

Coding for Sensing

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Traditional Radar and Communications Separation

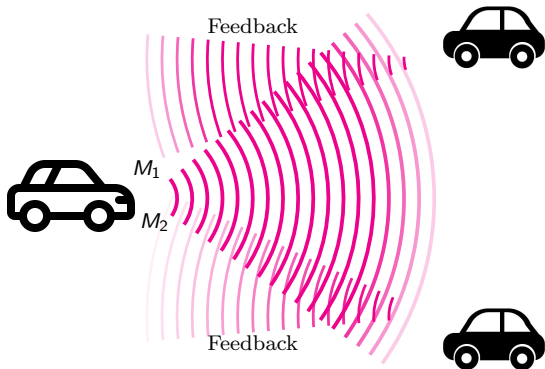
Communication

Sensing

Conventional approach: Resource splitting

Integrated Communication and Sensing

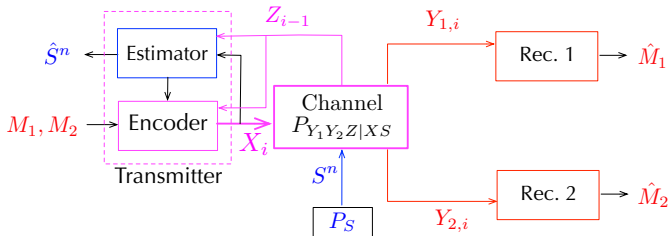
Communication and Sensing



- Synergistic Waveform for Sensing and Comm
- Coding for Sensing

Information Theoretic Model — The Channel

- State-dependent memoryless channels with generalized feedback



- Arbitrary forward and backward channels $P_{Y_1 Y_2 | X S}$ and $P_{Z | Y_1 Y_2 X S}$
- Model includes as special cases
 - Memoryless fading channels $Y_k = SX + N_k$ or $Y_k = \tilde{S}X + N_k$
 - Receiver CSI: Y_k can include S or imperfect versions of S
 - Generalized feedback can be $Z = SX + W_k$ or $Z = Y_k + W_k$ or ..

- Sensing Performance Measured by Average Block-Distortion:

$$\overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[d(S_i, \hat{S}_i)] \leq D.$$

- Arbitrary distortion functions $d(\cdot, \cdot)$, even **functions** of the state
- Can handle **vector states** $S = (S_1, S_2)$, e.g., 2 objects/environments
- Extends to multiple distortions:
 $\overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[d_k(S_i, \hat{S}_i)] \leq D_k, \quad k = 1, 2, \dots$

Fundamental Capacity-Distortion Tradeoff for BC

Definition

A rate-distortion tuple (R_1, R_2, D) is achievable if there exist encoder, decoders and estimator s.t.

$$\Pr \left(\hat{M}_1 \neq M_1 \quad \text{or} \quad \hat{M}_2 \neq M_2 \right) \rightarrow 0 \quad \text{as} \quad n \rightarrow \infty$$

and

$$\overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \mathbb{E}[d(S_i, \hat{S}_i)] \leq D$$

Optimal Estimator for BC

Lemma

The optimal estimator operates symbolwise on (X^n, Z^n)

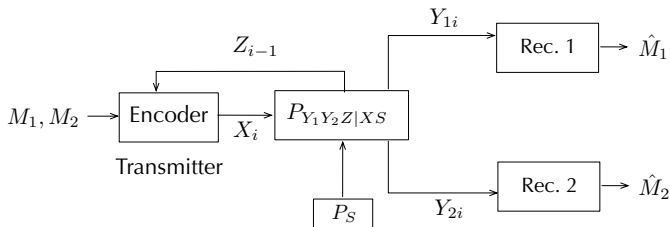
$$h^*(x^n, z^n) := (\hat{s}^*(x_1, z_1), \hat{s}^*(x_2, z_2), \dots, \hat{s}^*(x_n, z_n)),$$

where the optimal per-symbol estimator is

$$\hat{s}^*(x, z) := \arg \min_{s' \in \hat{\mathcal{S}}} \sum_{s \in \mathcal{S}} P_{S|XZ}(s|x, z) d(s, s').$$

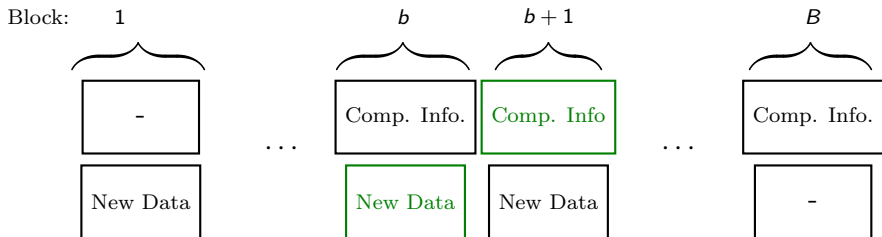
- Given (X^n, Z^n) , min. distortion $D_{\min} = \frac{1}{n} \sum_{i=1}^n \mathbb{E}[d(S_i, \hat{s}^*(X_i, Z_i))]$
- Finding an optimal integrated **sensing & comm. scheme** reduces to finding optimal **comm. scheme** with desired X^n statistics
- **Tradeoff between comm and sensing stems from X^n statistics**

State-Dependent BCs with Generalized Feedback



- Feedback does not increase capacity of degraded BCs (El Gamal,'79)
- Achievable scheme for general BCs (Shayevitz et al'12, Venkataramanan et al'13)
- Capacity of several BCs with full Receiver-CSI (Kim et al'16)

Intuition about the Shayevitz-Wigger BC Scheme



- 1 Block-Markov strategy:
 - Compression info sent in block $b + 1$: info about channel in block b learned via feedback
 - Block- b outputs improved with compression info sent in block $b + 1$
- 2 New data and compression info sent with Marton's BC scheme (without feedback)

Results on Capacity-Distortion Region of BCs

Degraded Broadcast Channels $X \rightarrow Y_1 \rightarrow Y_2$

Capacity-distortion region \mathcal{CD} : set of triples (R_1, R_2, D) satisfying

$$\begin{aligned}R_1 &\leq I(X; Y_1 | U) \\R_2 &\leq I(U; Y_2), \\ \mathbb{E}[d(S, \hat{s}^*(X, Z))] &\leq D\end{aligned}$$

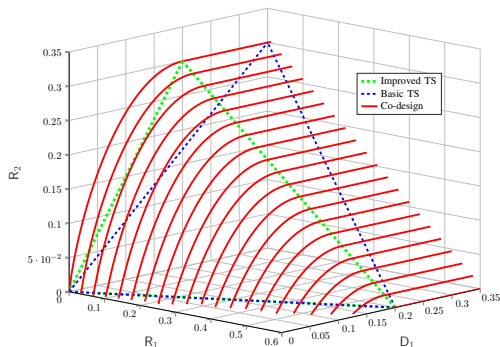
for some pmf $P_{UX}P_S P_{Y_1 Y_2 Z|XS}$.

General Broadcast Channels

Inner and outer bounds (feasible and infeasible regions) based on Shayevitz-W. scheme and genie-aided bound

Binary Fading Example: Capacity-Distortion Region

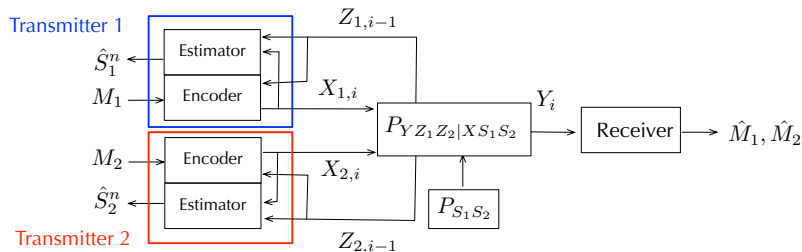
- Double-State $S = (S_1, S_2)$ with corr. components, known at Rxs!
- Fading outputs $Y_k = S_k X$, for $k = 1, 2$ (without noise)
- Perfect Rx CSI and both outputs fed back $Z = (Y_1, Y_2)$
- When $X = 1$ Tx learns S_1, S_2 ; when $X = 0$ it learns nothing



- Resource/time-sharing approaches sub-optimal
- Tradeoff between optimal sensing and comm. performances

Multiple-Access Channels

- Two Tx's sense (correlated) states and also send data messages
- Feedback allows for communication/coordination between Tx's



- Sensing at Tx k based only on (X_k^n, Z_k^n) is suboptimal!
- Use coding not only for communication but also for sensing!

Coding for Sensing – A Toy Example

- Via feedback, Tx 1 can get info about S_2 that Tx 2 does not get
→ use coding and send it to Tx 2 through the feedback link
- Example: $Y = S_2 X_2$ $Z_1 = S_2$ $Z_2 = X_1$

Can achieve distortion 0 at Tx 2 if Tx 1 repeats its feedback

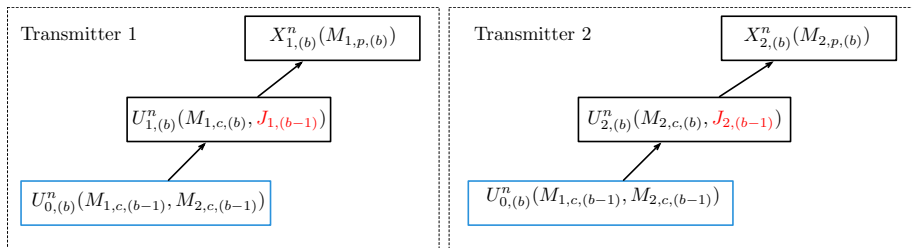
$$X_{1,i} = Z_{1,i-1} = S_{2,i-1} \implies Z_{2,i} = S_{2,i-1}$$

Without Tx 1 repeating the feedback this is not possible!

Coding for sensing can help if Tx 1 has info. on S_2 that is not available at Tx 2 (or vice versa)

Coding Scheme

- Extends [Willems'83] scheme (MAC with generalized feedback)
- Block-Markov coding and backwards decoding
- Feedback allows to create cooperative U_0 -codeword
- **Txs exchange** message parts and **compression info for sensing**



- $J_{k,(b)}$: compression info about $(Z_{k,(b)}^n, X_{k,(b)}^n, U_{k,(b)}^n)$

Results on MAC with Integrated Sensing and Communications

- New idea of coding-for-sensing
- Improved achievable region (better sensing performance) over state of the art
- Capacity-distortion region for the MAC remains an open problem

Summary

- Presented information-theoretic framework for integrated sensing and communication (Kobayashi,Caire,Kramer'18)
- Single Tx: sensing performance depends only on X^n statistics, i.e., on the chosen waveform
- Tradeoff between rates and distortion(s)
- Multiple Txs: *fully integrated coding for sensing and comm.* → we included coding for sensing in Willems'83 scheme
- Interesting future research direction: channels with memory