

# Variable-length Feedback Codes with Several Decoding Times for the Gaussian Channel

Recep Can Yavas

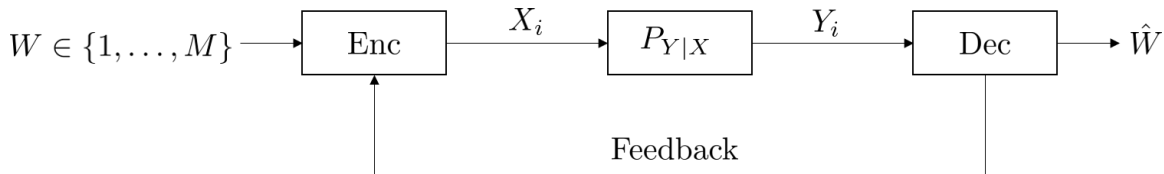
California Institute of Technology

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Joint work with Victoria Kostina and Michelle Effros  
ISIT 2021

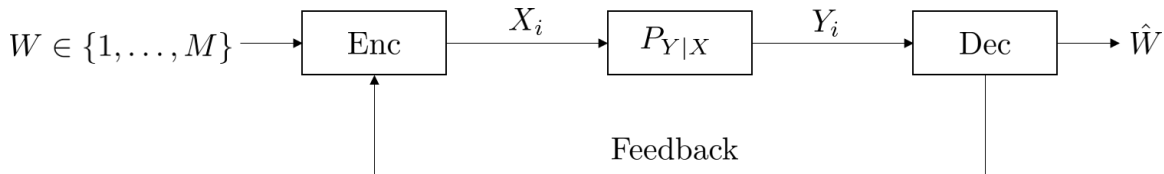
This work was supported in part by the National Science Foundation (NSF) under grant CCF-1817241 and CCF-1956386.

# Variable-length stop-feedback (VLSF) codes



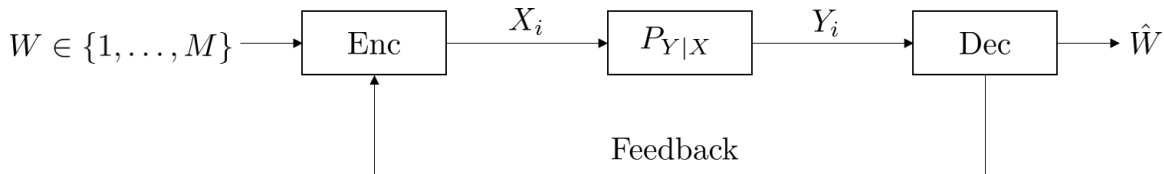
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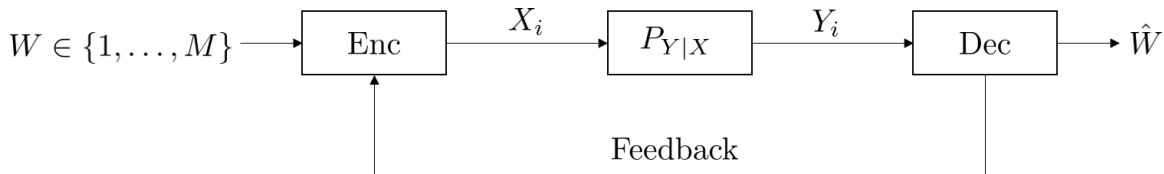
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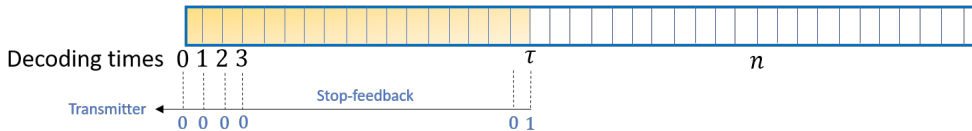
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- Feedback at each time instant is impractical!

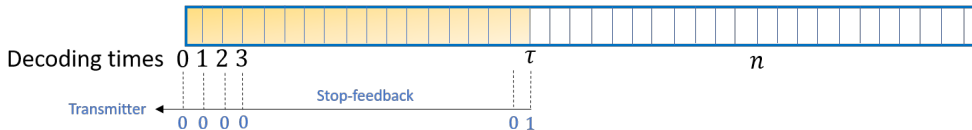
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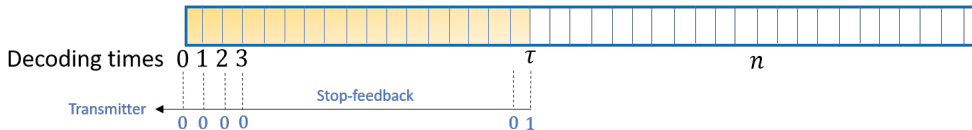
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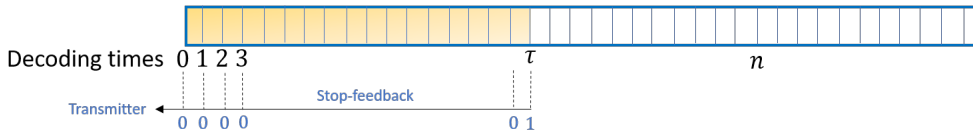


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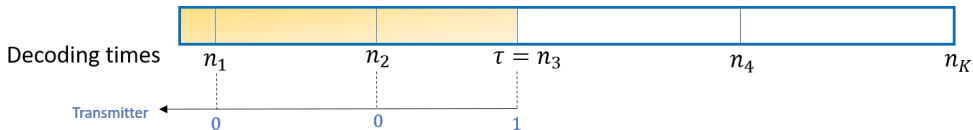


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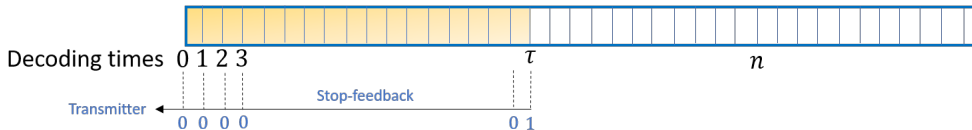


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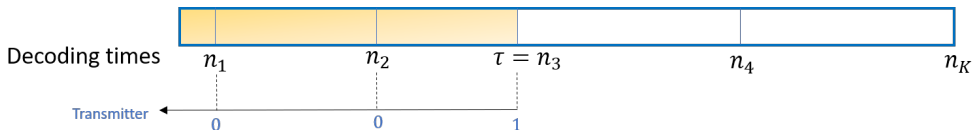


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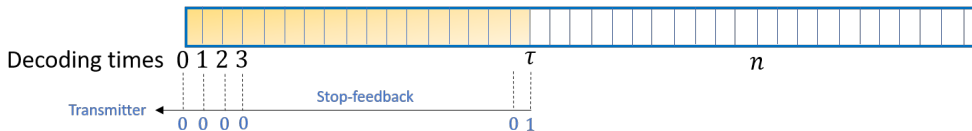
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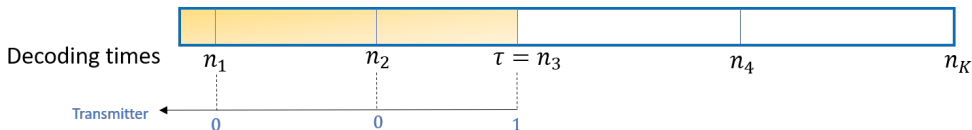
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- Sporadic feedback
- Practical codes: Incremental redundancy hybrid automatic repeat request codes

- [Burnashev (1976)]: error exponent  $\frac{-\ln P_e}{\mathbb{E}[\tau]}$  for DMCs
- [Polyanskiy et al. (2011)]: VLSF codes for DMCs under non-vanishing error value  $\epsilon$

$$\ln M^*(N, 1, \epsilon) = NC - \sqrt{NV}Q^{-1}(\epsilon) + O(\ln N)$$
$$\frac{NC}{1-\epsilon} - \ln N + O(1) \leq \ln M^*(N, \infty, \epsilon) \leq \frac{NC}{1-\epsilon} + O(1)$$

where  $C$  = capacity,  $V$  = dispersion,  $M^*(N, K, \epsilon)$  = maximum achievable message size compatible with average decoding time  $N$ , average error probability  $\epsilon$  and  $K$  decoding times

Prior work:  $K < \infty$

- [Vakilinia et al. (2016)]: VLSF codes with  $K$  decoding times for LDPC codes over the binary-input Gaussian channel

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- Then they search for the optimal  $n_1 \implies$  sequential differential optimization
- They do not solve the problem analytically  $\implies$  no second-order analysis

- Memoryless Gaussian channel: the channel output at time  $i$  is

$$Y_i = X_i + Z_i$$
$$Z_i \sim \mathcal{N}(0, 1),$$

where  $Z_i$ 's are i.i.d. and  $X_i$  and  $Z_i$  are independent.

## Definition

An  $(N, \{n_i\}_{i=1}^K, M, \epsilon, P)$  VLSF code comprises

- 1 encoding functions  $f_n: [M] \rightarrow \mathbb{R}$ ,  $n = 1, \dots, n_K$ :

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- such that

**Maximal power constraint:**  $\|f(m)^{n_k}\|^2 \leq n_k P \quad \forall m \in [M], k \in [K]$

**Average decoding time:**  $\mathbb{E}[\tau] \leq N$

**Average error probability:**  $\mathbb{P}[g_\tau(Y^\tau) \neq W] \leq \epsilon$

where the message  $W$  is uniformly distributed on the set  $[M]$ .

# Main Result

## Theorem (Achievability)

Fix  $K \geq 2$ ,  $P > 0$  and  $\epsilon \in (0, 1)$ . For the Gaussian channel

$$\ln M^*(N, K, \epsilon, P) \geq \frac{NC(P)}{1 - \epsilon} - \sqrt{N \ln_{(K-1)}(N) \frac{V(P)}{1 - \epsilon}} + o(\sqrt{N})$$

The decoding times satisfy  $n_1 = 0$  and the equations

$$\ln M^*(N, K, \epsilon, P) = n_k C(P) - \sqrt{n_k \ln_{(K-k+1)}(n_k) V(P)} - \ln n_k + O(1)$$

for  $k \in \{2, \dots, K\}$ .

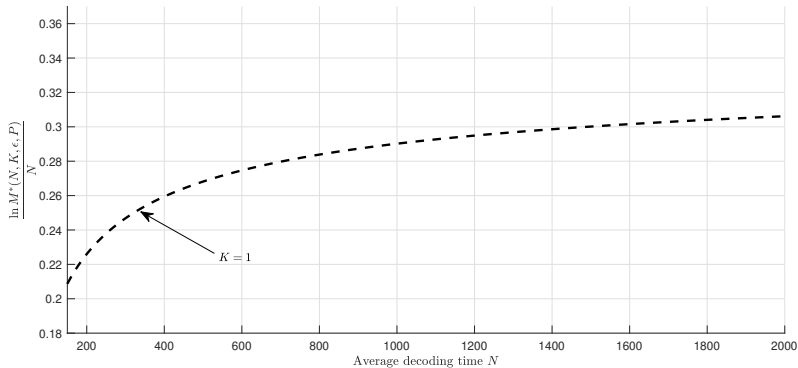
$C(P) = \frac{1}{2} \ln(1 + P) = \text{capacity}$ ,  $V(P) = \frac{P(P+2)}{2(1+P)^2} = \text{dispersion}$

$\ln_{(K)}(\cdot) \triangleq \underbrace{\ln(\ln(\dots \ln(\cdot)))}_{K \text{ times}}$

- **Bottom-line:** We derive an achievability bound and optimize the choices of the decoding times  $n_1, \dots, n_K$  to minimize average decoding time  $N$  for the given  $\epsilon$  and  $M$ .



Example:  $\epsilon = 10^{-3}$ ,  $P = 1$

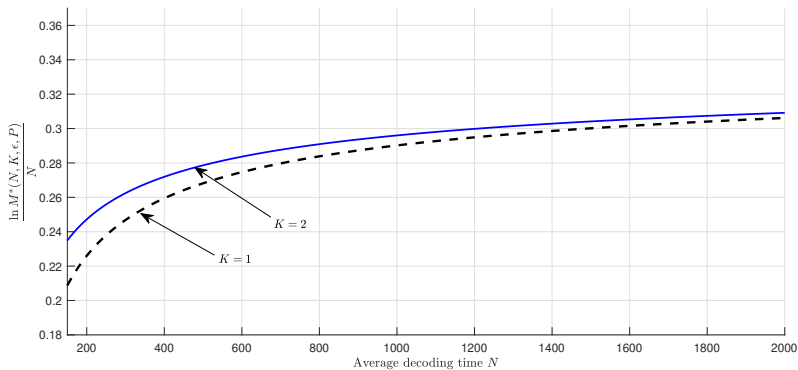


and



$n_1 = N$

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$$\mathbb{P}[\tau > n_i]$$

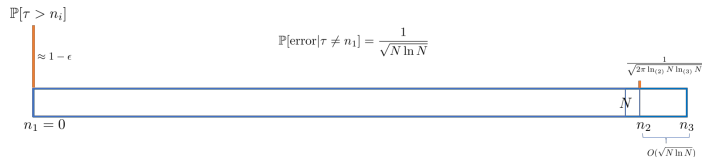
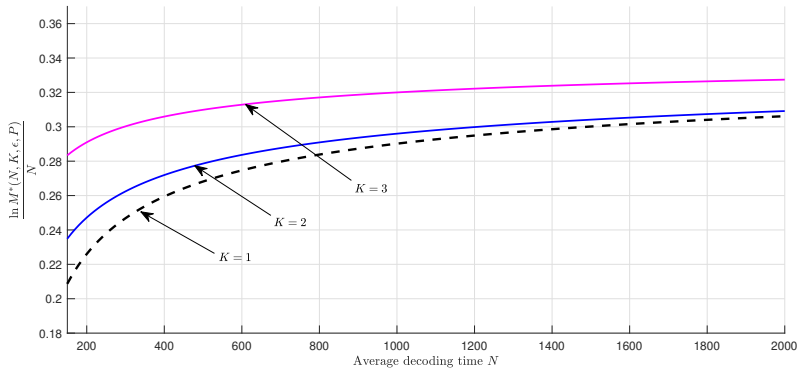
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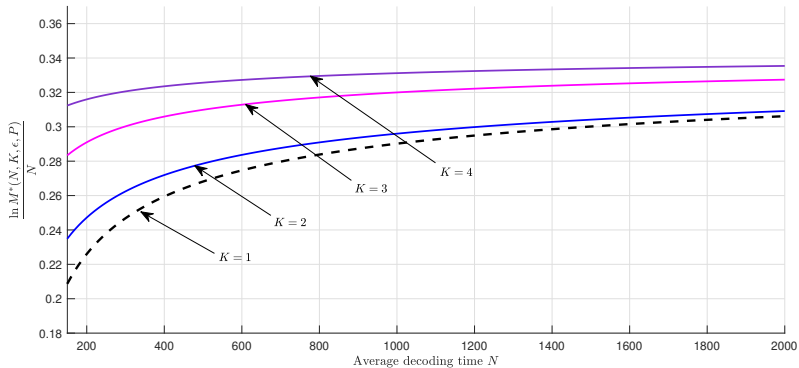
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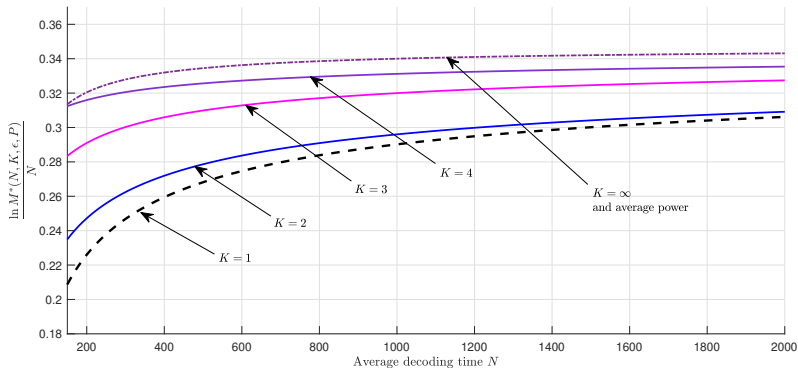
$$\frac{1}{\sqrt{2\pi \ln_{(2)} N \ln_{(3)} N}} \frac{1}{\sqrt{2\pi \ln N \ln_{(2)} N}}$$

$$N$$

$$n_2 \quad n_3 \quad n_4$$

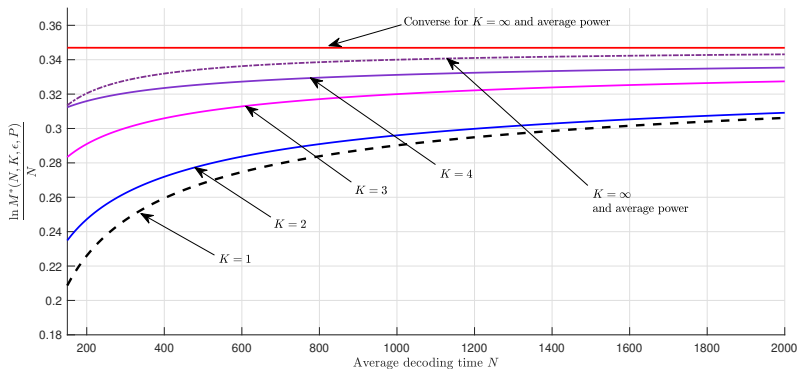
$$O\left(\sqrt{N \ln_{(2)} N}\right)$$

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# Comparison with prior work in extreme scenarios

- $2 \leq K < \infty$ , maximal power:

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[Polyanskiy et al. (2010) and Tan-Tomamichel (2015)]

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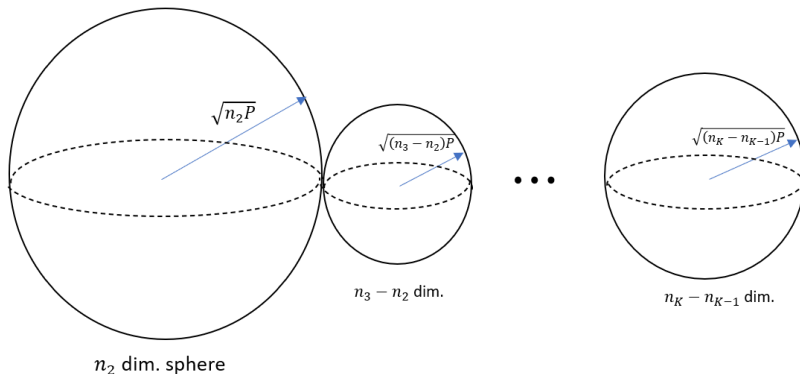
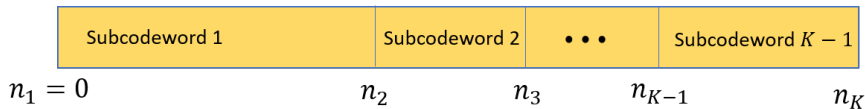
- $K = \infty$ , average power:  
[Truong-Tan (2018)]

$$\ln M^*(N, \infty, \epsilon, P)_{\text{ave}} \geq \frac{NC(P)}{1 - \epsilon} - \ln N + O(1)$$

$$\ln M^*(N, \infty, \epsilon, P)_{\text{ave}} \leq \frac{NC(P)}{1 - \epsilon} + \frac{h_b(\epsilon)}{1 - \epsilon}$$

# Random encoder design

- We generate  $M$  i.i.d. codewords of length- $n_K$  so that maximal power constraint is satisfied with equality for each  $n_k$ , and the subcodewords are drawn independent of each other.



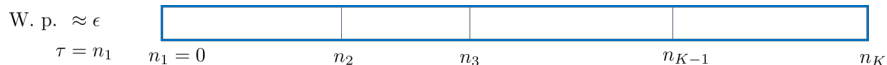
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**Threshold decoder:** Decode at the first time  $n_k \in \{n_2, \dots, n_K\}$  s.t.  $\iota(f(m)^{n_k}; Y^{n_k}) \geq \gamma$  for some  $m$ .

information density

$$\overbrace{\iota(x^n; y^n)} \triangleq \ln \frac{P_{Y^n|X^n}(y^n|x^n)}{P_{Y^n}(y^n)}$$



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- Goal: optimize  $n_2, \dots, n_K$ .

## Optimizing decoding times $n_2, \dots, n_K$ to minimize $N$

- $(N', \epsilon_N)$ : average decoding time and error probability given  $\tau > n_1$

$$\begin{aligned} \min \quad & N(n_2, \dots, n_K, \gamma) = \frac{N'(1 - \epsilon)}{1 - \epsilon_N} \\ \text{s.t.} \quad & N' = n_2 + \sum_{i=2}^{K-1} (n_{i+1} - n_i) \mathbb{P}[\tau > n_i] \\ & \epsilon_N = \mathbb{P}[i(X^{n_K}; Y^{n_K}) < \gamma] + M \exp\{-\gamma\} \end{aligned}$$

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Using moderate deviations theorem

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- Find the optimal  $(n_2^*, \dots, n_K^*, \gamma^*)$  by solving  $\nabla N = 0$ .



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- The optimal  $\epsilon_N^* = \frac{1}{\sqrt{N \ln N}}$ .

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- We drew independent subcodewords, each drawn uniformly on a power sphere.

# Future work

- Improve the converse result for  $K < \infty$  and maximal power constraint.

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- Investigate maximal power constraint vs. average power constraint for VLSF codes with  $K = \infty$ .

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We show for the maximal power constraint:

$$\ln M^*(N, \infty, \epsilon, P) \geq \frac{NC(P)}{1 - \epsilon} - O(\sqrt{N})$$

- ① V. Burnashev, "Data transmission over a discrete channel with feedback: Random transmission time," Problems of Information Transmission, vol. 12, no. 4, pp. 10–30, 1976.
- ② L. V. Truong and V. Y. F. Tan, "On Gaussian macs with variable-length feedback and non-vanishing error probabilities," IEEE Trans. on Inf. Theory., vol. 64, no. 4, p. 2333–2346, Apr. 2018.
- ③ Y. Polyanskiy, H. V. Poor, and S. Verdú, "Feedback in the non-asymptotic regime," IEEE Trans. Inf. Theory, vol. 57, no. 8, pp. 4903–4925, Aug. 2011.
- ④ Y. Polyanskiy, H. V. Poor, and S. Verdú, "Channel coding rate in the finite blocklength regime," IEEE Transactions on Information Theory, vol. 56, no. 5, pp. 2307–2359, May 2010.
- ⑤ V. Y. F. Tan and M. Tomamichel, "The third-order term in the normal approximation for the AWGN channel," IEEE Transactions on Information Theory, vol. 61, no. 5, pp. 2430–2438, May 2015.
- ⑥ K. Vakilinia, S. V. S. Ranganathan, D. Divsalar, and R. D. Wesel, "Optimizing transmission lengths for limited feedback with non-binary LDPC examples", IEEE Transactions on Communications, vol. 64, no. 6, pp. 2245–2257, 2016.
- ⑦ R. C. Yavas, V. Kostina, and M. Effros, "Variable-length Feedback Codes with Several Decoding Times for the Gaussian Channel," ISIT, 2021. Available at: <https://arXiv:2103.09373>.