

# Capacity of the Binomial Channel, or Minimax Redundancy for Memoryless Sources

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**Abstract** — This paper computes the capacity of the binomial channel of order  $n$ , for any finite  $n$  via convex optimization and duality. The binomial channel capacity is also the minimax coding redundancy for a class of memoryless sources. The exact capacity for finite  $n$  computed here is compared to the asymptotic expression found by Xie and Barron in [1].

## I. INTRODUCTION

Consider the “biased coin channel” or the binomial channel of order  $n$ . The input is a real random variable  $\Theta \in [0, 1]$ , and the output is the result of  $n$  coin flips with probability of heads  $\Theta$ . The number of heads  $X \in \{k, k = 0, 1, \dots, n\}$  follows the binomial distribution with parameters  $n$  and  $\theta$ :

$$t_k(\theta) \stackrel{\text{def}}{=} P(X = k | \theta) = \binom{n}{k} \theta^k (1 - \theta)^{n-k}, k = 0, \dots, n. \quad (1)$$

This paper presents an algorithm to determine the binomial channel capacity  $C_n = \max_{dP(\theta)} I_n(\Theta; X)$  for any finite  $n$ , based on convexity and duality. This problem was solved asymptotically (as  $n \rightarrow \infty$ ) in [1], treated from the dual viewpoint of universal encoding of a discrete memoryless Bernoulli source. The asymptotic minimax redundancy for universal encoding [1, p. 647] was computed:

$$\lim_{n \rightarrow \infty} \left( \max_{dP(\theta)} \int_0^1 D(\mathbf{t}(\theta) \| \mathbf{q}) dP(\theta) \right) = \frac{1}{2} \log \frac{n}{2\pi e} + \log \pi, \quad (2)$$

where  $\mathbf{q}$  is the pmf on  $X$  induced by the input distribution  $dP(\theta)$  and the “channel law”  $\mathbf{t}(\theta)$ . Fig. 1 presents the binomial capacity for finite  $n$  along with the asymptotic result of [1].

Although the input alphabet of the binomial channel is uncountably infinite (the unit interval), the capacity achieving input distribution is concentrated at a finite number  $J$  of points  $\theta_1, \theta_2, \dots, \theta_J$  for any  $n$ , and  $J \leq n + 1$ . This was proven in [2] using Dubins’ theorem (for a similar result see also Gallager’s Cor. 4.5.3).

Despite the discreteness of the optimum “capacity achieving” input distribution, the well-known Blahut-Arimoto algorithm cannot be used to compute the capacity, because neither the input support  $\{\theta_1, \theta_2, \dots, \theta_J\}$  nor its cardinality  $J$  are known *a-priori*. Furthermore,  $I_n(\Theta; X)$  is not convex in the variables  $\theta_1, \dots, \theta_J$ . Our solution provides the optimum input support as a by-product: the support points  $\theta_1, \dots, \theta_J$  are located where the slack variables of the Lagrange dual problem become zero.

## II. BINOMIAL CHANNEL CAPACITY

Consider  $\mathcal{U}$  a  $\sigma$ -field on  $[0, 1]$  and  $dF(\theta) \in \mathcal{F}$ , the space of signed measures on  $([0, 1], \mathcal{U})$ . Then the binomial channel capacity becomes a convex problem in Euclidean space:

$$\text{minimize} \quad \sum_{k=0}^n q_k \log q_k + \int_0^1 h(\theta) dF(\theta) \quad (3)$$

subject to the inequality constraint  $dF(\theta) \geq 0$  and the equality constraints  $q_k = \int_0^1 t_k(\theta) dF(\theta)$ ,  $k = 0, 1, \dots, n$ , and  $\int_0^1 dF(\theta) = 1$ , where we define  $h(\theta) = H(X|\Theta = \theta) = -\sum_{k=0}^n t_k(\theta) \log t_k(\theta)$ .

Taking the Lagrange dual function and eliminating the slack variable of the inequality constraint, leads to the dual problem:

$$\min_{\mathbf{z}} \left\{ \sum_{k=0}^n 2^{-z_k} + \frac{e}{\log e} \cdot \max_{0 \leq \theta \leq 1} \left[ \sum_{k=0}^n z_k t_k(\theta) - h(\theta) \right] \right\}. \quad (4)$$

This problem is finite-dimensional in the dual variable  $\mathbf{z}$  in the Euclidean space  $\mathbf{R}^{n+1}$ , and may be solved via standard convex optimization techniques, such as the ellipsoid algorithm [3].

The optimal output pmf  $\mathbf{q}^*$  is recovered from the dual optimal  $\mathbf{z}^*$ :  $q_k^* = 1/e \cdot 2^{-z_k^*}$ ,  $k = 0, \dots, n$ . The “complementary slackness” condition for inequality constraints [5, Thm. 2.1] (similar to [4, p. 218]) forces the input support to consist of exactly those  $\theta_m$  where the inner function  $\sum_{k=0}^n z_k^* t_k(\theta) - h(\theta)$  of (4) attains its maxima. By [2], there can only be  $J \leq n + 1$  such  $\theta_m$ , and usually there are significantly fewer. Since  $\sum_{k=0}^n z_k^* t_k(\theta) - h(\theta) = -\log(e) + D(\mathbf{t}(\theta) \| \mathbf{q}^*)$ , complementary slackness is another way of stating Csiszár’s Min-Max Capacity Theorem [6], that  $C = \min_{\mathbf{q}} \max_{dF(\theta)} D(\mathbf{t}(\theta) \| \mathbf{q})$ . In other words, the optimum prior  $dF^*(\theta)$  induces an output pmf  $\mathbf{q}^*$  that is equidistant in the relative entropy sense from  $\mathbf{t}(\theta)$  at those points  $\theta$  where  $D(\mathbf{t}(\theta) \| \mathbf{q}^*)$  is maximized.

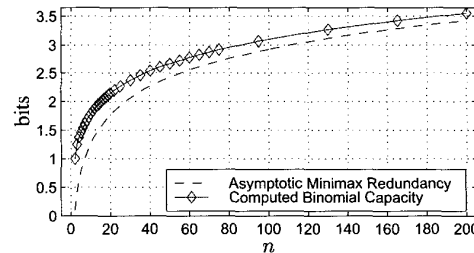


Figure 1: Asymptotic minimax redundancy in [1], and the binomial channel capacity for finite  $n$ .

The above method can also be applied to other problems with uncountably infinite input alphabet, where the capacity achieving prior distribution, as in the binomial channel, has finite support.

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