

# Integrated Sensing and Communication (ISAC)

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# Traditional Sensing and Communications Separation

Communication



Sensing

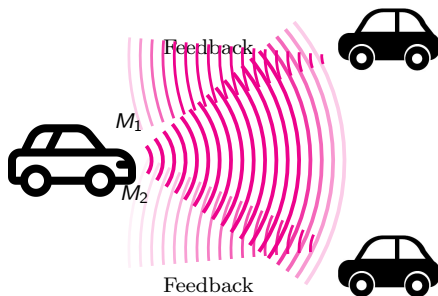


## Conventional approach

- Individual hardware with own antenna and own RF chain for each of the two tasks
- Separate bandwidths for the two tasks

# Integrated Sensing and Communication (ISAC)

## Sensing and Communication



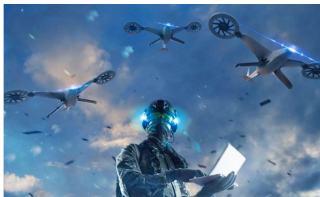
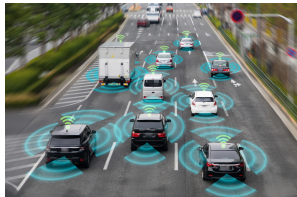
- Synergistic hardware, bandwidth, and waveform performing both tasks: Sensing and Communications

# Motivation for Integrating Sensing and Communication

The most immediate benefits of ISAC:

- Cellular communication move up in frequencies, even to the THz regime  
→ radar and cellular communication occupy similar bandwidths
- Integrating radar and communication will allow to free up precious bandwidth
- Savings in hardware costs, resources, and energy consumption

# Important Use Cases



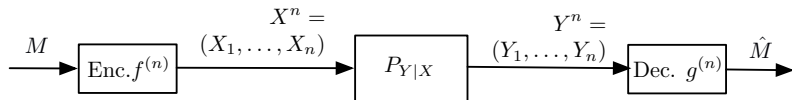
# Current Status of ISAC

- Predicted to be crucial building block of future 6G networks
- Heavily investigated in the communications and signal processing societies
- First prototypes available

## Information-theoretic angle of attack

Determine the optimal performances of ISAC systems. And the inherent tradeoffs between sensing and communications

# The Discrete Memoryless Channel (DMC)



- Discrete-time and stationary memoryless channel law:

$$\mathbb{P}[Y_t = y | X^t = x^t, Y_{t-1} = y^{t-1}] = P_{Y|X}(y|x_t)$$

- Finite input and output alphabets  $\mathcal{X}$  and  $\mathcal{Y}$
- Error probability  $P_e^{(n)} = \mathbb{P}[\hat{M} \neq M]$

# A Refined Shannon Theorem

*What Shannon says:*

- There is no family of encodings/decodings  $\{(f^{(n)}, g^{(n)})\}_{n=1}^{\infty}$  of rate  $R > C$  such that  $P_e^{(n)} \rightarrow 0$  as  $n \rightarrow \infty$
- For any rate  $R < C$  there does exist a family of encodings/decodings  $\{f^{(n)}, g^{(n)}\}_{n=1}^{\infty}$  s.t.  $P_e^{(n)} \rightarrow 0$  as  $n \rightarrow \infty$

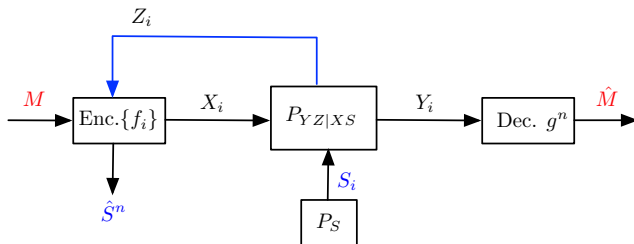


# A Refined Shannon Theorem

*A stronger version:*

For any distribution  $P_X$  over  $\mathcal{X}$ :

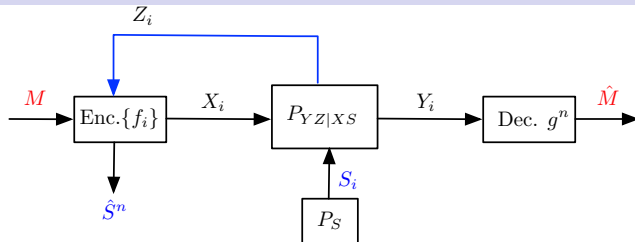
- There is no family of encodings/decodings  $\{(f^{(n)}, g^{(n)})\}_{n=1}^{\infty}$  of rate  $R > I(X; Y)$  and with codebook statistics  $P_X$  such that  $P_e^{(n)} \rightarrow 0$  as  $n \rightarrow \infty$
- For any rate  $R < I(X; Y)$  there does exist a family of encodings/decodings  $\{(f^{(n)}, g^{(n)})\}_{n=1}^{\infty}$  with codebook statistics  $P_X$  s.t.  $P_e^{(n)} \rightarrow 0$  as  $n \rightarrow \infty$



- state sequence  $S^n = (S_1, \dots, S_n)$  i.i.d.  $\sim P_S$
- Behaviour of the channel depends on the state  $S^n$  (for example the acceleration of an object)
- Sensing Performance measured by Average Block-Distortion:

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

# Distortion as a Sensing Performance



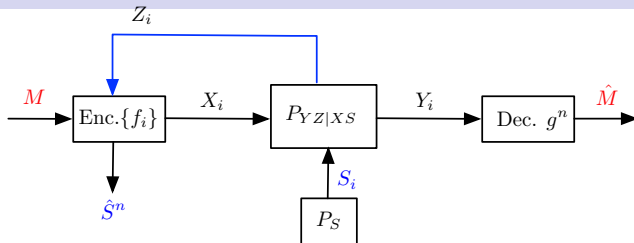
- Sensing Performance Measured by Average Block-Distortion:

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- Examples of distortion measures

- Mean-Squared Error  $d(s, \hat{s}) = (s - \hat{s})^2$
- Hamming weight  $d(s, \hat{s}) = \mathbb{1}\{s \neq \hat{s}\}$
- Distortion on a function of the state  $d(s, \hat{s}) = d'(f(s), \hat{s})$

# Information-Theoretic Fundamental Limit



## Definition

Capacity-distortion tradeoff  $C(D)$  is largest rate  $R$  such that there exist encoders, decoders and estimators with

$$\Pr(\hat{M} \neq M) \rightarrow 0 \quad \text{as} \quad n \rightarrow \infty$$

and

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

# Capacity-Distortion Tradeoff $C(D)$

Theorem (Kobayashi et al.)

*Capacity-distortion tradeoff*

$$C(D) := \max I(\textcolor{red}{X}; Y)$$

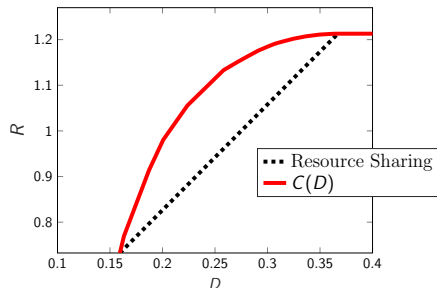
where maximum is over  $P_{\textcolor{red}{X}}$  satisfying

$$\mathbb{E}[d(S, \hat{s}^*(\textcolor{red}{X}, Z))] \leq D.$$

- Tradeoff between communication and sensing stems from  $P_{\textcolor{red}{X}}$
- Generalized feedback not used for coding. Simple point-to-point codes are sufficient. It suffices to adjust input pmf  $P_{\textcolor{red}{X}}$  to desired sensing performance.

## Ex.: Rayleigh Fading Channel

- Standard Gaussian state and noises  $S, N, N_{\text{fb}}$
- Rayleigh fading channel  $Y' = SX + N$
- Rx observes  $Y = (Y', S)$  and Tx  $Z = Y' + N_{\text{fb}}$
- Input power constraint  $P = 10\text{dB}$
- Quadratic distortion  $d(s, \hat{s}) = (s - \hat{s})^2$ .



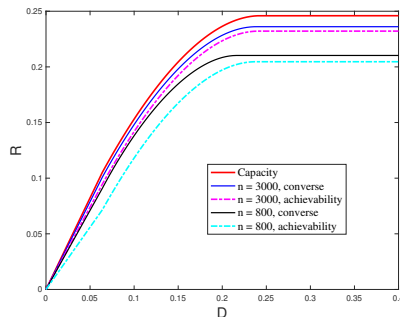
- $X \sim \mathcal{N}(0, P)$  achieves capacity
- $X \pm \sqrt{P}$  optimal for sensing

# The Finite Blocklength Regime

- Given blocklength  $n$ , triple  $(R, D, \epsilon)$  is called achievable if  $\exists$  encoder, decoder, and estimator with

$$\Pr(\hat{M} \neq M) \leq \epsilon \quad \text{and} \quad \frac{1}{n} \sum_{i=1}^n \mathbb{E}[d(S_i, \hat{S}_i)] \leq D$$

- Can reuse the optimal estimator  $s^*(x, z)$  from the capacity-problem!
- Ex.:  $Z = Y = XS$  and  $S \sim \mathbb{B}(0.4)$  and  $\epsilon = 10^{-3}$



# Information-Theoretic Finite-Blocklength Bounds

## Theorem

Given  $n$ . Triple  $(R, D, \epsilon)$  is achievable if  $\exists P_X$  and  $K > 0$  s.t.:

$$R \leq I(\textcolor{red}{X}; Y) - \sqrt{\frac{V}{n}} \mathbb{Q}^{-1}(\epsilon - \beta_u) - K \frac{\log(n)}{n}, \quad (1)$$

$$D \geq \mathbb{E}[d(S, \hat{s}^*(\textcolor{red}{X}, Z))] \quad (2)$$

with  $\beta_u := \frac{1}{n^K} + \frac{0.7975T}{\sqrt{nV^3}}$  and  $V / T$  the 2nd / 3rd cent. mom. of  $i(\textcolor{red}{X}; Y)$ .  
Triple  $(R, D, \epsilon)$  not achievable if  $\forall \delta > 0$  and pmfs  $P$  satisfying (2):

$$R \geq I(\textcolor{red}{X}; Y) - \sqrt{\frac{V}{n}} \mathbb{Q}^{-1}(\epsilon + \beta_l) + \frac{\log(n)}{2n} - \frac{\log \delta}{n}, \quad (3)$$

where  $\beta_l := \frac{6T}{\sqrt{nV^3}} + \frac{\delta}{\sqrt{n}}$ .

[3] H. Nikbakht et al., “Integrated Sensing and Communication in the Finite Blocklength Regime”, ISIT 2024.



# Beyond the Memoryless Assumption

- *arbitrary* state sequence  $S^n = (S_1, \dots, S_n)$  (for each  $n$ )
- No feedback coding  $X^n = f^n(M)$
- *arbitrary* channel law  $P_{Z^n Y^n | X^n S^n}$  (for each  $n$ )
- *general* distortion constraint  $\overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \mathbb{E}[d(S^n, \hat{S}^n)] \leq D$

## Theorem (Capacity-distortion tradeoff)

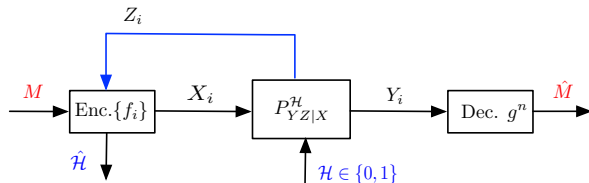
$$C(D) := \sup_{\{P_{X^n}\}_n} p - \underline{\lim}_{n \rightarrow \infty} \frac{1}{n} i(X^n; Y^n)$$

where supremum over all  $\{P_{X^n}\}$  s.t.  $\overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \mathbb{E}[d(S^n, \hat{S}^n(X^n, Z^n))] \leq D$

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[4] Chen et al. “On general capacity-distortion formulas of integrated sensing and communication,” Arxiv 2023.

# ISAC Models with Detection Exponents



- Single sensing parameter  $\mathcal{H} \in \{0, 1\}$  constant for all times
- Sensing performance measured in detection-error exponents

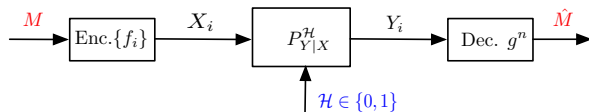
- Symmetric detection exponent:

$$E_{\text{Sym}} := \lim_{n \rightarrow \infty} -\frac{1}{n} \log \left( \max \left\{ \mathbb{P} \left[ \hat{\mathcal{H}} = 1 \mid \mathcal{H} = 0 \right], \mathbb{P} \left[ \hat{\mathcal{H}} = 0 \mid \mathcal{H} = 1 \right] \right\} \right)$$

- Stein's exponent requires

$$\lim_{n \rightarrow \infty} \mathbb{P} \left[ \hat{\mathcal{H}} = 0 \mid \mathcal{H} = 1 \right] \leq \epsilon$$
$$\theta_{\text{Stein}} := \lim_{n \rightarrow \infty} -\frac{1}{n} \log \mathbb{P} \left[ \hat{\mathcal{H}} = 0 \mid \mathcal{H} = 1 \right]$$

# Relation to a Compound Channel

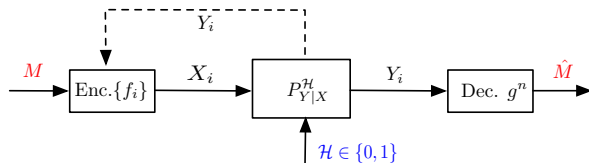


- Without the sensing it is a compound channel
- Compound capacity without feedback:

$$C_{\text{compound}} \leq \max_{P_X} \min_{\mathcal{H}} I(X; Y | \mathcal{H})$$

Cannot adjust codebook statistics  $P_X$  to the channel  $\mathcal{H}$ !

# Relation to a Compound Channel



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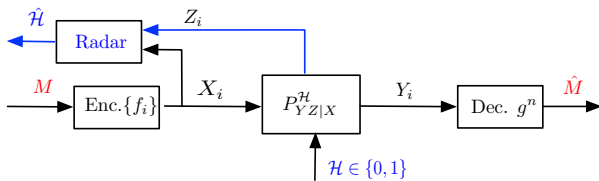
Cannot adjust codebook statistics  $P_X$  to the channel  $\mathcal{H}$ !

- Compound capacity with feedback:

$$C_{\text{compound,fb}} \leq \min_{\mathcal{H}} \max_{P_X} I(X; Y|\mathcal{H}) = \min_{\mathcal{H}} C(P_{Y|X}^{\mathcal{H}}).$$

Can learn channel and adapt codebook statistics!

# ISAC Model with Stein Error Exponent



## Theorem (Ahmadipour et al.)

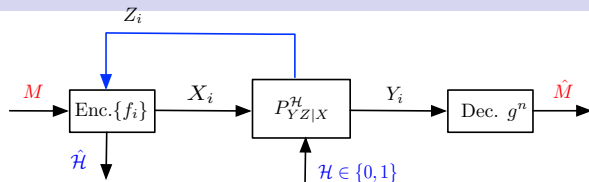
$(R, E_{\text{Stein}})$  pairs are achievable iff for some  $P_X$ :

$$\begin{aligned} R &\leq \min_{\mathcal{H}} I(X; Y | \mathcal{H}), \\ E_{\text{Stein}} &\leq \sum_x P_X(x) D(P_{Z|X}(\cdot|x) \| Q_{Z|X}(\cdot|x)) \end{aligned}$$

- Tradeoff between sensing and communication due to common  $P_X$ ! (input statistics)

[5] M. Ahmadipour, M. Kobayashi, M. W. and G. Caire, “An Information-Theoretic Approach to Joint Sensing and Communication,” *Trans. IT*, 2022.

# Adaptive Channel Coding/Sensing

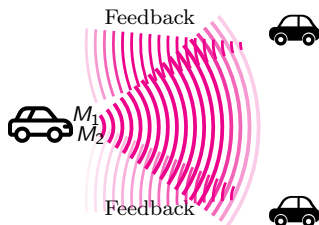


- Feedback allows the encoder to “learn” the channel parameter  $\mathcal{H}$  and to adapt its coding to the correct channel
- Sensing (detection) problem is still open when  $\mathcal{H}$  non-binary  
→ adaptive inputs also improve detection performance
- Chang et al. propose joint sensing and communication schemes
- Problem seems difficult and is open!

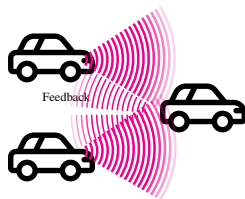
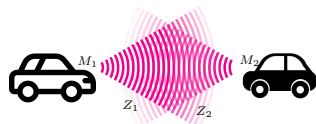
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[[7] M.-C. Chang, et al. “Rate and detection-error exponent tradeoff for joint communication and sensing of fixed channel states,” *JSAIT* 2023.

# Network ISAC



- Data sent to both receivers
- Fundamental limits partly characterized



- Both Transmitters sense and send data
- Comm. path between Txs!  $\rightarrow$  Collaborative comm. and sensing

# Summary

- Presented information-theoretic framework for integrated sensing and communication [Kobayashi,Caire,Kramer'18] and [Joudeh&Willems'22]
- Information-theoretic limits have been derived for various sensing criteria and discrete-memoryless channels/state sequences
- Single Tx: optimal sensing performance depends only on  $x^n$  statistics.
- Tradeoff between rates and distortion(s)/exponents .
- Multiple Txs: *Fully integrate coding for collaborative sensing and comm.*



# Interesting future research directions

- Simplified capacity-expressions/coding schemes for channels with memory
- Continuous-time channels
- Other sensing criteria
- Further investigations on secrecy constraints

# IT References on ISAC with Distortion

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- T. Welling, O. Günlü, and A. Yener, “Transmitter actions for secure integrated sensing and communication,” *ISIT* 2024.

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