

Distributed Beamforming in Adversarial Environments

Yagiz Savas and Ufuk Topcu

in collaboration with Abolfazl Hashemi, Abraham P. Vinod, Brian M. Sadler

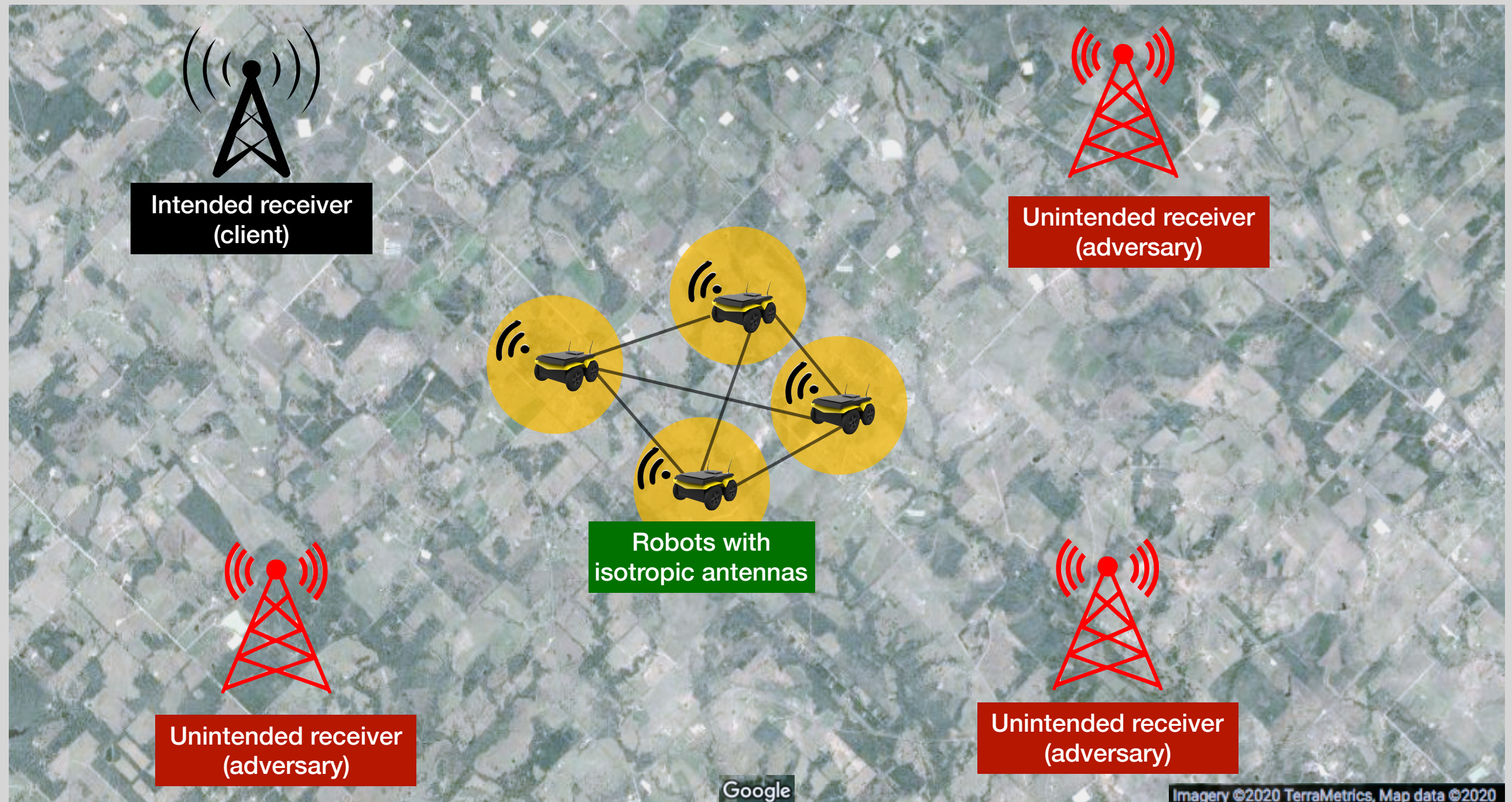
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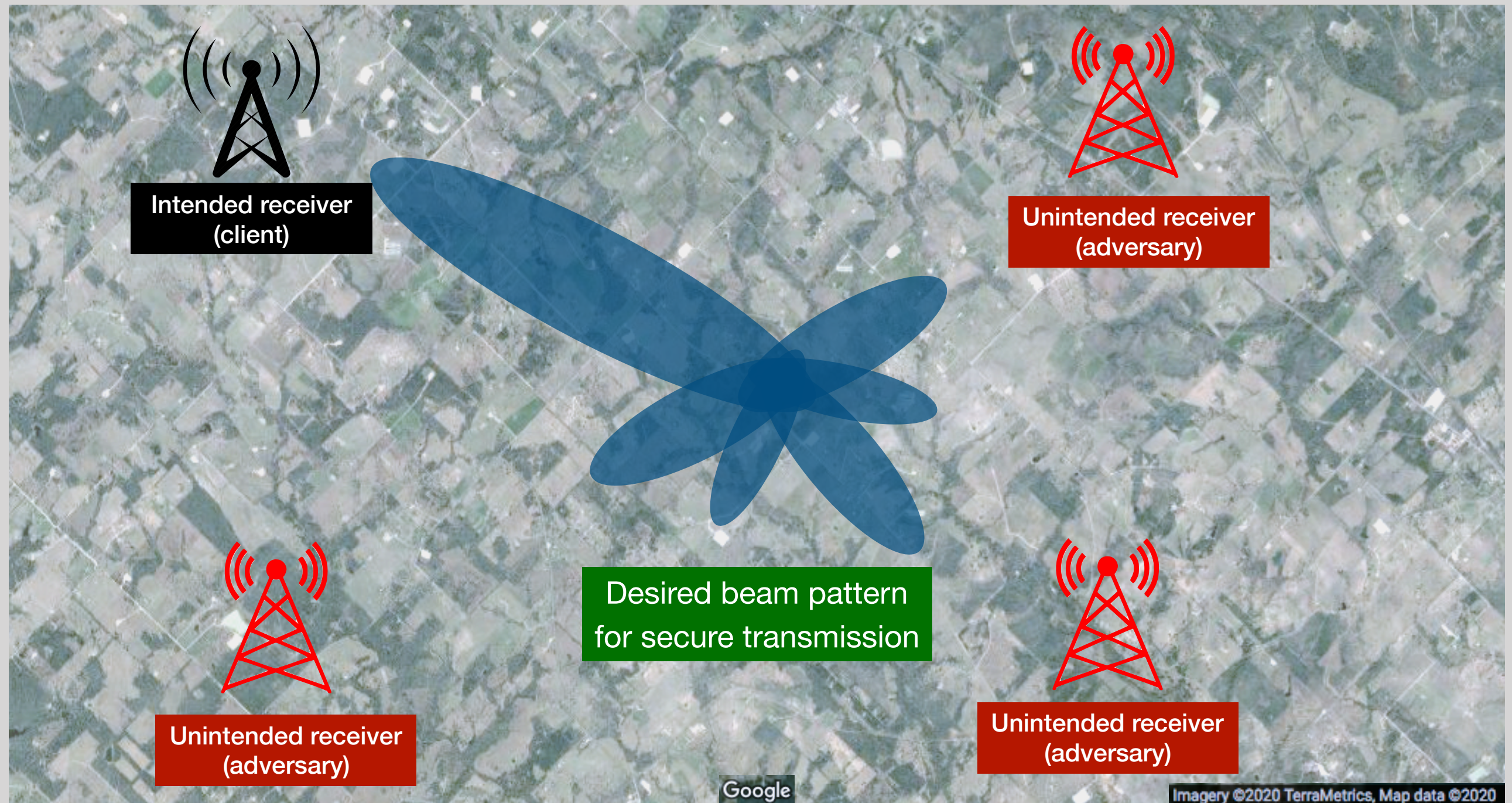
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Wireless Communications in the Presence of Adversaries



Wireless Communications in the Presence of Adversaries



Objective and Structure

Develop a **time-varying** transmission strategy
that enables **secure** communication

**Distributed
beamforming**

**Physical-layer
security**

**Semi-definite
programming**

Beamforming as a Wireless Communication Technique: **Main Idea**

Message signal: $s(t) = a e^{j\phi}$

amplitude \swarrow a \nwarrow phase ϕ

Adjust phase and amplitude: $w_i s(t) = \bar{a} e^{j\bar{\phi}}$

Transmit collectively: $y(t) = \sum_i w_i s(t)$

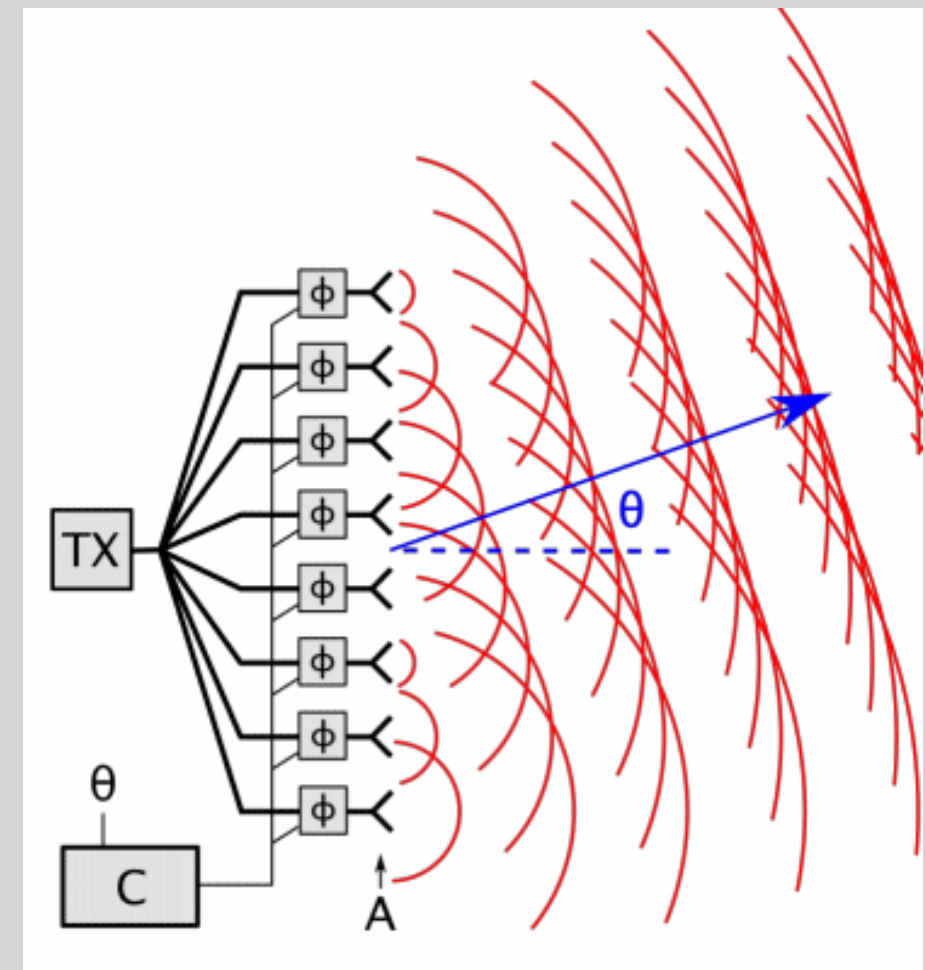
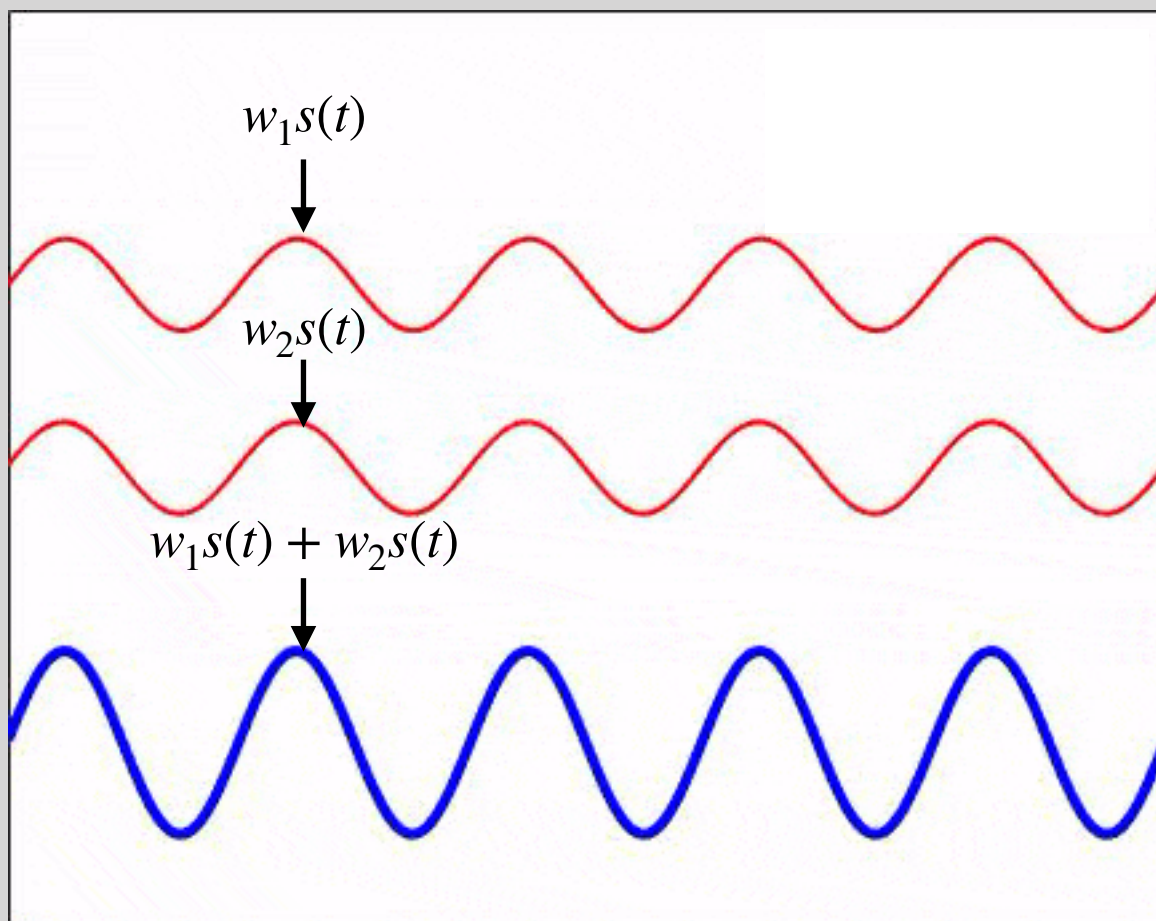
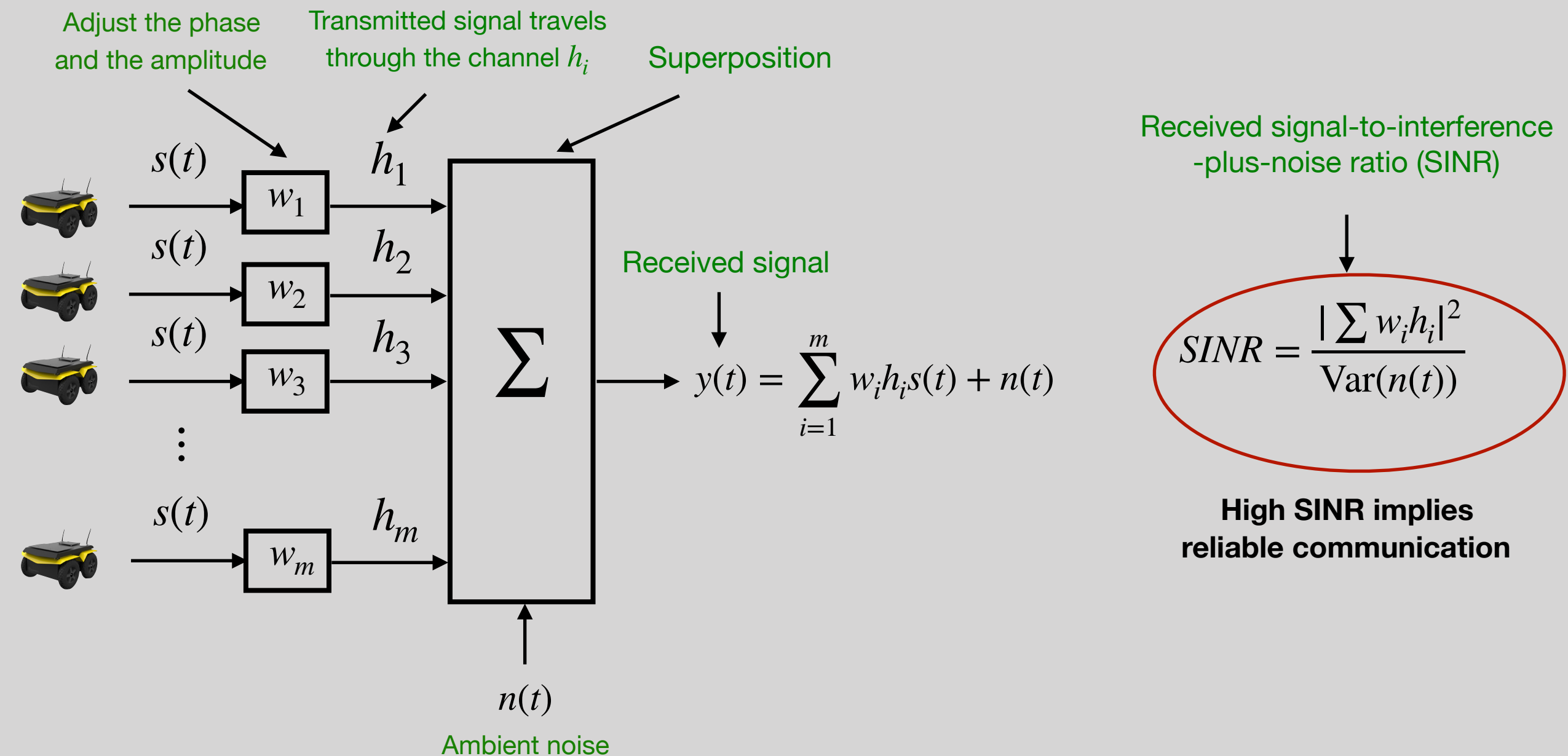
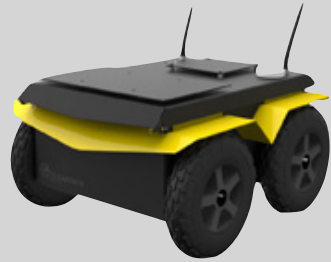


Illustration: https://en.wikipedia.org/wiki/Phased_array

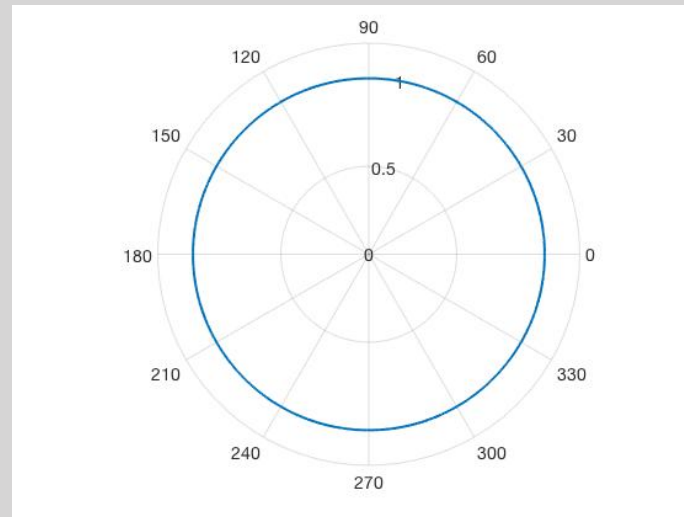
Beamforming as a Wireless Communication Technique: **Main Idea**



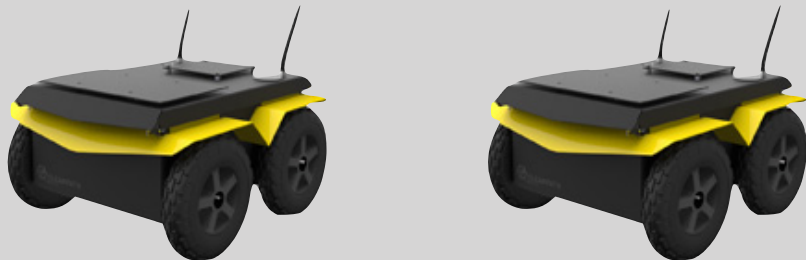
Beamforming as a Wireless Communication Technique: **Benefits**



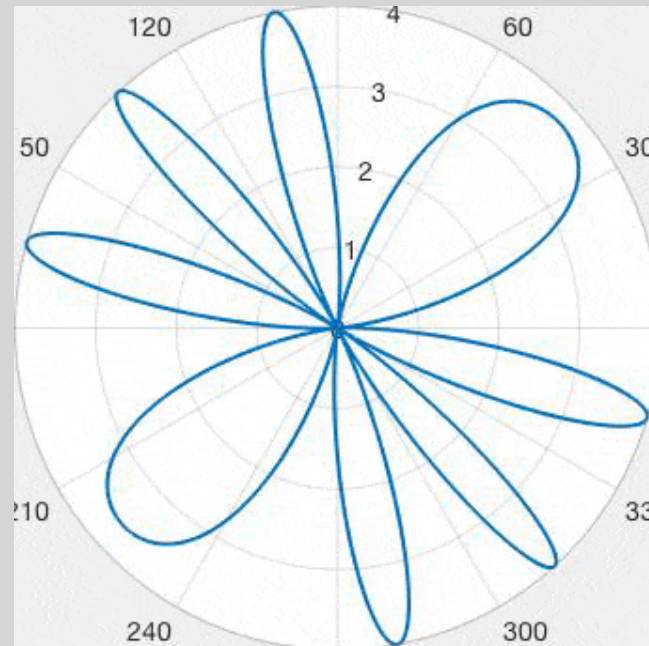
Single robot equipped with an isotropic antenna



- **No directionality**
- **Low SINR**

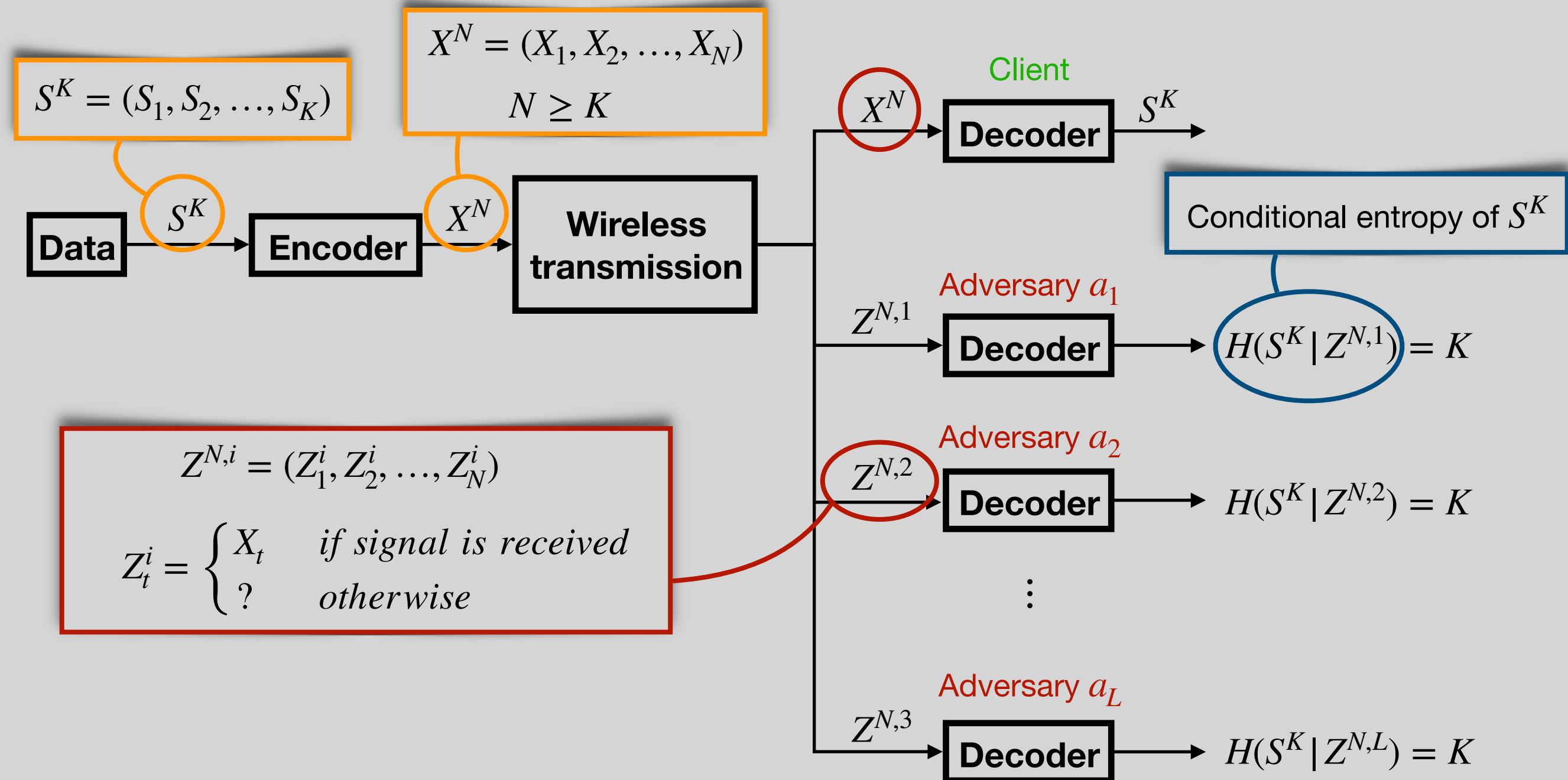


Two robots each equipped with an isotropic antenna



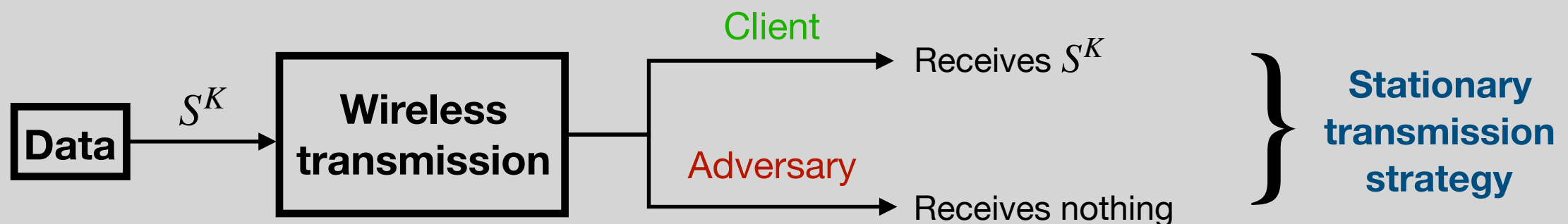
- **Improved directionality**
- **Improved SINR**

Secure Communication Problem: An Informal Problem Statement



Related Work

- **No adversaries:** optimal beamformer can be found analytically ^[1]
- **Adversaries with known locations:** convex optimization-based beamformers ^[2]
- **Adversaries with unknown locations:** minimize SINR in all directions by broadcasting artificial noise ^[3]



- Ozarow and Wyner ^[4] showed in 1984 that if S^K is encoded into X^N , then

μ_i : number of symbols
received by adversary a_i

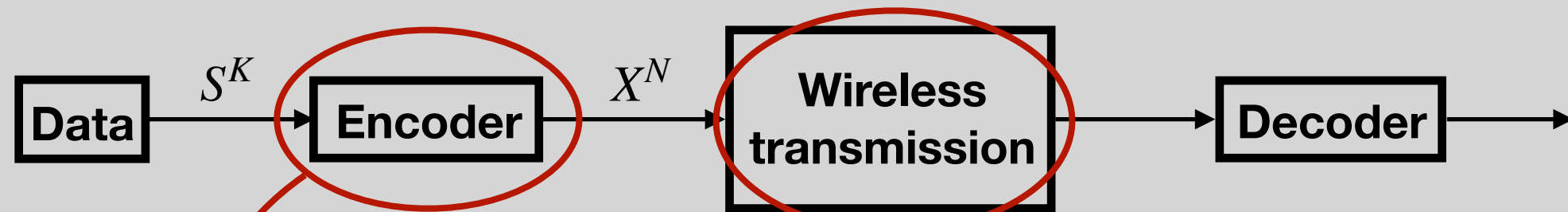
$$\mu_i \leq N - K \implies H(S^K | Z^{N,i}) = K$$

Implication: We can let each adversary receive $N - K$ symbols and still establish a secure communication

[1] Lorenz, R. G. and Boyd, S. P., "Robust minimum variance beamforming", IEEE Transactions on Signal Processing, 2005
 [2] Liao et al, "QoS-Based Transmit Beamforming in the Presence of Eavesdroppers", IEEE Transactions on Signal Processing, 2010
 [3] Goel, S. And Negi, R., "Guaranteeing Secrecy Using Artificial Noise", IEEE Transactions on Wireless Communications, 2008
 [4] Ozarow, L. H. and Wyner, A. D., "Wire-Tap Channel II", AT&T Bell Laboratories technical journal, 1984

Contributions

We approach the problem from a **sequential decision-making** perspective

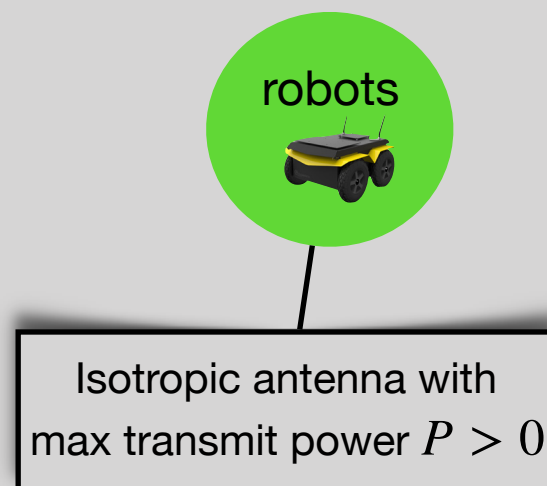


Maps K bits into $N = KL$ symbols where $L \in \mathbb{N}$ is the number of adversaries

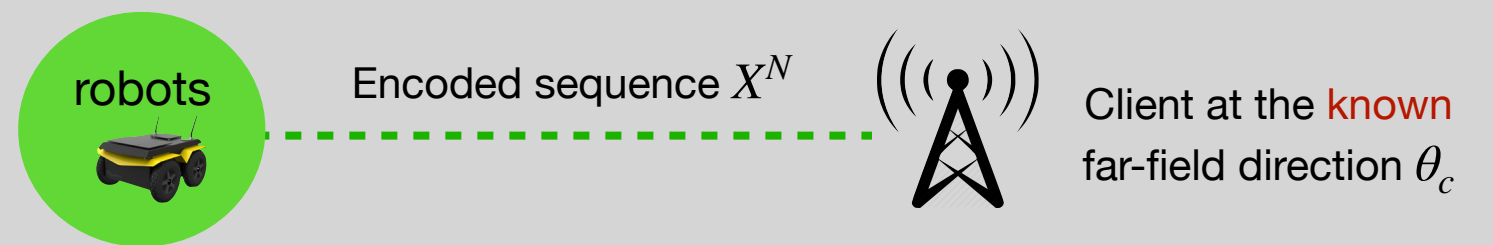
Periodic transmission strategy ensures that each adversary receives **at most $K(L - 1)$** symbols

The proposed periodic strategy enables the agents to securely communicate with the client in scenarios in which all stationary strategies fail to ensure security

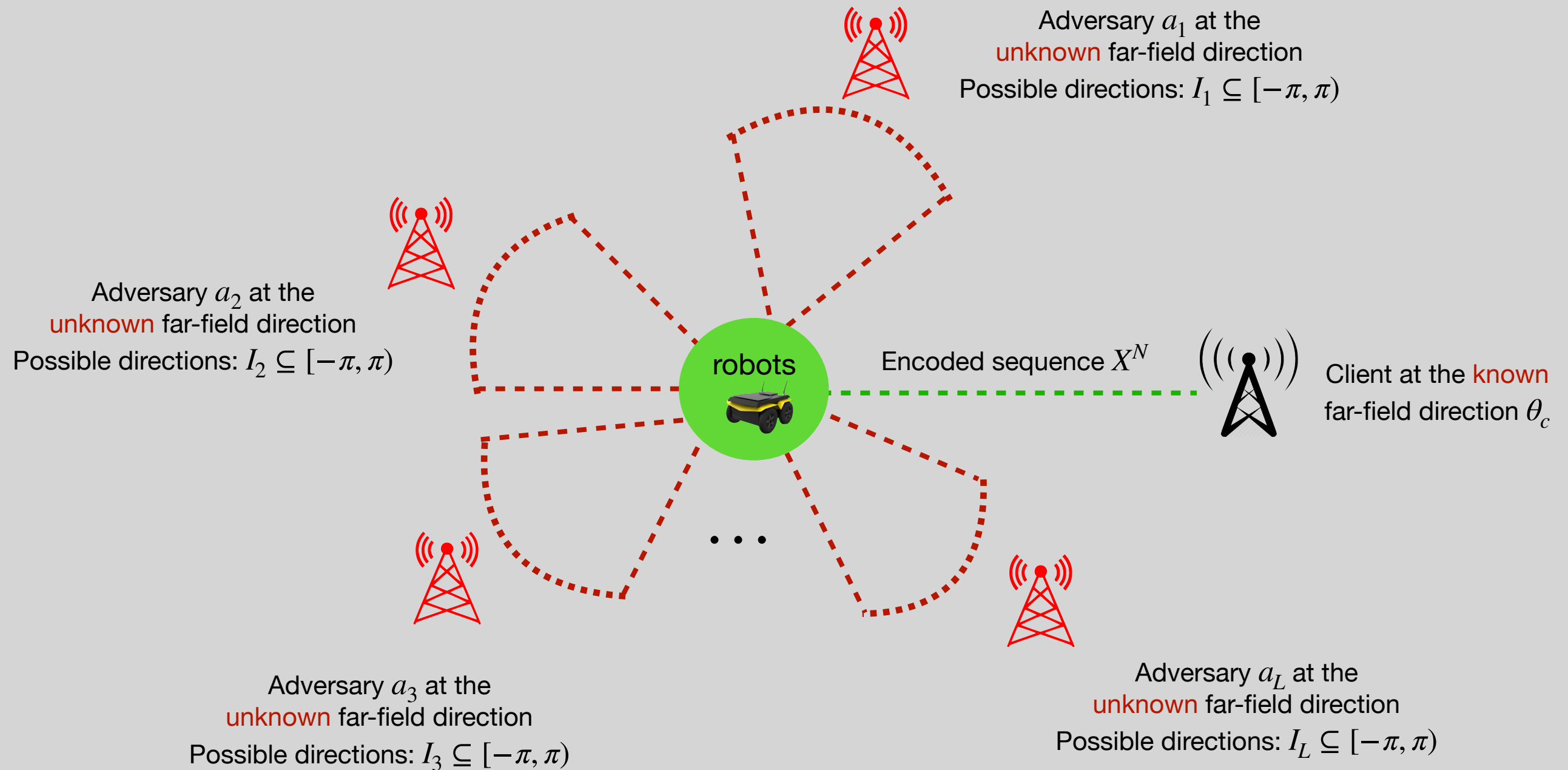
Environment Model



Environment Model



Environment Model



Transmission model

At time $t \in [N]$, the agents transmit the encoded symbol X_t as a continuous signal s_t .

The vector of signals transmitted by the agents is

$$y_{\text{transmit}}[t] = \mathbf{w}_t s_t + \mathbf{v}_t$$

Beamforming vector $\mathbf{w}_t = [w_1, w_2, \dots, w_m]'$

Artificial noise $\mathbf{v}_t \sim \mathcal{CN}(0, \Sigma_t)$

Transmission model

At time $t \in [N]$, the agents transmit the encoded symbol X_t as a continuous signal s_t .

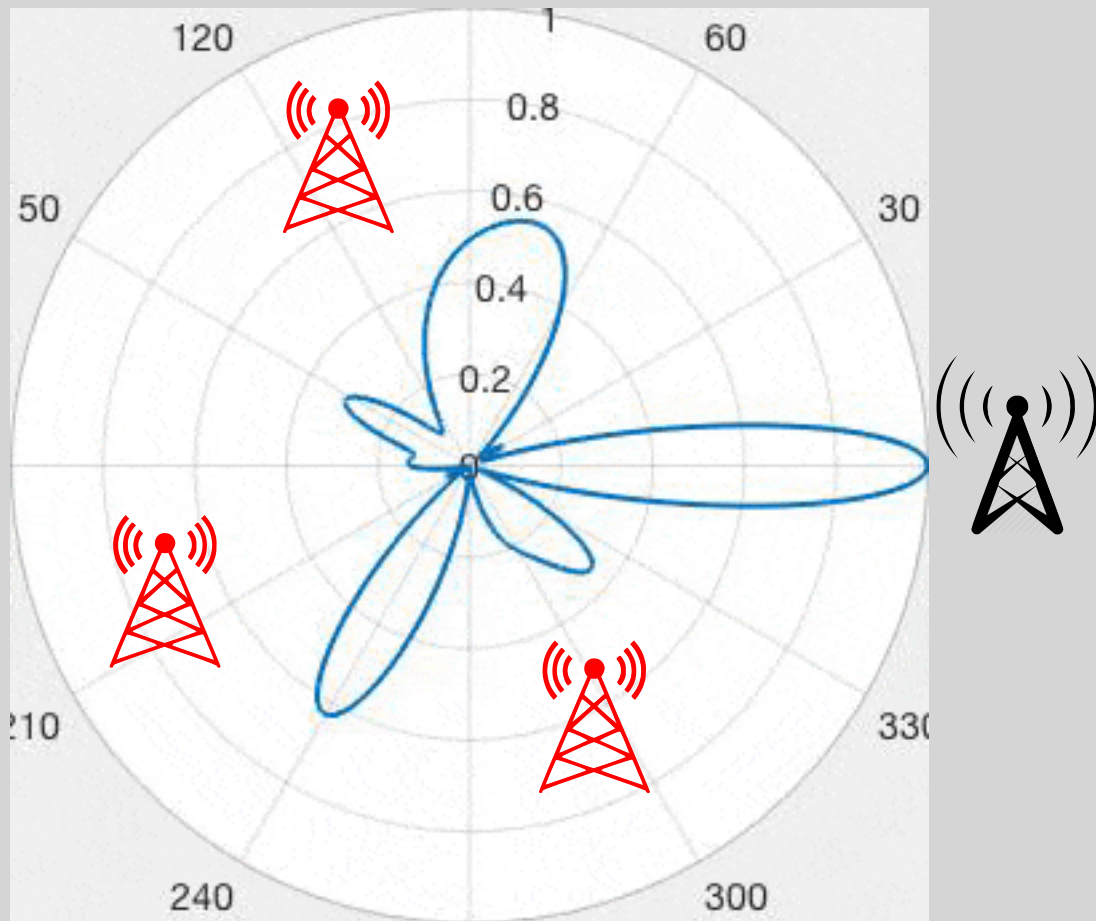
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What is the effect of artificial noise? ^[1]



If the agents had **infinite** transmit power, they would minimize the SINR in all adversary directions **simultaneously**

Transmission model

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The vector of signals transmitted by the agents is

$$y_{transmit}[t] = \mathbf{w}_t s_t + \mathbf{v}_t$$

Beamforming vector $\mathbf{w}_t = [w_1, w_2, \dots, w_m]'$

Artificial noise $\mathbf{v}_t \sim \mathcal{CN}(0, \Sigma_t)$

Since the maximum transmit power is P , we have $w_t(i) + \Sigma_t(i, i) \leq P$.

The known narrowband channel between the agent $i \in [m]$ and a receiver in the direction $\theta \in [-\pi, \pi)$ is denoted by $h_i(\theta) \in \mathbb{C}$.

- Finally, the SINR received from the direction θ is

$$SINR_t(\theta) = \frac{\mathbf{w}_t^H \mathbf{H}(\theta) \mathbf{w}_t}{Tr(\mathbf{H}(\theta) \Sigma_t) + \sigma_t^2}$$

Channel matrix $\mathbf{H}(\theta) = \mathbf{h}(\theta)\mathbf{h}(\theta)^H$

Variance of the ambient noise

$Tr(M)$ denotes the trace of the matrix M

Ensuring Security with a Periodic Transmission Strategy


The objective is to find a sequence $((\mathbf{w}_1, \Sigma_1), (\mathbf{w}_2, \Sigma_2), \dots, (\mathbf{w}_N, \Sigma_N))$ of pairs (\mathbf{w}_t, Σ_t) such that


- (I) The client receives all transmitted symbols X_t
- (II) Each adversary receives at most $N - K$ symbols


STEP 1: Encoding by $[N, N - K]$ linear maximum-distance-separable codes.

STEP 2: Transmission by the periodic strategy

$$((\mathbf{w}_1, \Sigma_1), (\mathbf{w}_2, \Sigma_2), \dots, (\mathbf{w}_L, \Sigma_L), (\mathbf{w}_1, \Sigma_1), (\mathbf{w}_2, \Sigma_2), \dots, (\mathbf{w}_L, \Sigma_L), \dots, (\mathbf{w}_1, \Sigma_1), (\mathbf{w}_2, \Sigma_2), \dots, (\mathbf{w}_L, \Sigma_L))$$


 First cycle


 Second cycle


 K-th cycle

$\min_{\mathbf{w}_k \in \mathbb{C}^m, \Sigma_k \succeq 0}$	$Tr(\Sigma_k) + \ \mathbf{w}_k\ _2^2$	Minimize total transmit power
subject to:	$SINR_k(\theta_c) \geq \gamma_c$	Client's SINR constraint
	$\forall \theta \in I_k, \quad SINR_k(\theta) \leq \gamma_a$	Adversary a_k 's SINR constraint
	$\forall i \in [m], \quad \mathbf{w}_k(i) + \Sigma_k(i, i) \leq P$	Agents' power constraints

Semi-Definite Program Relaxation and Probabilistic Approximation

Semi-infinite nonconvex optimization problem

$$\begin{aligned} \min_{\mathbf{W}_k \succeq 0, \Sigma_k \succeq 0} \quad & Tr(\Sigma_k) + Tr(\mathbf{W}_k) \\ s.t. \quad & Tr(\mathbf{H}(\theta_c)\mathbf{W}_k) \geq \gamma_c \left(Tr(\mathbf{H}(\theta_c)\Sigma_k) + \sigma_k^2 \right) \\ & \forall \theta \in I_k, Tr(\mathbf{H}(\theta)\mathbf{W}_k) \leq \gamma_a \left(Tr(\mathbf{H}(\theta)\Sigma_k) + \sigma_k^2 \right) \\ & \forall i \in [m], \mathbf{W}_k(i, i) + \Sigma_k(i, i) \leq P \\ & rank(\mathbf{W}_k) = 1 \end{aligned}$$

Source of infiniteness

Source of nonconvexity

Semi-Definite Program Relaxation and Probabilistic Approximation

Semi-infinite **convex** optimization problem

$$\begin{aligned} \min_{\mathbf{W}_k \succeq 0, \Sigma_k \succeq 0} \quad & Tr(\Sigma_k) + Tr(\mathbf{W}_k) \\ s.t. \quad & Tr(\mathbf{H}(\theta_c)\mathbf{W}_k) \geq \gamma_c \left(Tr(\mathbf{H}(\theta_c)\Sigma_k) + \sigma_k^2 \right) \\ & \forall \theta \in I_k, \quad Tr(\mathbf{H}(\theta)\mathbf{W}_k) \leq \gamma_a \left(Tr(\mathbf{H}(\theta)\Sigma_k) + \sigma_k^2 \right) \\ & \forall i \in [m], \quad \mathbf{W}_k(i, i) + \Sigma_k(i, i) \leq P \\ & \text{rank}(\mathbf{W}_k) = 1 \end{aligned}$$

Result: The convex relaxation is exact

Semi-Definite Program Relaxation and Probabilistic Approximation

Semi-infinite convex optimization problem

$$\begin{aligned} \min_{\mathbf{W}_k \succeq 0, \Sigma_k \succeq 0} \quad & Tr(\Sigma_k) + Tr(\mathbf{W}_k) \\ s.t. \quad & Tr(\mathbf{H}(\theta_c)\mathbf{W}_k) \geq \gamma_c \left(Tr(\mathbf{H}(\theta_c)\Sigma_k) + \sigma_k^2 \right) \\ & \forall \theta \in I_k, Tr(\mathbf{H}(\theta)\mathbf{W}_k) \leq \gamma_a \left(Tr(\mathbf{H}(\theta)\Sigma_k) + \sigma_k^2 \right) \\ & \forall i \in [m], \mathbf{W}_k(i, i) + \Sigma_k(i, i) \leq P \end{aligned}$$

Source of infiniteness

Semi-Definite Program Relaxation and Probabilistic Approximation

Finite convex optimization problem

$$\min_{\mathbf{W}_k \succeq 0, \Sigma_k \succeq 0} \text{Tr}(\Sigma_k) + \text{Tr}(\mathbf{W}_k)$$

$$s.t. \quad \text{Tr}(\mathbf{H}(\theta_c)\mathbf{W}_k) \geq \gamma_c \left(\text{Tr}(\mathbf{H}(\theta_c)\Sigma_k) + \sigma_k^2 \right)$$

$$\forall \theta \in \Theta_B, \quad \text{Tr}(\mathbf{H}(\theta)\mathbf{W}_k) \leq \gamma_a \left(\text{Tr}(\mathbf{H}(\theta)\Sigma_k) + \sigma_k^2 \right)$$

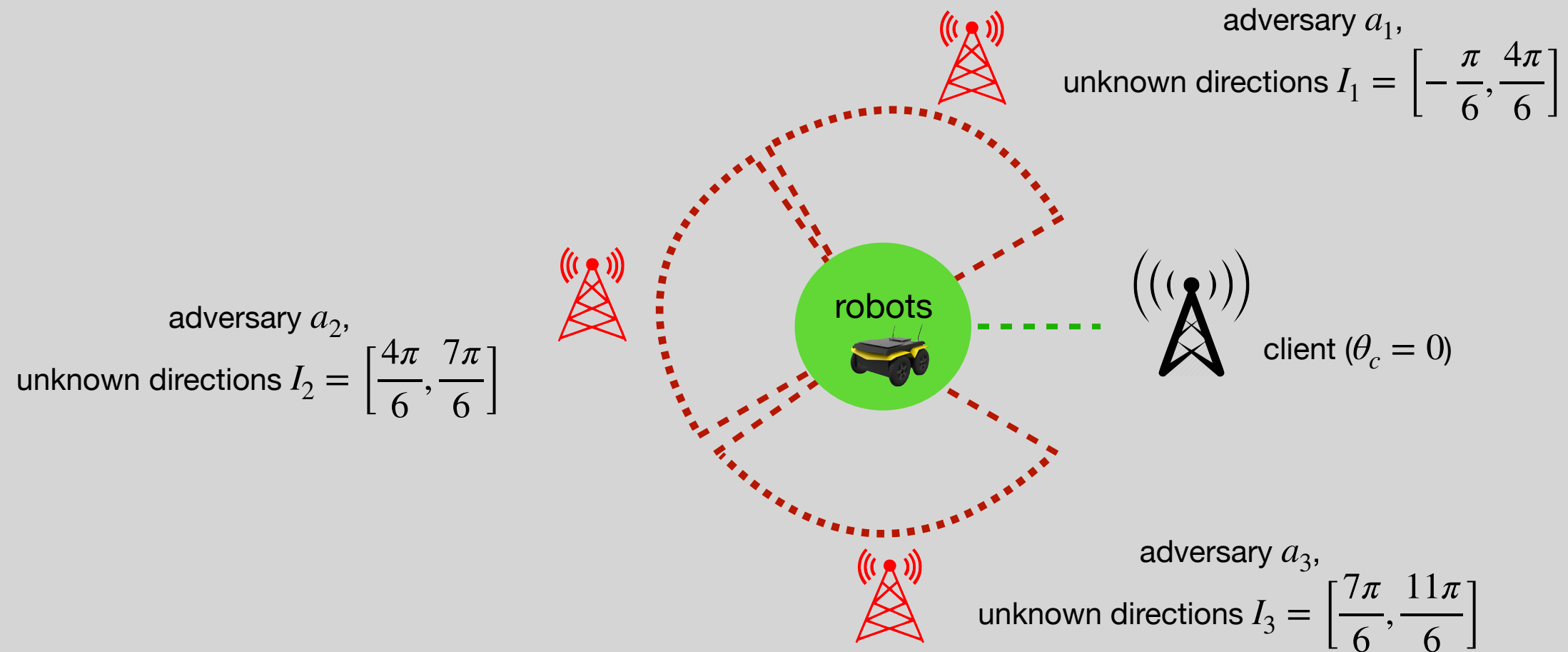
$$\forall i \in [m], \quad \mathbf{W}_k(i, i) + \Sigma_k(i, i) \leq P$$

Randomly sample $B \in \mathbb{N}$
points from the set I_k

Result [1]: The following statement is true with probability $1 - \beta_2$:

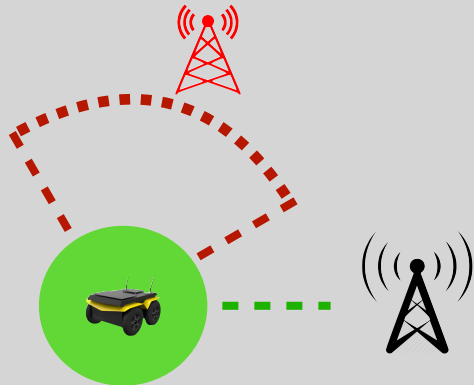
If $B \geq (2 \log_e(\beta_2^{-1}) + 16m^2)/\beta_1$, the problems are **equivalent** with probability $1 - \beta_1$.

A Numerical Example

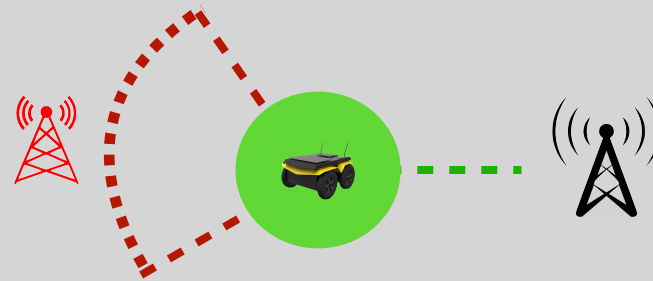


A Numerical Example

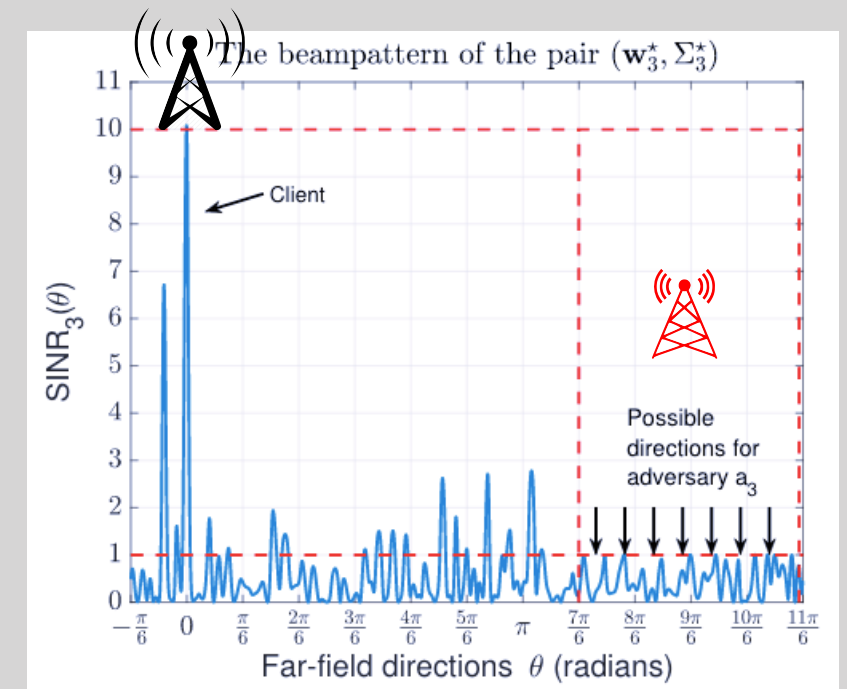
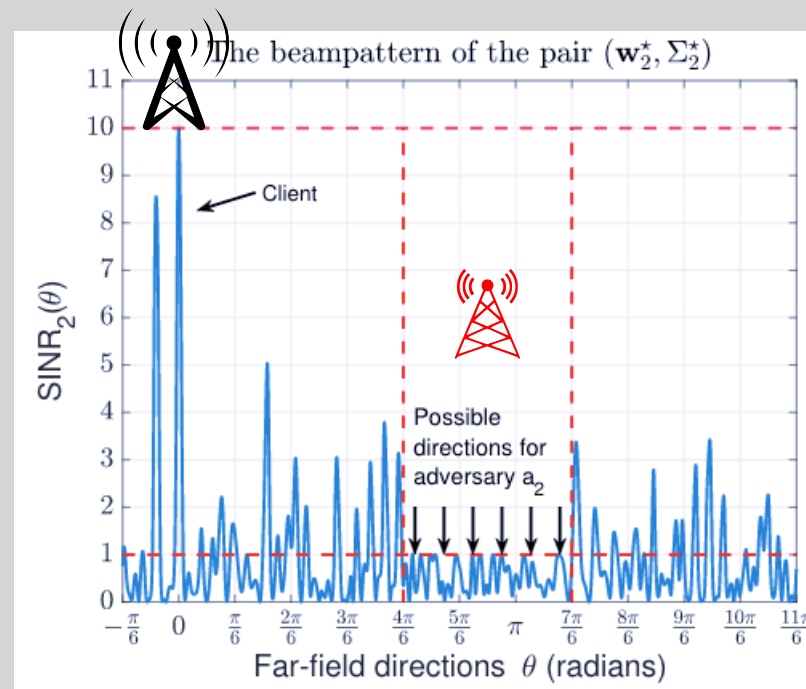
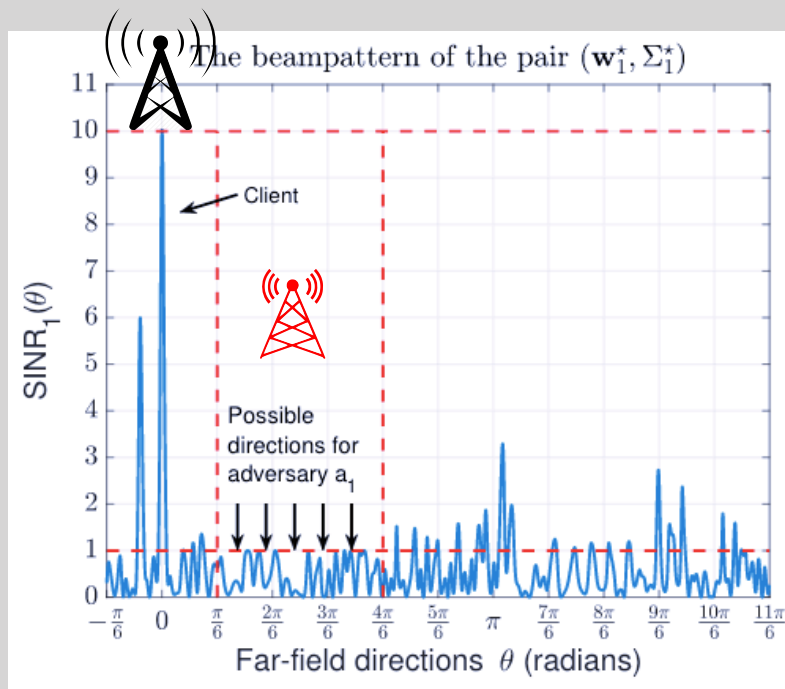
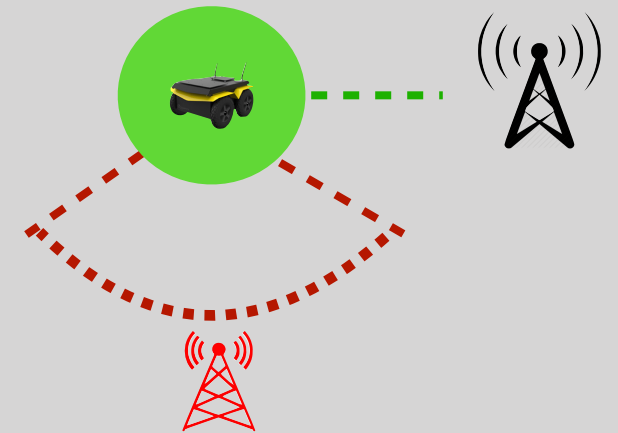
First time step:



Second time step:

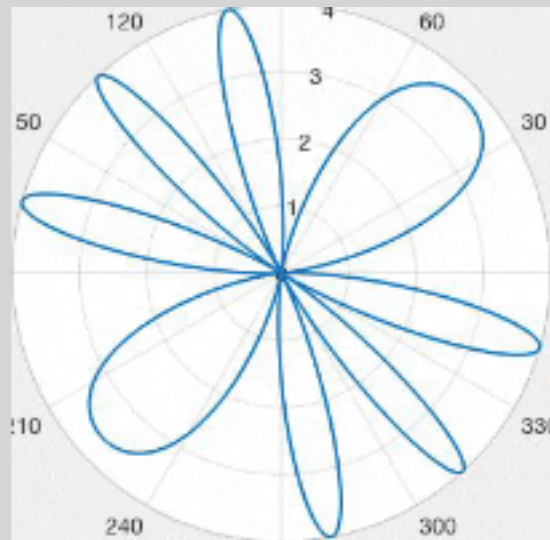


Third time step:

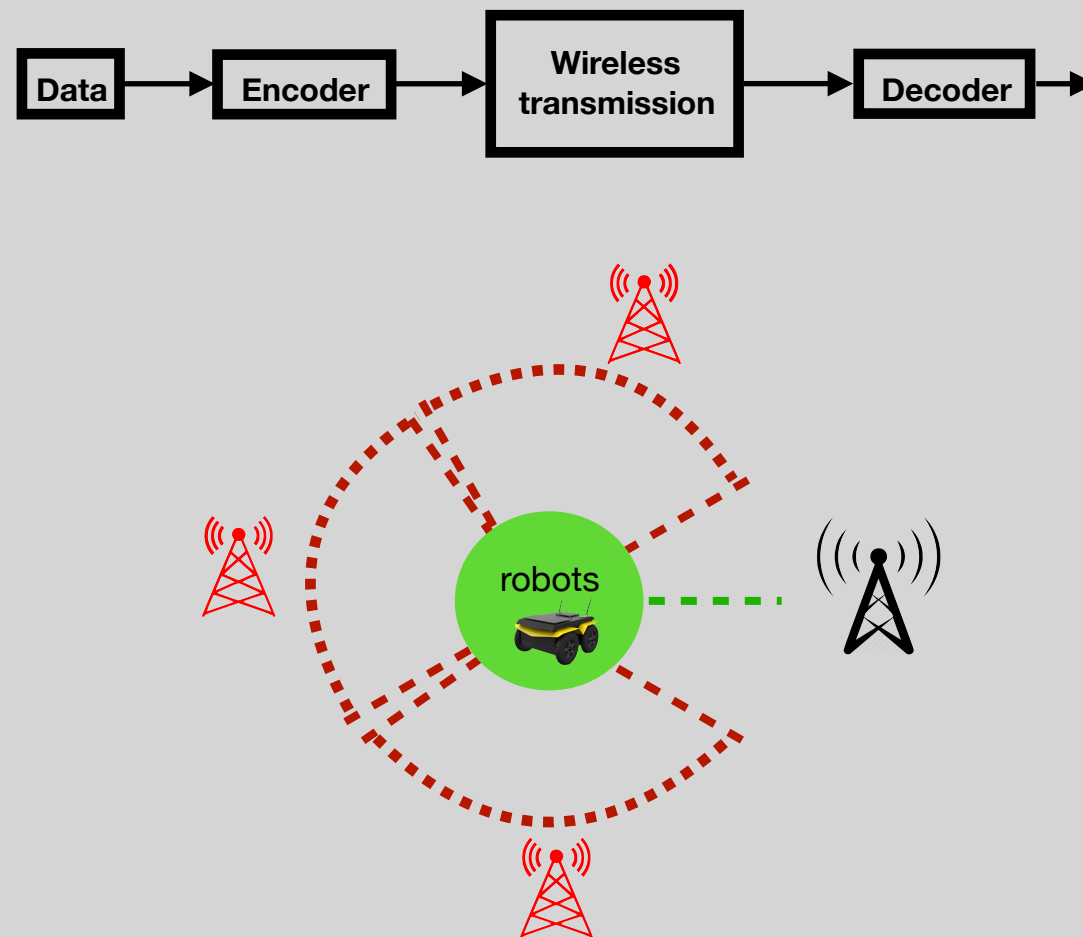


Conclusions

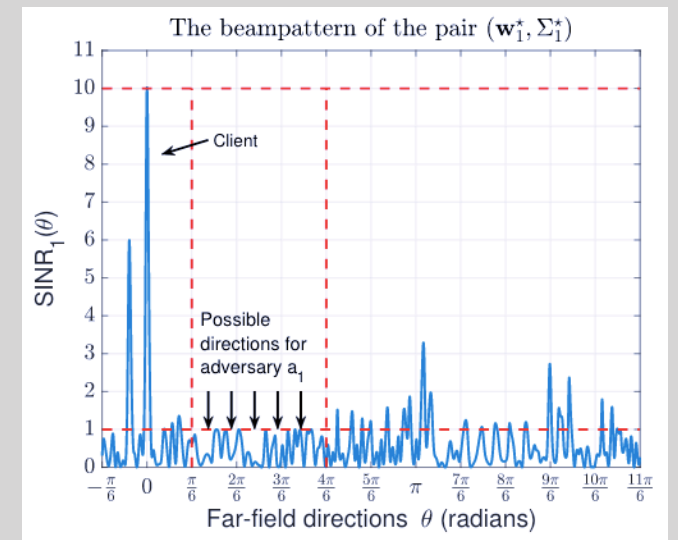
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Thank you for listening

E-mail: yagiz.savas@utexas.edu

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