

Efficient Single/Multiple Unimodular Waveform Design With Low Weighted Correlations

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Motivations

- Waveform design approaches *manipulate with correlation properties*. Focuses: 1) waveform quality itself; 2) mismatched filter design.
- Corresponding problems are *non-convex* and can grow to *large scale* as the code length and number of waveforms increase.
- The difficulty is also that the *required accuracy is high*.

Contributions

- Fast and efficient algorithm utilizing inherent algebraic structures in weighted integrated sidelobe level (WISL) expressions.
- Deriving the objective into an alternative quartic form which allows to apply the quartic-quadratic transformation.
- The non-convex quartic optimization into quadratic form by means of majorization-minimization (MaMi).
- Proposed algorithm shows faster convergence and better correlation properties compared to its counterparts.

WISL and Problem Formulation

- \mathbf{Y} : $P \times M$ waveform matrix (M : waveform number; P : code length).

- WISL expression:

$$\zeta = \sum_{m=1}^M \sum_{\substack{p=-P+1 \\ p \neq 0}}^{P-1} \gamma_p^2 |r_{mm}(p)|^2 + \sum_{m=1}^M \sum_{\substack{m'=1 \\ m' \neq m}}^M \sum_{p=-P+1}^{P-1} \gamma_p^2 |r_{mm'}(p)|^2.$$

- Unimodular waveform design problem:

$$\min_{\mathbf{y}} \zeta \quad \text{s.t. } |y_m(p)| = 1, \quad m = 1, \dots, M; \quad p = 1, \dots, P.$$

Main Results

- Transforming into frequency domain and performing derivations:

$$\zeta = \frac{1}{2P} \sum_{p=1}^{2P} \left\| \mathbf{Y}^H ((\mathbf{a}_p \mathbf{a}_p^H) \odot \boldsymbol{\Gamma}) \mathbf{Y} - \gamma_0 P \mathbf{I}_M \right\|^2.$$

- Alternative objective function in quartic form:

$$\zeta = \sum_{p=1}^{2P} \sum_{k=1}^K \sum_{k'=1}^K \left(\mathbf{y}^H ((\mathbf{A}_p \mathbf{A}_p^H) \odot \boldsymbol{\Gamma}_{kk'}^{\text{real}}) \mathbf{y} \right)^2 + \left(\mathbf{y}^H ((\mathbf{A}_p \mathbf{A}_p^H) \odot \boldsymbol{\Gamma}_{kk'}^{\text{img}}) \mathbf{y} \right)^2.$$

- Generalized majorization function for $f(x)$ at point x_0 :

$$g(\mathbf{x}) = f(\mathbf{x}_0) + \Re \left\{ (\nabla f(\mathbf{x}_0))^H (\mathbf{x} - \mathbf{x}_0) \right\} + \frac{1}{2} (\mathbf{x} - \mathbf{x}_0)^H \mathbf{G} (\mathbf{x} - \mathbf{x}_0).$$

- Objective after 1st majorization in quadratic form:

$$(\text{vec}^H(\tilde{\mathbf{Y}}))^H \left(\tilde{\boldsymbol{\Phi}} - \frac{\lambda_{\tilde{\boldsymbol{\Phi}}}}{2} \mathbf{I}_{M^2 P^2} \right) \text{vec}(\tilde{\mathbf{Y}}^{(k)}).$$

- Objective after 2nd majorization in quadratic form:

$$\mathbf{y}^H \left(\mathbf{B}^{(k)} - \frac{\lambda_{\tilde{\boldsymbol{\Phi}}}}{2} \mathbf{y}^{(k)} (\mathbf{y}^{(k)})^H \right) \mathbf{y}$$

$$\begin{aligned} \mathbf{B}^{(k)} &\triangleq \begin{bmatrix} \mathbf{B}_{11}^{(k)} & \dots & \mathbf{B}_{1M}^{(k)} \\ \vdots & \ddots & \vdots \\ \mathbf{B}_{M1}^{(k)} & \dots & \mathbf{B}_{MM}^{(k)} \end{bmatrix}; \quad \rho_{mm'}^{(k)} \triangleq \begin{cases} \gamma_p^2 \mathbf{1}_{P-p}^T \mathcal{U}_p(\mathbf{Z}_{mm'}^{(k)}), & p \in \Omega \\ 0, & p \in \bar{\Omega} \end{cases} \\ \mathbf{B}_{mm'}^{(k)} &= 2P\mathcal{T}(\rho_{mm'}^{(k)}, \eta_{mm'}^{(k)}); \quad \eta_{mm'}^{(k)} \triangleq \begin{cases} \gamma_p^2 \mathbf{1}_{P-p}^T \mathcal{D}_p(\mathbf{Z}_{mm'}^{(k)}), & p \in \Omega \\ 0, & p \in \bar{\Omega} \end{cases} \end{aligned}$$

$\Omega \triangleq \{0\} \cup \{p \mid \gamma_p \neq 0, p > 0\}; \quad \bar{\Omega} \triangleq \{p \mid \gamma_p = 0, p > 0\};$
 $\mathcal{U}_p(\cdot), \mathcal{D}_p(\cdot)$: p th off-diagonal vector element in the upper/lower triangular part of a matrix; $\mathcal{T}(\cdot, \cdot)$: construction of Hermitian matrix from vector(s).

- Final optimization problem:

$$\begin{aligned} \min_{\mathbf{y}} \quad & \mathbf{y}^H \left(\mathbf{B}^{(k)} - \frac{\tau^{(k)} + MP\lambda_{\tilde{\boldsymbol{\Phi}}}}{2} \mathbf{I}_{MP} \right) \mathbf{y}^{(k)} \\ \text{s.t.} \quad & |\mathbf{y}(p')| = 1, \quad p' = 1, \dots, MP. \end{aligned}$$

- Closed-form solution:

$$\mathbf{y}(p') = \exp\{j \cdot \arg(\mathbf{z}^{(k)}(p'))\}, \quad p' = 1, \dots, MP$$

with $\mathbf{z}^{(k)} \triangleq ((\tau^{(k)} + MP\lambda_{\tilde{\boldsymbol{\Phi}}}) \mathbf{I}_{MP}/2 - \mathbf{B}^{(k)}) \mathbf{y}^{(k)}$.

Fast WISL Minimization-Based Algorithm

- Algorithm

Algorithm 1 Fast WISL Minimization-Based Algorithm

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1:  $k \leftarrow 0$ ,  $\mathbf{y} \leftarrow$  unimodular sequence with random phases.
2:  $\lambda_{\tilde{\boldsymbol{\Phi}}} \triangleq 2MP\lambda_{\max}^2(\boldsymbol{\Gamma})$ 
3: repeat
4:   procedure WISLMAMI( $\mathbf{y}^{(k)}$ )
5:     Calculate  $\rho_{mm'}^{(k)}$  and  $\eta_{mm'}^{(k)}$ ,  $m = 1, \dots, M; m' = m, \dots, M$ .
6:     Construct  $\mathbf{B}^{(k)}$  through  $\mathbf{B}_{mm'}^{(k)} = (\mathbf{B}_{m'm}^{(k)})^H = 2P\mathcal{T}(\rho_{mm'}^{(k)}, \eta_{mm'}^{(k)})$ 
7:      $\tau^{(k)} = \|\mathbf{B}^{(k)} - \frac{1}{2}\lambda_{\tilde{\boldsymbol{\Phi}}}\mathbf{y}^{(k)}(\mathbf{y}^{(k)})^H\|$ 
8:      $\mathbf{z}^{(k)} = \left( \frac{1}{2}(\tau^{(k)} + MP\lambda_{\tilde{\boldsymbol{\Phi}}}) \mathbf{I}_{MP} - \mathbf{B}^{(k)} \right) \mathbf{y}^{(k)}$ 
9:      $\mathbf{y}^{(k+1)}(p') = e^{j \cdot \arg(\mathbf{z}^{(k)}(p'))}, \quad p' = 1, \dots, MP$ 
10:     $k \leftarrow k + 1$ 
11:  end procedure
12: until convergence

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- Computational complexity:

- Assume Ω consists of N_P ($0 < N_P \leq P$) elements.
- Both $\rho_{mm'}^{(k)}$ and $\eta_{mm'}^{(k)}$ need at most $N_P P$ operations if $\mathbf{Z}_{mm'}^{(k)}$ is given.
- Calculation of $\mathbf{Z}_{mm'}^{(k)}$ costs P^2 operations.
- Calculations are repeated only for subscripts $m = 1, \dots, M$ and $m' = m, \dots, M$.
- Calculation of the vector $\mathbf{z}^{(k)}$ needs $M^2 P^2$ operations.
- Total number of operations is upper bounded by $((3M^2 - M)P^2 + (M^2 - M)N_P P)/2$, i.e., at most of order $\mathcal{O}((M - 1)MP^2)$ suitable for large-scale optimization.

Simulation Results (Convergence speed and Correlation properties)

- WISL performance comparisons of the algorithms tested versus code length for $M = 2$ waveforms for stopping criterion (i):

	$P = 32, M = 2$	$P = 128, M = 2$	$P = 512, M = 2$	$P = 1024, M = 2$	$P = 2048, M = 2$							
	Min. ^a	Ave. ^b	Time	Iter.	Min.	Ave.	Time	Iter.	Min.	Ave.	Time	Iter.
WeCAN	29.08	29.84	1.47	314	19.84	19.83	5.82	200	19.38	20.44	17.65	67
WISLSong	27.80	28.65	0.57	238	-34.88	-27.18	0.78	41	-47.12	-21.17	7.82	21
WISLNew	27.75	28.44	1.05	94	-65.74	-31.76	0.37	26	-80.31	-50.37	2.05	7

^aMin.: Obtained minimum WISL value (in dB). ^bAve.: Obtained average WISL value (in dB). ^cIter.: Number of conducted iterations.

- WISL performance comparisons of the algorithms tested versus code length for $M = 2$ waveforms for stopping criterion (ii):

	$P = 32, M = 2$	$P = 128, M = 2$	$P = 512, M = 2$	$P = 1024, M = 2$	$P = 2048, M = 2$							
	Min. ^a	Ave. ^b	Time	Iter.	Min.	Ave.	Time	Iter.	Min.	Ave.	Time	Iter.
WeCAN	28.20	28.92	1.58	374	18.70	19.61	7.59	267	20.72	20.84	18.47	84
WISLSong	27.48	28.59	0.84	298	-40.31	-39.38	0.80	51	-47.63	-46.42	8.55	14
WISLNew	27.67	28.25	0.30	149	-49.91	-37.68	0.20	24	-59.52	-55.09	1.22	7

^aMin.: Obtained minimum WISL value (in dB). ^bAve.: Obtained average WISL value (in dB). ^cIter.: Number of conducted iterations.

- Auto- and cross-correlations of $M = 2$ designed waveforms with code length $P = 32$:

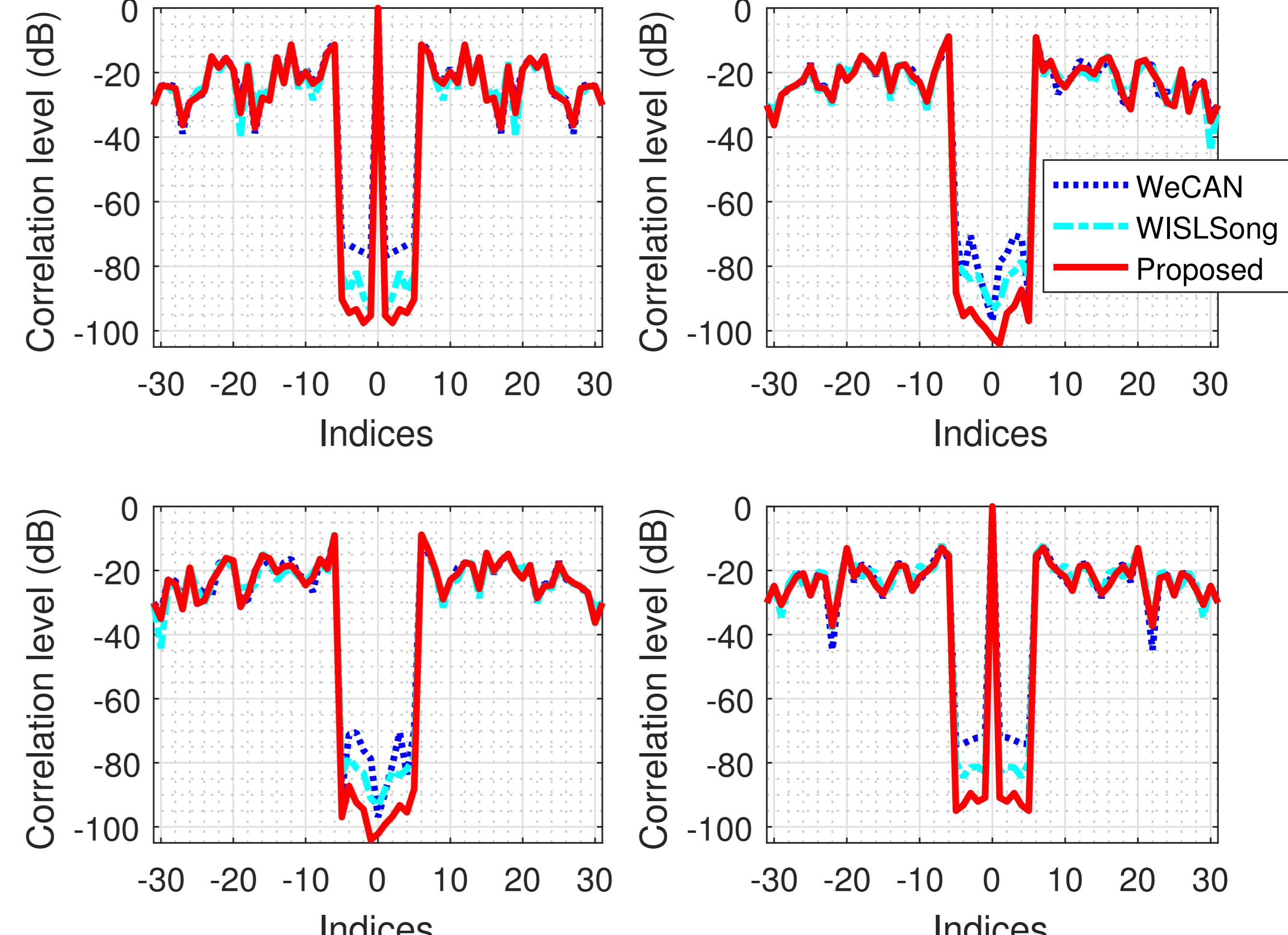


Figure 1: Evaluation of correlation properties.

References

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