

MIMO physical layer security using multiple Reconfigurable Intelligent Surfaces

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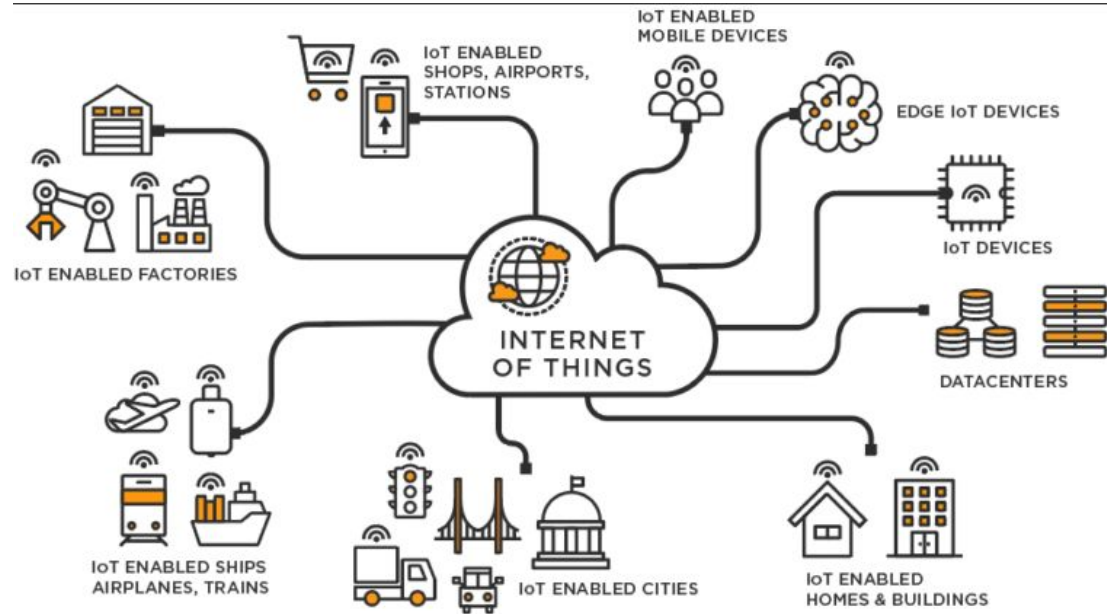


MANTA
Mobile Adaptive
and Pervasive
Network Communications



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Background and motivation



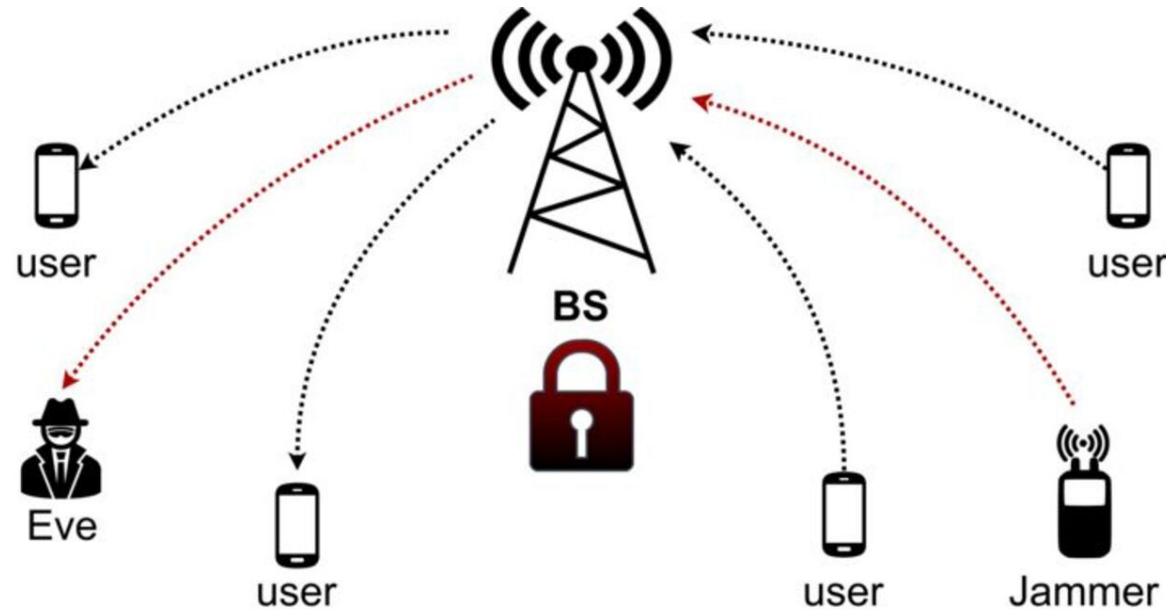
<https://businesstech.bus.umich.edu/uncategorized/tech-101-internet-of-things/>



<https://www.itf-oecd.org/co-operative-mobility-systems-automated-driving-roundtable>

- Our lives depend more and more on various devices
- They need to communicate fast, reliably and securely with each other
- These requirements cannot be mutually exclusive anymore

Physical Layer Security



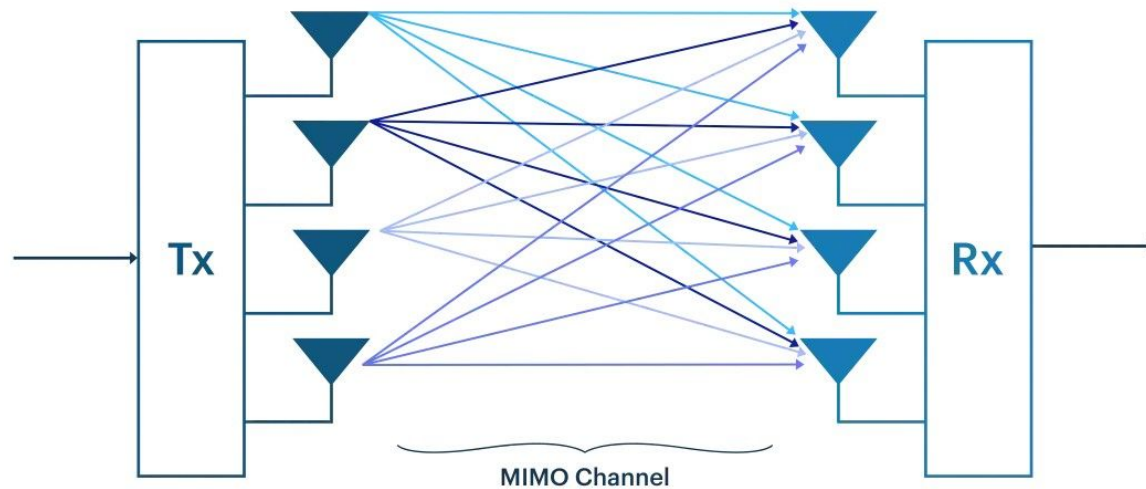
<https://liu.se/en/research/physical-layer-security-in-massive-mimo>



<https://quantumai.google/discover/whatisqc>

- With our lives depending more and more on technology, we need to be protected from malicious actors that may hear or disrupt our communications
- Quantum computing could break encryption
- We need new, low latency security schemes

Multiple Input Multiple Output (MIMO)



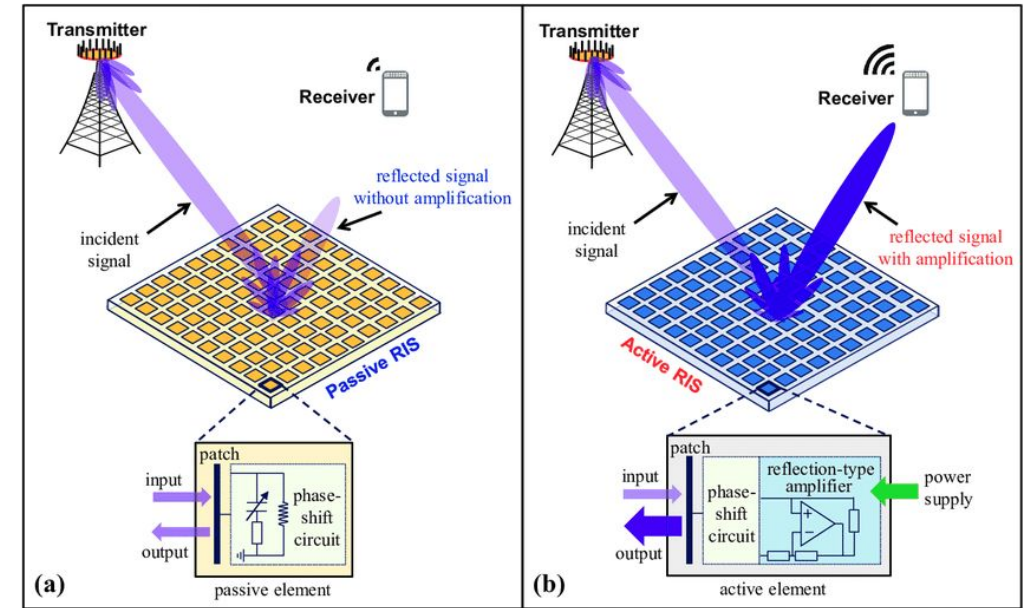
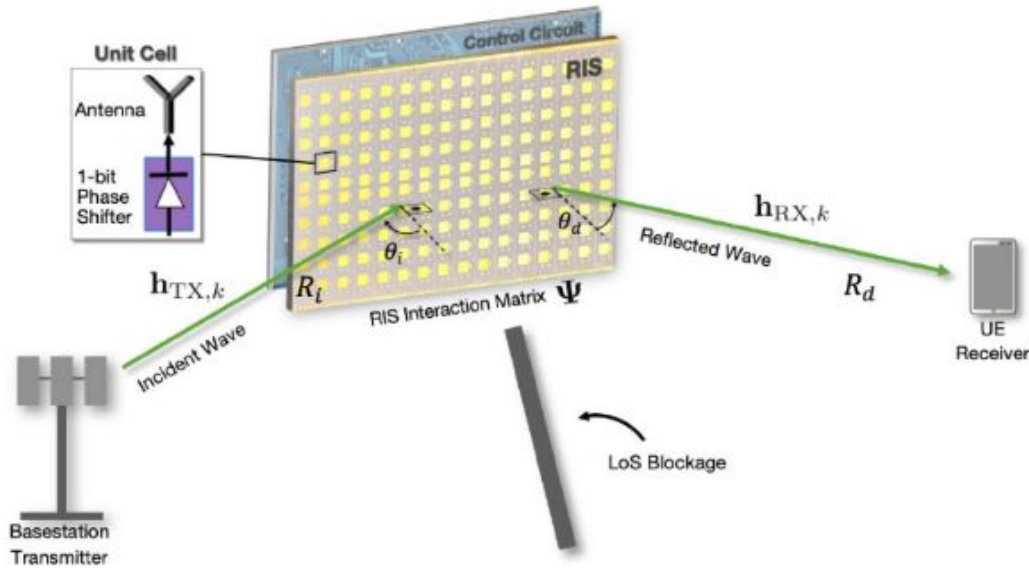
$$\mathbf{B} = \begin{bmatrix} \beta_{11} & \beta_{12} & \cdots & \beta_{1K} \\ \beta_{21} & \beta_{22} & \cdots & \beta_{2K} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{K1} & \beta_{K2} & \cdots & \beta_{KK} \end{bmatrix} \in \mathbb{C}^{K \times K}$$

$$\textcircled{c} \quad y = \mathbf{B}x = \begin{bmatrix} \beta_{11} & \beta_{12} & \cdots & \beta_{1K} \\ \beta_{21} & \beta_{22} & \cdots & \beta_{2K} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{K1} & \beta_{K2} & \cdots & \beta_{KK} \end{bmatrix} \begin{bmatrix} \xi_1 \\ \xi_2 \\ \vdots \\ \xi_K \end{bmatrix} = \begin{bmatrix} \delta_1 \\ \delta_2 \\ \vdots \\ \delta_K \end{bmatrix} \in \mathbb{C}^K$$

<https://www.linkedin.com/pulse/enhancing-wireless-communication-mimo-technology-cavliwireless-wzo0e/>

- We can use multiple antennas to communicate
- The signals from each transmitter antenna to each receiver antenna form a matrix of complex numbers, the *channel gain matrix*
- The received total signal at the receiver is the transmitter signal multiplied by the transmitted signal

Reconfigurable Intelligent Surfaces (RISs)

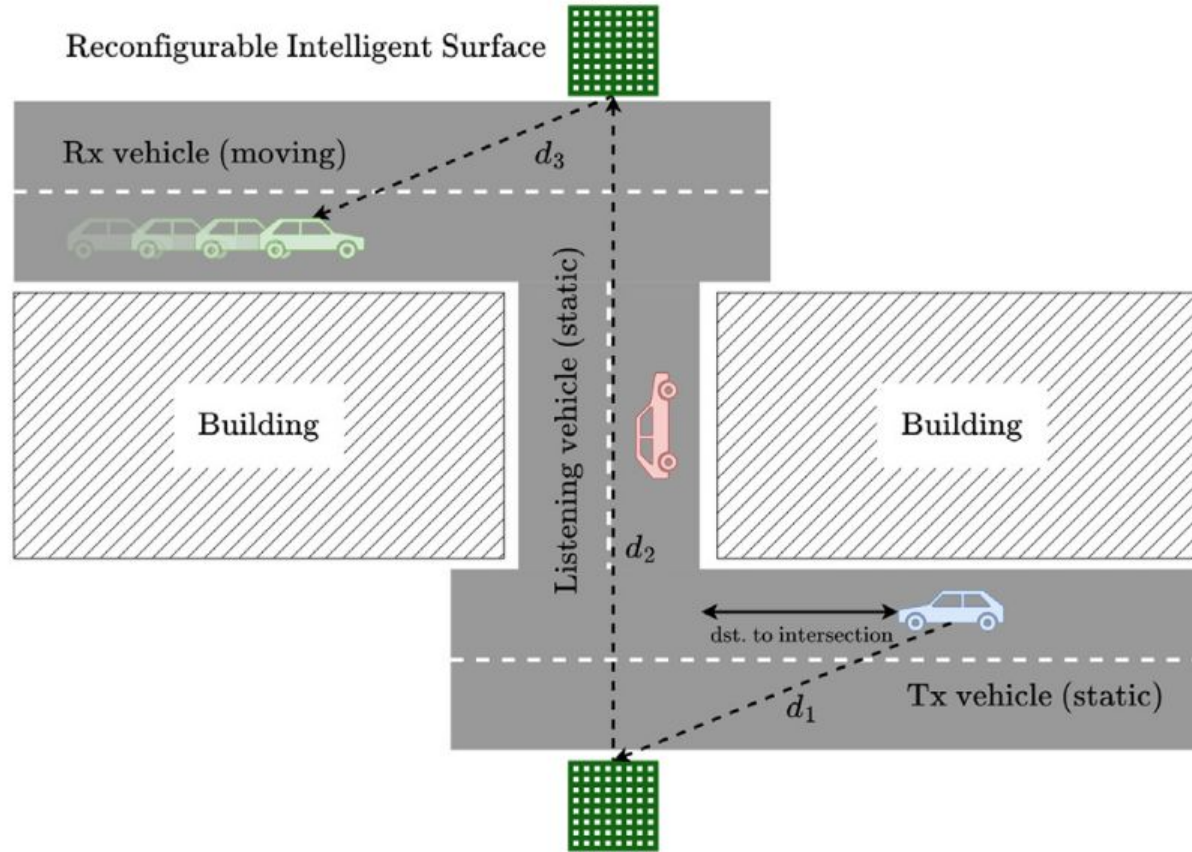


https://www.researchgate.net/figure/Comparison-between-the-existing-passive-RIS-a-and-the-proposed-active-RIS-b_fig1_350484632

- RISs are a promising technology that can help us expand the signal reach
- We can modulate the reflection to better suit our needs
- They can be either passive or active

G. C. Trichopoulos *et al.*, "Design and Evaluation of Reconfigurable Intelligent Surfaces in Real-World Environment,"

Using RISs for physical layer security



(b) Z-intersection scenario

- Our objective is to find a way to hide our signal in NLoS scenarios, using the same RISs we use for actors to communicate
- This is crucial in crossroads used by cooperative autonomous vehicles

Michele Segata, Paolo Casari, Marios Lestas, Alexandros Papadopoulos, Dimitrios Tyrovolas, Taqwa Saeed, George Karagiannidis, Christos Liaskos, CoopeRIS: A framework for the simulation of reconfigurable intelligent surfaces in cooperative driving environments,

Space Shift Keying (SSK) Modulation

$\mathbf{b} = [b_1 \quad b_2]$	symbol	antenna index j	$\mathbf{x} = [x_1 \quad \dots \quad x_4]^T$
$\begin{bmatrix} 0 & 0 \end{bmatrix}$	0	1	$\begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix}^T$
$\begin{bmatrix} 0 & 1 \end{bmatrix}$	1	2	$\begin{bmatrix} 0 & 1 & 0 & 0 \end{bmatrix}^T$
$\begin{bmatrix} 1 & 0 \end{bmatrix}$	2	3	$\begin{bmatrix} 0 & 0 & 1 & 0 \end{bmatrix}^T$
$\begin{bmatrix} 1 & 1 \end{bmatrix}$	3	4	$\begin{bmatrix} 0 & 0 & 0 & 1 \end{bmatrix}^T$

- Communicate using the antenna index
- Resistant to noise and signal variations
- More complex schemes exist to use multiple antennas at the same time

$$j = \arg \max_j p_y(y|x_j, \mathbf{B}) = \arg \min_j ||y - b_j||^2$$

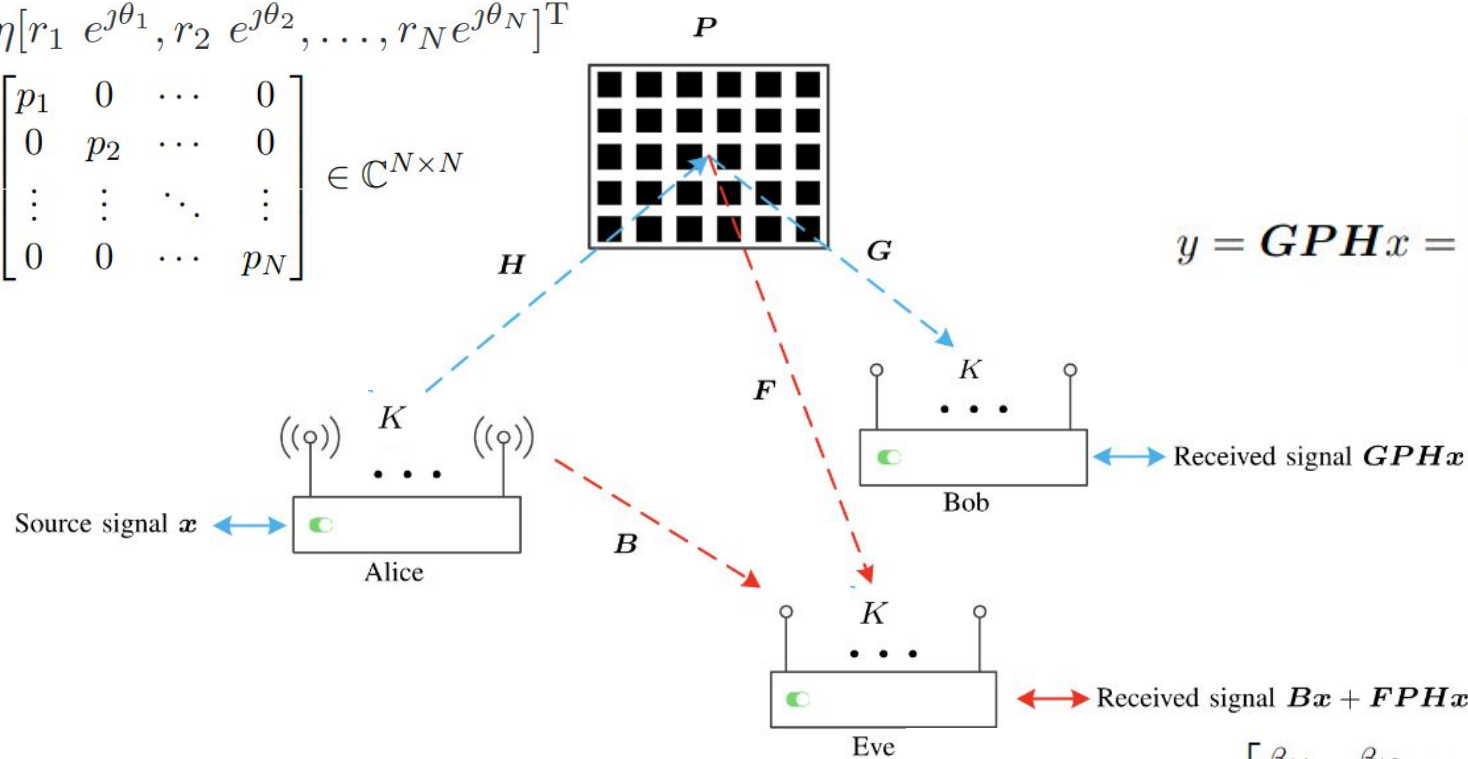
$$y = \mathbf{B}x = \begin{bmatrix} \beta_{11} & \beta_{12} & \dots & \beta_{1K} \\ \beta_{21} & \beta_{22} & \dots & \beta_{2K} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{K1} & \beta_{K2} & \dots & \beta_{KK} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} \beta_{11} \\ \beta_{12} \\ \vdots \\ \beta_{1K} \end{bmatrix}$$

J. Jeganathan, A. Ghrayeb, L. Szczecinski and A. Ceron,
"Space shift keying modulation for MIMO channels,"

Reference scenario

$$\mathbf{p} = \eta[r_1 e^{j\theta_1}, r_2 e^{j\theta_2}, \dots, r_N e^{j\theta_N}]^T$$

$$\mathbf{P} = \begin{bmatrix} p_1 & 0 & \cdots & 0 \\ 0 & p_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & p_N \end{bmatrix} \in \mathbb{C}^{N \times N}$$



$$j = \arg \max_j y_j$$

$$\mathbf{y} = \mathbf{GPHx} = \begin{bmatrix} \alpha_{11} & 0 & \cdots & 0 \\ 0 & \alpha_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \alpha_{KK} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} \alpha_{11} \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\mathbf{y} = \mathbf{Bx} + \mathbf{FPHx} = \begin{bmatrix} \beta_{11} & \beta_{12} & \cdots & \beta_{1K} \\ \beta_{21} & \beta_{22} & \cdots & \beta_{2K} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{K1} & \beta_{K2} & \cdots & \beta_{KK} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + \begin{bmatrix} \gamma_{11} & \gamma_{12} & \cdots & \gamma_{1K} \\ \gamma_{21} & \gamma_{22} & \cdots & \gamma_{2K} \\ \vdots & \vdots & \ddots & \vdots \\ \gamma_{K1} & \gamma_{K2} & \cdots & \gamma_{KK} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} \delta_1 \\ \delta_2 \\ \vdots \\ \delta_K \end{bmatrix}$$

J. Luo, F. Wang, S. Wang, H. Wang and D. Wang,
"Reconfigurable Intelligent Surface: Reflection Design Against Passive Eavesdropping,"

RIS parametrization

<https://ieeexplore.ieee.org/document/9328149>

$$\|GPH - [GPH]_{diag}\|^2 = 0$$

$$\mathbf{W} = \sum_{i,k=1, i \neq k}^K (g_k \odot h_i^T)^H (g_k \odot h_i^T)$$

$$\mathbf{W}p = 0$$

$$\mathbf{W} = \mathbf{R}\mathbf{\Sigma}\mathbf{V}^H$$

$$N(\mathbf{W}^H) = [r_{N-K(K-1)}, \dots, r_N] = \mathbf{U}$$

$$p = \frac{\eta \mathbf{U}q}{\max(|\mathbf{U}q|)}$$

- The idea is finding a RIS vector to satisfy our condition
- By analyzing the null space of \mathbf{W} , we have a family of solutions in the last $N-K(K-1)$ columns of \mathbf{R}
- We can multiply this matrix \mathbf{U} by a random complex vector q to get the RIS configuration
- We are now able to communicate between two stationary antennas

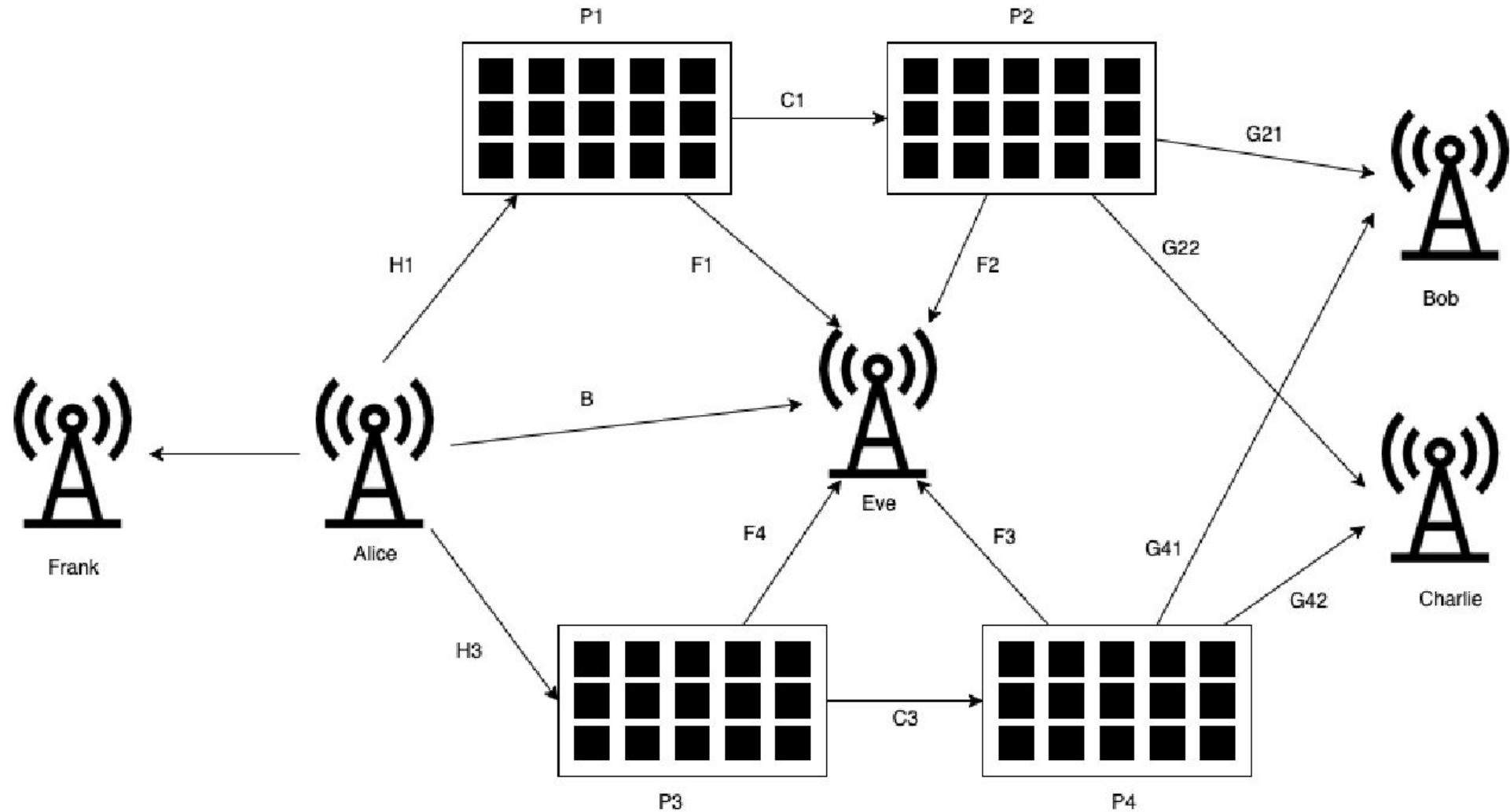
J. Luo, F. Wang, S. Wang, H. Wang and D. Wang,

"Reconfigurable Intelligent Surface: Reflection Design Against Passive Eavesdropping,"

Contribution

- Can we send the message to multiple receivers at the same time?
- Can we concatenate multiple RIS together (in series)?
- Can we send the message through multiple paths (in parallel)?
- What are the performance of all these cases?
- Is our framework valid for realistic communications?

Multi-user and multi-RIS scenario



Multi-user and multi-RIS scenario

$$\forall j \in \{1, 2, \dots, J\} \rightarrow \|G_j P H - [G_j P H]_{diag}\|^2 = 0$$

$$\forall j \in \{1, 2, \dots, J\} \rightarrow \mathbf{W}_j = \sum_{i,k=1, i \neq k}^K (g_{jk,:} \odot h_i^T)^H (g_{jk,:} \odot h_i^T)$$

$$\forall j \in \{1, 2, \dots, J\} \rightarrow \mathbf{W}_j p = 0$$

$$\begin{bmatrix} \mathbf{W}_1 \\ \mathbf{W}_2 \\ \dots \\ \mathbf{W}_J \end{bmatrix} p = 0$$

$$\begin{bmatrix} \mathbf{W}_1 \\ \mathbf{W}_2 \\ \dots \\ \mathbf{W}_J \end{bmatrix} = \mathbf{W} \in \mathbb{C}^{JN \times N}, \mathbf{W} = \mathbf{R} \mathbf{\Sigma} \mathbf{V}^H$$

- By having the condition of using multiple receivers, our matrix \mathbf{W} is not square
- We cannot use \mathbf{R} to find the null space of dimension N anymore
- We can however use the last $N-K(K-1)$ rows of $\mathbf{V}^H \mathbf{H}$

$$N(\mathbf{W}) = \begin{bmatrix} v_{N-K(K-1)}^H \\ \dots \\ v_N^H \end{bmatrix}^H = \mathbf{U}$$

$$p = \frac{\eta \mathbf{U} q}{\max(|\mathbf{U} q|)}$$

Multi-user and multi-RIS scenario

- *RIS in parallel*

$$\sum_{m=1}^M \mathbf{G}_j \mathbf{P}_m \mathbf{H}_m x = \left(\sum_{m=1}^M \mathbf{G}_j \mathbf{P}_m \mathbf{H}_m \right) x$$

- For RISs in parallel, the sum of multiple diagonal matrices is still a diagonal matrix

- *RIS in series*

$$\| \mathbf{G} \mathbf{P}_1 \mathbf{C}_1 \dots \mathbf{P}_M \mathbf{H} - [\mathbf{G} \mathbf{P}_1 \mathbf{C}_1 \dots \mathbf{P}_M \mathbf{H}]_{diag} \|^2 = 0$$

$$\forall m \in [1, M-1] : p_m[i] = \eta r_i e^{j\theta_i}$$

$$\mathbf{G}' = \mathbf{G} \mathbf{P}_1 \mathbf{C}_1 \dots \mathbf{P}_{M-1} \mathbf{C}_{M-1} \in \mathbb{C}^{K \times N}$$

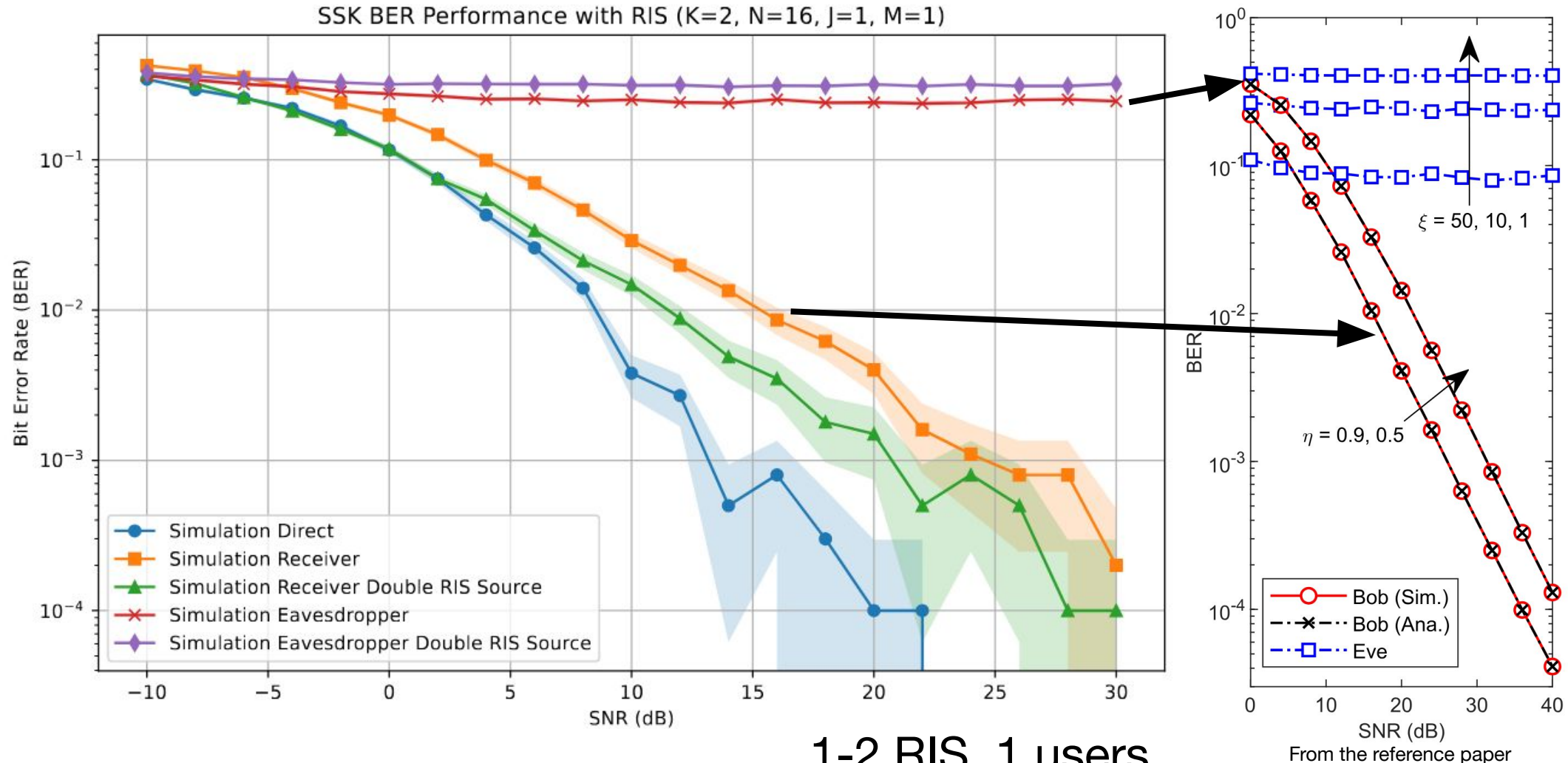
$$\| \mathbf{G}' \mathbf{P}_M \mathbf{H} - [\mathbf{G}' \mathbf{P}_M \mathbf{H}]_{diag} \|^2 = 0$$

- For RISs in series, we can setup the first (or last) M-1 RISs randomly, and setup only the last one
- Complex configuration can also be set up

Performance evaluation

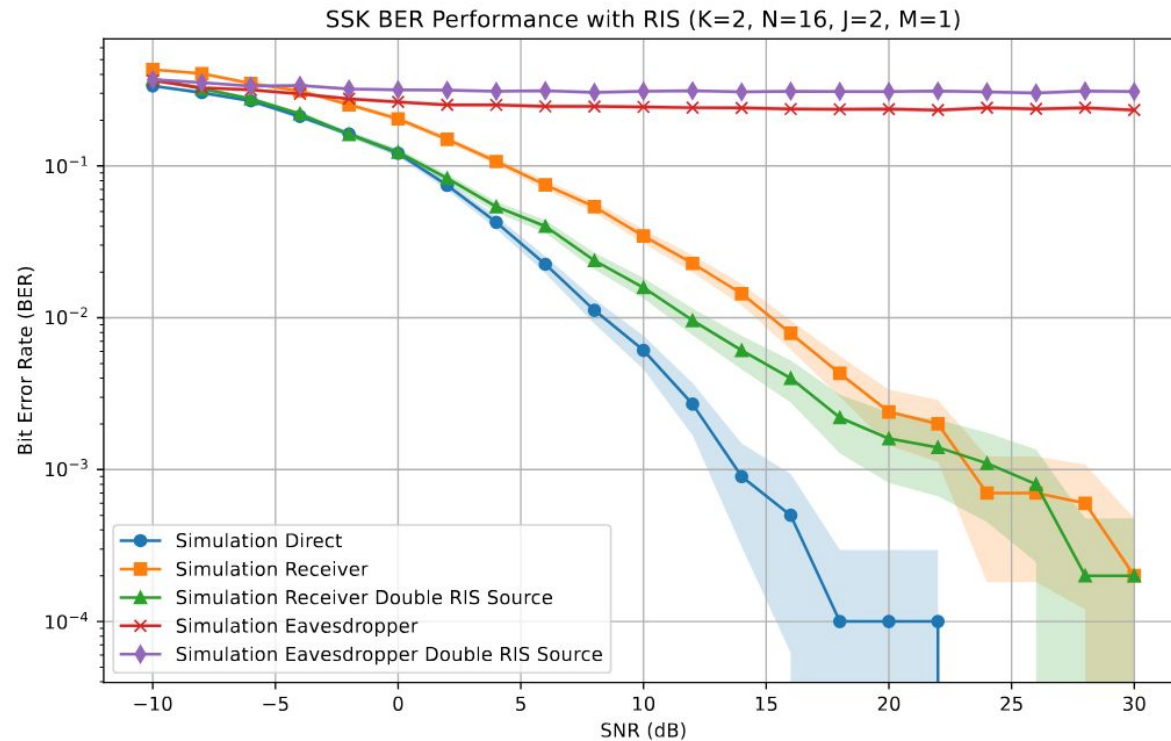
- We first make simulations correlating the Bit Error Rate (BER) to the Signal to Noise Ratio (SNR)
 - The BER indicates the percentage of wrong bits in a message
 - The SNR indicates how strong the noise is in relation to the correct signal
 - We will show that for higher SNR, the receiver reception gets clearer, while the eavesdropper error rate remains constant due to the RIS noise
-
- We will then realistically model the channel gains and the path loss considering the distance between two points
 - We will create an heatmap showing for each point the received BER
 - We will discuss how the type of the RIS influence the signal received by an eavesdropper

Bit Error Rate (BER) simulations

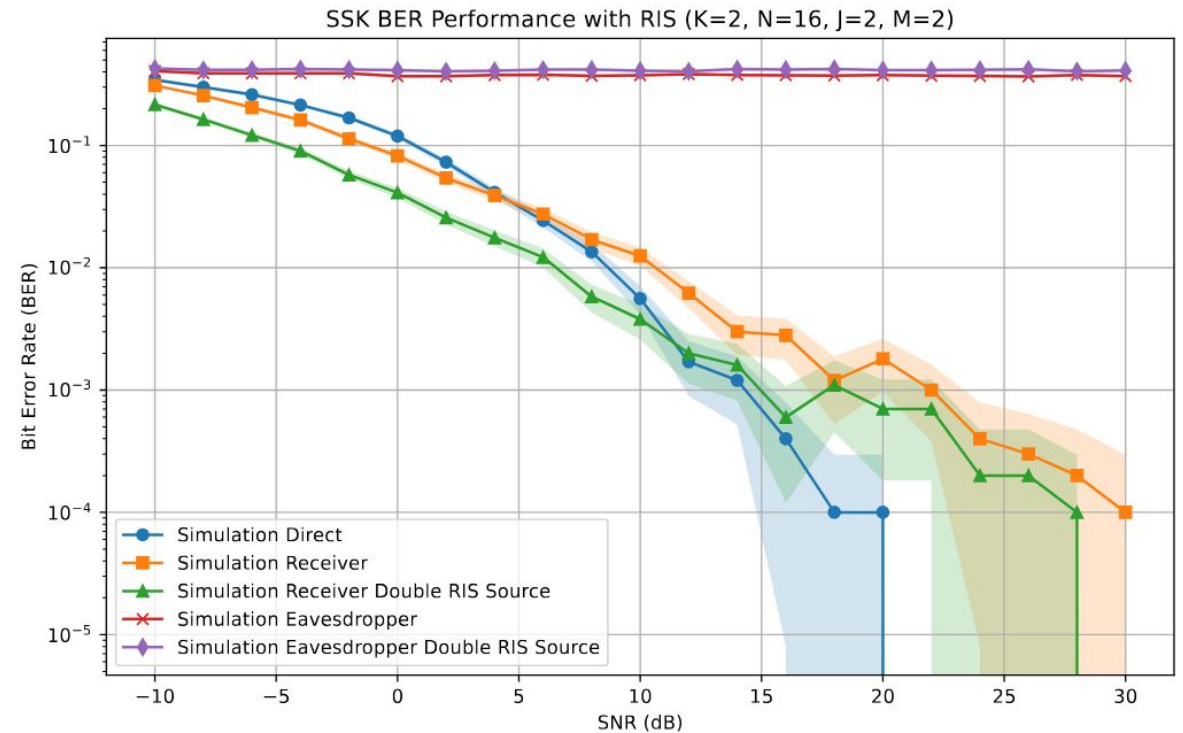


Bit Error Rate (BER) simulations

1-2 RIS, 2 users



2-4 RIS, 2 users



Realistic channel model

$$\nu^2 = \frac{\tau\xi}{1+\tau} \quad \sigma^2 = \frac{\xi}{2(1+\tau)} \quad \Xi \sim C\left(\frac{\nu}{\sqrt{2}}, \sigma\right)$$

$$e_r(\Omega) = \frac{1}{\sqrt{n_r}} \begin{bmatrix} 1 \\ \exp(-j2\pi\Delta\Omega) \\ \exp(-j2\pi2\Delta\Omega) \\ \vdots \\ \exp(-j2\pi(n_r-1)\Delta\Omega) \end{bmatrix}$$

$$\mathbf{H} = \Xi \odot \sqrt{n_t n_r} \exp(-j2\pi d/\lambda) e_r(\Omega_r) e_t(\Omega_t)^H$$

$$PL = ((4\pi/\lambda)^2 d^k)^{-\frac{1}{2}} \quad y = PL_B \cdot \mathbf{B}x$$

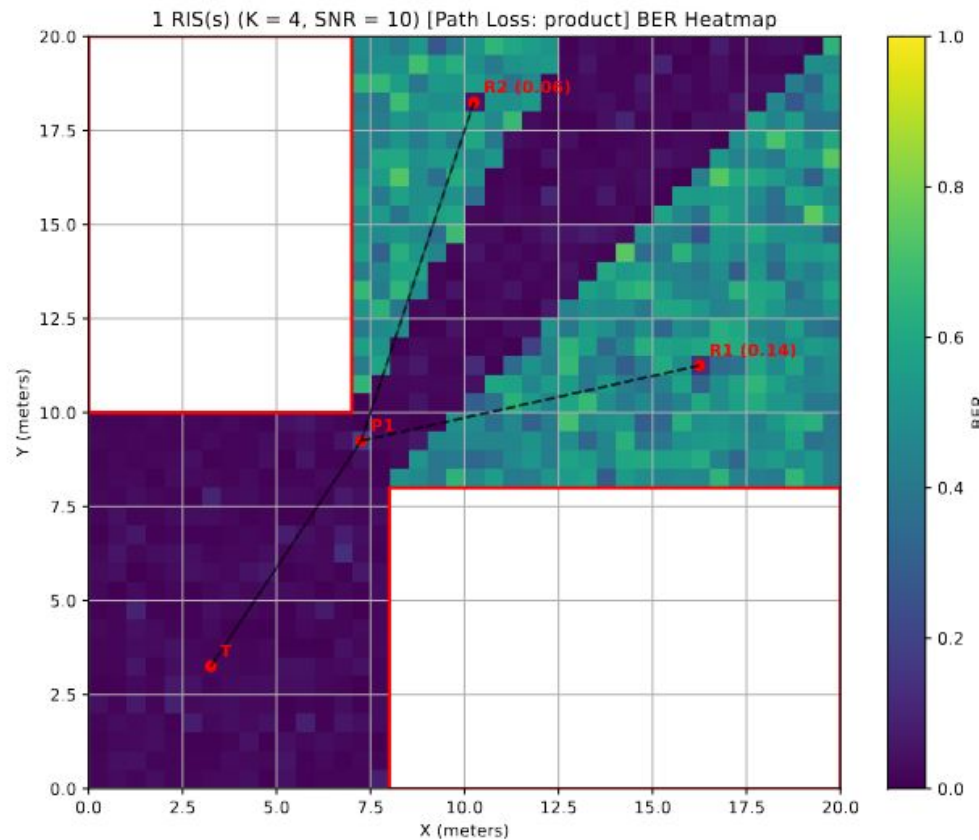
David Tse, Pramod Viswanath,
“Fundamentals of Wireless Communication”

- We generate the fading matrix from a complex distribution using the Shape and Scale parameters
- We calculate the channel gain matrix from the distance and the incidence angle between the antennas arrays
- We use the ideal Free Space path loss

BER Heatmaps - scenario 1

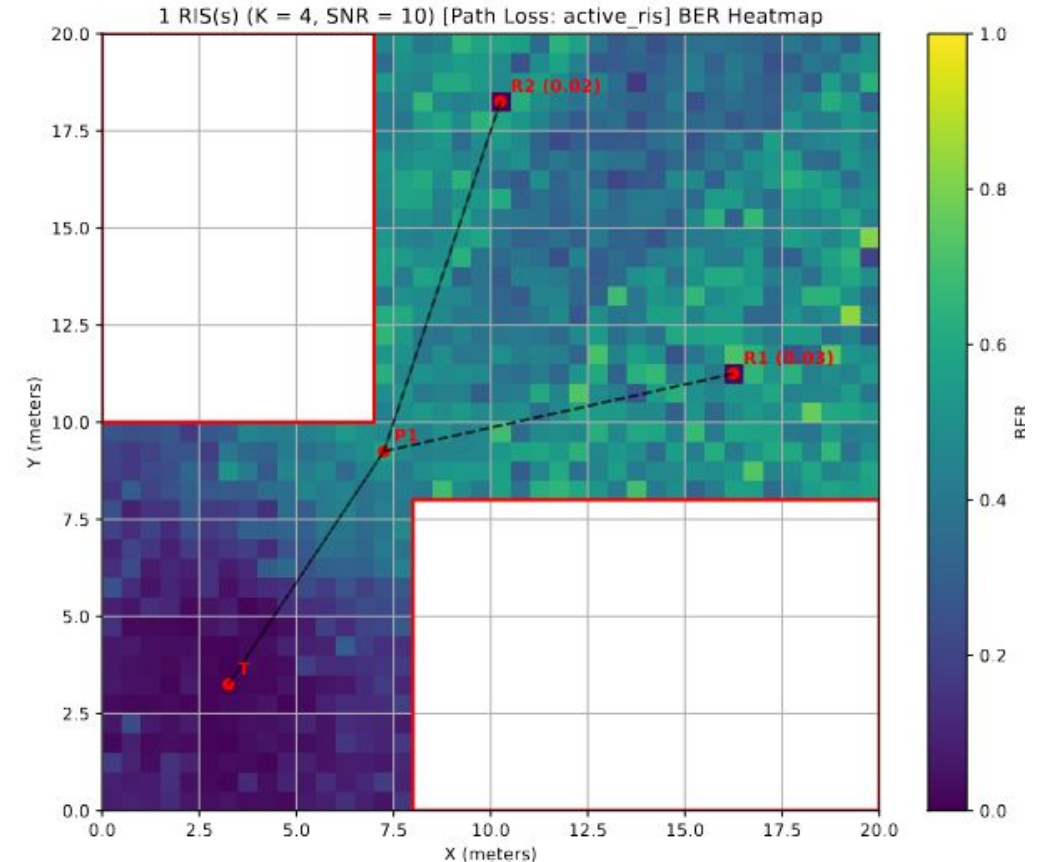
Passive RIS

$$y = PL_G \cdot PL_H \cdot GPHx$$



Active RIS

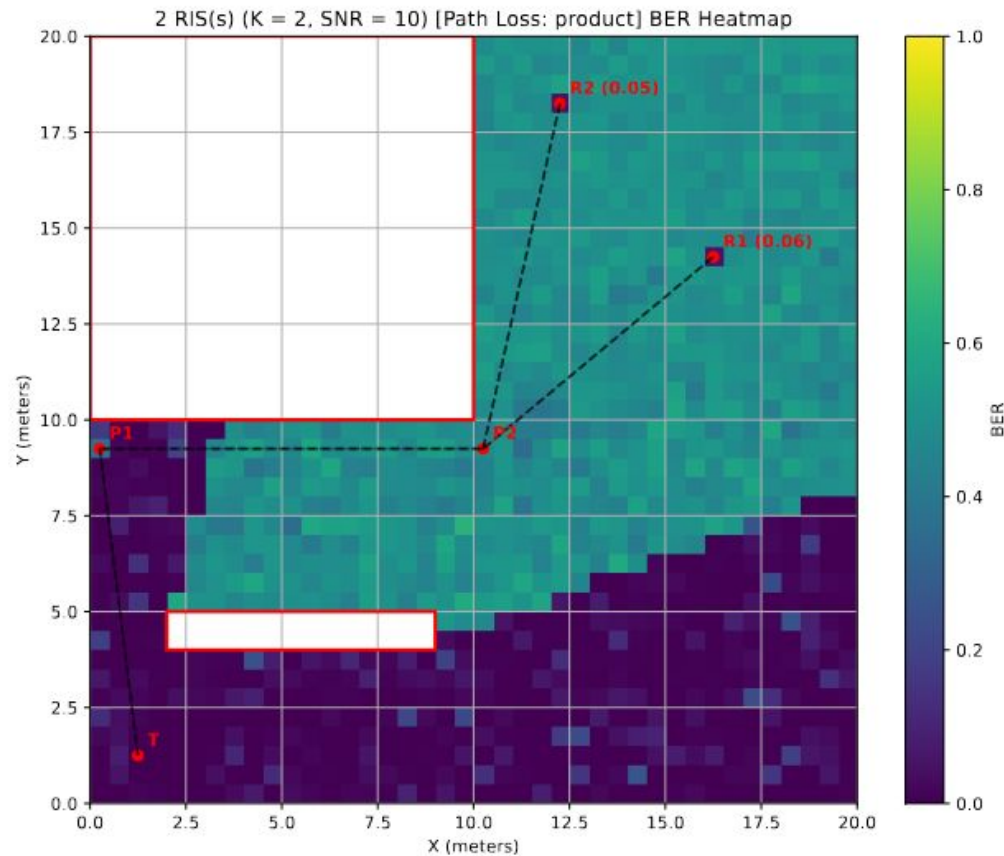
$$y = PL_H \cdot GPHx.$$



BER Heatmaps - scenario 2

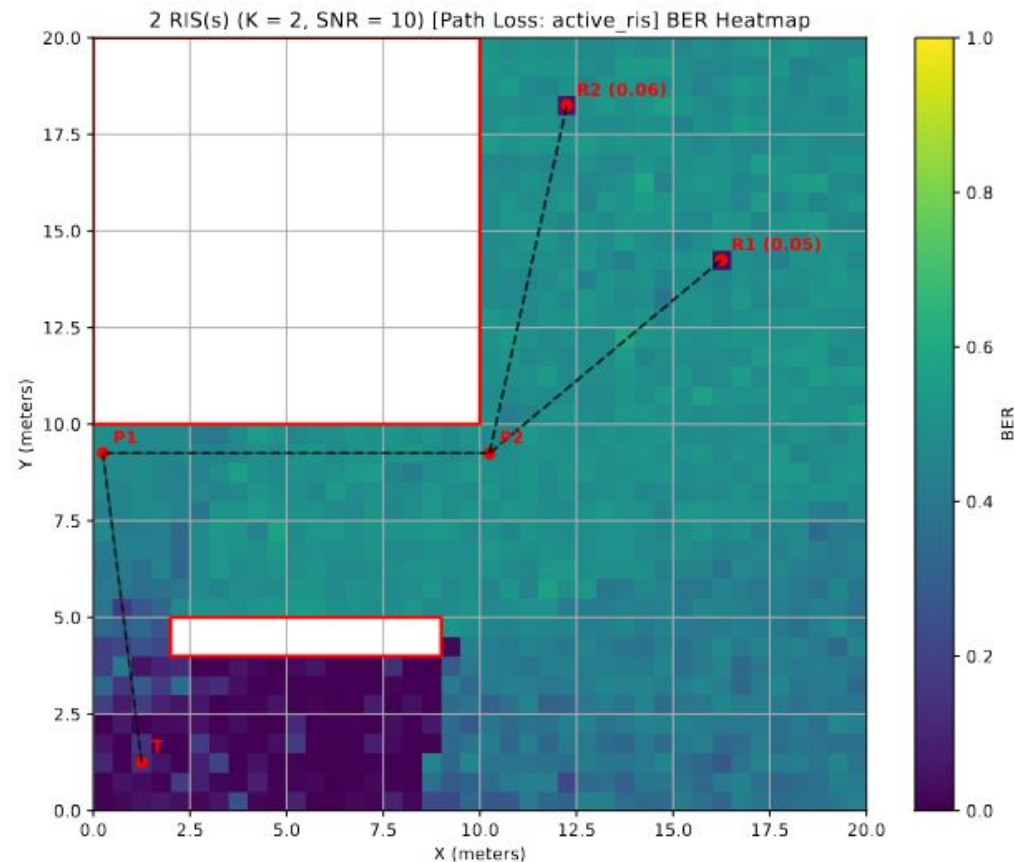
Passive RIS

$$y = PL_G \cdot PL_H \cdot \mathbf{GPH}x$$



Active RIS

$$y = PL_H \cdot \mathbf{GPH}x.$$



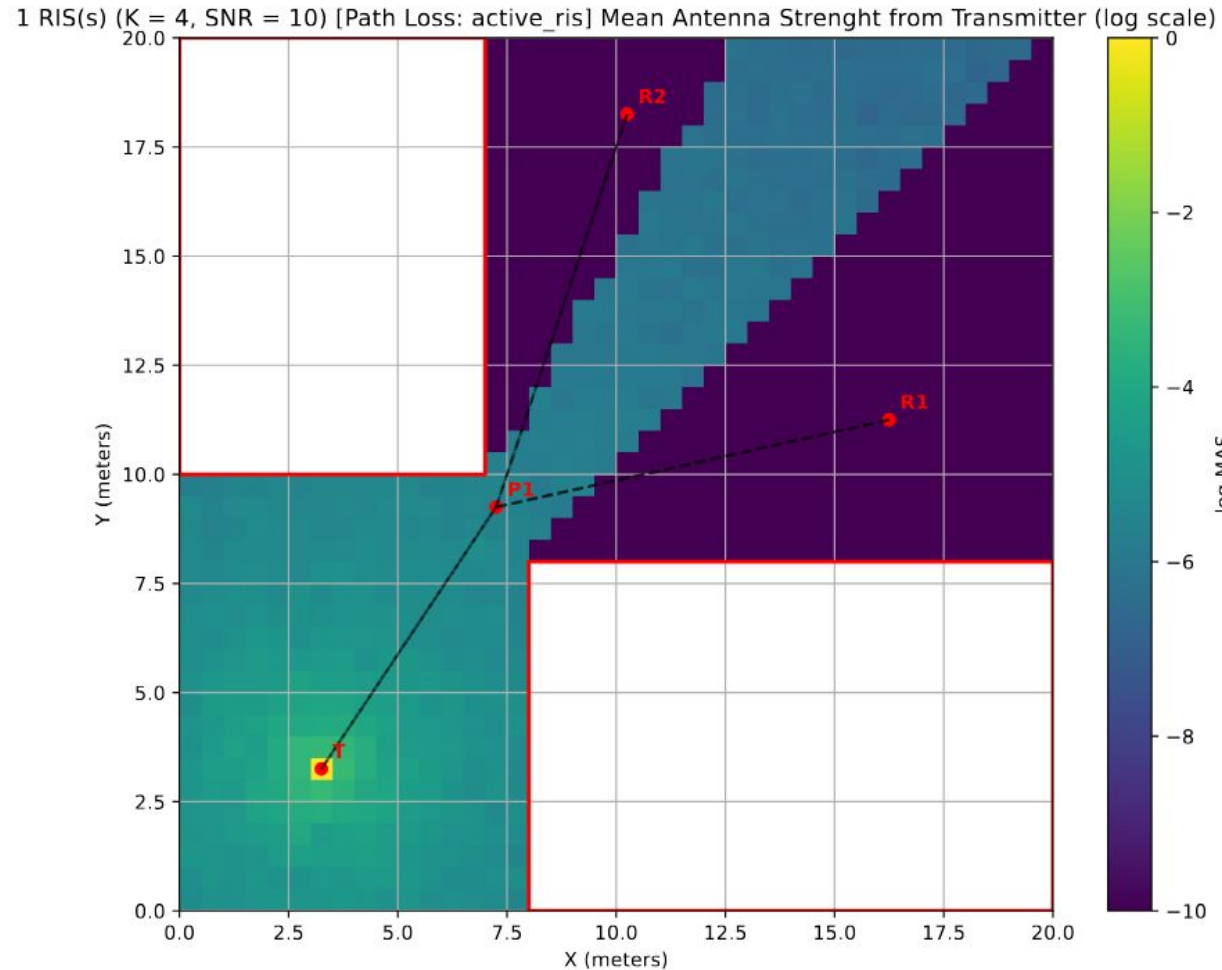
Conclusions and future directions

- We expanded on the work *RECONFIGURABLE INTELLIGENT SURFACE: REFLECTION DESIGN AGAINST PASSIVE EAVESDROPPING*
 - We proved the correctness of our contribution
 - We validated our solution in realistic scenarios
 - Still, the area when the message is received is limited in size
-
- Future work should be studied in channel gain estimation for moving vehicles
 - Low level language implementations should be made to calculate the latency of communication
 - Complex schemes that expand on SSK modulation should be considered

Thank you

Questions?

Mean Antenna Strength



$$MAS_H = \sum_{x=1}^X ||h_x||^2 / X$$

$$\log MAS_H = \log_{10} MAS_H$$

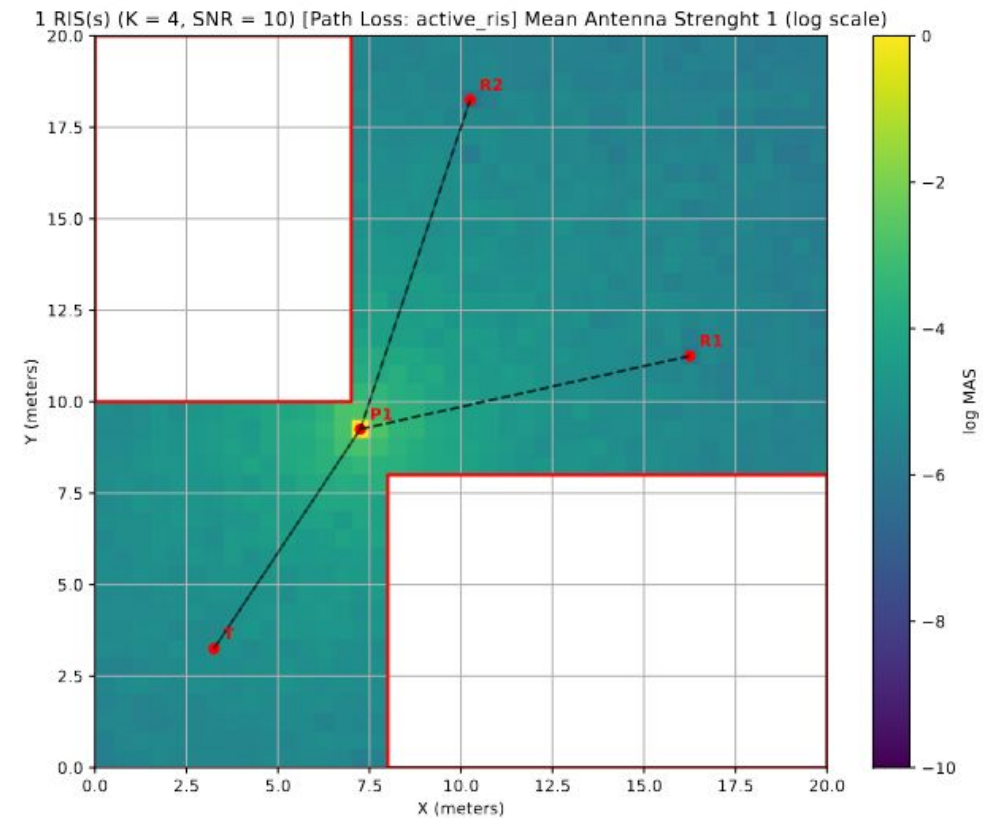
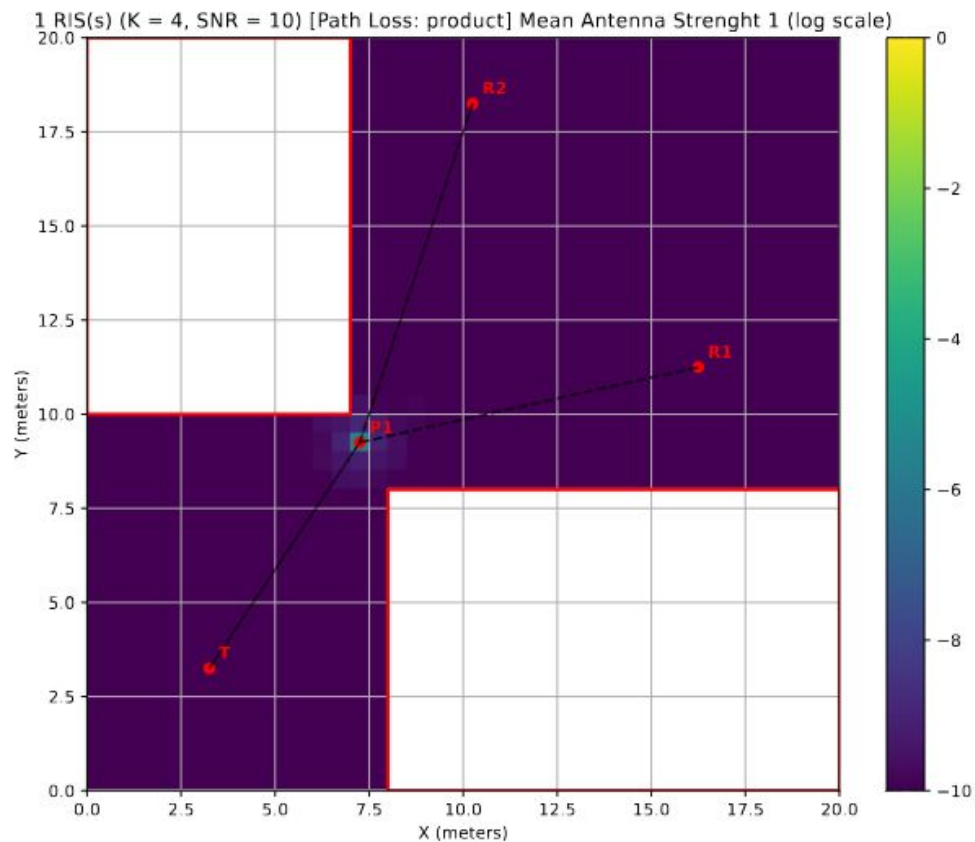
BER Heatmaps - scenario 1

Passive RIS

$$y = PL_G \cdot PL_H \cdot GPHx$$

Active RIS

$$y = PL_H \cdot GPHx.$$



Channel Gain Estimation

- The transmitter communicates to the RIS controller a setup message x' that it will send to the receiver
- The RIS will set a random P'
- The receiver gets a signal y' (which will mean nothing), and sends it back to the RIS controller
- Based on x' , y' , P' the RIS controller estimates G , H and correctly sets up P
- The transmitter sends x , and the receiver gets y which it can correctly convert back
- If transmitter and receiver are moving, the procedure will start all over. Otherwise, G and H remain the same, and the RIS controller can just create a new P for the next messages

<https://ieeexplore.ieee.org/document/8879620>

Algorithm 1: JBF-MC algorithm

Input: Y, S, X , prior distributions $p(G)$ and $p(Z)$

% sparse matrix factorization via BiG-AMP

```

1: Initialization:  $\forall l, n, t$ : generate  $g_{l,n}$  from  $p(g_{l,n})$ ,  $v_{l,n}^g(1) = \nu_g$ ,
    $\hat{z}_{n,t}(1) = \mathbb{E}(z_{n,t})$ ,  $v_{n,t}^z(1) = \lambda \nu_z$ , and  $\hat{u}_{l,t}(1) = 0$ 
2: for  $i = 1, \dots, I_{\max}$  % outer iteration
3:   for  $j = 1, \dots, J_{\max}$  % inner iteration
4:      $\forall l, t$ :  $\bar{v}_{l,t}^p(i) = \sum_{n=1}^N |\hat{g}_{l,n}(i)|^2 v_{n,t}^z(i) + v_{l,n}^g(i) |\hat{z}_{n,t}(i)|^2$ 
5:      $\forall l, t$ :  $\bar{p}_{l,t}(i) = \sum_{n=1}^N \hat{g}_{l,n}(i) \hat{z}_{n,t}(i)$ 
6:      $\forall l, t$ :  $v_{l,t}^p(i) = \bar{v}_{l,t}^p(i) + \sum_{n=1}^N v_{l,n}^g(i) v_{n,t}^z(i)$ 
7:      $\forall l, t$ :  $\hat{p}_{l,t}(i) = \bar{p}_{l,t}(i) - \hat{u}_{l,t}(i - 1) \bar{v}_{l,t}^p(i)$ 
8:      $\forall l, t$ :  $v_{l,t}^p(i) = \sigma^2 v_{l,t}^p(i) / [v_{l,t}^p(i) + \sigma^2]$ 
9:      $\forall l, t$ :  $\hat{b}_{l,t}(i) = v_{l,t}^p(i) [y_{l,t} - \hat{p}_{l,t}(i)] / [v_{l,t}^p(i) + \sigma^2] + \hat{p}_{l,t}(i)$ 
10:     $\forall l, t$ :  $v_{l,t}^u(i) = [1 - v_{l,t}^z(i) / v_{l,t}^p(i)] / v_{l,t}^p(i)$ 
11:     $\forall l, t$ :  $\hat{u}_{l,t}(i) = [\hat{b}_{l,t}(i) - \hat{p}_{l,t}(i)] / v_{l,t}^p(i)$ 
12:     $\forall l, n$ :  $v_{l,n}^q(i) = [\sum_{t=1}^T |\hat{z}_{n,t}(i)|^2 v_{l,t}^u(i)]^{-1}$ 
13:     $\forall l, n$ :  $\hat{q}_{l,n}(i) = \hat{g}_{l,n}(i) [1 - v_{l,n}^q(i) \sum_{t=1}^T v_{n,t}^z(i) v_{l,t}^u(i)]$ 
    $+ v_{l,n}^q(i) \sum_{t=1}^T \hat{z}_{n,t}^*(i) \hat{u}_{l,t}(i)$ 
14:     $\forall n, t$ :  $v_{n,t}^r(i) = [\sum_{l=1}^L |\hat{g}_{l,n}(i)|^2 v_{l,t}^u(i)]^{-1}$ 
15:     $\forall n, t$ :  $\hat{r}_{n,t}(i) = \hat{z}_{n,t}(i) (1 - v_{n,t}^r(i) \sum_{l=1}^L v_{l,n}^q(i) v_{l,t}^u(i))$ 
    $+ v_{n,t}^r(i) \sum_{l=1}^L \hat{g}_{l,n}^*(i) \hat{u}_{l,t}(i)$ 
16:     $\forall l, n$ :  $\hat{g}_{l,n}(i+1) = \mathbb{E}\{g_{l,n} | \hat{q}_{l,n}(i), v_{l,n}^q(i)\}$ 
17:     $\forall l, n$ :  $v_{l,n}^g(i+1) = \text{Var}\{g_{l,n} | \hat{q}_{l,n}(i), v_{l,n}^q(i)\}$ 
18:     $\forall n, t$ :  $\hat{z}_{n,t}(i+1) = \mathbb{E}\{z_{n,t} | \hat{r}_{n,t}(i), v_{n,t}^r(i)\}$ 
19:     $\forall n, t$ :  $v_{n,t}^z(i+1) = \text{Var}\{z_{n,t} | \hat{r}_{n,t}(i), v_{n,t}^r(i)\}$ 
20:    if a certain stopping criterion is met, stop
21:  end for
22:   $\forall l, n, t$ :  $\hat{g}_{l,n}(i) = \hat{g}_{l,n}(i+1)$ ,  $\nu_{m,n}^g(i) = \nu_{l,n}^g(i+1)$ ,
    $\hat{z}_{n,t}(i) = \hat{z}_{n,t}(1)$ ,  $\nu_{n,t}^z(i) = v_{n,t}^z(1)$ 
23: end for
24:  $\hat{G} \leftarrow \hat{G}(i+1)$ ,  $\hat{Z} \leftarrow \hat{Z}(i+1)$ 
% matrix completion via RGrad
25: Initialization:  $A(0) = 0$ 
26: for  $k = 1, \dots, K_{\max}$ 
27:    $Q(k) = S^* \odot (\hat{Z} - A(k))$ 
28:    $\alpha(k) = \frac{\|\mathcal{P}_{S(k)}(Q(k))\|_F^2}{\|S^* \odot (\mathcal{P}_{S(k)}(Q(k)))\|_F^2}$ 
29:    $W(k) = A(k) + \alpha_k \mathcal{P}_{S(k)}(Q(k))$ 
30:    $A(k+1) = \mathcal{H}_r(W(k))$ 
31:   if a certain stopping criterion is met, stop
32: end for
33:  $\hat{H} \leftarrow \hat{A} X^\dagger$  with  $\hat{A} = A(k+1)$ 
Output:  $\hat{G}$  and  $\hat{H}$ 

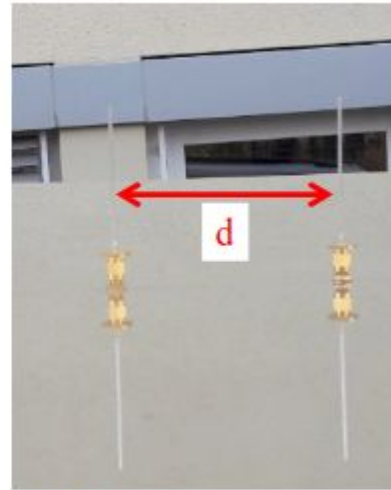
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Channel Gain Estimation for Vehicles

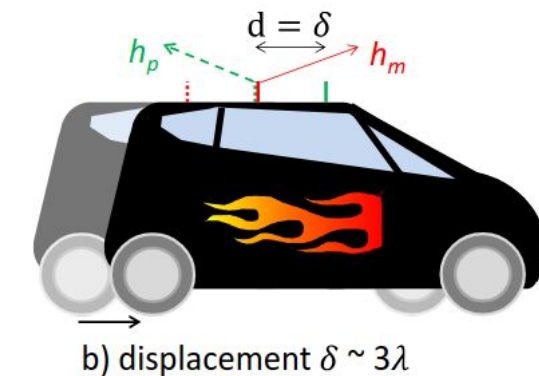
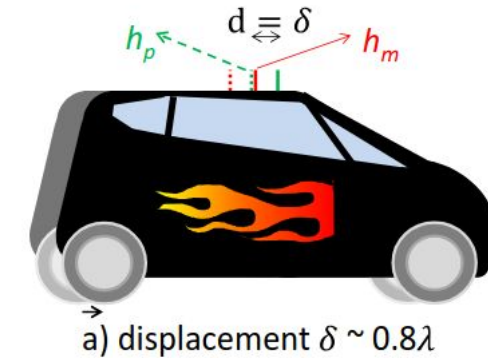


a) Car with mounted metallic plane



b) Monopole antennas on the metallic plane

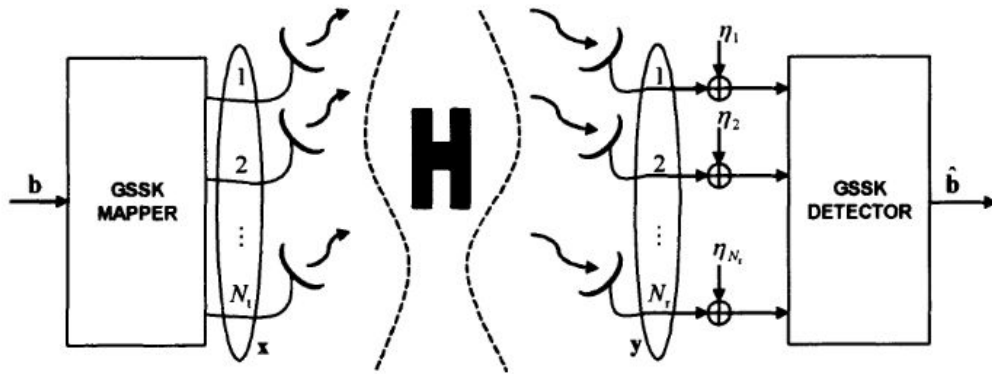
- | Current position of Predictor antenna
- | Position of Predictor antenna during prediction
- | Current position of antenna
- | Position of Predictor antenna during prediction



<https://ieeexplore.ieee.org/document/8385489>

Generalized Space Shift Keying Modulation

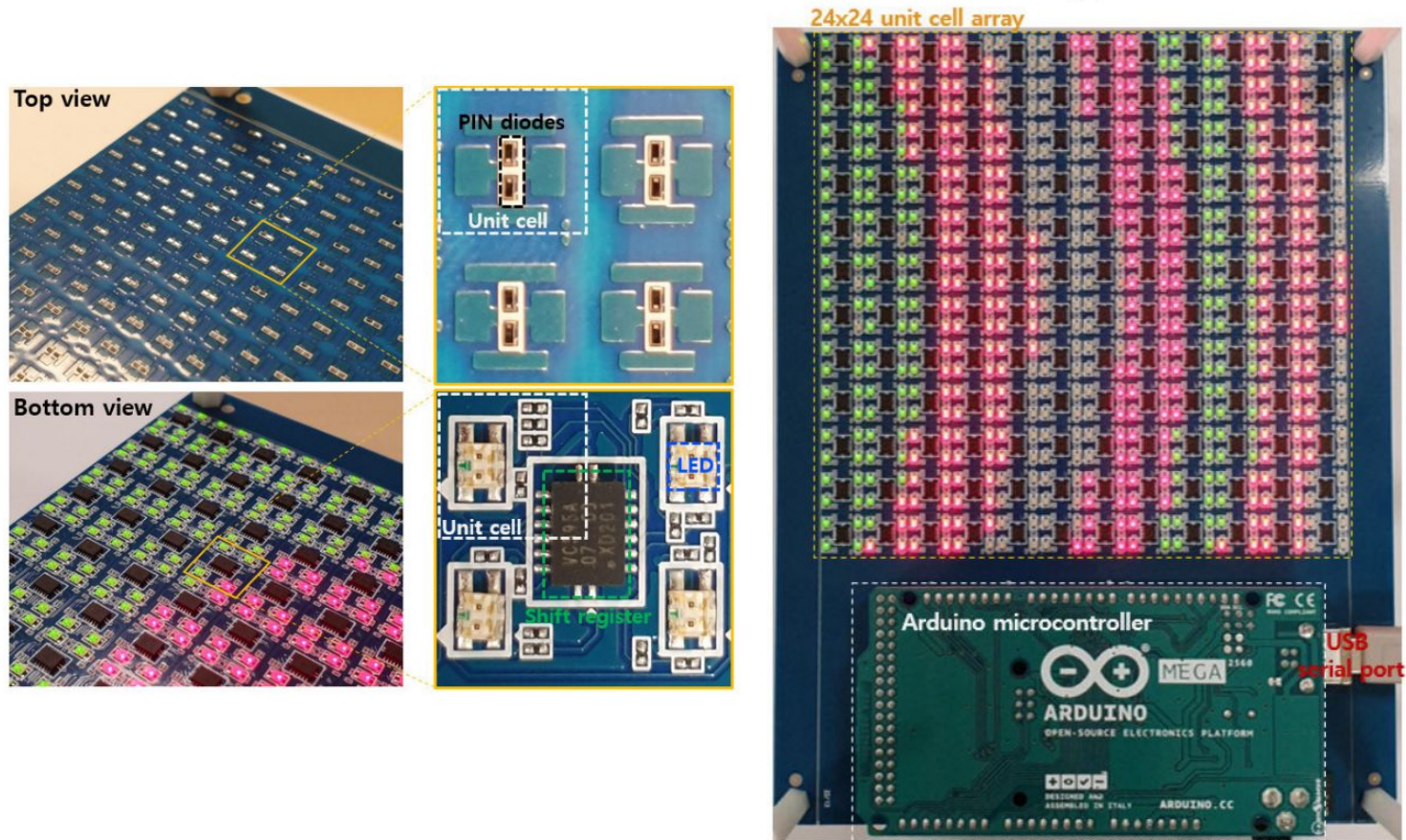
TABLE I
EXAMPLE OF THE GSSK MAPPER RULE.



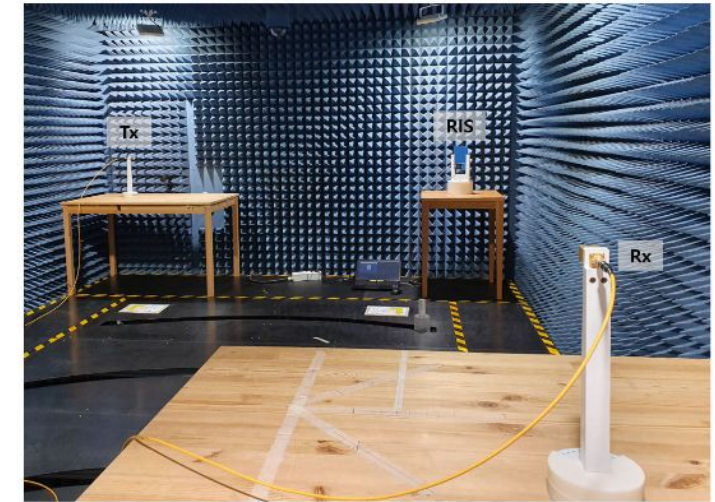
$\mathbf{b} = [b_1 \ b_2 \ b_3]$	\mathbf{j}	$\mathbf{x} = [x_1 \ x_2 \ \dots \ x_5]^T$
$[0 \ 0 \ 0]$	$(1, 2)$	$\begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 & 0 & 0 \end{bmatrix}^T$
$[0 \ 0 \ 1]$	$(1, 3)$	$\begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} & 0 & 0 \end{bmatrix}^T$
$[0 \ 1 \ 0]$	$(1, 4)$	$\begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & 0 & \frac{1}{\sqrt{2}} & 0 \end{bmatrix}^T$
$[0 \ 1 \ 1]$	$(1, 5)$	$\begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & 0 & 0 & \frac{1}{\sqrt{2}} \end{bmatrix}^T$
$[1 \ 0 \ 0]$	$(2, 3)$	$\begin{bmatrix} 0 & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 & 0 \end{bmatrix}^T$
$[1 \ 0 \ 1]$	$(2, 4)$	$\begin{bmatrix} 0 & \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} & 0 \end{bmatrix}^T$
$[1 \ 1 \ 0]$	$(2, 5)$	$\begin{bmatrix} 0 & \frac{1}{\sqrt{2}} & 0 & 0 & \frac{1}{\sqrt{2}} \end{bmatrix}^T$
$[1 \ 1 \ 1]$	$(3, 4)$	$\begin{bmatrix} 0 & 0 & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \end{bmatrix}^T$

J. Jeganathan, A. Ghayeb, L. Szczecinski
"Generalized Space shift keying modulation for MIMO channels,"

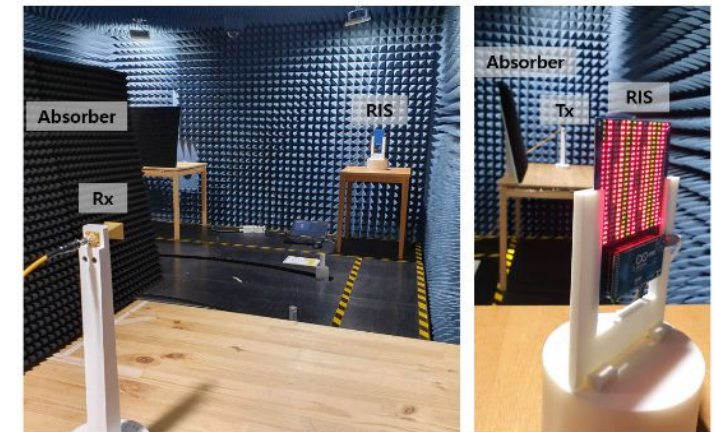
RIS implementations



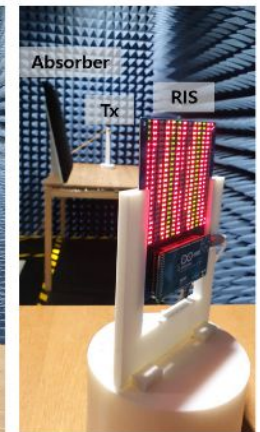
<https://ieeexplore.ieee.org/document/9881509>



(a)



(b)



(c)