

HIGHLIGHTS

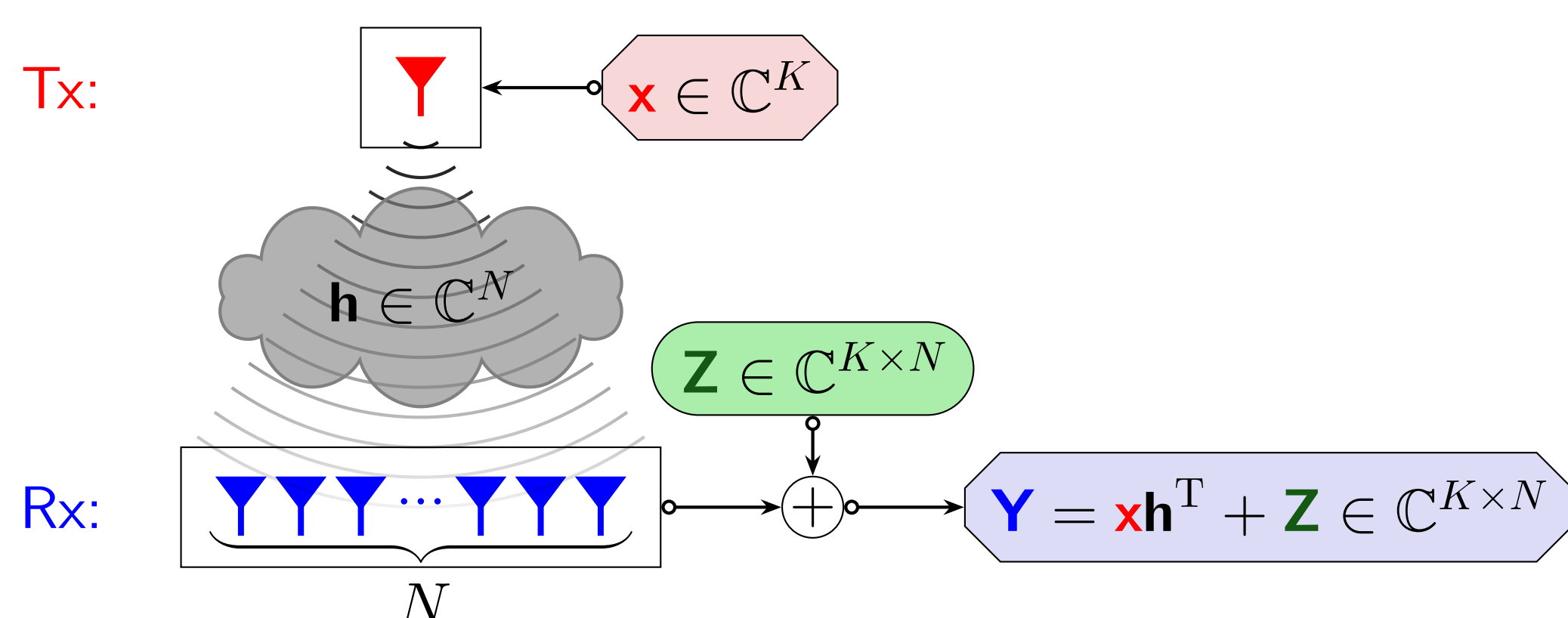
- **Objective:** To characterize statistical properties of the received signal in a noncoherent massive SIMO setting.
- **Approach:**
 - Vectorizing the time-space signal reveals a cyclostationary (CS) process.
 - Its asymptotic structure can be effectively exploited in the Karhunen-Loève (KL) domain.
- **Outcome:** This spectral representation provides insights on noncoherent detection with large arrays and unveils the fundamental parameters that play a role in it, while pointing at simplified implementations.

BACKGROUND

- It is agreed among the scientific community that future wireless communications will be built upon massive MIMO technologies.
- In many scenarios, the acquisition of reliable instantaneous channel state information (CSI) becomes cumbersome and can degrade spectral efficiency of coherent systems.
- On the contrary, noncoherent communications do not require instantaneous CSI, instead relying on statistical knowledge of the channel.
- To effectively exploit this *statistical CSI*, the receiver can leverage various properties of fading, such as its spatial stationarity.
- Through the use of massive arrays, we access the asymptotic spectral properties of noncoherent SIMO systems, mirroring well-known ideas from time-series analysis in the spatial domain.
- We develop a technique to uncorrelate received data and distill the signal properties that are essential to the detection problem.

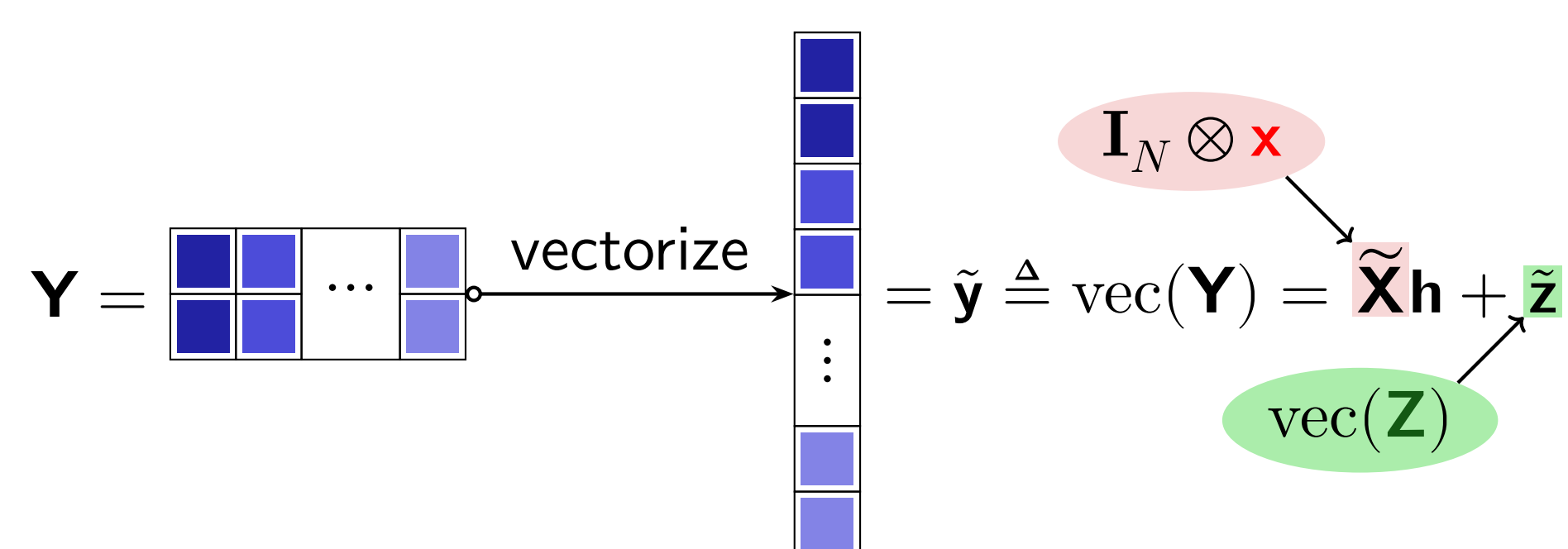
PROBLEM STATEMENT

- Point-to-point SIMO, Rx has N antennas and statistical CSI.
- Correlated Rayleigh channel ($\mathbf{h} \sim \mathcal{CN}(\mathbf{0}_N, \mathbf{C}_h)$), block flat fading, coherence time K .
- Tx sends an equiprobable codeword $\mathbf{x}_i \in \mathbb{C}^K$ from a finite alphabet \mathcal{X} during a coherence block. For convenience, we parameterize it as $\mathbf{x}_i \triangleq \sqrt{K}\gamma_i \mathbf{u}_i$.
- Additive noise \mathbf{Z} with i.i.d. components $[\mathbf{Z}]_{r,c} \sim \mathcal{CN}(0, P_Z)$.
- Received signal in complex baseband time-space notation:



Maximum likelihood (ML) detection:

- To derive the (unconditional) likelihood function of the received signal, we must *vectorize* \mathbf{Y} columnwise:



- The resulting received signal is $\tilde{\mathbf{y}}|\mathbf{x}_i \sim \mathcal{CN}(\mathbf{0}_{KN}, \mathbf{C}_i \triangleq \mathbf{C}_h \otimes \mathbf{x}_i \mathbf{x}_i^H + P_Z \mathbf{I}_{KN})$.
- ML error probability is commonly bounded by the simpler pairwise error probability between codewords \mathbf{x}_a and \mathbf{x}_b : $P_{a \rightarrow b} \triangleq \Pr\{L_{a,b}(\tilde{\mathbf{y}}) \leq 0 | \mathbf{x} = \mathbf{x}_a\}$. The term $L_{a,b}$ is the log-likelihood ratio (LLR) between hypotheses:

$$L_{a,b}(\tilde{\mathbf{y}}) \triangleq \ln \frac{f_{\tilde{\mathbf{y}}|\mathbf{x}_a}(\tilde{\mathbf{y}})}{f_{\tilde{\mathbf{y}}|\mathbf{x}_b}(\tilde{\mathbf{y}})} = \tilde{\mathbf{y}}^H (\mathbf{C}_b^{-1} - \mathbf{C}_a^{-1}) \tilde{\mathbf{y}} + \ln \frac{|\mathbf{C}_b|}{|\mathbf{C}_a|}. \quad (1)$$

- The most relevant signal characteristics in the detection problem are not directly recognizable from the time-space formulation (1).

💡 The fundamental structure of the problem will emerge if we represent (1) in the KL domain asymptotically ($N \rightarrow \infty$).

Statistical properties of $\tilde{\mathbf{y}}$:

- Assuming channel fading is *spatially stationary* (Figure 1), the resulting process $y_i(n) \triangleq [\tilde{\mathbf{y}}|\mathbf{x}_i]_n$ will display a CS behavior of period K (Figure 2).

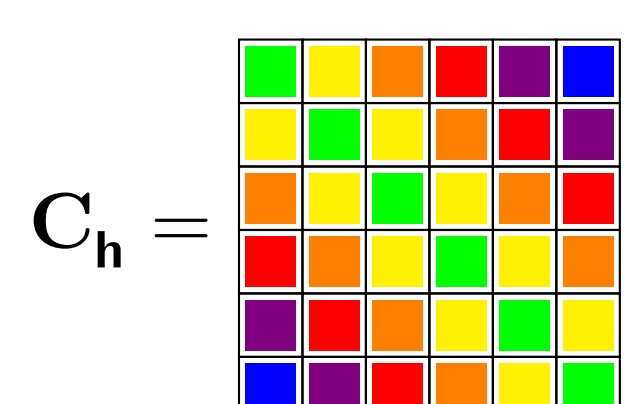


Figure 1. Toeplitz covariance matrix corresponding to a stationary process.

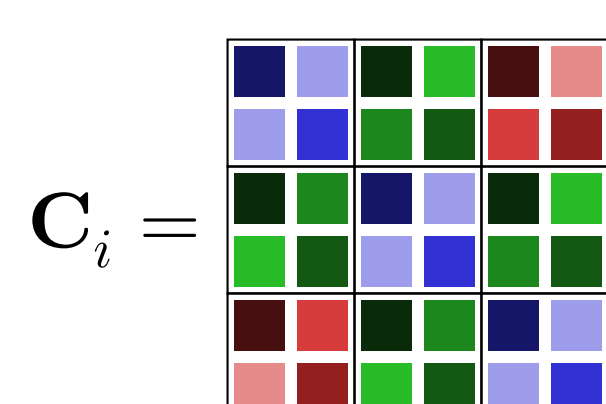


Figure 2. K -Toeplitz ($K=2$) covariance matrix corresponding to a CS process.

SPECTRAL ANALYSIS OF THE RECEIVED SIGNAL

- A CS process $y_i(n)$ of period K can be expanded onto its KL basis $\{\phi_i^{(k)}(n, \sigma)\}$ as:

$$y_i(n) \triangleq \sum_{k=0}^{K-1} \int_0^{\frac{1}{K}} \phi_i^{(k)}(n, \sigma) dy_i^{(k)}(\sigma), \quad (2)$$

where $dy_i^{(k)}(\sigma)$ are uncorrelated increments in the KL spectral domain.

- In [1], [2] the asymptotic KL basis of $y_i(n)$ was derived in terms of its *cyclic spectrum matrix (CSM)*, $\mathbf{S}_i(\sigma) \in \mathbb{C}^{K \times K}$. It can be constructed from its spectral correlation in the Cramér-Loève (CL) domain, as shown in Figure 3.

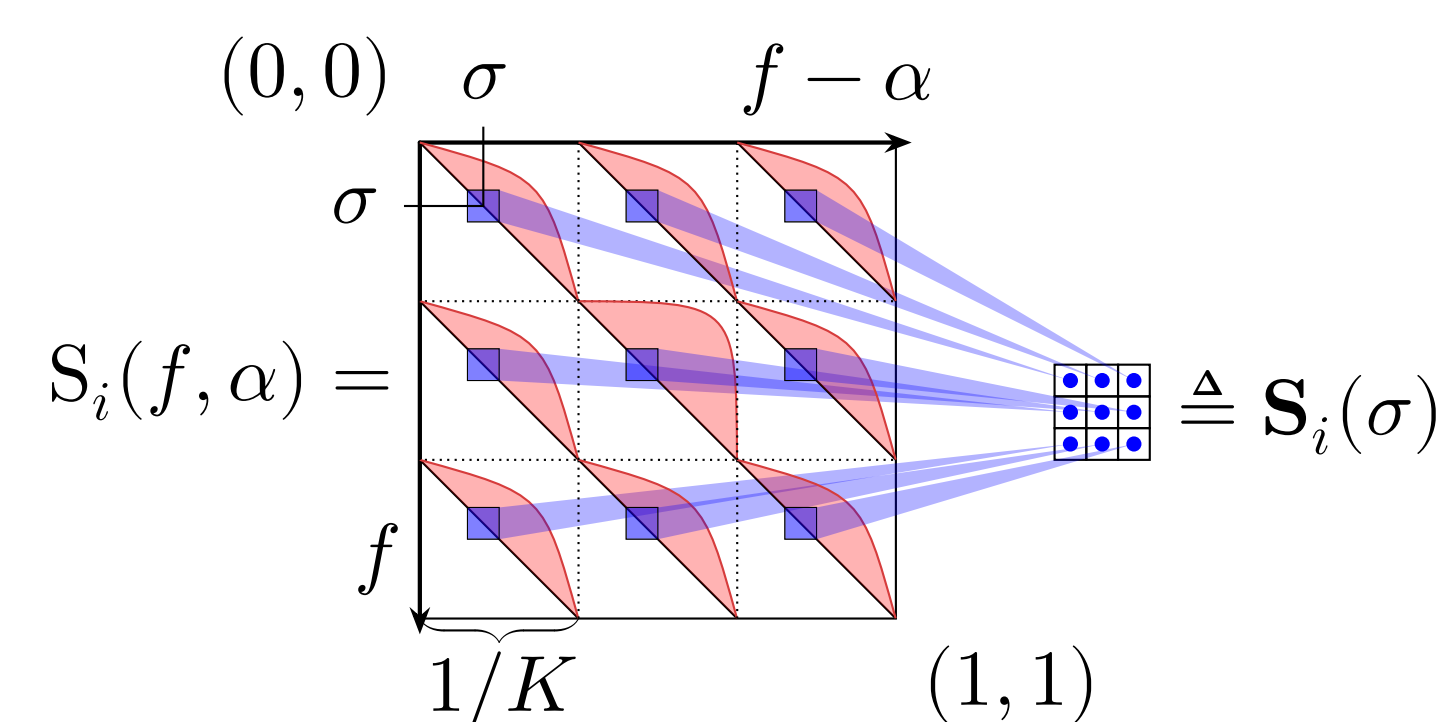


Figure 3. Graphical representation of the construction of $[\mathbf{S}_i(\sigma)]_{r,c}$ from the spectral correlation of $y_i(n)$ at frequencies $(\sigma + \frac{r}{K}, \sigma + \frac{c}{K})$, i.e. $S_i(\sigma + \frac{r}{K}, \frac{r-c}{K})$. It is only defined across δ -ridges spaced by $1/K$ both horizontally and vertically.

- Using these ideas, we can express the LLR (1) as $N \rightarrow \infty$ in the KL domain:

$$L_{a,b}(d\mathbf{y}_i) = \int_0^{\frac{1}{K}} (d\mathbf{y}_i^H \Psi_i^H (\mathbf{D}_b^{-1} - \mathbf{D}_a^{-1}) \Psi_i d\mathbf{y}_i)(\sigma) + \ln \frac{|\mathbf{D}_b(\sigma)|}{|\mathbf{D}_a(\sigma)|} d\sigma \quad (3)$$

where $d\mathbf{y}_i(\sigma) \triangleq [dy_i^{(0)}(\sigma), dy_i^{(1)}(\sigma), \dots, dy_i^{(K-1)}(\sigma)]$ and $\Psi_i \triangleq [\mathbf{u}_i \mathbf{U}_i] \in \mathbb{C}^{K \times K}$.

- This representation displays the fundamental structures involved in the detection problem; mainly

$$\mathbf{D}_\diamond(\sigma) \triangleq \gamma_\diamond \mathbf{S}_h(K\sigma) \mathbf{K} \mathbf{P}_\diamond + P_Z \mathbf{I}_K, \quad \diamond = a, b. \quad (4)$$

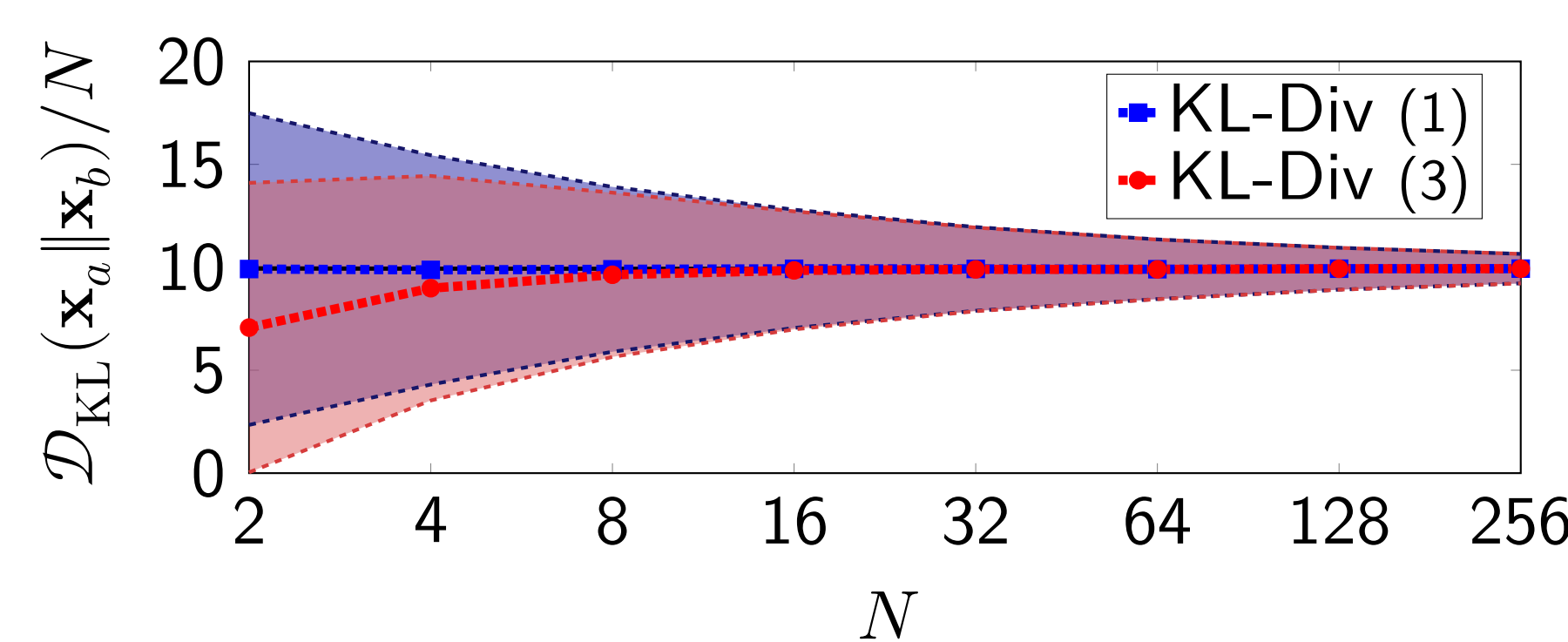


Figure 4. Normalized KL-Div estimated from (1) and (3) by averaging 10^4 Monte Carlo rounds.

LIKELIHOOD RATIO STRUCTURE ANALYSIS

- Using the relationship between the CL and KL representations of $y_i(n)$ [2], expression (3) can be decomposed into four clear constituents:

$$L_{a,b}(d\mathbf{y}_i) = \ell_{a,b}(\mathbf{x}_i) + \ell_{a,b}(\mathbf{z}) + \ell_{a,b}(\mathbf{x}_i, \mathbf{z}) + \kappa_{a,b} \quad (5)$$

➔ **Signal term:** $E[\ell_{a,b}(\mathbf{x}_i)] = \int_0^{\frac{1}{K}} \mathbf{u}_i^H (|\rho_a(\sigma)|^2 \mathbf{P}_a - |\rho_b(\sigma)|^2 \mathbf{P}_b) \mathbf{u}_i \Gamma_i(\sigma) d\sigma$

➔ **Noise term:** $E[\ell_{a,b}(\mathbf{z})] = \int_0^{\frac{1}{K}} (|\rho_a(\sigma)|^2 - |\rho_b(\sigma)|^2) d\sigma$

➔ **Signal-noise term:** $E[\ell_{a,b}(\mathbf{x}_i, \mathbf{z})] = 0$

➔ **Constant term:** $\kappa_{a,b} = \int_0^{\frac{1}{K}} \ln \frac{\Gamma_i(\sigma)+1}{\Gamma_i(\sigma)+1} d\sigma$

- Terms $\Gamma_\diamond(\sigma)$ and $|\rho_\diamond(\sigma)|^2$ are, respectively, the SNR and *squared spectral coherence* of process $y_\diamond(n)$ at point σ in the KL spectrum.

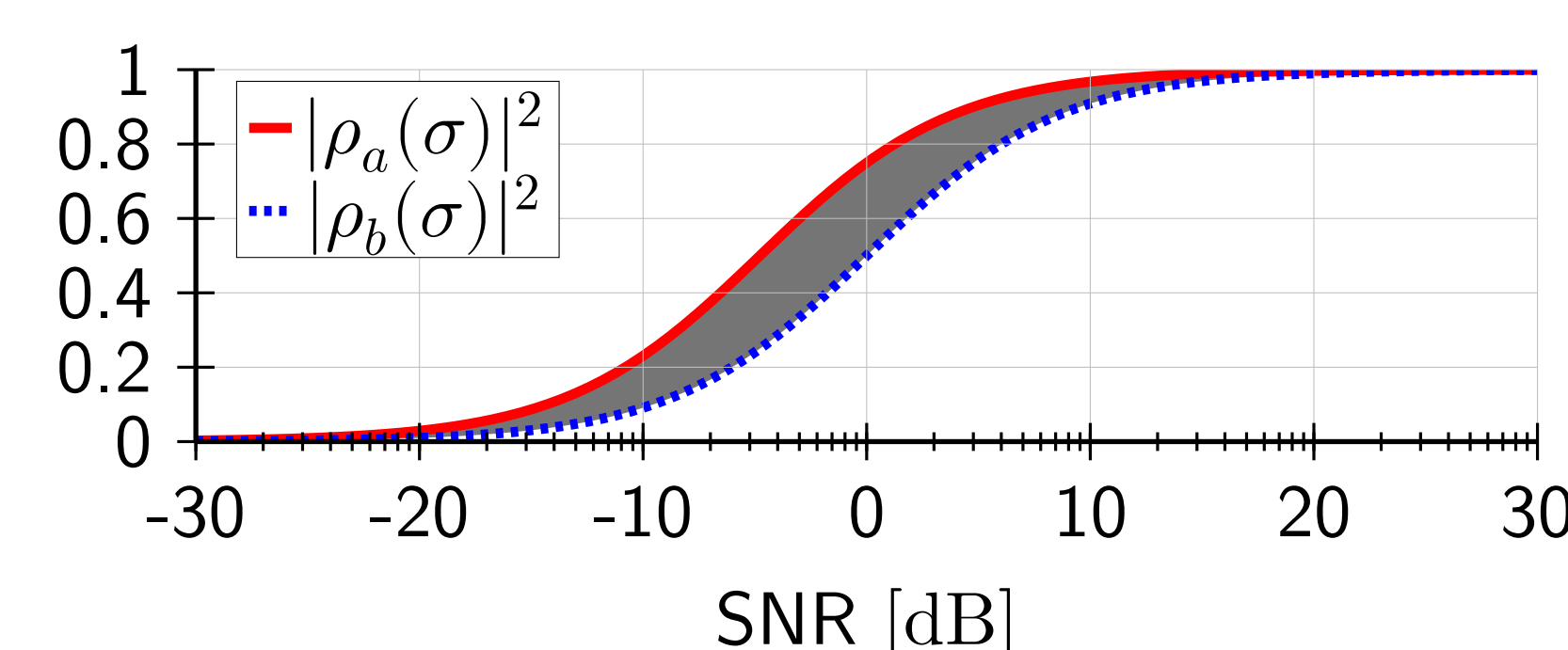


Figure 5. Squared spectral coherences in terms of SNR.

References

- [1] J. Riba and M. Vila, "On infinite past predictability of cyclostationary signals," *IEEE Signal Processing Letters*, vol. 29, pp. 647–651, 2022. DOI: 10.1109/1sp.2022.3149705.
- [2] M. Vilà-Insa and J. Riba, "Asymptotic analysis of synchronous signal processing," *preprint [arXiv:2403.18445]*, 2024. arXiv: 2403.18445 [eess.SP].

Funding

This work was funded by project MAYTE (PID2022-136512OB-C21) by MICIU/AEI/10.13039/501100011033 and ERDF/EU, grant 2021 SGR 01033 and grant 2022 FI SDUR 00164 by Departament de Recerca i Universitats de la Generalitat de Catalunya.

