

From the book

Imre Csizsar - János Körner - Information Theory*
Coding Theorems for Discrete Memoryless Systems

Types

2 Types and typical sequences

LEMMA

Proof

we have

Hence

Most of the proof techniques used in this book will be based on a few simple combinatorial lemmas, summarized below.

Drawing k times independently with distribution Q from a finite set X , the probability of obtaining the sequence $\mathbf{x} \in X^k$ depends only on how often the various elements of X occur in \mathbf{x} . In fact, denoting by $N(a|\mathbf{x})$ the number of occurrences of $a \in X$ in \mathbf{x} , we have

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$$Q^k(\mathbf{x}) = \prod_{a \in X} Q(a)^{N(a|\mathbf{x})}. \quad (2.1)$$

DEFINITION 2.1 The *type* of a sequence $\mathbf{x} \in X^k$ is the distribution $P_{\mathbf{x}}$ on X defined by

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$$P_{\mathbf{x}}(a) \triangleq \frac{1}{k} N(a|\mathbf{x}) \quad \text{for every } a \in X.$$

For any distribution P on X , the set of sequences of type P in X^k is denoted by T_P^k or simply T_P . A distribution P on X is called a *type of sequences in X^k* if $T_P^k \neq \emptyset$. \square

Sometimes the term "type" will also be used for the sets $T_P^k \neq \emptyset$ when this does not lead to ambiguity. These sets are also called *type classes* or *composition classes*.

Applyi

REMARK In mathematical statistics, if $\mathbf{x} \in X^k$ is a sample of size k consisting of the results of k observations, the type of \mathbf{x} is called the *empirical distribution* of the sample \mathbf{x} . \square

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By (2.1), the Q^k -probability of a subset of T_P is determined by its cardinality. Hence the Q^k -probability of any subset A of X^k can be calculated by combinatorial counting arguments, looking at the intersections of A with the various sets T_P separately. In doing so, it will be relevant that the number of different types in X^k is much smaller than the number of sequences $\mathbf{x} \in X^k$.

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LEMMA 2.2 (*Type counting*) The number of different types of sequences in X^k is less than $(k+1)^{|X|}$. \square

→ 2.1

Proof For every $a \in X$, $N(a|\mathbf{x})$ can take $k+1$ different values. \square

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The next lemma explains the role of entropy from a combinatorial point of view, via the asymptotics of a multinomial coefficient.

* Second Edition, Cambridge University Press, 2011.

LEMMA 2.3 For any type P of sequences in X^k

→ 2.2

$$(k+1)^{-|X|} \exp[kH(P)] \leq |\mathcal{T}_P| \leq \exp[kH(P)]. \quad \circ$$

Proof Since (2.1) implies

$$P^k(\mathbf{x}) = \exp[-kH(P)] \quad \text{if } \mathbf{x} \in \mathcal{T}_P$$

we have

$$|\mathcal{T}_P| = P^k(\mathcal{T}_P) \exp[kH(P)].$$

Hence it is enough to prove that

$$P^k(\mathcal{T}_P) \geq (k+1)^{-|X|}.$$

This will follow by the Type counting lemma if we show that the P^k -probability of $\mathcal{T}_{\hat{P}}$ is maximized for $\hat{P} = P$.

By (2.1) we have

$$P^k(\mathcal{T}_{\hat{P}}) = |\mathcal{T}_{\hat{P}}| \cdot \prod_{a \in X} P(a)^{k\hat{P}(a)} = \frac{k!}{\prod_{a \in X} (k\hat{P}(a))!} \prod_{a \in X} P(a)^{k\hat{P}(a)}$$

for every type \hat{P} of sequences in X^k .

It follows that

$$\frac{P^k(\mathcal{T}_{\hat{P}})}{P^k(\mathcal{T}_P)} = \prod_{a \in X} \frac{(kP(a))!}{(k\hat{P}(a))!} P(a)^{k(\hat{P}(a)-P(a))}.$$

Applying the obvious inequality $\frac{n!}{m!} \leq n^{n-m}$, this gives

$$\frac{P^k(\mathcal{T}_{\hat{P}})}{P^k(\mathcal{T}_P)} \leq \prod_{a \in X} k^{k(P(a)-\hat{P}(a))} = 1. \quad \square$$

If X and Y are two finite sets, the *joint type* of a pair of sequences $\mathbf{x} \in X^k$ and $\mathbf{y} \in Y^k$ is defined as the type of the sequence $\{(x_i, y_i)\}_{i=1}^k \in (X \times Y)^k$. In other words, it is the distribution $P_{\mathbf{x}, \mathbf{y}}$ on $X \times Y$ defined by

$$P_{\mathbf{x}, \mathbf{y}}(a, b) \triangleq \frac{1}{k} N(a, b | \mathbf{x}, \mathbf{y}) \quad \text{for every } a \in X, b \in Y.$$

Joint types will often be given in terms of the type of \mathbf{x} and a stochastic matrix $V: X \rightarrow Y$ such that

$$P_{\mathbf{x}, \mathbf{y}}(a, b) = P_{\mathbf{x}}(a) V(b|a) \quad \text{for every } a \in X, b \in Y. \quad (2.2)$$

Note that the joint type $P_{\mathbf{x}, \mathbf{y}}$ uniquely determines $V(b|a)$ for those $a \in X$ which do occur in the sequence \mathbf{x} . For conditional probabilities of sequences $\mathbf{y} \in Y^k$, given a sequence $\mathbf{x} \in X^k$, the matrix V of (2.2) will play the same role as the type of \mathbf{y} does for unconditional probabilities.