

Optimizing Backscattering Coefficient Design for Minimizing BER at Monostatic MIMO Reader

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Introduction

- **Multiantenna technology** can help low-power backscatter communications (BSC)
- Realizing **sustainable wireless networking** by optimally designing the detection protocols for reader [1]
- BSC thrives on its capability to use low-power **passive** devices like:
 - a) envelope detectors,
 - b) dividers,
 - c) comparators,
 - d) Impedance controllers and others
- Avoids **costly and bulkier** conventional RF chain components like the radio unit, local oscillator, mixer, etc
- **Challenges**: Tags in BSC do not have dedicated radio resources [2,3]
- **Focus**: Specially designed **optimal detection protocols** are needed at multiantenna readers keeping in mind the underlying channel estimation and tag signal detection **errors**

[1] C. Xu, L. Yang, and P. Zhang, "Practical backscatter communication systems for battery-free Internet of Things: A tutorial and survey of recent research," IEEE Signal Process. Mag., vol. 35, no. 5, pp. 16–27, Sept. 2018.

[2] D. Bharadia, K. R. Joshi, M. Kotaru, and S. Katti, "Backfi: High throughput WiFi backscatter," in Proc. ACM SIGCOMM, London, UK, Oct. 2015, pp. 283–296.

[3] D. Mishra and E. G. Larsson, "Optimal channel estimation for reciprocity-based backscattering with a full-duplex MIMO reader," IEEE Trans. Signal Process., vol. 67, no. 6, pp. 1662–1677, Mar. 2019.

Literature Review

- Existing works on backscattering detection mostly investigated **ambient** BSC settings [4–5]
- Robust inference algorithms not requiring any channel statistics [6] to detect the sensing values of multiple single-antenna backscatter sensors for **bistatic BSC** model
- **Maximum likelihood** (ML) based optimal detector and suboptimal linear combiners for recovering the signals from the ambient emitter and the desired tag at the MIMO reader [7]

[4] J. Qian, A. N. Parks, J. R. Smith, F. Gao, and S. Jin, "IoT communications with M-PSK modulated ambient backscatter: Algorithm, analysis, and implementation," *IEEE Internet Things J.*, vol. 6, no. 1, pp. 844–855, Feb. 2019.

[5] A. Bekkali, S. Zou, A. Kadri, M. Crisp, and R. V. Penty, "Performance analysis of passive UHF RFID systems under cascaded fading channels and interference effects," *IEEE Trans. Wireless Commun.*, vol. 14, no. 3, pp. 1421–1433, Mar. 2015.

[6] G. Zhu, S. Ko, and K. Huang, "Inference from randomized transmissions by many backscatter sensors," *IEEE Trans. Wireless Commun.*, vol. 17, no. 5, pp. 3111–3127, May 2018.

[7] G. Yang, Q. Zhang, and Y. C. Liang, "Cooperative ambient backscatter communications for green internet-of-things," *IEEE Internet Things J.*, vol. 5, no. 2, pp. 1116–1130, Apr. 2018.

Monostatic BSC with MIMO Reader

- Existing works considered **perfect** channel state information (CSI) availability [8, 9] at the multiantenna monostatic reader for the maximum ratio combining (MRC) based detection [10, 11]
- Pairwise error probability and diversity order achieved by the orthogonal space-time block codes over the dyadic backscatter channel were derived in [10]
- **Research Gap:** Investigation on an optimal detection protocol for MIMO reader-based monostatic BSC

[8] Y. Zhang, F. Gao, L. Fan, X. Lei, and G. K. Karagiannidis, "Backscatter communications over correlated Nakagami-m fading channels," *IEEE Trans. Commun.*, vol. 67, no. 2, pp. 1693–1704, Feb. 2019.

[9] D. Kim, H. Jo, H. Yoon, C. Mun, B. Jang, and J. Yook, "Reverse-link interrogation range of a UHF MIMO-RFID system in Nakagami-m fading channels," *IEEE Trans. Ind. Electron.*, vol. 57, no. 4, pp. 1468–1477, Apr. 2010.

[10] C. Boyer and S. Roy, "Space time coding for backscatter RFID," *IEEE Trans. Wireless Commun.*, vol. 12, no. 5, pp. 2272–2280, May 2013.

[11] D. Mishra and E. G. Larsson, "Multi-tag backscattering to MIMO reader: Channel estimation and throughput fairness," *IEEE Trans. Wireless Commun.*, vol. 18, no. 12, pp. 5584–5599, Dec. 2019.

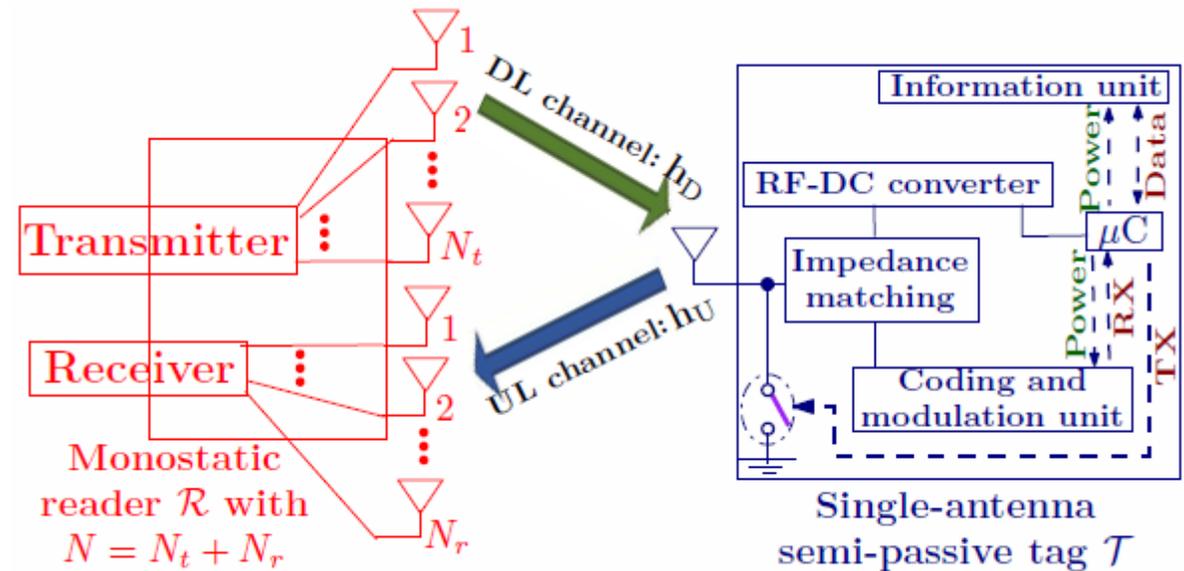
Motivation and Contributions

- **Large antenna array at reader** for enhancing the detection performance in monostatic BSC settings
- Noting **the resource-constraints** of tags
- **Practical preamble designing** that neither requires pilot transmission from tag nor any help in channel estimation

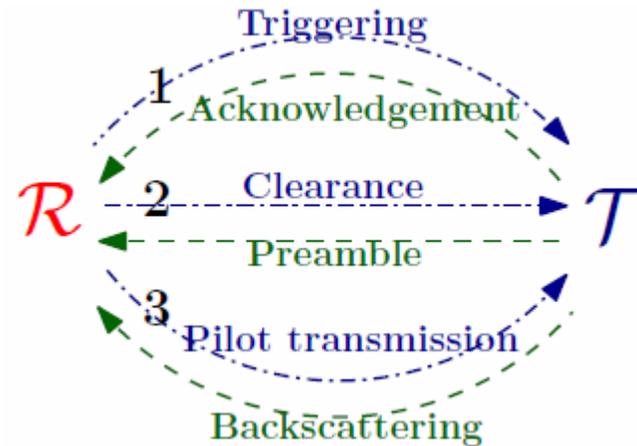
- The **key contribution** of this work is four-fold:
 - a) **Novel BSC transmission protocol** for a MIMO reader to detect the backscattered signals from a single-antenna semi-passive tag
 - b) Adopting ML detector at reader, **tight analytical approximations** for the optimal detection threshold and bit error rate (BER) are derived
 - c) **Globally-optimal value for tag's backscattering coefficients** (BC) is derived in closed-form for minimizing the underlying non-convex BER
 - d) The key analytical claims are **validated** via simulations, while demonstrating the **gains** achieved via optimal BC on the practical BER

Adopted Monostatic Backscattering Model

- The adopted **BSC model**:
 - a) N_t antennas at reader for transmission and N_r for receiving the backscattered signals
 - b) single-antenna tag
- Uniform linear array at the multi-antenna reader
- Rician block fading model is adopted
- **Binary modulated backscatter design** involving amplitude shift keying between two states:
 - a) To transmit bit '0', tag adjusts its impedance to move the antenna in absorbing state for ensuring that a **very little fraction** of incident signal is reflected to reader
 - b) For transmitting '1', tag sets its impedances to ensure that the transistor is shorted and **maximum amount of signal is reflected back**



Three-Phase Transmission Protocol for BSC Detection



- Three-phase transmission protocol:
 - a) First phase:** R **triggers** T from the sleep state to get ready for backscattering.
 - b) Second phase:**
 - i) After receiving T's **acknowledgment**, R sends a **clearance-to-send** while following the Electronic-Product-Code Class-1 Generation-2 protocol
 - ii) Reacting to it, T sends a known **preamble** sequence to aid R in estimating the long-term channel statistics like σ_h^2
 - c) Third phase:** **Pilot signal transmission** from R and thereby detecting the resulting **backscattered** from T
- First two phases setting-up or initializing BSC and providing **estimates for statistical values** of channel parameters
- Third phase performs the **targeted ML detection**

Backscattered Signal Model

- First $N/2$ antennas for transmitting $N/2$ orthogonal pilots **isotropically** from reader
- The **received signal matrix** at reader for tag's n th symbol period can be written as:

$$\mathbf{Y}(n) = \left(\mathbf{h}_U \mathbf{b}(n) \mathbf{h}_D^T + \mathbf{U} \right) \mathbf{S} + \mathbf{W} = \mathbf{H}_b(n) \mathbf{S} + \mathbf{W}$$

- Using **pseudo-inverse** of pilot matrix, signal available for detection at reader in vectorized form:

$$\mathbf{y}_v(n) = \text{vec} \left\{ \mathbf{Y}(n) \mathbf{S}^\dagger \right\} = \mathbf{b}(n) \mathbf{h}_v + \mathbf{u}_v + \hat{\mathbf{w}}_v = \begin{cases} a_0 \mathbf{h}_v + \mathbf{w}_v, & \text{bit '0'}, \\ (a_0 + a_1) \mathbf{h}_v + \mathbf{w}_v, & \text{bit '1'}, \end{cases}$$

where $\mathbf{h}_v \triangleq \text{vec} \left\{ \mathbf{h}_U \mathbf{h}_D^T \right\}$ and $\mathbf{w}_v \triangleq \text{vec} \left\{ \mathbf{U} + \frac{\mathbf{W} \mathbf{S}^H}{E_p} \right\}$

- To derive the expression for BER in **closed-form**, we consider two **approximations**:
 - a) **Strong LoS component** in BSC \rightarrow Point-wise product $\mathbf{h}_U \odot \mathbf{h}_D$ follows complex Gaussian with i.i.d entries
 - b) All entries of \mathbf{h}_v are **i.i.d. complex Gaussain** and it can be approximated as:

$$\mathbf{h}_v \sim \text{CN} \left(\boldsymbol{\mu}_{\mathbf{h}_v} \triangleq \text{vec} \left\{ \mathbb{E} \left\{ \mathbf{h}_U \right\} \mathbb{E} \left\{ \mathbf{h}_D^T \right\} \right\}, \frac{\beta^2(1+2K)}{(1+K)^2} \mathbf{I}_{\frac{N^2}{4}} \right), \forall K \gg 1$$

Detection Hypothesis

- Using [proposed approximations](#), ML detection problem can be written as the **hypothesis testing**:

$$\begin{cases} \mathcal{H}_0 : \mathbf{y}_v \sim \mathbb{CN} \left(\boldsymbol{\mu}_0, \sigma_0^2 \mathbf{I}_{\frac{N^2}{4}} \right), & \text{bit '0'}, \\ \mathcal{H}_1 : \mathbf{y}_v \sim \mathbb{CN} \left(\boldsymbol{\mu}_1, \sigma_1^2 \mathbf{I}_{\frac{N^2}{4}} \right), & \text{bit '1'}, \end{cases}$$

where $\boldsymbol{\mu}_0 \triangleq a_0 \boldsymbol{\mu}_{h_v}$, $\boldsymbol{\mu}_1 \triangleq (a_0 + a_1) \boldsymbol{\mu}_{h_v}$, $\sigma_0^2 \triangleq |a_0|^2 \beta^2 + \sigma_I^2$, $\sigma_1^2 \triangleq |a_0 + a_1|^2 \beta^2 + \sigma_I^2$, and $\sigma_I^2 \triangleq \sigma_u^2 + \frac{\sigma_w^2}{E_p}$

- The probability density function (PDF) for \mathbf{y}_v under proposed hypothesis is defined as:

$$p(\mathbf{y}_v; \mathcal{H}_i) = \frac{e^{-\frac{1}{2}(\mathbf{y}_v - \boldsymbol{\mu}_i)^H \left[\sigma_i^2 \mathbf{I}_{\frac{N^2}{4}} \right]^{-1} (\mathbf{y}_v - \boldsymbol{\mu}_i)}}{(2\pi)^{\frac{N^2}{8}} \det^{\frac{1}{2}} \left\{ \sigma_i^2 \mathbf{I}_{\frac{N^2}{4}} \right\}} = (2\pi\sigma_i^2)^{-\frac{N^2}{8}} e^{-\frac{\|\mathbf{y}_v - \boldsymbol{\mu}_i\|^2}{2\sigma_i^2}}$$

- Using this PDF under the proposed hypothesis, **the ML decision rule and hypotheses** can be **approximated** to:

$$\frac{p(\mathbf{y}_v; \mathcal{H}_1)}{p(\mathbf{y}_v; \mathcal{H}_0)} \underset{\mathcal{H}_0}{\overset{\mathcal{H}_1}{\gtrless}} 1, \quad \begin{cases} \mathcal{H}_0 : \frac{2\|\mathbf{y}_v - \boldsymbol{\mu}_0\|^2}{\sigma_0^2} \sim \chi_{\frac{N^2}{2}}^2, & \text{bit '0'}, \\ \mathcal{H}_1 : \frac{2\|\mathbf{y}_v - \boldsymbol{\mu}_1\|^2}{\sigma_1^2} \sim \chi_{\frac{N^2}{2}}^2, & \text{bit '1'}. \end{cases}$$

Approximation for BIT ERROR RATE (BER)

- Values for means of the underlying channel gains **are difficult to obtain** due to the practical constraints of the tags
- So, we set the value for these **means to be zero** for analytical tractability
- ML detection adopted at R, the **approximated** log-likelihood ratio to be set to zero and solved, is:

$$\log \left(\frac{p(\mathbf{y}_v; \mathcal{H}_0)}{p(\mathbf{y}_v; \mathcal{H}_1)} \right) \approx \frac{N^2}{8} \log \left(\frac{\sigma_1^2}{\sigma_0^2} \right) + \frac{\mathcal{Z}}{2\sigma_1^2} - \frac{\mathcal{Z}}{2\sigma_0^2} = 0$$

- Based on above conditionally defined key statistic, **the ML decision rule** can be approximated to:

$$\|\mathbf{y}_v\|^2 \underset{\mathcal{H}_0}{\overset{\mathcal{H}_1}{\gtrless}} \mathcal{Z}_{\text{th}} \triangleq \frac{N^2}{4} \left(\frac{\sigma_1^2 \sigma_0^2}{\sigma_1^2 - \sigma_0^2} \right) \log \left(\frac{\sigma_1^2}{\sigma_0^2} \right)$$

- Finally, the desired **BER expression can be approximated** as below:

$$p_{\text{be}} \triangleq \frac{1}{2} - \frac{\Gamma \left(\frac{N^2}{4}, \frac{\mathcal{Z}_{\text{th}}}{\sigma_1^2} \right) - \Gamma \left(\frac{N^2}{4}, \frac{\mathcal{Z}_{\text{th}}}{\sigma_0^2} \right)}{2 \Gamma \left(\frac{N^2}{4} \right)}$$

where $\Gamma(s, x) = \int_x^\infty t^{s-1} e^{-t} dt$ is upper incomplete gamma function with $\Gamma(s) = \Gamma(s, 0)$ being ordinary gamma function

OPTIMIZING BC DESIGN FOR MINIMIZING BER

- As BER is a function of **squared-magnitude of backscattering coefficients (BC)** -- a_0 and a_1
- We focus on optimizing the **real values of a_0 and a_1** because T can then set its BC to underlying complex number
- Optimization problem for **minimizing BER** can be outlined as:

$$\begin{aligned} \mathcal{OP}_1 : \operatorname{argmin}_{a_0, a_1} p_{be}, \quad & \text{subject to} \\ & \text{(C1) : } a_0, a_1 \geq 0, \quad \text{(C2) : } a_0^2 + a_1^2 \leq a_{ub} \end{aligned}$$

where a_{ub} is an **upper bound** on the reflection strength of BC at tag

Equivalent Univariate Transformation

- BER involves only two terms in the form of upper **incomplete gamma functions** which are dependent on BC variables (a_0 and a_1)
- Upper gamma function is **monotonically-decreasing** in its second argument
- Minimizing BER is equivalent to maximizing the **difference between these two upper incomplete gamma** terms, which yields to below:

$$\frac{Z_{\text{th}}}{\sigma_1^2} - \frac{Z_{\text{th}}}{\sigma_0^2} = \log \left(\frac{\sigma_1^2}{\sigma_0^2} \right) = \log \left(1 + \frac{(2a_1 a_0 + a_1^2) \beta^2}{\beta^2 a_0^2 + \sigma_I^2} \right)$$

- Utilizing the monotonicity of the logarithmic function along with the fact that highest BC strength has to be exploited for minimizing BER, we can rewrite an **equivalent single-variable optimization**:

$$\begin{aligned} \mathcal{OP}_2 : \quad & \underset{a_0}{\operatorname{argmax}} \quad \frac{\beta^2 \left(2a_0 + \sqrt{a_{\text{ub}} - a_0^2} \right) \sqrt{a_{\text{ub}} - a_0^2}}{\beta^2 a_0^2 + \sigma_u^2 + \frac{\sigma_w^2}{E_p}} \\ & \text{subject to} \quad (\text{C3}) : 0 \leq a_0 \leq a_{\text{ub}}. \end{aligned}$$

Globally-Optimal BC Design

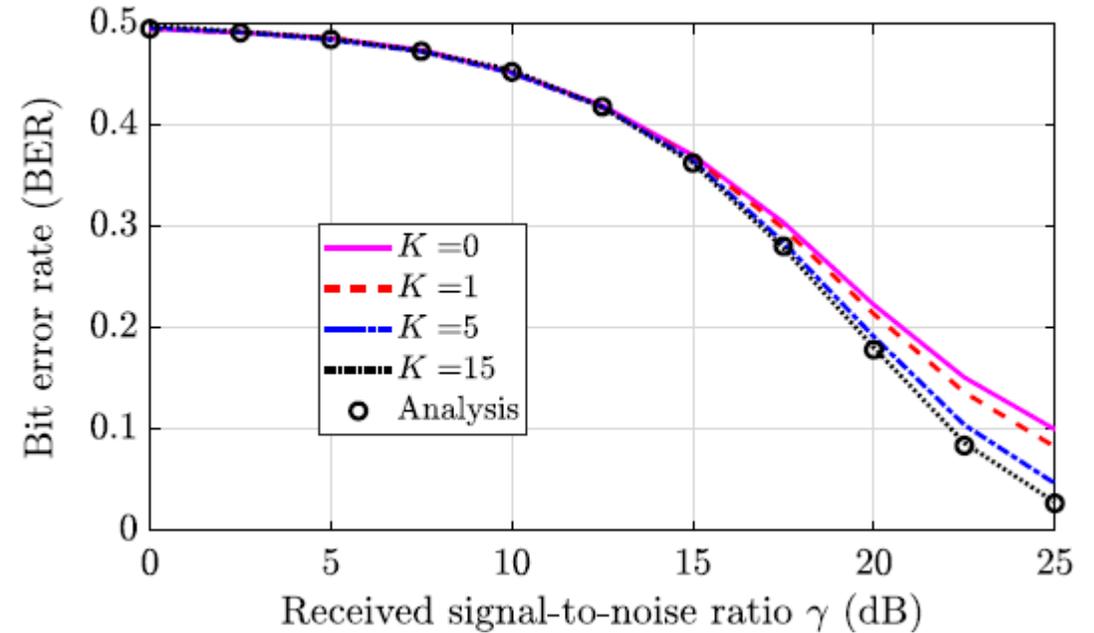
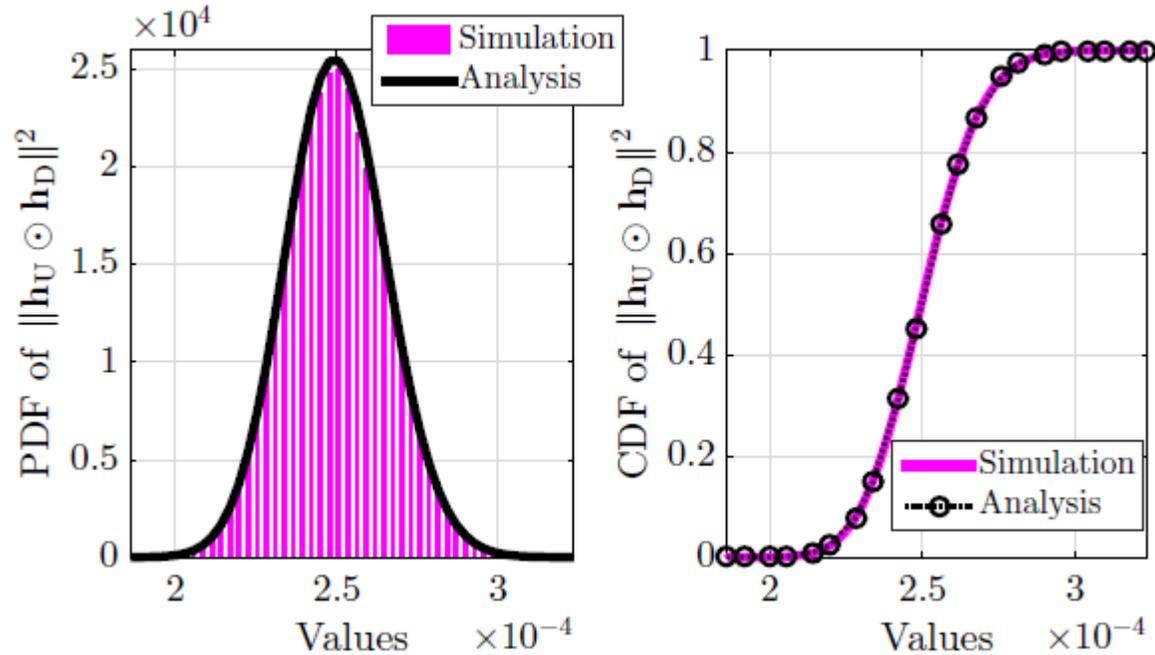
- Objective is **non-concave** in the variable \rightarrow Simplified optimization problem is non-convex
- Taking **partial derivative** of objective with respect to the variable α_0 and setting it to zero:

$$\sigma_I^2 (a_{\text{ub}} - a_0^2) - \left(a_0 \sqrt{a_{\text{ub}} - a_0^2} + a_0^2 \right) (\beta^2 a_{\text{ub}} + \sigma_I^2) = 0$$

- Above is a **Quartic equation** having two negative (or infeasible) real roots and two positive (or feasible) roots
- One of them leads to maxima and other to its minima
- **Globally-optimal solution** over the feasible region is given by:

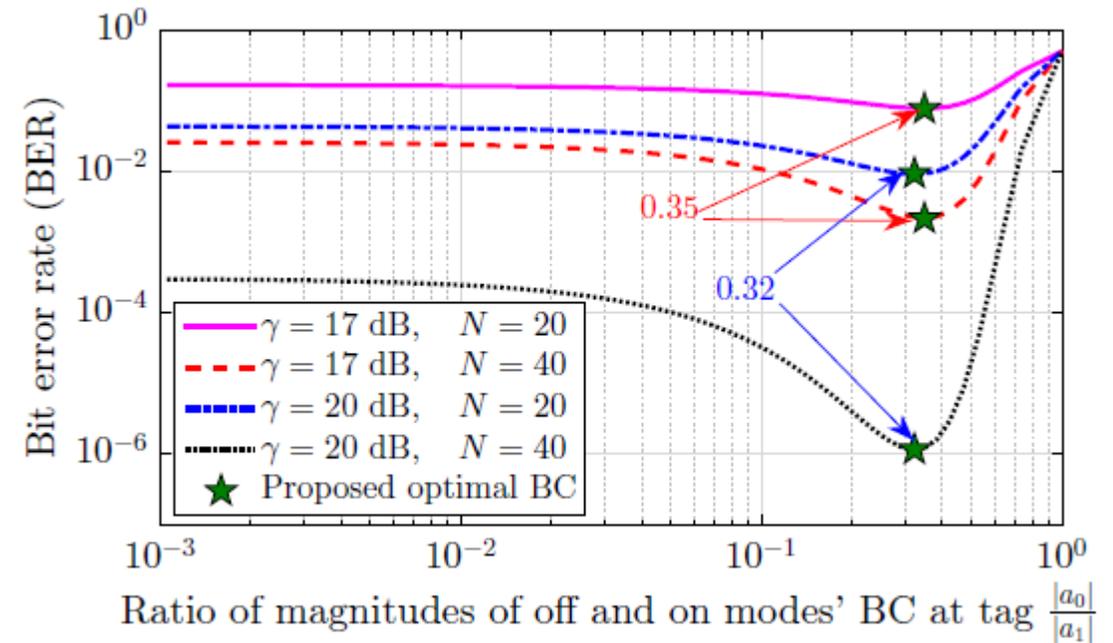
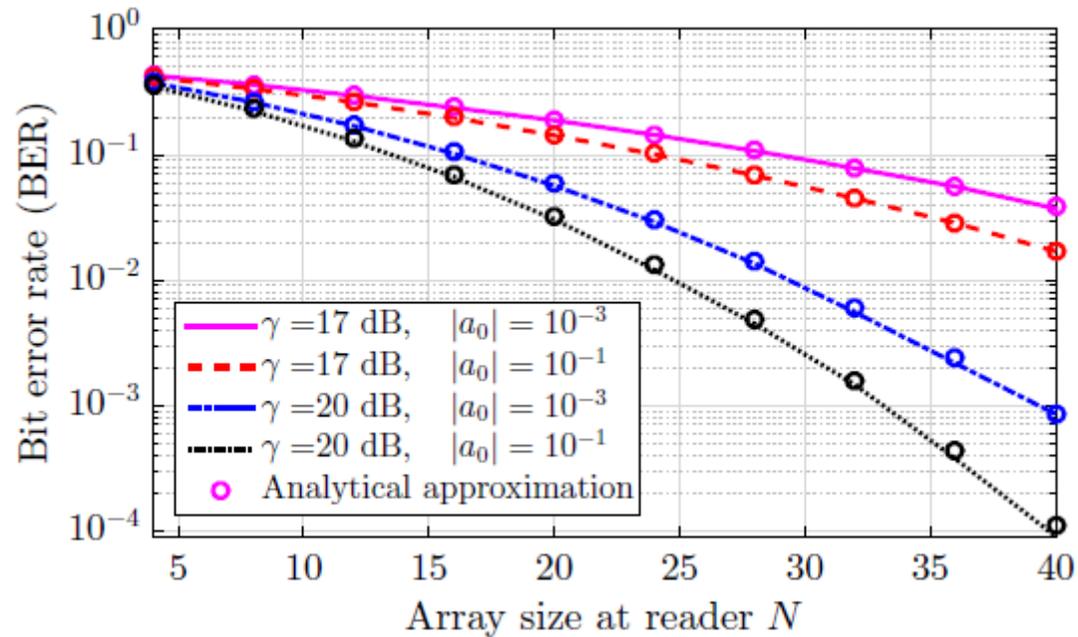
$$a_0^* \triangleq \sqrt{\frac{a_{\text{ub}} (\beta^4 a_{\text{ub}}^2 + \beta^2 a_{\text{ub}} (4\sigma_I^2 - \sqrt{\varpi}) + \sigma_I^2 (5\sigma_I^2 - \sqrt{\varpi}))}{2(\beta^4 a_{\text{ub}}^2 + \varpi)}} \quad \text{with } a_1^* \triangleq \sqrt{a_{\text{ub}} - (a_0^*)^2}$$

Validation of Analysis



- A **very close match** between the analytical and simulated results of both PDF and cumulative distribution function (CDF)
- It signifies the **goodness** of the proposed analytical approximation for distribution of $\|\mathbf{h}_U \odot \mathbf{h}_D\|^2$ for high Rice factor $K = 100$
- The **RMSE** value between the simulated and analytical results for BER over the considered range of SNR improves from 0.0336 to 0.0017 for $K = 0$ to $K = 15$
- Likewise, **R-square** statistics also respectively improves from 0.9924 to 0.9999
- This **verifies the goodness** of the proposed analytical approximation for BER

BER Performance Comparison



- **Average gap** between the simulation and analytical approximation results for BER over 200 kbits is **< 0.14dB**
- Smart selection of BC to $|a_0| = 10^{-1}$, a ten-fold increase in N can lead to **36dB improvement** in BER for SNR = 20dB
- Low BC value during off mode (or bit '0') transmission may **not necessarily** lead to a lower BER
- **Optimally-set BC** using proposed design can yield **33dB improvement in BER** over arbitrarily selected ones
- Optimal value of BC ratio $\frac{|a_0|}{|a_1|}$ gets **reduced** for higher SNR and is **independent** of the underlying array-size N at reader

Concluding Remarks

- A novel **ML-based backscattering detection protocol** for multiantenna reader-assisted monostatic BSC
- A three-phase transmission protocol and backscattering signal model that considers **practical BSC constraints** like UAR, strong LoS component, and tag's resource-limitations
- Exploiting specific BSC features to come up with **tight analytical approximation for BER**
- **Closed-form globally-optimal BC design** at the tag that can provide significant improvement in BER
- **Nontrivial insights** from system engineering perspectives
- Observations can be used for designing **sustainable low-power next-generation networks**

Thank you for your attention!

For questions and feedback, please contact:
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