

EMF-Aware Design of Cellular Networks with Reconfigurable Intelligent Surfaces



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EMF Exposure

- Concerns about EMF exposure were raised recently due to:
 - The electromagnetic spectrum is already congested.
 - 5G Networks depends on dense deployment of base stations.
 - Although the exposure is dominated in the uplink.
 - There is no clear evidence of the EMF harness from cellular networks.
 - As a precautionary action, Cellular networks EMF is considered as Possibly carcinogenic (*Group 2B*) according to International Agency on Research on Cancer



Reconfigurable Intelligent Surface (RIS)

- What is it?
 - A surface a surface that is segmented to form a grid of discrete elements.
 - Each element can reflect the incident electromagnetic wave to a certain direction (generalized Snell's law).
 - A phase shift can be induced on the signal.
 - This surface is controlled by the base station (BS).



Applications

- Main purpose of the RIS:
 - Overcome the line of sight (LOS) problem in mm Waves.
- Other applications of the RIS include:
 - Energy efficiency maximization.
 - Sum rate maximization.
 - Localization using RIS.
- In this work, RIS is used to minimize the EMF exposure.



Our Contribution

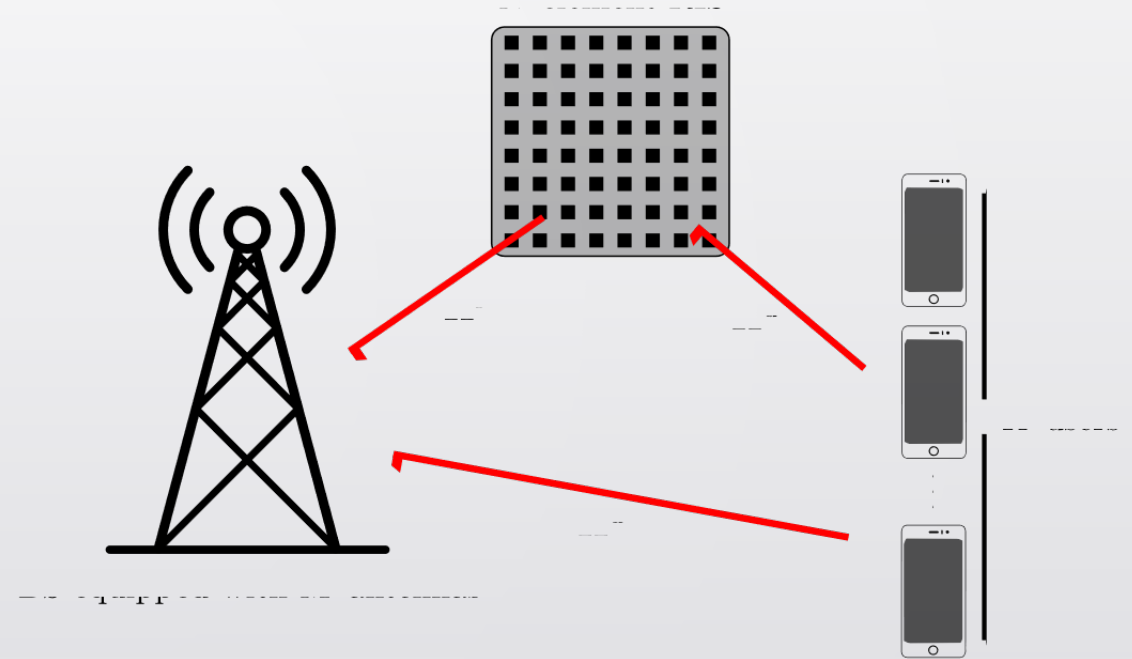
- We develop a dual gradient descent algorithm for the joint design of the RIS phases and the BS beamformer.
- We adopt a general model:
 - The direct path is included in the model.
 - Each user may have a different reference exposure metric.
- The gradient, Hessian matrix, and the suboptimal step size are derived in closed-form.

System Model

- The received signal at the BS for a MISO system can be represented as:

$$\mathbf{y} = (\mathbf{H}^r \Phi(\boldsymbol{\theta}) \mathbf{H}^u + \mathbf{H}^d) \mathbf{x} + \mathbf{n},$$

where $\mathbf{x} \in \mathbb{C}^{k \times 1}$ represents users' transmitted signal, $\mathbf{H}^u \in \mathbb{C}^{N \times k}$, $\mathbf{H}^r \in \mathbb{C}^{M \times N}$ and $\mathbf{H}^d \in \mathbb{C}^{K \times M}$ represent the channel between (users and the RIS), (RIS and BS) and (users and the BS), respectively. $\Phi(\boldsymbol{\theta})$ is a diagonal matrix where $\Phi_{i,i} \triangleq e^{j\theta_i}$ represents the phase shift induced by the i^{th} element. Finally, $\mathbf{n} \in \mathbb{C}^{M \times 1}$ is the AWGN noise at the BS with zero mean and σ^2 variance.





Exposure Index (EI)

- The EMF exposure can be quantified using the exposure index which is defined as

$$EI(\boldsymbol{\theta}) \triangleq \sum_{k=1}^K SAR_k = \sum_{k=1}^K SAR_k^{\text{ref}} p_k [W/Kg].$$

Specific Absorption Rate (SAR) represents the absorbed power for a unit of mass.

- SAR_k^{ref} is user k 's SAR when the transmitted power is unity.



Optimization Problem

$$\underset{\mathbf{p}, \boldsymbol{\theta}, \mathbf{G}}{\text{minimize}} \quad EI(\mathbf{p}),$$

s.t.

$$\begin{aligned} r_k(\boldsymbol{\theta}, \mathbf{G}, \mathbf{p}) &\triangleq \log_2(1 + SNIR_k) \geq r_k^{\text{th}} && \forall k \\ p_k &< p_{\max} && \forall k \end{aligned}$$

where $r_k(\boldsymbol{\theta}, \mathbf{G})$ in bits/Hz is the actual spectral efficiency for user k and r_k^{th} is the minimum spectral efficiency required by user k .



Problem Reduction

- First, we design the receiver beamformer in such a way that it eliminates the cross interference (Zero Forcing beamformer) which is represented as

$$\mathbf{G}(\boldsymbol{\theta}) = (\mathbf{H}^r \boldsymbol{\Phi}(\boldsymbol{\theta}) \mathbf{H}^u + \mathbf{H}^d)^+.$$

- Since the objective function is monotonically increasing in the users' power, the optimal power allocation strategy is such that the data rate constraint holds with equality. p_k can be represented as

$$p_k(\boldsymbol{\theta}) = \left(2^{r_k^{\text{th}}} - 1\right) \sigma^2 \|g_{(k)}\|^2.$$



The Reduced Optimization Problem

After making \mathbf{G} and \mathbf{p} as variables of $\boldsymbol{\theta}$, $\boldsymbol{\theta}$ becomes the only optimization variable to consider. Therefore, we can rewrite our problem as

$$\underset{\boldsymbol{\theta}}{\text{minimize}} \sum_{k=1}^K SAR_k^{\text{ref}} p_k(\boldsymbol{\theta}),$$

s.t.

$$p_k(\boldsymbol{\theta}) < p_{max}.$$



The Lagrangian

- The Lagrangian of our problem can be written as

$$\mathcal{L}(\boldsymbol{\theta}, \boldsymbol{\lambda}) = EI(\boldsymbol{\theta}) + \boldsymbol{\lambda}^T (\mathbf{p}(\boldsymbol{\theta}) - p_{max}),$$

where $\boldsymbol{\lambda} \in \mathbb{R}^{K \times 1}$ represents the KKT multipliers of the Lagrangian

- The optimal point must be a saddle point within the Lagrangian according to the KKT conditions.

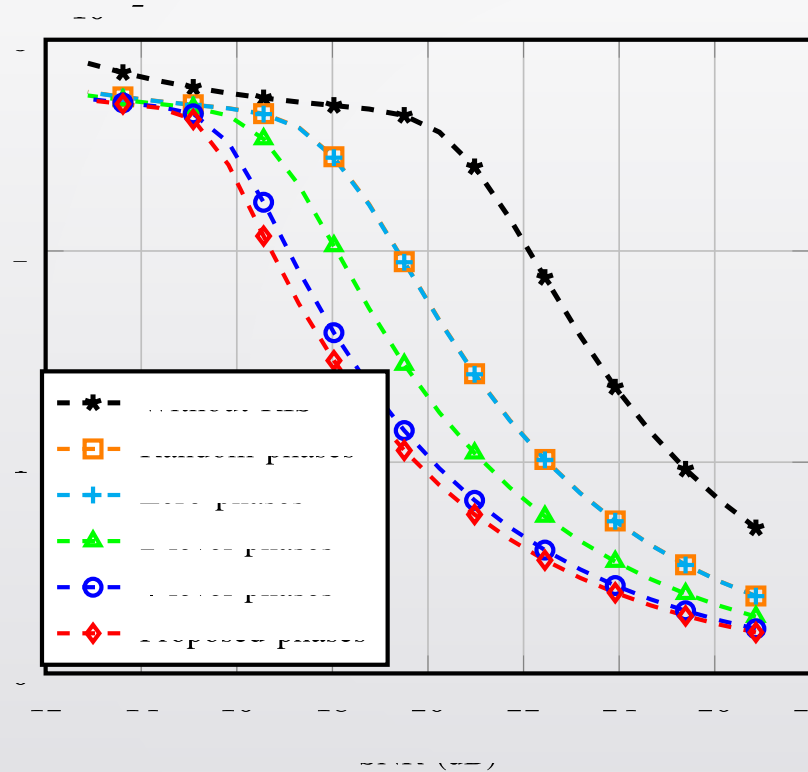


Dual Gradient Descent Algorithm

- One way to find the saddle point in the Lagrangian we apply the dual gradient descent algorithm as follows

$$\begin{aligned}\boldsymbol{\theta}^{(t)} &\leftarrow \boldsymbol{\theta}^{(t-1)} - \alpha_{\boldsymbol{\theta}} \nabla_{\boldsymbol{\theta}} \mathcal{L}(\boldsymbol{\theta}^{(t-1)}, \boldsymbol{\lambda}^{(t-1)}), \\ \boldsymbol{\lambda}^{(t)} &\leftarrow \boldsymbol{\lambda}^{(t-1)} + \alpha_{\boldsymbol{\lambda}} \nabla_{\boldsymbol{\lambda}} \mathcal{L}(\boldsymbol{\theta}^{(t)}, \boldsymbol{\lambda}^{(t-1)}).\end{aligned}$$

Proposed Algorithm Performance



Proposed Algorithm Performance

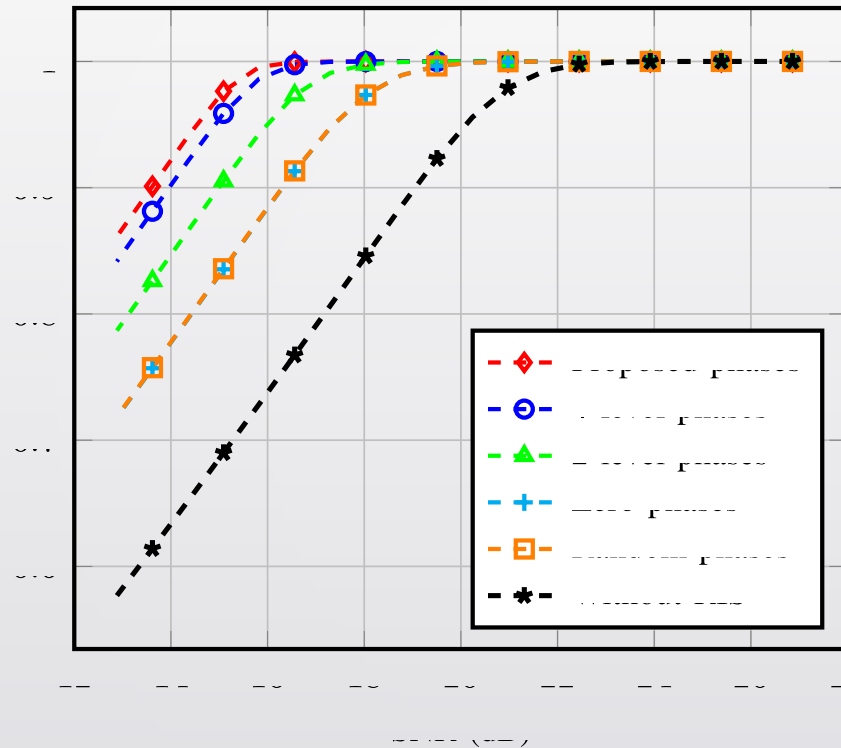


Figure 1: Performance comparison of the proposed algorithm with other methods. The proposed algorithm (black asterisk) shows superior performance, achieving a higher F1 score faster than the other methods.

Thank you

