

Deterministic Identification for Molecular Communications over the Poisson Channel

Mohammad Javad Salariseddigh

Institute for Communications Engineering



Joint work with:

Uzi Pereg (TUM-LNT), Holger Boche (TUM-LTI), Christian Deppe (TUM-LNT),
Vahid Jamali (Princeton, ECE), and Robert Schober (FAU-IDC)

February 24rd, 2022

Outline

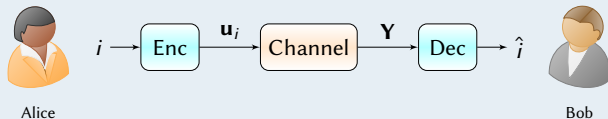
- 1 Motivation
- 2 Main Contributions
- 3 Definitions
- 4 Main Results
- 5 Conclusions

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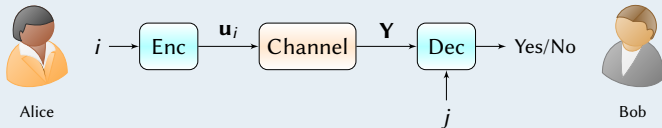
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Transmission vs. Identification

- **Shannon's setting:** Bob recover the message.



- **Identification setting:** Bob asks if a message was sent or not?



- **Event-triggered scenarios**
- **Molecular communication**

Transmission (TR) ¹

- Originally introduced by Shannon (1948)
- Capacity was established without randomness at encoder
- Codebook size $\sim 2^{nR}$

¹Shannon, "A Mathematical Theory of Communication", Bell Sys. Tech. J., 1948

Transmission (TR) ¹

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Theorem (Shannon, 1948)

Transmission capacity of a DMC \mathcal{W} with exponential codebook size, i.e., $L(n, R) = 2^{nR}$ is given by

$$\mathbb{C}_T(\mathcal{W}, L) = \max_{p_X} I(X; Y)$$

¹Shannon, "A Mathematical Theory of Communication", Bell Sys. Tech. J., 1948

Randomized Identification (RI)²

- Originally introduced by Ahlswede and Dueck (1989)
- Capacity was established with randomness at encoder
- Encoder employs distribution to select codewords

Remarkable Property

- Reliable identification is possible with codebook size $\sim 2^{2^{nR}}$

Theorem (Ahlswede and Dueck, 1989)

RI capacity of a DMC \mathcal{W} in double exponential codebook size, i.e., $L(n, R) = 2^{2^{nR}}$ is given by

$$\mathbb{C}_{RI}(\mathcal{W}, L) = \max_{p_X} I(X; Y)$$

²Ahlswede and Dueck, "Identification via Channels", T-IT, 1989

Deterministic Identification (DI)^{3 4}

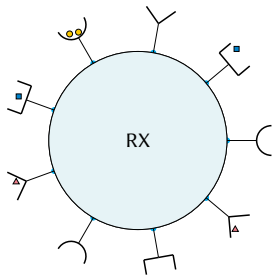
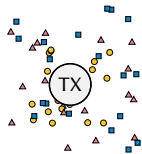
- Encoder uses deterministic mapping for coding
- Simpler implementation (random resource not required)
- Achievable rates **higher** than transmission^a
- Suitable for molecular communication
 - Olfactory-inspired MC system
 - Cancer treatment and target drug delivery

^a Salarisedigh *et al.*, "Deterministic Identification Over Channel With Power Constraints", T-IT, 2022

³ Ahlswede and Cai, "Identification Without Randomization", T-IT, 1999

⁴ Salarisedigh *et al.*, "Deterministic Identification Over Channels With Power Constraints," Proc. ICC, 2021

Olfactory-inspired MC System



- A codeword is molecular mixture
- Dedicated type receptors ensure spacial orthogonality

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Main Contributions

- We develop lower and upper bounds on the DI capacity for the memoryless discrete time Poisson channels (DTPC) subject to both average and peak power constraints
- We use the bounds to determine the **correct size** of codebook
- We show that the optimal codebook size scales as $\sim 2^{(n \log n)R}$

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DI Codes

Definition

An $(L(n, R), n, \lambda_1, \lambda_2)$ -DI code for DTPC \mathcal{W} is a system $\{(\mathbf{u}_i, \mathcal{D}_i)\}_{i \in [1:L(n,R)]}$ subject to

- 1 Codebook size: $L(n, R) = 2^{(n \log n)R}$
- 2 Codeword: $\mathbf{u}_i \in \mathcal{X}^n$, decoding sets: $\mathcal{D}_i \subset \mathcal{Y}^n$
- 3 Input constraints:
 - $0 \leq u_{i,t} \leq P_{\max}$
 - $n^{-1} \sum_{t=1}^n u_{i,t} \leq P_{\text{avg}}$
- 4 Error requirement type I: $W^n(\mathcal{D}_i | \mathbf{u}_i) > 1 - \lambda_1$
- 5 Error requirement type II: $W^n(\mathcal{D}_i | \mathbf{u}_j) < \lambda_2$
 $i \neq j$

DI Codes (Cont.)

Definition

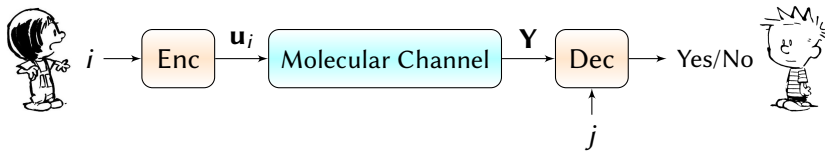
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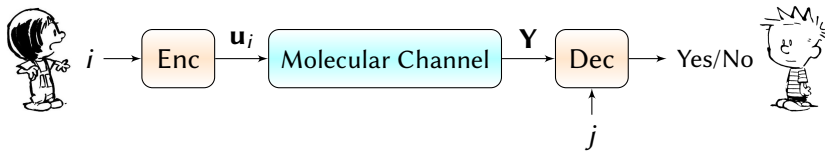
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DI for Poisson Channel



DI for Poisson Channel



Definitions

- $Y_t \sim \text{Pois}(\lambda + \rho u_{i,t})$
- $\lambda \in \mathbb{R}_{>0} \rightarrow$ Expected # of interfering molecules
- $\rho X \rightarrow$ Expected # observed molecules due to release by TX
- $\mathbf{y} \in \mathbb{N}_0^n \rightarrow$ Output vector
- Power const. $0 \leq u_{i,t} \leq P_{\max}$ and $\frac{1}{n} \sum_{t=1}^n u_{i,t} \leq P_{\text{avg}}$
- $W^n(\mathbf{y}|\mathbf{u}_i) = \prod_{t=1}^n \frac{e^{-(\lambda + \rho u_{i,t})} (\lambda + \rho u_{i,t})^{y_t}}{y_t!}$

DI for Poisson Channel ⁵

Theorem

Let \mathcal{W} be a DTPC with expected interference $\lambda \in \mathbb{R}_{>0}$. Then the DI capacity subject to power constraints $n^{-1} \sum_{t=1}^n u_{i,t} \leq P_{avg}$ and $0 \leq u_{i,t} \leq P_{max}$ for $L(n, R) = 2^{(n \log n)R}$ is bounded by

$$\frac{1}{4} \leq \mathbb{C}_{DI}(\mathcal{W}, L) \leq \frac{3}{2}$$

⁵ Salarisiddigh *et al.*, "Deterministic Identification Over Poisson Channels," Proc. GC, 2021

DI for Poisson Channel⁵

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- Achiev. proof: sphere pkg. of rad. $\sqrt[4]{n} \Rightarrow 2^{\frac{1}{4}(n \log n)}$ codewords

⁵Salarisiddigh *et al.*, "Deterministic Identification Over Poisson Channels," Proc. GC, 2021

Proof Sketch. (Achievability)

- Dense sphere packing arrangement with radius $\sqrt{n\epsilon_n}$
- *Minkowski-Hlawka Theorem* guarantees a density $\Delta_n \geq 2^{-n}$
- $2^{(n \log n)R} \geq \Delta_n \cdot \frac{\text{Vol}(\mathcal{Q}_0[n, A])}{\text{Vol}(\mathcal{S}_{\mathbf{u}_1}(n, \sqrt{n\epsilon_n}))} \geq 2^{-n} \cdot \frac{A^n}{\text{Vol}(\mathcal{S}_{\mathbf{u}_1}(n, \sqrt{n\epsilon_n}))}$
- $R \geq \frac{1}{n \log n} \left[\left(\frac{1-b}{4} \right) n \log n + n \log \left(\frac{A}{e\sqrt{a}} \right) + o(n) \right] \xrightarrow[b \rightarrow 0]{n \rightarrow \infty} \frac{1}{4}$
 - 1 $A = \min(P_{\text{ave}}, P_{\text{max}})$
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Chebyshev's inequality gives following error bounds:

- ❶ $P_{e,1}(i) \leq \frac{c_1}{n\epsilon_n^2} = \mathcal{O}\left(\frac{1}{n^b}\right)$
- ❷ $P_{e,2}(i, j) \leq \frac{c_2}{n\epsilon_n^2} = \mathcal{O}\left(\frac{1}{n^b}\right)$

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- Error bounds (1 and 2) gives: $\epsilon_n = an^{\frac{1}{2}(b-1)}$

Proof Sketch. (Converse)

Lemma

- For every pair of codewords \mathbf{u}_i and \mathbf{u}_j $1 \leq \exists t \leq n$ such that

$$\left| 1 - \frac{\rho u_{i,t} + \lambda}{\rho u_{j,t} + \lambda} \right| > \epsilon'_n$$

then exploiting *continuity* of Poisson distribution gives:

$$P_{e,1}(i) + P_{e,2}(i, j) \geq 1 - \kappa_n \quad \zeta$$

- $\epsilon'_n, \kappa_n \xrightarrow{n \rightarrow \infty} 0$

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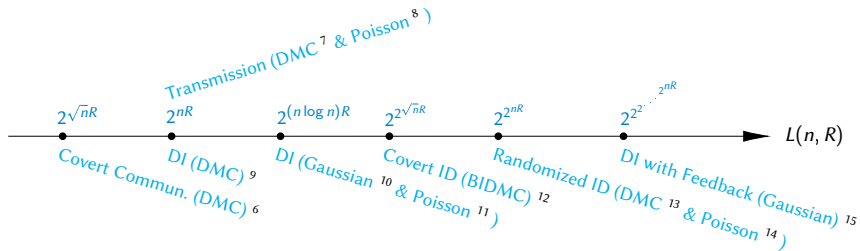
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- $\left| 1 - \frac{\rho u_{i,t} + \lambda}{\rho u_{j,t} + \lambda} \right| > \epsilon'_n \Rightarrow \rho |u_{i,t} - u_{j,t}| > \lambda \epsilon'_n \Rightarrow \|\mathbf{u}_i - \mathbf{u}_j\| > \frac{\lambda \epsilon'_n}{\rho}$
- $R \leq \frac{1}{n \log n} \left[\left(\frac{1}{2} + (1 + b) \right) n \log n + o(n) \right] \xrightarrow[n \rightarrow \infty]{b \rightarrow 0} \frac{3}{2}$

Codebook Sizes Spectrum



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- ⁶ Bloch, "Covert Communication over Noisy Channels: A Resolvability Perspective", T-IT, 2016
 - ⁷ Shannon, "A Mathematical Theory of Communication", Bell Sys. Tech. J., 1948
 - ⁸ Lapidoth and Moser, "On the Capacity of the Discrete-Time Poisson Channel", T-IT, 2009
 - ⁹ Salarisedigh *et al.*, "Deterministic Identification Over Channel With Power Constraints", T-IT, 2021
 - ¹⁰ Salarisedigh *et al.*, "Deterministic Identification Over Fading Channels", Proc. ITW, 2020
 - ¹¹ Salarisedigh *et al.*, "Deterministic Identification Over Poisson Channels", Proc. GC, 2021
 - ¹² Zhang and Tan, "Covert Identification over Binary-Input Discrete Memoryless Channels", arXiv, 2021
 - ¹³ Ahlswede and Dueck, "Identification via Channels", T-IT, 1989
 - ¹⁴ Burnashev, "On Identification Capacity of Infinite Alphabets or Continuous-Time Channels", T-IT, 2000
 - ¹⁵ Labidi *et al.*, "Identification over the Gaussian Channel in the Presence of Feedback", Proc. ISIT, 2021

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Conclusions

- We have determined bounds on the DI capacity for DTPC
- We have found optimal codebook size $\sim 2^{(n \log n)C}$
- We observed that DI codebook size for **DTPC** and **Gaussian** channels is the same
- Future directions:
 - ① Extension to multi-user scenarios (e.g., broadcast and multiple access channels) or multiple-input multiple-output channels
 - ② Extension to **non-orthogonal** molecule reception / ISI Memory
 - ③ To obtain better bounds:
 - Enhance sphere packing arrangement
 - Try a different decoder
 - Improve error analysis techniques (sharper conc. inequal.)

Question and Discussion

