

# Security Issues in Spectrum Sharing Between Radar and Communication Systems

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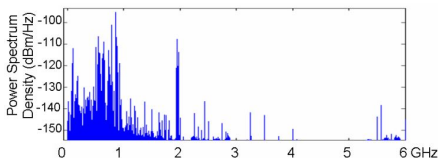


- ① Introduction
- ② System Model
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# The Need for Spectrum Sharing

- Radar and communications jointly consume most of the spectrum below 6 GHz.
- Until recently, allocated spectrum for commercial and non-commercial purposes (i.e. military radar) were on distinct bands.
- S-band radar (2 – 4 GHz) partially overlaps with LTE and WiMax systems.

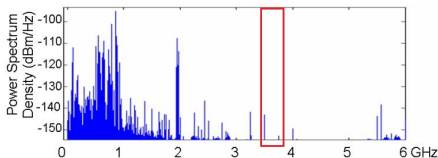


Freq (GHz)	0~1	1~2	2~3	3~4	4~5	5~6
Utilization(%)	54.4	35.1	7.6	0.25	0.128	4.6

Figure: Spectrum utilization in downtown Berkeley (UC Berkeley, 2007).

# The Need for Spectrum Sharing

- As the number of connected devices grows, these band distinctions limit a more efficient use of the spectrum.
- Spectrum regulators have proposed to make the 3.55 – 3.7 GHz band (used for military radar) available to commercial cellular systems.
- The need arises for an efficient use of the spectrum for both systems, without one interfering with the other → Spectrum sharing approaches.



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# Existing Approaches for Spectrum Sharing

- 1 Avoid interference by large spatial separation.



Figure: Shipborne radar exclusion zones in 3.5 GHz band (NTIA 2015).

- 2 Dynamic spectrum access based on spectrum sensing.

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- 1 Avoid interference by large spatial separation.



Figure: Shipborne radar exclusion zones in 3.5 GHz band (NTIA 2015).

- 2 Dynamic spectrum access based on spectrum sensing.
- 3 **Spatial multiplexing enabled by the multiple antennas at both the radar and communication systems.**

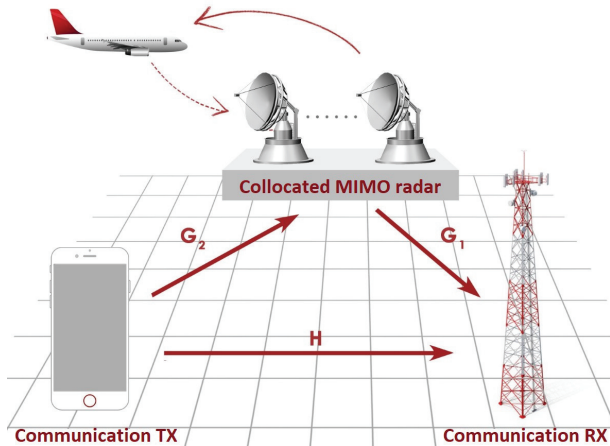
- Methods that address the objectives of one or the other system but not both.
  - Nullspace projection implemented by radar to reduce interference towards the communication system [Sodagari et al. 2012] or vice versa.
  - Nullspace projection precoding to avoid interference is possible on either the radar or the communication systems but not on both [Mahal et al.,2017].
- Co-design methods that address the constraints of both systems.
  - Communication system and/or radar precoding schemes are co-designed in order to maximize an objective function of one user (typically the radar), subject to meeting certain constraints for the other (typically the communication system)  
[Li, Kumar, and Petropulu, 2016] [Li, Petropulu, Trappe, 2016]  
[Li and Petropulu, 2017]



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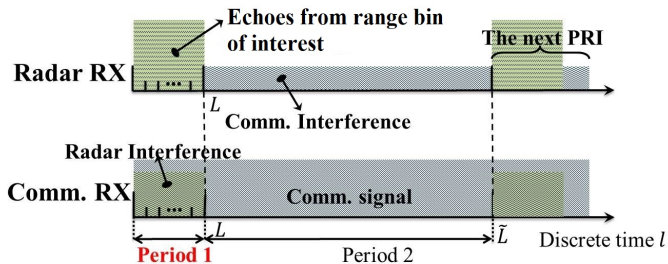
# Spectrum Sharing Formulation

- $M_R^r \times M_R^t$  MIMO radar
- $M_C^r \times M_C^t$  MIMO communication system



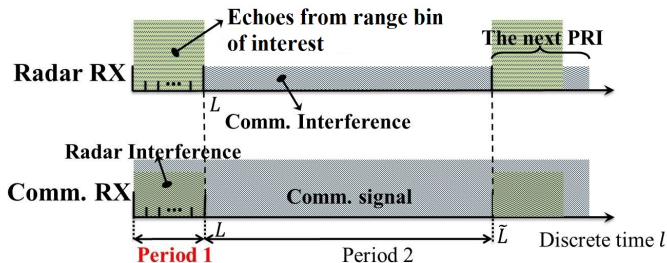
# Interference During Spectrum Sharing

- Interference at the radar occurs when the radar is listening, or forwarding the obtained samples to the radar fusion center.



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- Use transmit precoding to limit the interference to the radar.

# The Coexistence Signal Model

- The received signals at the radar and communication RX are

**Radar fusion center:** (1a)

$$\Omega \circ \mathbf{Y}_R = \Omega \circ \left( \underbrace{\mathbf{DPS}}_{\text{signal}} + \underbrace{\mathbf{CPS} + \mathbf{G}_2 \mathbf{X} \Lambda_2}_{\text{interference}} + \underbrace{\mathbf{W}_R}_{\text{noise}} \right), \quad (1b)$$

**Communication receiver:** (1c)

$$\mathbf{Y}_C = \underbrace{\mathbf{H}\mathbf{X}}_{\text{signal}} + \underbrace{\mathbf{G}_1 \mathbf{P} \mathbf{S} \Lambda_1}_{\text{interference}} + \underbrace{\mathbf{W}_C}_{\text{noise}}, \quad (1d)$$

where

- $\mathbf{P}, \mathbf{S}, \Omega$ : radar precoder, waveforms, subsampling matrix
- $\mathbf{D} = \sum_{k=1}^K \sigma_{\beta_0}^2 \mathbf{v}_t^*(\theta_k) \mathbf{v}_t^T(\theta_k)$
- $\mathbf{C} \triangleq \sum_{k=1}^{K_c} \beta_k^c \mathbf{v}_r(\theta_k^c) \mathbf{v}_t^T(\theta_k^c)$ : clutter response matrix
- $\mathbf{X} \triangleq [\mathbf{x}(1), \dots, \mathbf{x}(L)]$ : comm codewords  $\mathbf{x}(l) \sim \mathcal{CN}(0, \mathbf{R}_{xl})$
- $\Lambda_1, \Lambda_2$ : diagonal matrices denoting random phase offsets

- Radar SINR:

$$\text{SINR} = \frac{m \text{Tr}(\mathbf{P}\mathbf{P}^H \mathbf{D})}{m \text{Tr}(\mathbf{P}\mathbf{P}^H \mathbf{C}) + \sum_{l=1}^L \text{Tr}(\mathbf{G}_{2l} \mathbf{R}_{xl} \mathbf{G}_{2l}^H) + m\sigma_R^2}. \quad (2)$$

## Constraints:

- The power budget at the communication transmitter:  
$$\sum_{l=1}^L \text{Tr}(\mathbf{R}_{xl}) \leq P_t,$$
- The requirement on the average communication rate achieved during the  $L$  symbol periods

$$C_{\text{avg}}(\{\mathbf{R}_{xl}\}) \triangleq \frac{1}{L} \sum_{l=1}^L \log_2 |\mathbf{I} + \mathbf{R}_{\text{Cin}}^{-1} \mathbf{H} \mathbf{R}_{xl} \mathbf{H}^H| \geq C \quad (3)$$

$$\mathbf{R}_{\text{Cin}} = \mathbf{G}_1 \mathbf{P} \mathbf{P}^H \mathbf{G}_1^H + \sigma_C^2 \mathbf{I}.$$

# The Co-Design Problem

- Cooperate on estimating  $\mathbf{G}_1$ ,  $\mathbf{G}_2$ . Share  $\mathbf{H}$ ,  $\mathbf{G}_1$ , and  $\mathbf{G}_2$  with the controller.
- The controller designs  $\Phi = \mathbf{P}\mathbf{P}^H$ ,  $\Omega$  and  $\{\mathbf{R}_{xl}\}$  as

$$\begin{aligned} & \max_{\{\mathbf{R}_{xl}\} \succeq 0, \Phi \succeq 0, \Omega} \text{SINR}(\{\mathbf{R}_x\}, \Omega, \Phi), \\ \text{s.t. } & C_{\text{avg}}(\{\mathbf{R}_{xl}\}, \Phi) \geq C, \end{aligned} \quad (4a)$$

$$\sum_{l=1}^L \text{Tr}(\mathbf{R}_{xl}) \leq P_C, L\text{Tr}(\Phi) \leq P_R, \quad (4b)$$

$$\Omega \text{ is proper} \quad (4c)$$

# The Interference Channel

- The interference channel matrix is directly related to the radar location, as seen in the following model [Heath, 2017] [Molisch, 2012]

$$\mathbf{G}_2 = \frac{\sqrt{E_x} \lambda_c}{4\pi d \sqrt{M_C^t}} \left( \sqrt{\frac{K}{1+K}} \mathbf{S}_{\text{LoS}} + \sqrt{\frac{1}{1+K}} \mathbf{S}_{\text{NLoS}} \right) \quad (5)$$

- $\lambda_c$ : carrier wavelength;  $E_x$ : transmit energy;  $d$ : the radar distance from the smartphone;  $K$  is the Rician factor.
- $\mathbf{S}_{\text{LoS}} = \mathbf{e}_r(\Omega_r) \mathbf{e}_t(\Omega_t)^T$  and  $\mathbf{S}_{\text{NLoS}}$  a matrix of i.i.d.  $\mathcal{N}_{\mathbb{C}}(0, 1)$  entries.
- $\Omega_t = \sin(\phi_t)$  and  $\Omega_r = \sin(\phi_r)$  the angles of incidence of the Line-of-Sight path on the TX and RX steering vectors

$$\mathbf{e}_t(\Omega_t) = \left[ 1, e^{-j \frac{2\pi \Delta_t}{\lambda_c} \Omega_t}, \dots, e^{-j(M_C^t - 1) \frac{2\pi \Delta_t}{\lambda_c} \Omega_t} \right]^T,$$

$$\mathbf{e}_r(\Omega_r) = \left[ 1, e^{-j \frac{2\pi \Delta_r}{\lambda_c} \Omega_r}, \dots, e^{-j(M_R^r - 1) \frac{2\pi \Delta_r}{\lambda_c} \Omega_r} \right]^T$$



[Li and Petropulu, 2017]

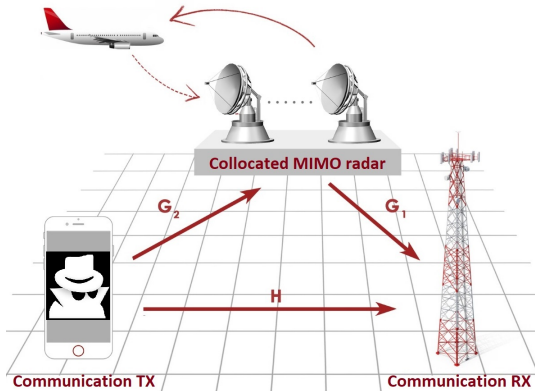
- The controller is incorporated into the MIMO radar.
  - This avoids interference during communication with the radar.
  - Also, the controller is a trusted node.
- The controller collects information from the two systems and designs the precoders so that some performance objective is met.
- The computed precoder is passed to the communication system.

[Li and Petropulu, 2017]

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- The controller collects information from the two systems and designs the precoders so that some performance objective is met.
- The computed precoder is passed to the communication system.
- The precoder contains *implicit* information about the radar.

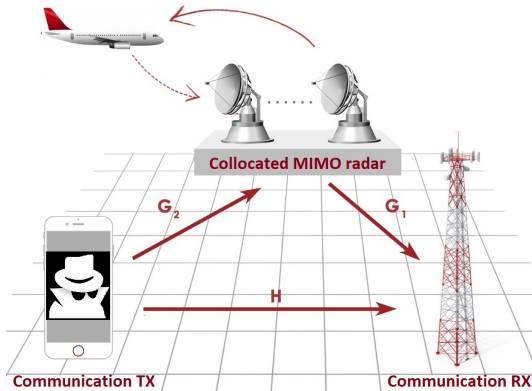
# Security Concern

- Can the precoder be used by an adversary to launch an inference attack?



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- Can the adversary reverse engineer the precoder matrix to infer the radar location?

- Two precoders are examined here:
  - **Null Space Precoder** - Zero forces the interference at the radar receive antennas

$$\mathbf{P}_n = \text{nullspace}(\mathbf{G}_2)$$

Assumes more comm system TX antennas than radar RX antennas  
[Sodagari et al. 2012, Babaei et. al., 2013, Khawar et. al.]

- **Optimized Precoder** - Designed to minimize interference at the radar RX, subject to the comm system meeting certain rate and power constraints.

[Li, Petropulu, Trappe, 2016],[Li, Kumar, and Petropulu, 2016],  
[Li and Petropulu, 2017]

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# Adversary Inference Attack

- Suppose an adversary is operating  $S$  independent smartphones, and observes at every point in time  $t = 1, \dots, T$  all precoder matrices  $\mathcal{P}^t = \{\mathbf{P}_1^t, \dots, \mathbf{P}_S^t\}$  sent to the smartphones by the controller.
- For simplicity, each precoder is obtained independently of the others.
- The adversary is not capable of estimating  $\mathbf{G}_2$ ; otherwise it would easily locate the radar.
- The adversary treats the unknown radar location as a random variable  $R$ , and attempts to create an estimate of its pdf,  $p_R$ , based on the observed precoders sent by the controller.

- This can be formulated as a Bayesian inference problem, where the conditional pdf of a sequence of  $T$  candidate radar locations given a sequence of  $T$  precoders equals

$$p_R(R^1, \dots, R^T | \mathcal{P}^1, \dots, \mathcal{P}^T) = \frac{p_{P|R}(\mathcal{P}^1, \dots, \mathcal{P}^T | R^1, \dots, R^T)}{p_P(\mathcal{P}^1, \dots, \mathcal{P}^T)} p_R(R^1, \dots, R^T) \quad (6)$$

- $p_{P|R}$  is the probability of the observed precoder matrices given a specific radar location.
- May assume that all candidate locations are equally likely, i.e., the a priori pdf  $p_R(R^1, \dots, R^T)$  is a constant.



- May also assume that the controller assignments are memoryless, i.e.,

$$p_R(R^1, \dots, R^T | \mathcal{P}^1, \dots, \mathcal{P}^T) = \frac{\prod_{t=1}^T p_{P|R}(\mathcal{P}^t | R^t)}{\sum_{\mathcal{R}} \prod_{t=1}^T p_{P|R}(\mathcal{P}^t | R^t)} \quad (7)$$

$\mathcal{R}$  is the set of all candidate location sequences.

- If the adversary knew  $p_{P|R}(\mathcal{P}^t | R^t)$ , it could compute (7) for every possible combination of candidate locations.

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- If the adversary knew  $p_{P|R}(\mathcal{P}^t | R^t)$ , it could compute (7) for every possible combination of candidate locations.
- Optimal estimation is computationally prohibitive  $\rightarrow$  use a supervised machine learning approach for radar location estimation.

- Adversary divides search area into cells.
- Adversary trains a classifier for every separate cell, using training data and their corresponding labels.
- Features in the classification problem are the precoding matrices, separated into real and imaginary parts, and stacked in a long vector.
- Once training has been completed, the adversary can decide which cell a new precoder corresponds to. This task can be parallelized.

- One way to quantify the amount of information a precoder reveals about the radar location is via the *Mutual Information (MI)*.
- $R = (R_x, R_y) \sim p(R_x, R_y)$  denote radar coordinates, and  $P = [P_1, \dots, P_n]^T \sim p(P_1, \dots, P_n)$  the precoder vector.

**Mutual Information**  $I(R; P) \triangleq$

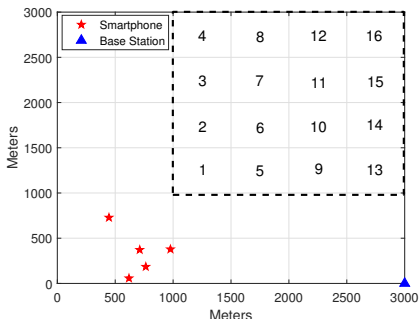
$$\int \cdots \int p(R_x, \dots, P_n) \log_2 \frac{p(R_x, \dots, P_n)}{p(R_x, R_y)p(P_1, \dots, P_n)} dR_x \cdots dP_n \quad (8)$$

- MI can be estimated numerically using multi-dimensional histograms.

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# Simulation Setup

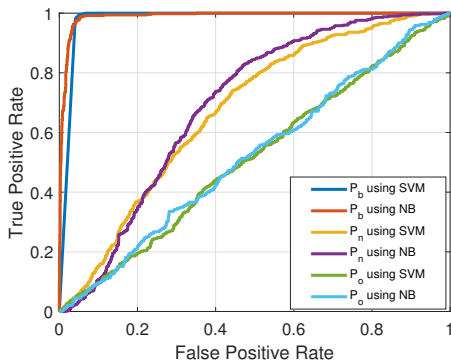
- The adversary will test all cells and make a binary decision on the presence of the radar in a particular cell.
- We assume the adversary is controlling  $S = 5$  smartphones.
- The radar has  $M_R^r = M_R^t = 6$  antennas and the communication system has  $M_C^t = M_C^r = 8$  antennas.



# Simulation Setup

- Baseline approach ( $\mathbf{P}_b$ )  $\rightarrow$  the adversary observes  $\mathbf{G}_2$ .
- Three separate balanced training sets  $\mathcal{L}_b^c$ ,  $\mathcal{L}_n^c$ ,  $\mathcal{L}_o^c$ , of 6000 samples each were created for cell  $c = 4$ , for the cases where the adversary observes  $\mathbf{P}_b$ ,  $\mathbf{P}_n$ , and  $\mathbf{P}_o$ , respectively.
- A separate test set  $\mathcal{T}^c$  for  $c = 4$  was created, consisting of 2375 samples; 500 samples correspond to precoders for radar locations in  $c = 4$ , and 1,875 samples for the radar in all other cells (125 samples for each  $c \neq 4$ ).
- To avoid over-fitting, the radar locations used for training were different than those used in testing.
- For training we used the Support Vector Machine (SVM) and Naive Bayes (NB) classifiers (Matlab functions *fitcsvm* and *fitcnb*, respectively).

# Simulation Setup

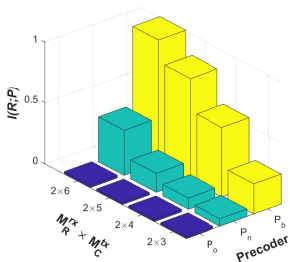


- ROC for cell  $c = 4$
- $P_b$  results in almost perfect radar location prediction.
- Using  $P_o$  results in a random adversary guess  $\rightarrow$   $P_o$  a better option in protecting the radar privacy.

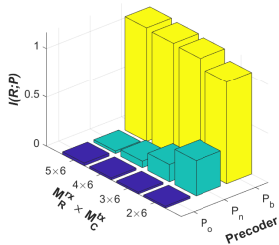


# Mutual Information

- Numerically computed mutual information for all assumed precoders.
- Depending on precoder, the bins of the multi-dimensional histogram where created from the positive samples of  $\mathcal{L}_b^c$ ,  $\mathcal{L}_n^c$ , or  $\mathcal{L}_o^c$ , using the K-means clustering algorithm.



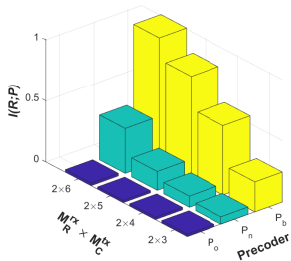
(a)



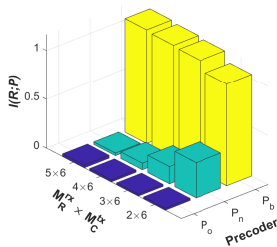
(b)

# Mutual Information

- Notice that  $I(R; P_o) < I(R; P_n) < I(R; P_b) \rightarrow$  greater reduction in the uncertainty of  $R$  when observing  $P_b$  than when observing  $P_o$ .
- In other words,  $\mathbf{P}_b$  reveals the most information about a radar location while  $\mathbf{P}_o$  the least.



(a)



(b)

# Mutual Information

- For  $\mathbf{P}_b$  or  $\mathbf{P}_n$ , an increase in the # of transmit antennas at the communication system results in an increase to the mutual information  $\rightarrow$  respective increase in the column space of  $\mathbf{P}_b$  directly affects the size of  $\mathbf{P}_n$  as well.
- The value of  $I(R; P_o)$  is very small  $\rightarrow R$  and  $P_o$  are close to being independent, with most of the radar information being suppressed in the optimized precoder.
- $\mathbf{P}_n$  is only a function of  $\mathbf{G}_2$  but  $\mathbf{P}_o$  is additionally a function of  $\mathbf{H}, \mathbf{G}_1$ .
- $\mathbf{P}_o$  is obtained as the solution of a constrained optimization problem  $\rightarrow$  contribution of  $\mathbf{G}_1$  to the final solution less transparent and  $\mathbf{H}$  by definition has no information regarding the radar position.
- The optimal precoder  $\mathbf{P}_o$  seems to be better for the radar privacy but involves more computational complexity.

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- We examined the extent to which the adversary can infer radar location information from the communication system precoder matrix, using a machine learning based inference attack.
- Depending on the used precoder scheme, our simulations indicated that this was indeed possible, a result further supported by our estimation of the mutual information between the precoder matrix and radar location.

The precoder  $\mathbf{P} = \sqrt{\mathbf{R}_{xl}}$  is the solution to:

$$\min_{\mathbf{R}_{xl}} \sum_{l=1}^L \text{Tr}(\mathbf{G}_2 \mathbf{R}_{xl} \mathbf{G}_2^H)$$

$$\text{s.t.} \quad \sum_{l=1}^L \text{Tr}(\mathbf{R}_{xl}) \leq P_C \quad (\text{restricts comm. TX antenna power})$$

$$\frac{1}{L} \sum_{l=1}^L \log_2 |\mathbf{I} + \mathbf{R}_{wl}^{-1} \mathbf{H} \mathbf{R}_{xl} \mathbf{H}^H| \geq C \quad (\text{restricts comm. average capacity})$$

$$I(\mathbf{R}; \mathbf{P}) \leq M$$

where  $\mathbf{R}_{xl}$  is the transmit covariance matrix,  $M$  an accepted scalar value for which we assume privacy is achieved.

Thank you!  
Questions?



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