

# Channel Dispersion Thoughts

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Define the information density etc as per the Verdu, Polyanski paper. We have an input alphabet  $\mathcal{X}$  with  $n$  symbols, and an output alphabet,  $\mathcal{Y}$  with 2 symbols. There is also some discrete spaces where the probability distributions live,  $S(\mathcal{X})$  and  $S(\mathcal{Y})$ .

In particular the channel dispersion is:

$$V = \sum_{x,y \in \mathcal{X}, \mathcal{Y}} p(x,y) \left( \log \frac{p(y|x)}{p(x)} \right)^2 - I(\mathcal{X}; \mathcal{Y})^2 \quad (1)$$

This can be rewritten as:

$$V = \sum_{x,y \in \mathcal{X}, \mathcal{Y}} I(\mathcal{X}; \mathcal{Y}) \log \frac{p(y|x)}{p(x)} - I(\mathcal{X}; \mathcal{Y})^2 \quad (2)$$

Define  $Q$  as:

$$Q = \sum_{x,y \in \mathcal{X}, \mathcal{Y}} I(\mathcal{X}; \mathcal{Y}) \log \frac{p(y|x)}{p(x)} \quad (3)$$

Now  $Q^2$  is:

$$Q^2 = \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} I(\mathcal{X}; \mathcal{Y}) \log \frac{p(y|x)}{p(x)} \right)^2 \quad (4)$$

By Cauchy-Swartz:

$$Q^2 \leq \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} I(\mathcal{X}; \mathcal{Y}) \right)^2 \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} \log \frac{p(y|x)}{p(x)} \right)^2 \quad (5)$$

Using the bound on the mutual information:

$$I(\mathcal{X}; \mathcal{Y}) \leq 1 \quad (6)$$

So,

$$Q^2 \leq \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} 1 \right)^2 \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} \log \frac{p(y|x)}{p(x)} \right)^2 \quad (7)$$

The first term is at most  $n$ , so:

$$Q^2 \leq n \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} \log \frac{p(y|x)}{p(x)} \right)^2 \quad (8)$$

So

$$Q \leq \sqrt{n \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} \log \frac{p(y|x)}{p(x)} \right)^2} \quad (9)$$

$$Q \leq \sqrt{n} \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} \log \frac{p(y|x)}{p(x)} \right) \quad (10)$$

Then we are left with:

$$V \leq \sqrt{n} \left( \sum_{x,y \in \mathcal{X}, \mathcal{Y}} \log \frac{p(y|x)}{p(x)} \right) - 1 \quad (11)$$

So I don't think the channel dispersion can be bounded in anything better than  $O(\sqrt{n} \log n)$ .