscript for the publication. Moreover, we have modified the convergence analysis with additional technical materials to address some missing details in the convergence of beamformer iterates.

Response to second reviewer comments

I have no further comments.

Reply: We thank the reviewer for all constructive comments and recommending the revised manuscript for the publication. Moreover, we have modified the convergence analysis with additional technical materials to address some missing details in the convergence of beamformer iterates.

Response to third reviewer comments

The authors have modified the manuscript taken into account the reviewers' comments. However, the proof of convergence still lacks of convincing argumentation to make this paper suitable for publication. Two main aspects are arguable as follows:

Reply: We thank the reviewer for pointing out the flaw in our previous convergence proof. Based on the reviewer's suggestion, we have rewritten Appendix A-D with the subsequence convergence instead of claiming the sequence convergence. We hope that our current changes will be convincing enough to make the convergence proof complete.

In order to discuss the convergence of the sequence of iterates generated by the centralized method in Algorithm 1, we use a unified superscript index t to represent both alternating optimization (AO) index i and successive convex approximation (SCA) indexing k . By doing so, we represent iterate \mathbf{x}^t as the solution variable produced by the iterative algorithm in every SCA k and AO step i . Using this notation, we have revised our convergence discussions in Appendices A-C and A-D.

As an observation, we have noted that the feasible set of nonconvex problem (16) need not be bounded, even if the feasible set of the original nonconvex problem in (6) and the relaxed mean squared error (MSE) reformulated problem in (26) is bounded. It happens in problem (16) because of the newly introduced optimization variables in (16b) and (16c) to relax the signal-to-interference-plus-noise ratio (SINR) expression, since $\beta_{l,k,n}$ can assume any value satisfying (16c) without violating any other constraints in (16) when $\gamma_{l,k,n} = 0$. Therefore, to ensure boundedness of the feasible set of problem (16), we can include an additional bounding constraint on $\beta_{l,k,n}$ as $\beta_{l,k,n} \leq B_{\max}$, where B_{\max} can be a maximum interference threshold seen by any user in the network for a given maximum transmit power budget P_{\max} . Note that the additional constraint on $\beta_{l,k,n}$ will not alter the solution space of other optimization variables. However, as a consequence of considering the regularized objective in (46b), the additional constraint on $\beta_{l,k,n}$ is not required, since $\beta_{l,k,n}$ is bounded due to the proximal term in (46b). We have highlighted this discussion on page 6, after (28).

- When discussing the convergence of the SCA algorithm (Appendix A-C), the authors claim that the SCA algorithm converges because the objective function is strictly decreasing and the minimizer in each iteration is unique. However, this is in general not enough to guarantee the convergence of the whole sequence. Instead, one can only claim that every limit point of the whole sequence is a stationary point, or globally optimal solution if the problem is convex. As an example, simply consider the gradient projection algorithm, where each subproblem is strongly convex and the objective function is strictly decreasing but the whole sequence may not converge. The authors are referred to Bertsekas' book: *Nonlinear Programming*.

Reply: We agree with the reviewer's comment. Convergence of iterates cannot be guaranteed based on the strict monotonicity of the objective sequence and the uniqueness of minimizer in each step. We apologize for our emphasis on the convergence of sequence of iterates. Based on the reviewer's suggestion, we have modified the convergence analysis for the sequence of iterates generated by the centralized method in Appendices A-C and A-D. The revised discussion on the convergence of iterates is presented briefly here for reference. Since the feasible set is bounded, the iterates generated by the iterative algorithm are bounded. Therefore, there exists at least one subsequence that converges to a limit point. Now, by using a convergent subsequence of the sequence of iterates, we argue that every limit point is a stationary point of the original nonconvex problem.

- Even under the assumption that the inner loop SCA converges, there are doubts with respect to the convergence of the alternating optimization algorithm described by (54). The authors cite [27] to justify convergence, but it is not straightforward to claim that the algorithmic model considered in this manuscript is the same as in [27]. More specifically, let us consider the variable \mathbf{x} (so it is consistent with the authors' notation in Appendix A-D). In the algorithmic model of [27], at iteration $t + 1$, all elements of \mathbf{x} are updated simultaneously, based on \mathbf{x}^t . But in the authors' model, the update of \mathbf{x} is performed in two phases: in the first phase, only the elements of \mathbf{m} and γ in \mathbf{x} are updated based on

\mathbf{x}^t . In the second phase, the remaining elements of \mathbf{x} are updated, i.e. \mathbf{w} and γ , based on both \mathbf{x}^t and the elements that have been updated in the first phase. This is not a trivial difference and thus the direct application of the conclusion from [27] in the current context cannot be taken for granted. Due to the open concerns after several revision rounds, it is recommended to reject the paper.

Reply: We understand the reviewer’s concern on using [R1] to our problem involving both SCA and AO updates, which solves only for a subset of optimization variables by keeping others fixed. Even though we have revised our convergence proof based on a convergent subsequence of the original sequence without referring [R1], nevertheless, we clarify our usage of [R1] in our previous manuscript for the convergence of iterates. We refer to [R2] for a similar usage of [R1] while discussing the convergence of AO method used in designing tight frames by solving matrix nearness problem. The usage of [R1] to address the convergence of AO method in [R2] is based on the composition of two sub-algorithms and the compactness of the feasible set, where each sub-algorithm updates only a subset of optimization variables. It is discussed briefly in [Appendix I and II, R2] and in particular for AO algorithm in [Appendix I-F, R2].

However, we note that in our previous manuscript, convergence of the sequence of beamformer iterates claim based on [R1] was incomplete and we apologize for that. In [Theorem. 3.1, R1], it has been shown that if the strict monotonicity of the objective sequence is ensured and the iterates are bounded, then *either the sequence of iterates converges or the accumulation points of $\{\mathbf{x}^t\}$ is a continuum*. Therefore, our earlier claim on the guaranteed convergence of the sequence of iterates is not correct as pointed out by the reviewer. Similar approach of restricting the solution set with a proximal operator to ensure strict monotonicity of the objective sequence is highlighted in [Section 4, R1] under restrictive mapping. Therefore, by using the discussions in [R1], we can only claim that the sequence of iterates $\{\mathbf{x}^t\}$ converges to a continuum of accumulation points, and every limit point is a stationary point.

We have revised Appendix A-D titled “*Stationarity of Limit Points*” to discuss the convergence of the sequence of iterates. We have shown that every limit point of the sequence of iterates generated by the centralized algorithm is a stationary point.

- R1. R.R. Meyer, “Sufficient Conditions for the Convergence of Monotonic Mathematical Programming Algorithms,” *Journal of Computer and System Sciences*, vol. 12, no. 1, pp. 108-121, 1976
- R2. J. Tropp, I. Dhillon, R. Heath, and T. Strohmer, “Designing structured tight frames via an alternating projection method,” *IEEE Trans. Inform. Theory*, vol. 51, no. 1, pp. 188-209, Jan. 2005.

Response to fourth reviewer comments

This reviewer's concerns have been addressed, and this manuscript is now deemed fit for publication.

Reply: We thank the reviewer for all constructive comments and recommending the revised manuscript for the publication. Moreover, we have modified the convergence analysis with additional technical materials to address some missing details in the convergence of beamformer iterates.