

Erasure and Undetected Error Probabilities in the Moderate Deviations Regime

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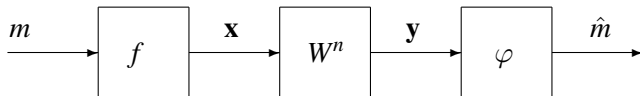
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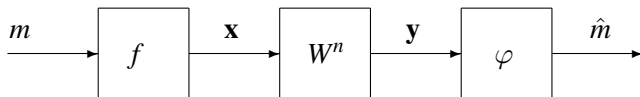
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Channel Coding with the Erasure Option



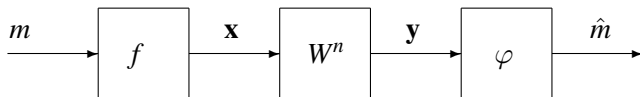
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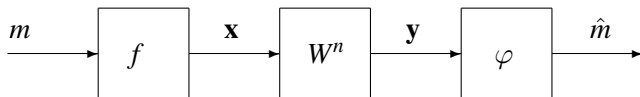
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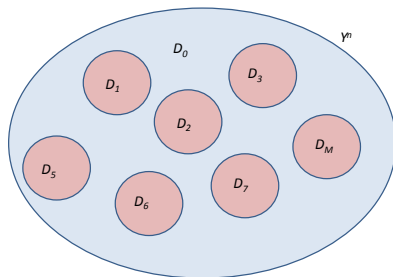
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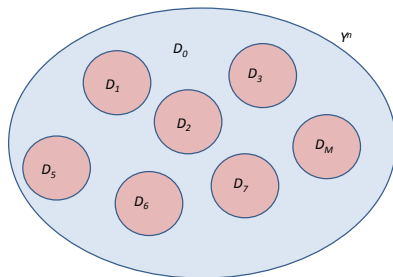
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- If $\mathbf{y} \in \mathcal{D}_m$, declare m was sent; if $\mathbf{y} \in \mathcal{D}_0$, declare an erasure
- We assume W^n is a DMC, i.e.,

$$W^n(\mathbf{y} | \mathbf{x}) = \prod_{i=1}^n W(y_i | x_i)$$

Channel Coding with the Erasure Option : Illustration



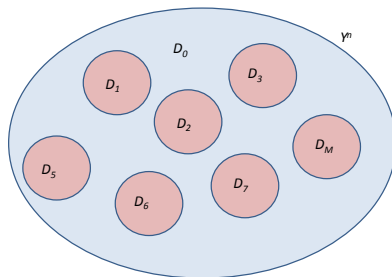
Channel Coding with the Erasure Option : Illustration



- Probability of **total error**

$$\Pr(\mathcal{E}_1 | \mathcal{C}_n) = \frac{1}{M_n} \sum_{m=1}^{M_n} \sum_{\mathbf{y} \in \mathcal{D}_m^c} W^n(\mathbf{y} | f_n(m)).$$

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- Probability of **undetected error**

$$\Pr(\mathcal{E}_2 | \mathcal{C}_n) = \frac{1}{M_n} \sum_{m=1}^{M_n} \sum_{\mathbf{y} \in \mathcal{D}_m} \sum_{m' \neq m} W^n(\mathbf{y} | f_n(m')).$$

Some Historical Remarks on Erasure Decoding

- Forney (1968) derived **exponential** error bounds for erasure and list decoding using Gallager-type bounding techniques

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- Tan-Moulin (2014) considered **non-vanishing** total and undetected error version of this problem.

Various Asymptotic Regimes

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Strassen (1962), Hayashi (2009) and PPV (2010).

- When $0 < t < 1/2$, this is the **moderate deviations** regime and

$$\epsilon_n^* \approx \exp\left(-n^{1-2t} \frac{a^2}{2V}\right), \quad \text{sub-exponential}$$

Altuğ-Wagner (2014) and Polyanskiy-Verdú (2010)

Our Contribution : Two Asymptotic Regimes

- **Moderate Deviations** Regime with $0 < t < 1/2$:

$$\Pr(\mathcal{E}_1 | \mathcal{C}_n) \approx \exp(-\Theta(n^{1-2t})), \quad \Pr(\mathcal{E}_2 | \mathcal{C}_n) \approx \exp(-\Theta(n^{1-t}))$$

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- Similar to Somekh-Baruch and Merhav (2011), we seek **ensemble-tight** results, i.e., we seek **asymptotic equalities** for

$$\mathbb{E}_{\mathcal{C}_n}[\Pr(\mathcal{E}_1 | \mathcal{C}_n)], \quad \text{and} \quad \mathbb{E}_{\mathcal{C}_n}[\Pr(\mathcal{E}_2 | \mathcal{C}_n)]$$

where $\mathbb{E}_{\mathcal{C}_n}[\cdot]$ denotes expectation over a random codebook.

- Mutual information

$$I(P, W) = \sum_x P(x) \sum_y W(y|x) \log \frac{W(y|x)}{PW(y)}$$

- Conditional information variance

$$V(P, W) = \sum_x P(x) \sum_y W(y|x) \left[\log \frac{W(y|x)}{PW(y)} - D(W(\cdot|x) \| PW) \right]^2$$

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- Always assume $V_{\min}(W) > 0$.

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$$\liminf_{n \rightarrow \infty} -\frac{1}{n^{1-t}} \log \Pr(\mathcal{E}_2 | \mathcal{C}_n) \geq b$$

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Decoding regions are reminiscent of **information spectrum** analysis:

$$\tilde{\mathcal{D}}_m := \left\{ \mathbf{y} : \log \frac{W^n(\mathbf{y} | \mathbf{x}_m)}{(PW)^n(\mathbf{y})} \geq \log M_n + bn^{1-t} \right\}$$

Remarks on Direct Results : Moderate

$$\text{Total} \quad \Pr(\mathcal{E}_1 | \mathcal{C}_n) \approx \exp\left(-n^{1-2t} \frac{(a-b)^2}{2V_{\min}(W)}\right)$$

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and

$$\Pr(\mathcal{E}_2 | \mathcal{C}_n) \lesssim \exp(-\sqrt{n}b)$$

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- Also need a stronger decoder; c.f. Forney (1968)

$$\mathcal{D}_m := \left\{ \mathbf{y} : \frac{W^n(\mathbf{y}|\mathbf{x}_m)}{\sum_{m' \neq m} W^n(\mathbf{y}|\mathbf{x}_{m'})} \geq \exp(nT_n) \right\}$$

T_n : **Blocklength-varying threshold** depending on regime.

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Then,

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$$\lim_{n \rightarrow \infty} -\frac{1}{n^{1-t}} \log \mathbb{E}_{\mathcal{C}_n} [\Pr(\mathcal{E}_2 | \mathcal{C}_n)] = b$$

Remarks on Ensemble Tight Result

- Similar result for mixed regime, i.e.,

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- As b **increases** towards a , decoding regions become **smaller**; erasure probability **higher**; undetected probability **smaller**

Main Analysis Technique

- Recall Forney's optimum decoding rule

$$\mathcal{D}_m := \left\{ \mathbf{y} : \frac{W^n(\mathbf{y}|\mathbf{x}_m)}{\sum_{m' \neq m} W^n(\mathbf{y}|\mathbf{x}_{m'})} \geq \exp(nT_n) \right\}$$

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- Suffices to understand the **cumulant generating function**

$$\phi_n(s) := \log \mathbb{E} [\exp(sF_n)]$$

Here s doesn't have to be a constant; can vary with n

Cumulant Generating Function Asymptotics

Lemma

Consider $\phi_n(s) := \log \mathbb{E} [\exp(sF_n)]$. Let $0 < t \leq 1/2$. We have

$$\phi_n \left(\frac{u}{n^t} \right) = \left(-au + u^2 \frac{V(P)}{2} \right) n^{1-2t} + o(1)$$

for any constant $u > 0$.

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- Type class enumerator method (cf. Merhav)
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- For example, if $t = 1/2$, we have

$$\lim_{n \rightarrow \infty} \phi_n \left(\frac{u}{\sqrt{n}} \right) = -au + u^2 \frac{V(P)}{2} \quad \Rightarrow \quad \text{Gaussian}$$

Modified Gärtner-Ellis Theorem

- Exponent of **expected total error probability** $\mathbb{E}_{\mathcal{C}_n}[\Pr(\mathcal{E}_1 | \mathcal{C}_n)]$ is

$$-\log \mathbb{E}_{\mathcal{C}_n}[\Pr(\mathcal{E}_1 | \mathcal{C}_n)] = -\log \Pr(F_n > -bn^{1-t})$$

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- Asymptotic behavior of the **cumulant generating function** of F_n

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- We analyzed the total and undetected errors whose scalings are

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- Full version: <http://arxiv.org/abs/1407.0142>.