

# Variable blocklength communication with feedback

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# Motivation

- Shannon - feedback does not improve capacity
- Can help in the non-asymptotic regime

Length required to achieve 90% capacity on a  $C = 1/2$  BSC

- Fixed length with feedback,  $l > 3100$  bits
- Variable length with feedback,  $l < 200$  bits

Even a simple termination signal sent by the source indicating end of transmission helps a lot

# System Model

- Discrete Memoryless Channel

$$P_{Y_i|X_1^i, Y_1^{i-1}} = P_{Y_i|X_i} = P_{Y_1|X_1}$$

- Input and Output Alphabet  $\mathcal{A}$  and  $\mathcal{B}$
- Transition matrix  $\mathcal{P}$  -  $p_{i,j}$  is the probability of transmitting  $i^{\text{th}}$  input symbol and receiving  $j^{\text{th}}$  output symbol
- $W \in 1, 2, \dots, M$  equiprobable message to be transmitted - mapped to input alphabet  $\mathcal{A}$

# Part I

## Feedback in non-asymptotic regime

# Fixed Blocklength without feedback

An  $(l, M, \epsilon)$  code,

- Encoder  $X_n = f(W)$ , Decoder  $\hat{W} = g(Y^l)$
- The fundamental limit of coding is,

$$M^*(l, \epsilon) = \max\{M : \exists(l, M, \epsilon) \text{ code}\}$$

- Maximum information we can send is,

$$\log M^*(l, \epsilon) = lC - \sqrt{lV}Q^{-1}(\epsilon) + O(\log l)$$

- $V$  - Channel dispersion - measures the stochastic variability of the channel as compared to a deterministic channel of the same capacity.
- In presence of variable-length coding with feedback the  $\frac{1}{\sqrt{l}}$  penalty term is eliminated

# Fixed Blocklength with feedback

An  $(l, M, \epsilon)$  code

- Noiseless Feedback of  $Y$ 's to the encoder
- Encoder  $X_n = f(W, Y^{n-1})$
- Feedback does not help remove the penalty term

$$\log M_b^*(l, \epsilon) = nC - \sqrt{nV}Q^{-1}(\epsilon) + O(\log n)$$

- $\log M_b^*$  increases hardly by 2-3 bits as compared to  $\log M^*$ .

## Variable Blocklength without feedback

- Allow a non-vanishing probability of error  $\epsilon$
- The capacity increases to give  $\epsilon$ -capacity

### Theorem

*For any non-anticipatory channel with capacity  $C$  that satisfies the strong converse for fixed-blocklength codes (without feedback), the  $\epsilon$ -capacity under variable-length coding without feedback, is*

$$C_\epsilon = \frac{C}{1 - \epsilon}$$

# Variable blocklength with feedback

- Encoder  $X_n = f(W, Y^{n-1})$
- Decoder  $\hat{W} = g_\tau(Y^\tau)$
- Stopping time  $\tau$  on  $\sigma\{Y_1, \dots, Y_n\}$  such that  $\mathbb{E}(\tau) \leq l$
- $Pr(\hat{W} \neq W) \leq \epsilon$

The fundamental limit of VLF coding is,

$$M_f^*(l, \epsilon) = \max M : \exists(l, M, \epsilon) - \text{VLF code}$$



## VLF codes with termination

- In VLF codes, the decoder decides the stopping time  $\tau$  as a function of outputs  $Y^{\tau-1}$ . It is conveyed to source through feedback
- Source sends a termination signal to receiver on a separate reliable channel - VLFT code
- Stopping time  $\tau$  on  $\sigma\{W, Y_1, \dots, Y_n\}$  such that  $\mathbb{E}(\tau) \leq l$

The fundamental limit of VLFT coding is,

$$M_t^*(l, \epsilon) = \max M : \exists(l, M, \epsilon) - \text{VLFT code}$$

## Variable Blocklength with feedback

## Theorem

For an arbitrary DMC with capacity  $C$  we have for any  $0 < \epsilon < 1$

$$\log M_f^*(l, \epsilon) = \frac{lC}{1 - \epsilon} + O(\log l) \quad (1)$$

$$\log M_t^*(l, \epsilon) = \frac{lC}{1 - \epsilon} + O(\log l) \quad (2)$$

More precisely, we have,

$$\frac{lC}{1 - \epsilon} - \log l + O(1) \leq \log M_f^*(l, \epsilon) \leq \frac{lC}{1 - \epsilon} + O(1) \quad (3)$$

$$\log M_f^*(l, \epsilon) \leq \log M_t^*(l, \epsilon) \leq \frac{lC + \log l}{1 - \epsilon} + O(1) \quad (4)$$

# Proof of converse of the theorem

## Theorem

Consider an arbitrary DMC with capacity  $C$ . Then any  $(l, M, \epsilon)$  VLF code with  $0 \leq \epsilon \leq 1$  satisfies

$$\log M \leq \frac{Cl + h(\epsilon)}{1 - \epsilon},$$

whereas each  $(l, M, \epsilon)$  VLFT code with  $0 \leq \epsilon \leq 1$  satisfies

$$\begin{aligned} \log M &\leq \frac{Cl + h(\epsilon) + (l+1)h(\frac{1}{l+1})}{1 - \epsilon} \\ &\leq \frac{Cl + \log(l+1) + h(\epsilon) + \log(\epsilon)}{1 - \epsilon}, \end{aligned}$$

where  $h(x) = -x \log x - (1-x) \log(1-x)$  is the binary entropy function.

# Proof of converse of the theorem

By Fano's inequality we have,

$$\begin{aligned}
 (1 - \epsilon) \log M &\leq I(W; Y^\tau, \tau) + h(\epsilon) \\
 &= I(W; Y^\tau) + I(W; \tau | Y^\tau) + h(\epsilon) \\
 &\leq I(W; Y^\tau) + H(\tau) + h(\epsilon) \\
 &\leq I(W; Y^\tau) + (l + 1)h\left(\frac{1}{l + 1}\right) + h(\epsilon),
 \end{aligned}$$

where, we upper bound  $H(\tau)$  by solving the optimization problem:

$$\max_{\tau: \mathbb{E}[\tau] \leq l} H(\tau) = (l + 1)h\left(\frac{1}{l + 1}\right)$$

$\tau$  cannot convey more than  $O(\log l)$  bits of information about the message

## Proof of converse of the theorem

$$\begin{aligned}(1 - \epsilon) \log M &\leq I(W; Y^\tau) + (l + 1)h\left(\frac{1}{l + 1}\right) + h(\epsilon) \\ &\leq Cl + (l + 1)h\left(\frac{1}{l + 1}\right) + h(\epsilon),\end{aligned}$$

We use the result from Burnashev which says that,

$$I(W; Y^\tau) \leq C\mathbb{E}[\tau] \leq Cl.$$

## Concluding Remarks

- Variable length coding with feedback drastically reduces the average blocklength required to achieve a given probability of error by removing the  $\frac{1}{\sqrt{l}}$  penalty term.
- Even simple decision-feedback codes with just the termination signal have performance very close to the VLFT codes

## Part II

# Optimal Error Exponents

## Decision making method

- Find the posterior probability  $p_j(n)$  of the  $j^{\text{th}}$  input symbol after  $n$  observations
- Calculate the likelihood functions  $\log \frac{p_j(x_n)}{1-p_j(x_n)}$ .
- Make a decision in favor of symbol  $X_j$  if the likelihood crosses  $\log(1/\epsilon)$
- Probability of error,

$$P_e = \frac{1}{M} \sum_{j=1}^M (1 - p_j(x_n)) \leq \frac{\epsilon}{1 + \epsilon} \leq \epsilon$$



## Entropy of the posterior distribution

- Entropy of the posterior distribution  $p(n)$  is defined as  $H_n$

$$\mathbb{E}(H_n - H_{n+1}) \leq C$$
$$\mathbb{E}(\log H_n - \log H_{n+1}) \leq C_1$$

where  $C_1 > C$  is the maximal relative entropy between output distributions.

$$C_1 = \max_{i,k} \sum_{l=1}^K p_{i,l} \log \frac{p_{i,l}}{p_{k,l}} = \max_{i,k} D(\mathbf{p}_i || \mathbf{p}_k)$$

# Burnashev's Error exponent

- We know that without feedback, the error exponent is  $E(R) = (C - R)$ . i.e the probability of error with blocklength  $l$  is,  $P_e \leq e^{-El}$
- With variable length and feedback we get the error exponent,

$$E(R) = C_1 \left( 1 - \frac{R}{C} \right)$$

## Yamamoto Itoh scheme

- Simple two-phase coding scheme that achieves this error exponent
- Phase 1 - Transmit message for  $\gamma N$  symbols
- Phase 2 - Transmit correct/error signal for  $n = (1 - \gamma)N$  symbols
- If in error, retransmit the message in the next block
- Probability of error =  $P_E = P_{1e}P_{ce}$