

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

Recep Can Yavas

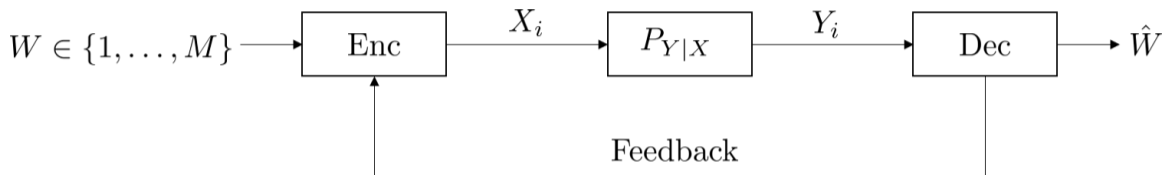
California Institute of Technology

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Joint work with Victoria Kostina and Michelle Effros
ISIT 2021

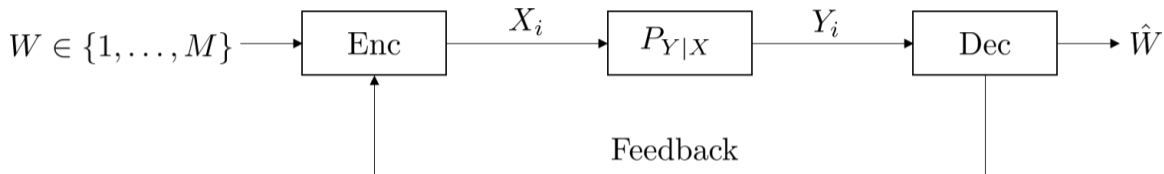
This work was supported in part by the National Science Foundation (NSF) under grant CCF-1817241 and CCF-1956386.

Variable-length stop-feedback (VLSF) codes



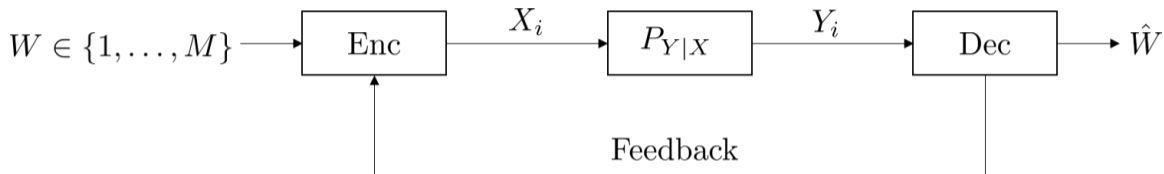
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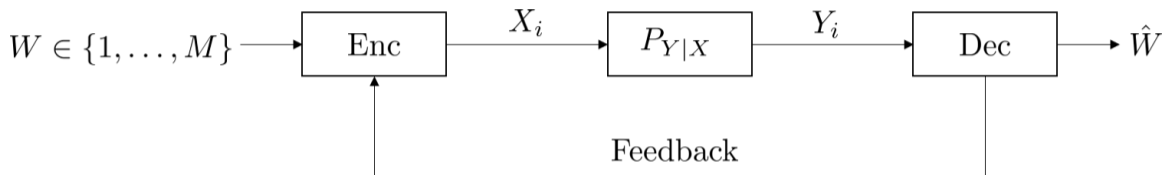
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 \implies High reliability

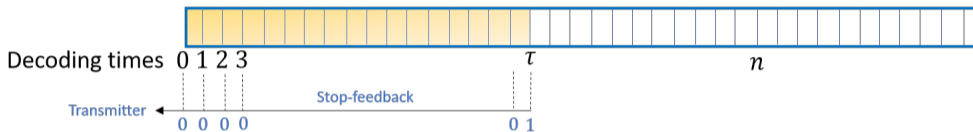
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- **Feedback at each time instant is impractical!**

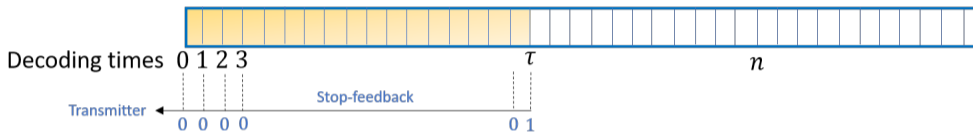
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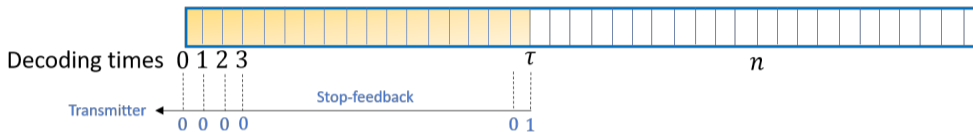
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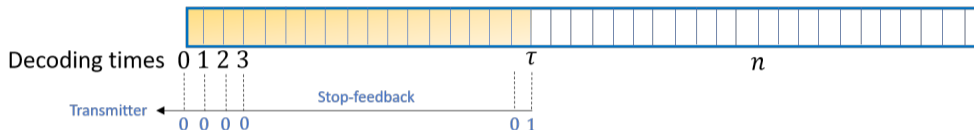
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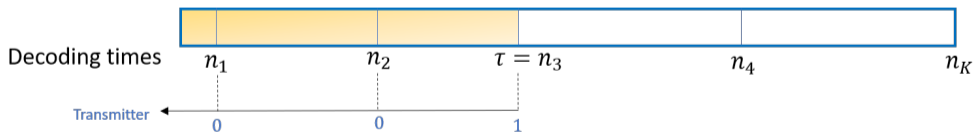
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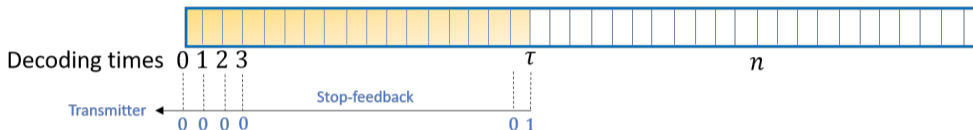


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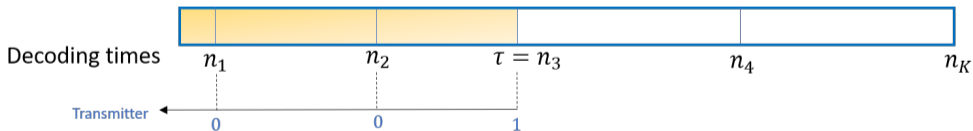


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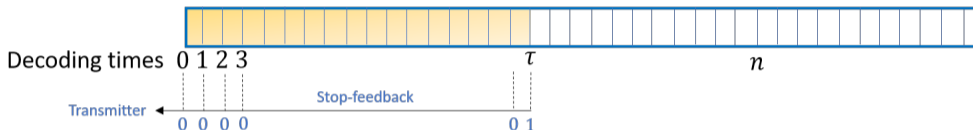
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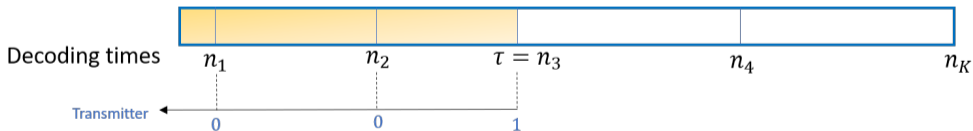
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- Sporadic feedback
- Practical codes: Incremental redundancy hybrid automatic repeat request codes

- [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 ϵ

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

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

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- 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 $[M]$.

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

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

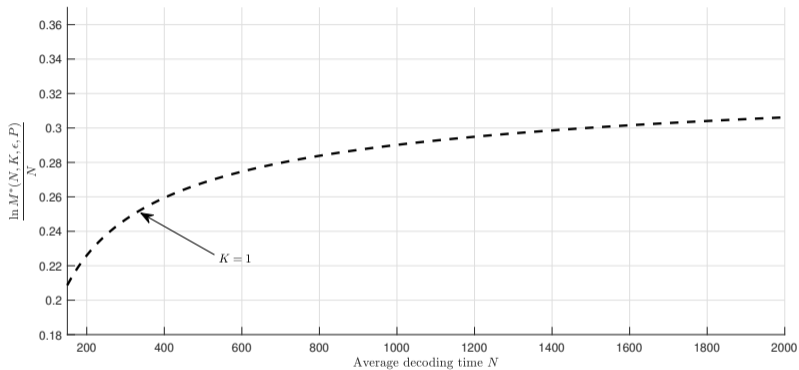
for $k \in \{2, \dots, K\}$.

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

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

- **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 ϵ and M .

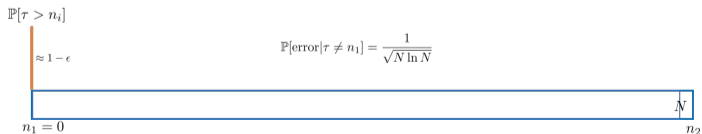
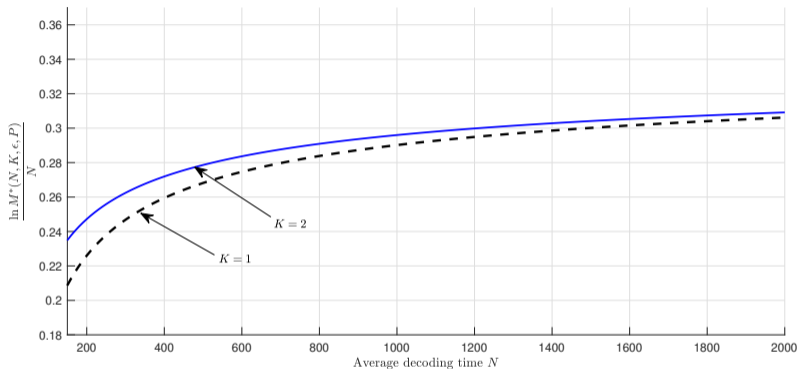
Example: $\epsilon = 10^{-3}$, $P = 1$



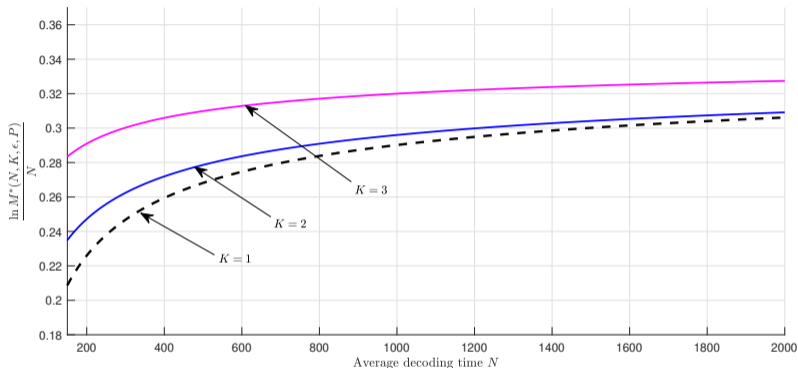
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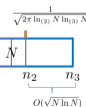


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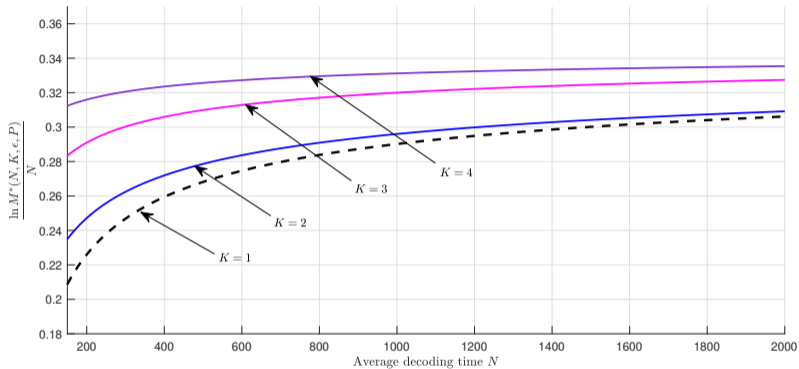
$\approx 1 - \epsilon$

$n_1 = 0$

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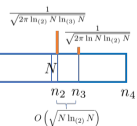
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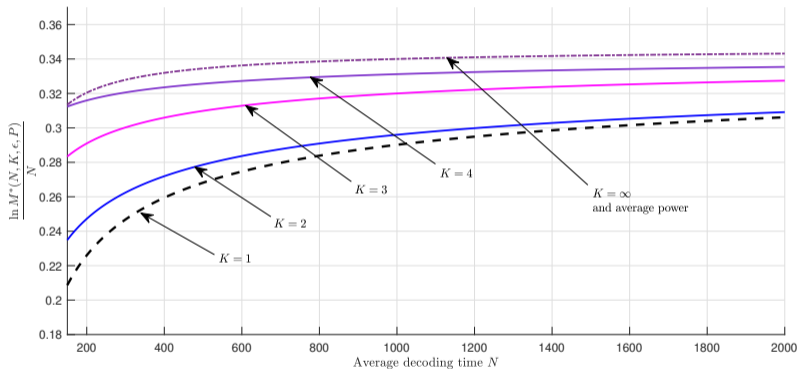
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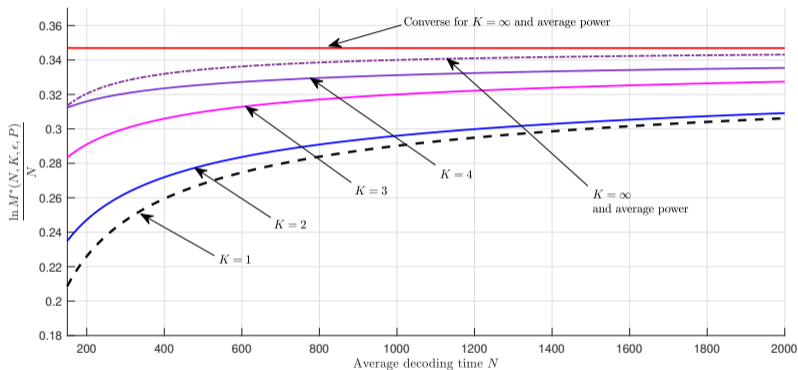


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Comparison with prior work in extreme scenarios

- $2 \leq K < \infty$, 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)]

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

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

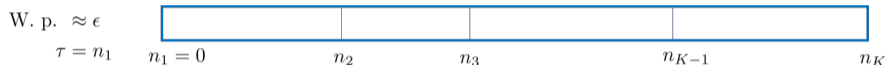
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Threshold decoder: Decode at the first time $n_k \in \{n_2, \dots, n_K\}$ s.t. $\iota(f(m)^{n_k}; Y^{n_k}) \geq \gamma$ for some m .

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', ϵ_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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- 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.

- 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 $K = \infty$.

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

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.
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