

Optimizing a Binary Intelligent Reflecting Surface for OFDM Communications



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IEEE SP Cup

- The most prestigious competition in signal processing for undergraduate students
- Solving real-world problems with signal processing methods
- The competition finals take place at the ICASSP conference

Goals

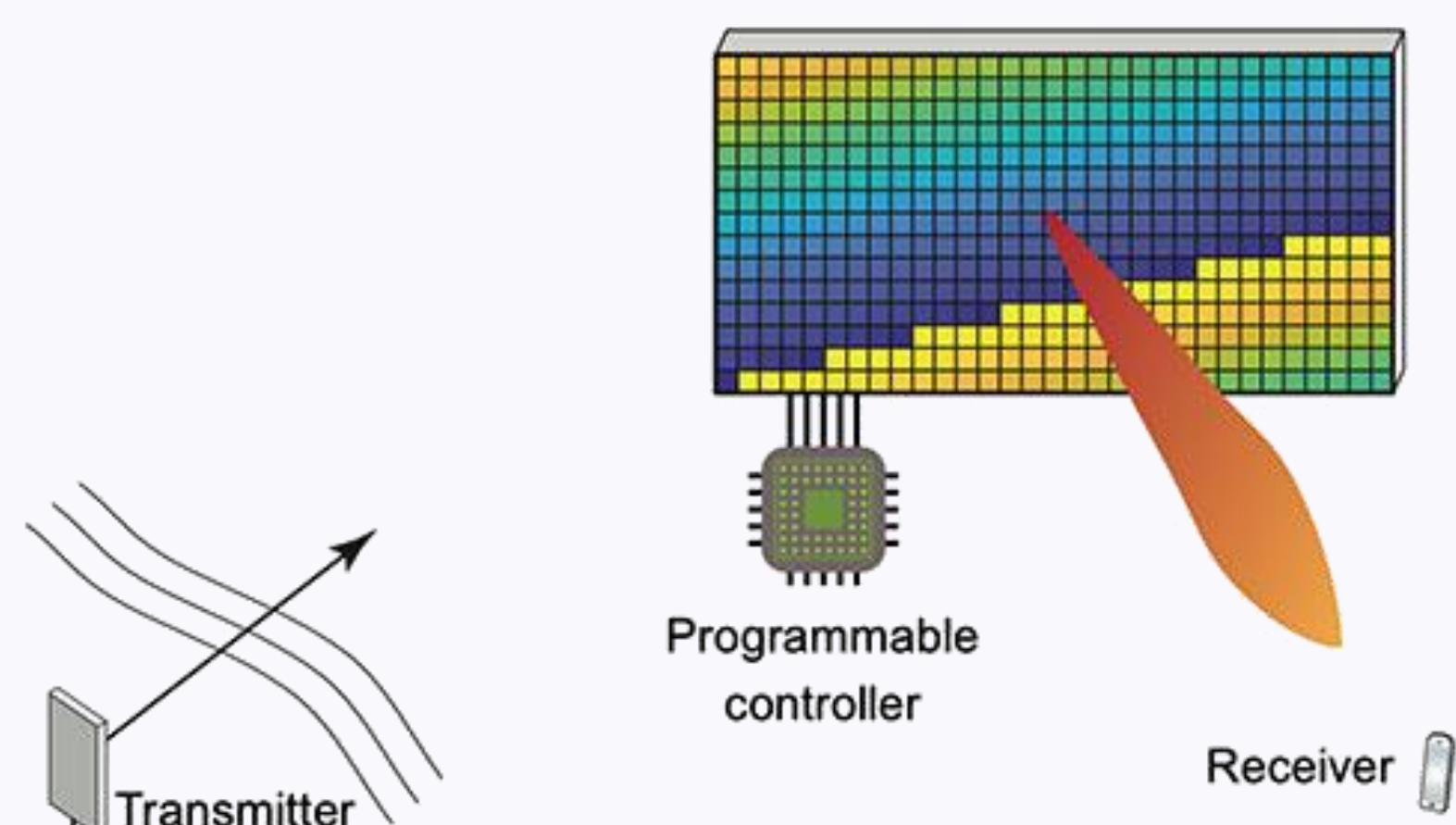
- Characterize the behavior of an intelligent reflecting surface
- Develop a control algorithm to configure the surface
- For each user, find a configuration that gives the highest data transmission rate
- Obtain highest score among all the competitors

Challenges

- Finding best configurations out of 2^{4096}
- No data provided on the IRS spatial shape
- Insufficient small dataset
- Limited number of papers on IRS discrete optimization

Intelligent Reflective Surface

- An Intelligent Reflecting Surface (IRS) is a two-dimensional array of metamaterial
- Consists of an array of controllable passive elements
- Can alter the amplitude and/or phase of the reflected signal
- Helps to overcome the problem of signal attenuation in 6G communication



$$z[k] = \sum_{\ell=0}^{M-1} (h_d[\ell] + v_\ell^T \omega_\theta) x[k - \ell] + w[k]$$

- $z[k]$ - Received signal
- $x[k]$ - Transmitted signal
- h_d - Direct channel BS \Rightarrow Receiver
- v_ℓ - Cascaded channel BS \Rightarrow IRS \Rightarrow Receiver
- $\omega_\theta \in \{\pm 1\}^{1 \times 4096}$ - IRS configuration
- $w[k]$ - AWGN

Data Set

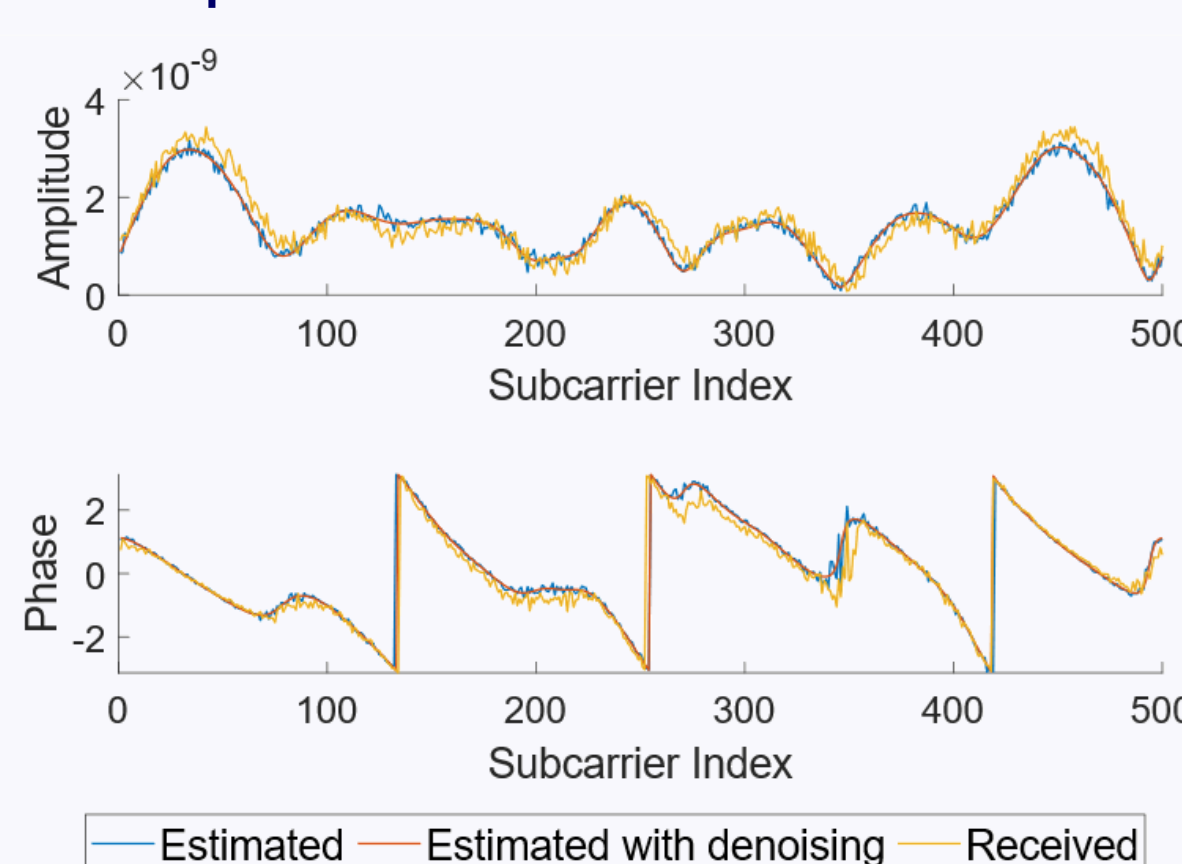
- A novel synthetic dataset, divided to two:
 - Dataset1 includes 4×4096 configurations, and their received signals for a certain user
 - Dataset2 includes 4096 configurations, and their received signals for 50 users
- The transmitted pilot signal is $x[k] \equiv \alpha$
- The pilot configurations create a 4096×4096 Hadamard matrix

Channel Estimation

- Direct channel estimation using the pilot matrix Hadamard formation:

$$\hat{h}_d = \frac{\sum_{i=1}^{4096} z_i}{4096 \cdot \alpha}$$
 - z_i - The received pilot signal
- BS \Rightarrow IRS \Rightarrow Receiver cascaded channel estimation

$$\hat{V} = \frac{1}{\alpha} ([z_1^f \dots z_{4096}^f] - [h_{d1}^f \dots h_{d4096}^f] \cdot \alpha) [\omega_{\theta 1} \dots \omega_{\theta 4096}]^{-1}$$
- Estimated the spectral noise density using redundancy in dataset 1
- Deeper understanding of the IRS geometrical shape helped us to overcome the lack of data

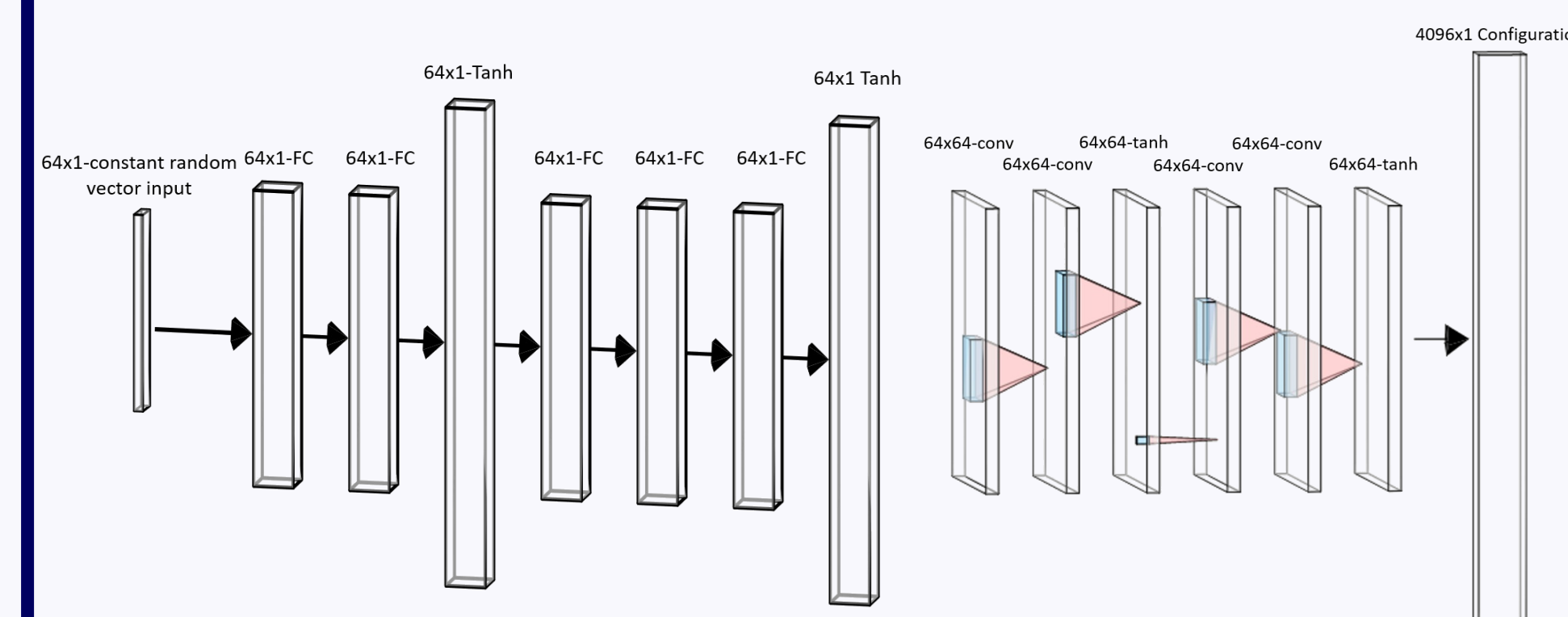


Gradient-Quantization Algorithm

- Efficient heuristic solution exist for the continuous phase case – Strongest Tap Maximization (STM) [Zheng & Zhang, 2020]
 - Initialize: STM continuous phase configuration
 - Find best separating line for the quantized phases
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- Our novel algorithm consists of two steps that repeat until all IRS elements are fixed :
 - Quantize of the most ambiguous element
 - Optimize the unquantized element using gradient-descent

Generative Neural Network

- A novel method for configuration optimization
- Based on the known paper “Deep Image Prior” [Ulyanov et al., 2018],
- Use an untrained CNN as a regulator
- Optimize the weights to get maximal data transmission rate



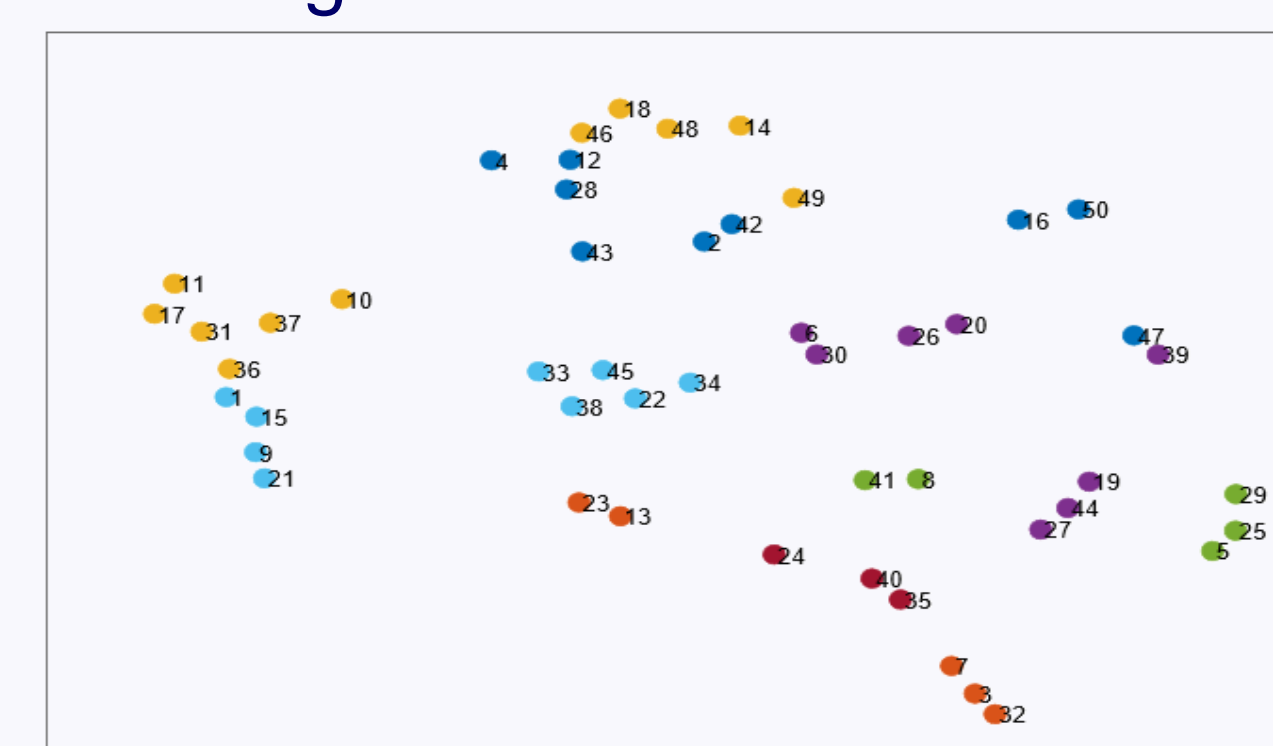
- The output of the first NN is a constant-columns steering configuration
- We used the second NN to fine tune the configuration



First NN output example Second NN output example

Localization of users

- It is possible to locate the relative positions of the users based on the similarity between their optimal configuration



Results

- Finished in the 6th place in the competition

Method	Mean Data Transmission Rate
Best configuration from the given dataset	103.55 $\frac{[Mb]}{[sec]}$
The competition instructor method	117.36 $\frac{[Mb]}{[sec]}$
Our method	117.73 $\frac{[Mb]}{[sec]}$

