Information Geometry and Capacity Limits

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Challenges in the Design of Structured Codes

- Generalize capacity region and scaling laws, parameterized by various system assumptions and constraints, characterized as a function of energy, delay, and outage

- Why is short and structured codes important in ITMANET?
 - Real-time applications, simple protocols, and wireless environments give sharp delay and reliability constraints;
 - Cooperative transmission requires early decoding (partial or soft information);
 - Network coding results suggest transmission of small pieces of information over the network;
 - Conjecture: joint source-channel-network codes require new performance metrics (eg. MDS);
- Why is this hard?
 - Shannon's results based on long block codes;
 - Randomization as the key step of information proofs;
 - Error exponents not well-understood:

$$E_{sp}(R) = \max_{P_x} \max_{\rho \ge 0} \left\{ -\log(\sum_{y} (\sum_{x} P_{\mathbf{x}}(x) W_{\mathbf{y}|\mathbf{x}}(y|x)^{(1/1+\rho)})^{1+\rho}) - \rho R \right\}$$

Advantage of Geometric Approach

i-Projection and Sanov's theorem:

Exponential family characterizing typical error event



Differential analysis allows layered coding structure with good reliability; eg. unequal error protection:

$$(M1, M2, ..., M_k) \Rightarrow X \to Y_1 \to Y_2 \to \ldots \to Y_k$$

Relation to estimation related quantities allowing new encoder/decoder designs;

$$D(P_t||P_0) = \int \int \mathbf{F}$$
isher Information

Useful Properties of Information Geometry



Capacity and Error Exponents



• For $R > R_{crit}$, + active,

$$E_r(R) = D(p_t || p_1),$$
 where $D(p_t || p_0) = R$

For $R < R_{crit}$, + not active,

$$E_r(R) = \left[\min_t D(p_t||p_0) + D(p_t||p_1)\right] - R$$

Capacity with Mismatched Decoders

Mismatched Receiver with decoding metric $F : \mathcal{X} \times \mathcal{Y} \to \Re$: pick the codeword with the largest value of

$$\sum_{i=1}^{n} F(x_i, y_i)$$

• ML decoder corresponds to
$$F(x, y) = \log \frac{P_{xy}(x, y)}{P_x(x)P_y(y)}$$
.

- Motivations
 - Wireless channel with non-perfect CSIR;
 - \checkmark F maybe much simpler to evaluate than likelihood.
- Known results (Kaplan et.al.'94, Csiszar et.al.'95): optimization over input distributions difficult.
- Slightly different question: best mismatch out of a linear family of decoding metrics.

Geometric Solutions

For given input distribution P_x , channel W,

$$R = \min_{Q_{xy}: E_Q[F] = E_{\mu_J}[F], Q_y = P_y} D(Q||\mu_P)$$



Very Noisy Sitation





- Euclidean approximation of probability model
- Quadratic approximation of divergence, oth order approximation of Fisher information

Mismatched capacity in very noisy channels:

$$C = \frac{1}{2} \frac{\langle \tilde{F}, L \rangle}{\|\tilde{F}\|^2}$$

Layered Codes

Two message tree code:



In very noisy case, at high rates,

$$\Pr(E_1) = \exp\left[-n\left(\sqrt{C} - \sqrt{aR_1} - \sqrt{(1-a)R_2}\right)^2\right]$$
$$\Pr(E_2) = \exp\left[-(1-a)n\left(\sqrt{C} - \sqrt{R_2/(1-a)}\right)^2\right]$$

Degraded broadcasting network (ISIT 07)