

THE CAPACITY OF A CLASS OF CHANNELS¹

BY DAVID BLACKWELL, LEO BREIMAN, AND A. J. THOMASIAN

University of California, Berkeley

1. Summary. Shannon's basic theorem on the capacity of a channel is generalized to the case of a class of memoryless channels. A generalized capacity is defined and is shown to be the supremum of attainable transmission rates when the coding and decoding procedure must be satisfactory for every channel in the class.

2. Definitions and Introduction. For any positive integer n and any set \mathcal{A} we denote by $\mathcal{A}^{(n)}$ the set of all n -tuples (x_1, \dots, x_n) with each $x_i \in \mathcal{A}$.

A channel, denoted by $(\mathcal{A}, \mathcal{B}, P(y|x))$ or by $P(y|x)$, consists of two finite sets \mathcal{A}, \mathcal{B} having $a \geq 2, b \geq 2$ elements, respectively, and a set of probability distributions $P(\cdot|x)$ on \mathcal{B} , one for each $x \in \mathcal{A}$. $P(y|x)$ is interpreted as the probability of receiving $y \in \mathcal{B}$ given that $x \in \mathcal{A}$ was transmitted.

The n -extension of a channel $(\mathcal{A}, \mathcal{B}, P(y|x))$ is the channel $(\mathcal{A}^{(n)}, \mathcal{B}^{(n)}, P(v|u))$ where $v = (y_1, \dots, y_n) \in \mathcal{B}^{(n)}, u = (x_1, \dots, x_n) \in \mathcal{A}^{(n)}$ and $P(v|u) = \prod_{i=1}^n P(y_i|x_i)$.

When considering a class of channels, $(\mathcal{A}, \mathcal{B}, P_\gamma(y|x))$ for $\gamma \in \mathcal{C}$, where \mathcal{C} is an index set, we shall always assume that the \mathcal{A}, \mathcal{B} sets are the same for each channel in the class. We shall sometimes denote such a class of channels by \mathcal{C} , the index set.

A (G, ϵ_n, n) code for a class \mathcal{C} of channels for $G \geq 1, \epsilon_n \geq 0, n$ a positive integer, is a sequence of $[G]$ distinct elements of $\mathcal{A}^{(n)}$; $u_1, \dots, u_{[G]}$; where $[G]$ is the largest integer $\leq G$, and a sequence of $[G]$ disjoint subsets of $\mathcal{B}^{(n)}$; $B_1, \dots, B_{[G]}$; such that

$$P_\gamma(B_i^c | u_i) \leq \epsilon_n \quad \text{for } i = 1, \dots, [G] \quad \text{and all } \gamma \in \mathcal{C}.$$

The set $\{u_1, \dots, u_{[G]}\}$ is called the set of input messages of the code and B_i is called the decoding set for u_i . We think of an input letter u_i of the code as being selected arbitrarily and transmitted over an unknown one of the channels $P_\gamma, \gamma \in \mathcal{C}$. The letter v is received with probability $P_\gamma(v|u)$ and if $v \in B_j$ it is decoded as u_j . Thus, the probability is $\leq \epsilon_n$ that any input message u_i will be transmitted so as to be not decoded as u_i ; regardless of which channel in the class \mathcal{C} is used.

An $R \geq 0$ is an attainable transmission rate for a class \mathcal{C} of channels if there exists a sequence of (e^{Rn}, ϵ_n, n) codes for \mathcal{C} with $\epsilon_n \rightarrow 0$. Since $\mathcal{A}^{(n)}$ has only a^n points we know that any attainable rate $R \leq \log a$. Clearly 0 is an attainable rate for any class of channels. For any class of channels \mathcal{C} we define $T = T(\mathcal{C})$ to be the supremum of the set of attainable rates for \mathcal{C} .

Received February 16, 1959.

¹ This research was supported by the Office of Naval Research under Contract Nonr-222(53).

If $(\mathcal{A}, \mathcal{B}, P_\gamma(y|x))$ for $\gamma \in \mathcal{C}$ is a class of channels and $Q(x)$ is a given probability distribution on \mathcal{A} then for each $\gamma \in \mathcal{C}$ we let $P_\gamma(x, y) = P_\gamma(y|x)Q(x)$ and we define on $\mathcal{A} \times \mathcal{B}$ the random variable J_γ by

$$J_\gamma(x, y) = \log \frac{P_\gamma(x, y)}{P_\gamma(x)P_\gamma(y)} \quad \text{if } P_\gamma(x, y) > 0$$

$$= 0 \quad \text{if } P_\gamma(x, y) = 0.$$

The dependence of P_γ and J_γ on Q will usually not be exhibited. Since we will often be interested in expressions of the form $x \log x$ it is natural to define $\log 0 = 0$. We will denote the expectation of a random variable X with respect to the P_γ distribution by $E_\gamma X$. If \mathcal{C} has only one element we may drop the subscript γ . Finally for any class \mathcal{C} of channels we define the capacity of the class \mathcal{C} by

$$C(\mathcal{C}) = C = \sup_{Q(x)} \inf_{\gamma \in \mathcal{C}} E_\gamma J_\gamma$$

where the sup is over all distributions Q on \mathcal{A} .

In the case considered by Shannon, \mathcal{C} has only one element and our formula reduces to $C = \sup_Q EJ$, which is the usual formula for the capacity of a memoryless channel. Shannon's theorem then states that $T = C$. $T \geq C$, $T \leq C$ are called the direct and converse halves, respectively. This theorem for a single channel has been proved in various ways and under various conditions by Shannon [12], [13], McMillan [11], Feinstein [6], Khinchin [9], Wolfowitz [14], Blackwell, Breiman, and Thomasian [1]. We will show that within the framework that has been set up

$$T(\mathcal{C}) = C(\mathcal{C})$$

always holds true. This result follows immediately from Theorem 1 which also gives an exponential error bound for any rate $R < C$.

THEOREM 1: *Let $(\mathcal{A}, \mathcal{B}, P_\gamma(y|x))$ for $\gamma \in \mathcal{C}$ be any class of channels.*

(a) *For any integer n and any $R > 0$ such that $0 \leq C - R \leq 1/2$ there is an (e^{Rn}, ϵ_n, n) code for \mathcal{C} with*

$$\epsilon_n = Ae^{-\frac{(C-R)^2}{B}n}$$

where

$$A = \left[\frac{2^{10} ab^3}{(C-R)^2} \right]^{2ab} \quad \text{and} \quad B = 2^7 ab.$$

(b) *For any integer n and $R > C$ if $e^{Rn} \geq 2$ then any (e^{Rn}, ϵ_n, n) code for \mathcal{C} must satisfy*

$$\epsilon_n \geq 1 - \frac{C + \frac{\log 2}{n}}{R - \frac{\log 2}{n}}.$$

The sequence of steps used in proving Theorem 1 will be outlined. Theorem 2 presents a basic inequality, for a single channel, which is contained implicitly

in Feinstein [8]. This inequality is of independent interest since it gives the same bound for the maximum probability of error that Shannon [13] gives for the average probability of error. Theorem 2 permits a simple proof of $T \geq C$ for a single channel. Lemma 2 shows that \sup_Q in the definition of $C(\mathcal{C})$ can be replaced by \max_Q . Theorem 3 gives an exponential bound on the error of a code for one channel, which depends only on $a, b, (C - R)^2$. This is convenient in that the particular probabilities $P(y | x)$ may not be known and, in any case, need not be computed with. Results related to Theorem 3 have been given by Elias [3] and [4], Feinstein [7], Shannon [13], and Wolfowitz [14].

Lemma 3 generalizes the inequality of Theorem 2 to the case when \mathcal{C} has a finite number of elements, and Theorem 4 generalizes the exponential error bound of Theorem 3 to this case.

Lemma 4 shows that for a given \mathcal{A}, \mathcal{B} there is a large finite number of channels on \mathcal{A}, \mathcal{B} such that any channel on \mathcal{A}, \mathcal{B} is close, in several senses, to one of them. Lemma 5 shows that if a channel has a sequence of codes (e^{Rn}, ϵ_n, n) with $\epsilon_n = e^{-Bn}$ for large n , with $B > 0$, then this same sequence of codes can be used for all channels in a certain neighborhood of the channel. This result justifies some of our attention to exponential error bounds. The technique of Lemma 5 can also be used to get some similar results when the channel probabilities vary from letter to letter.

At this point the direct half of Theorem 1 is demonstrated by approximating the class \mathcal{C} of channels by a certain finite set of channels \mathcal{C}' from Lemma 4; obtaining an exponential error bound code for \mathcal{C}' from Theorem 4; and using Lemma 5 to show that such a code must be satisfactory for \mathcal{C} .

The converse half of Theorem 1 is then proved.

Before proceeding to the proofs we pause to clear up one point. It is obvious that

$$C(\mathcal{C}) \leq \inf_{\gamma \in \mathcal{C}} \sup_{Q(x)} E_{\gamma} J_{\gamma},$$

i.e., $C(\mathcal{C}) \leq$ the capacity of every channel in \mathcal{C} . We now exhibit an example where $C(\mathcal{C}) \neq \inf$ of the capacities of channels in \mathcal{C} . Let $\mathcal{A} = \mathcal{B} = \{1, 2, 3, 4\}$, $\mathcal{C} = \{1, 2\}$, and let $P_1(y | x)$ and $P_2(y | x)$ be defined by the left and right following matrices, respectively.

$$\begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{pmatrix} \begin{pmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

Let $Q(x)$ be any distribution on \mathcal{A} and let $H_i(Y) = -\sum_y P_i(y) \log P_i(y)$, $H_i(Y | X) = -\sum_x Q(x) \sum_y P_i(y | x) \log P_i(y | x)$. Using the fact that $\log x = (\log 2) \log_2 x$ we see that $(\log 2)^{-1} H_1(Y | X) = Q(1) + Q(2) + 2Q(3) + 2Q(4) = 1 + Q(3) + Q(4)$. Also from Feinstein [8], p. 15 we have $(\log 2)^{-1} H_1(Y) \leq 2$ so that $E_1 J_1 = H_1(Y) - H_1(Y | X) \leq (\log 2)(Q(1) + Q(2))$. Similarly $E_2 J_2 \leq (\log 2)(Q(3) + Q(4))$ so that $C(\mathcal{C}) \leq (1/2) \log 2$. The case $Q(i) = 1/4$ for $i = 1, \dots, 4$ shows that $C(\mathcal{C}) = (1/2) \log 2$; the case

$Q(1) = Q(2) = 1/2$ shows the capacity of channel one to be $\log 2$; the case $Q(3) = Q(4) = 1/2$ shows the capacity of channel two to be $\log 2$. Thus for this example

$$\frac{1}{2} \log 2 = C(\mathfrak{C}) < \inf_{\gamma \in \mathfrak{C}} \sup_{Q(x)} E_{\gamma} J_{\gamma} = \log 2.$$

3. A basic inequality.

THEOREM 2: For any channel $(\mathfrak{A}, \mathfrak{B}, P(y | x))$, any distribution $Q(x)$ on \mathfrak{A} , $\alpha > 0, G \geq 1$ there is a $(G, \epsilon, 1)$ code for the channel with $\epsilon = Ge^{-\alpha} + P(J \leq \alpha)$.

PROOF: It is clearly sufficient to construct an $(M, \epsilon, 1)$ code with the same ϵ as in the theorem and with $M \geq G$. Let $A = \{J > \alpha\}$ and for any $x_0 \in \mathfrak{A}$ let $A_{x_0} = \{(x, y) | (x_0, y) \in A\}$. $P(J \leq \alpha) \leq \epsilon$ so that $P(A) \geq 1 - \epsilon$, hence there is an x_1 such that $P(A | x_1) \geq 1 - \epsilon$. Let $B_1 = A_{x_1}$. (Each B_k will be a cylinder set with base in \mathfrak{B} . The base of B_k will be the decoding set for x_k .) At the k th step select x_k such that $P(B_k | x_k) \geq 1 - \epsilon$ where

$$B_k = \bigcup_1^k A_{x_i} - \bigcup_1^{k-1} A_{x_i}.$$

This process will terminate at some $M \geq 1$. For every x

$$P\left(A - A \cap \left(\bigcup_1^M A_{x_i}\right) \middle| x\right) < 1 - \epsilon$$

otherwise we could add this x to x_1, \dots, x_M contradicting the definition of M . Thus

$$\begin{aligned} P(A) &= P\left(A \cap \left(\bigcup_1^M A_{x_i}\right)\right) + P\left(A - A \cap \left(\bigcup_1^M A_{x_i}\right)\right) \\ &\leq \sum_1^M P(A_{x_i}) + 1 - \epsilon. \end{aligned}$$

Now if $(x, y) \in A$ then $J(x, y) > \alpha$ so that $P(y | x) > P(y)e^{\alpha}$. For fixed x sum both sides of this inequality over all y such that $(x, y) \in A$. Then

$$1 \geq P(A | x) \geq P(A_x)e^{\alpha}.$$

Thus $P(A_x) \leq e^{-\alpha}$ for any $x \in \mathfrak{A}$ so that $P(A) \leq Me^{-\alpha} + 1 - \epsilon$. Since $P(A) = Ge^{-\alpha} + 1 - \epsilon$, we have $M \geq G$. Clearly the B_1, \dots, B_M are disjoint and

$$P(B_k | x_k) \geq 1 - \epsilon$$

for $k = 1, \dots, M$ so the proof is completed.

Consider a single channel $(\mathfrak{A}, \mathfrak{B}, P(y | x))$ and let $Q(x)$ be specified and determine $P(x, y), J(x, y)$. Applying Theorem 2 to $(\mathfrak{A}^{(n)}, \mathfrak{B}^{(n)}, P(v | u))$ and $Q(u) = Q(x_1) \cdots Q(x_n)$ with $\alpha = n(R + EJ)/2, G = e^{Rn}$ we see that for any R such that $0 < R < EJ$ there is an (e^{Rn}, ϵ_n, n) code for $(\mathfrak{A}, \mathfrak{B}, P(y | x))$ with

$$\epsilon_n = e^{-(EJ-R)n/2} + P\left(\frac{1}{n} J' \leq \frac{R + EJ}{2}\right).$$

Now

$$J'(u, v) = \log \frac{P(u, v)}{P(u)P(v)} \quad \text{if } P(u, v) > 0$$

$$= 0 \quad \text{otherwise.}$$

Let $J''(u, v) = \sum_1^n J_i(x_i, y_i)$ where

$$J_i(x_i, y_i) = \log \frac{P(x_i, y_i)}{P(x_i)P(y_i)} \quad \text{if } P(x_i, y_i) > 0$$

$$= 0 \quad \text{otherwise.}$$

Clearly $P(J' = J'') = 1$ and J'' is the sum of n independent random variables each having the distribution of $J(x, y)$. Since $EJ > (R + EJ)/2$ we see that $\epsilon_n \rightarrow 0$. Now it is easily seen (and we will shortly prove even more) that for a fixed channel EJ is a continuous function of $(Q(x_1), Q(x_2), \dots, Q(x_n))$ and since the domain of the function is a closed bounded subset of Euclidean space the supremum is actually achieved. Thus for any channel $(\mathfrak{A}, \mathfrak{B}, P(y | x))$ there is a distribution $Q(x)$ on \mathfrak{A} such that $C = EJ$. Using this $Q(x)$ in the earlier portions of this paragraph we obtain the direct half of Shannon's theorem for a memoryless channel: $T \geq C$.

By introducing a brief epsilon argument in the proof of the direct half of Shannon's theorem we could clearly have ignored the question of whether or not there is a maximizing $Q(x)$. Although the fact that there is a maximizing $Q(x)$ in the general case of a class of channels is not vital in the following work, we will pause to prove this fact now. The proof is based on Lemma 1 which will be needed later.

LEMMA 1: Let $Q(x), Q'(x)$ be any two distributions on \mathfrak{A} such that

$$|Q(x) - Q'(x)| \leq \epsilon \leq 1/e \text{ for all } x \in \mathfrak{A}.$$

Then

$$|H(X) - H'(X)| \leq a\epsilon^{1/2}$$

where $H(X) = -\sum_x Q(x) \log Q(x)$ and $H'(X) = -\sum_x Q'(x) \log Q'(x)$.

PROOF: Let

$$f(y) = [-(y + \epsilon) \log (y + \epsilon)] - [-y \log y]$$

where $0 < \epsilon \leq 1/e$ and $0 \leq y \leq 1 - \epsilon$. Then $f(0) = -\epsilon \log \epsilon > 0$ and $f(1 - \epsilon) = (1 - \epsilon) \log (1 - \epsilon) < 0$ also

$$f'(y) = -\log (y + \epsilon) - 1 + \log y + 1 = \log \frac{y}{y + \epsilon} < 0$$

so that $|f(y)| \leq \max \{-\epsilon \log \epsilon, -(1 - \epsilon) \log(1 - \epsilon)\}$. Now

$$(1 - \epsilon) \log \frac{1}{1 - \epsilon} \leq (1 - \epsilon) \left(\frac{1}{1 - \epsilon} - 1 \right) = \epsilon \leq -\epsilon \log \epsilon$$

since $\epsilon \leq 1/e$. Thus

$$|f(y)| \leq -\epsilon \log \epsilon = \frac{\epsilon^{\frac{1}{2} \log \frac{1}{\epsilon}}}{\left(\frac{1}{\epsilon}\right)^{\frac{1}{2}}} \leq \epsilon^{\frac{1}{2}}$$

since $x^{1/2} - \log x \geq 2 - \log 4 > 0$ for $x > 0$. Applying the result $|f(y)| \leq \epsilon^{1/2}$ to $y = p$, $\epsilon = q - p$ where $0 \leq p \leq q \leq 1$ and $|q - p| \leq 1/e$ we see that

$$|[-p \log p] - [-q \log q]| \leq (|p - q|)^{1/2}$$

which easily gives us the bound on $|H(X) - H'(X)|$ completing the proof.

LEMMA 2: For any class of channels $(\mathfrak{A}, \mathfrak{B}, P_\gamma(y|x))$ for $\gamma \in \mathfrak{C}$,

$$C = \max_{Q(x)} \inf_{\gamma \in \mathfrak{C}} E_\gamma J_\gamma.$$

PROOF: Let $(\mathfrak{A}, \mathfrak{B}, P(y|x))$ be a channel and $Q(x)$ a distribution on \mathfrak{A} determining $P(x, y) = P(y|x)Q(x)$ and $J(x, y)$. Clearly $EJ = H(X) + H(Y) - H(X, Y)$ where $H(X) = -\sum_x P(x) \log P(x)$, $H(Y) = -\sum_y P(y) \log P(y)$, $H(X, Y) = -\sum_{x,y} P(x, y) \log P(x, y)$. Let $Q'(x)$ be another distribution on \mathfrak{A} determining $P'(x, y) = P(y|x)Q'(x)$ and $J'(x, y)$, and note that $E'J' = H'(X) + H'(Y) - H'(X, Y)$ where the primed quantities have analogous definitions. Assume that $|Q(x) - Q'(x)| \leq \epsilon \leq 1/e$ for all $x \in \mathfrak{A}$. Clearly $|P(x, y) - P'(x, y)| \leq P(y|x) |Q(x) - Q'(x)| \leq \epsilon$ and $|P(y) - P'(y)| \leq \sum_x |P(x, y) - P'(x, y)| \leq a\epsilon$. Applying Lemma 1 we get

$$\begin{aligned} |EJ - E'J'| &\leq |H(X) - H'(X)| + |H(Y) - H'(Y)| \\ &\quad + |H(X, Y) - H'(X, Y)| \\ &\leq a\epsilon^{1/2} + b(a\epsilon)^{1/2} + ab\epsilon^{1/2} \leq (a + 2ab)\epsilon^{1/2}. \end{aligned}$$

Thus not only is EJ continuous in $Q(x)$ but it is continuous in $Q(x)$ uniformly in $Q(x)$ and $P(y|x)$. We easily take $\inf_{\gamma \in \mathfrak{C}}$ on the inequalities

$$E'_\gamma J'_\gamma - (a + 2ab)\epsilon^{1/2} \leq E_\gamma J_\gamma \leq E'_\gamma J'_\gamma + (a + 2ab)\epsilon^{1/2}$$

and see that $\inf_{\gamma \in \mathfrak{C}} E_\gamma J_\gamma$ is continuous in $Q(x)$ so that once again there is a maximizing $Q(x)$ and Lemma 2 is proved.

4. The error bound for one channel.

THEOREM 3: Let $(\mathfrak{A}, \mathfrak{B}, P(y|x))$ be any channel. For any integer n and any $R > 0$ such that $0 \leq C - R \leq 1/2$, there is an (e^{Rn}, ϵ_n, n) code for the channel with

$$\epsilon_n = 2e^{-\frac{(C-R)^2}{16ab} n}.$$

PROOF: Applying Theorem 2 to $(\mathfrak{Q}^{(n)}, \mathfrak{B}^{(n)}, P(v|u))$ with $Q(u) = Q(x_1) \cdots Q(x_n)$, where $Q(x)$ is any distribution on \mathfrak{A} , $G = e^{Rn}$, $\alpha = (R + \theta)n$ we see that for any $R > 0$, $\theta > 0$ there is an (e^{Rn}, ϵ_n, n) code for $(\mathfrak{A}, \mathfrak{B}, P(y|x))$ with

$$\epsilon_n = e^{-n\theta} + P(J'' \leq n(R + \theta))$$

where, as shown in Section 3, J'' is the sum of n independent random variables, each having the distribution of $J(x, y)$. Select $R > 0$, $0 \leq EJ - R \leq 1/2$ and let $\theta = (EJ - R)^2$. Then $R + \theta \leq R + (EJ - R)/2 = (EJ + R)/2$.

Thus it remains only to show that

$$P(J'' \leq n(EJ + R)\frac{1}{2}) \leq e^{-\frac{(EJ-R)^2}{16ab}n}$$

(we will need this result later) for we can then choose Q so that $C = EJ$.

A method due to Chernoff [2] will be used to bound the probability in question. Let $0 \leq t \leq 1$, then

$$\begin{aligned} P\left(0 \leq \frac{n(EJ + R)}{2} - J''\right) &\leq Ee^{t\left[\frac{n(EJ+R)}{2} - J''\right]} = e^{\frac{tn(EJ+R)}{2}} Ee^{-tJ''} \\ &= \left[e^{\frac{t(EJ+R)}{2}} Ee^{-tJ}\right]^n \end{aligned}$$

so that we need show only that for a proper selection of t ,

$$e^{\frac{t(EJ+R)}{2}} Ee^{-tJ} \leq e^{-\frac{(EJ-R)^2}{16ab}}$$

Now

$$Ee^{-tJ} = 1 - tEJ + \frac{t^2}{2} EJ^2 e^{-\theta t}, \quad 0 < \theta < 1.$$

We need consider only (x, y) with $P(x, y) > 0$. Terms in $EJ^2 e^{-\theta t}$ are of the form

$$\begin{aligned} P(x, y) \left(\frac{P(x)P(y)}{P(x,y)}\right)^{\theta t} \log^2 \frac{P(x, y)}{P(x)P(y)} &\leq P(x, y) \left(\frac{1}{P(x, y)}\right)^{\theta t} \log^2 \frac{P(x, y)}{P(x)P(y)} \\ &\leq (P(x, y))^{1-t} \log^2 \frac{P(x, y)}{P(x)P(y)} \leq (P(x, y))^{1-t} \log^2 P(x, y) \end{aligned}$$

where the last inequality followed from $P(x, y) \leq P(x)P(y)/P(x, y) \leq 1/P(x, y)$. Also

$$\begin{aligned} [(P(x, y))^{\frac{1-t}{2}} \log P(x, y)]^2 &= \left(\frac{2}{1-t}\right)^2 [(P(x, y))^{\frac{1-t}{2}} \log P(x, y)]^{\frac{1-t}{2}}]^2 \\ &\leq \left(\frac{2}{1-t}\right)^2 \frac{1}{e^2} \leq \frac{1}{(1-t)^2}. \end{aligned}$$

Thus

$$Ee^{-tJ} \leq 1 - tEJ + \frac{t^2}{2} \frac{ab}{(1-t)^2} \leq e^{-tEJ + \frac{t^2}{2} \frac{ab}{(1-t)^2}}$$

so that

$$e^{\frac{t(EJ+R)}{2}} Ee^{-tJ} \leq e^{-\frac{1}{2}J(t)}$$

where

$$f(t) = (EJ - R)t - t^2 \frac{ab}{(1 - t)^2}.$$

Let $t = (EJ - R)/4ab \leq 1/8$ so that $1/(1 - t)^2 \leq (8/7)^2$, then

$$f\left(\frac{EJ - R}{4ab}\right) \geq \frac{(EJ - R)^2}{4ab} \left[1 - \left(\frac{8}{7}\right)^2 \frac{1}{4}\right] \geq \frac{(EJ - R)^2}{8ab}$$

completing the proof.

5. The error bound for a finite set of channels. Lemma 3 is needed in the proof of Theorem 4.

LEMMA 3: Let $(\mathcal{A}, \mathcal{B}, P_\gamma(y | x))$ for $\gamma \in \mathcal{C} = \{1, 2, \dots, L\}$ be a finite class of channels and let $Q(x)$ be a distribution on \mathcal{A} , determining $P_\gamma(x, y), J_\gamma(x, y)$.

(a) Define a channel $(\mathcal{A}, \mathcal{B}, P(y | x))$ by $P(y | x) = (1/L) \sum_{\gamma=1}^L P_\gamma(y | x)$ and let $Q(x)$ determine $P(x, y), J(x, y)$. Then for all α, δ

$$P(J \leq \alpha) \leq \frac{1}{L} \sum_{\gamma=1}^L P_\gamma(J_\gamma \leq \alpha + \delta) + Le^{-\delta}.$$

(b) For any $\alpha > 0, G \geq 1, \delta > 0$ there is a $(G, \epsilon, 1)$ code for \mathcal{C} with

$$\epsilon = LGe^{-\alpha} + L^2e^{-\delta} + \sum_1^L P_\gamma(J_\gamma \leq \alpha + \delta).$$

PROOF: We first prove part (a).

$$P(J \leq \alpha) = \frac{1}{L} \sum P_\gamma(J \leq \alpha) \leq \frac{1}{L} \sum [P_\gamma(J_\gamma \leq \alpha + \delta) + P_\gamma(J_\gamma > \alpha + \delta; J \leq \alpha)]$$

so that we need only prove that $P_\gamma(A_\gamma) \leq Le^{-\delta}$ where $A_\gamma = (J_\gamma > \alpha + \delta; J \leq \alpha)$. For any $(x, y) \in A_\gamma$ with $P_\gamma(x, y) > 0$ we have

$$e^\alpha P(y) \geq P(y | x) \geq \frac{1}{L} P_\gamma(y | x) \geq \frac{1}{L} e^{\alpha + \delta} P_\gamma(y)$$

so that $P_\gamma(y) \leq Le^{-\delta} P(y)$. Summing this last inequality over all y such that there is an x with $(x, y) \in A_\gamma$ we get $P_\gamma(A_\gamma) \leq \sum P_\gamma(y) \leq Le^{-\delta}$ which completes the proof of part (a).

Applying Theorem 2 to the channel $P(y | x)$ defined in part (a) and then using part (a) to bound $P(J \leq \alpha)$ we find that there is a $(G, \epsilon_0, 1)$ code for $P(y | x)$ with

$$\epsilon_0 = Ge^{-\alpha} + P(J \leq \alpha) \leq Ge^{-\alpha} + \frac{1}{L} \sum_\gamma P_\gamma(J_\gamma \leq \alpha + \delta) + Le^{-\delta}.$$

Now $P_\gamma(y | x) \leq LP(y | x)$ so that if x_i is an input letter for the $(G, \epsilon_0, 1)$ code and B_i is its decoding set, then $P_\gamma(B_i^c | x_i) \leq L P_\gamma(B_i^c | x_i) \leq L\epsilon_0$. Thus the $(G, \epsilon_0, 1)$ code for $P(y | x)$ is a $(G, L\epsilon_0, 1)$ code for \mathcal{C} and the lemma is proved.

THEOREM 4: *Let $(\mathfrak{A}, \mathfrak{B}, P_\gamma(y|x))$ for $\gamma \in \mathfrak{C} = \{1, 2, \dots, L\}$ be a finite class of channels. For any $R > 0$ such that $0 \leq C - R \leq 1/2$ there is an (e^{Rn}, ϵ_n, n) code with*

$$\epsilon_n = 2L^2 e^{-\frac{(C-R)^2}{16ab}n}.$$

PROOF. Applying part (b) of Lemma 3 to the class of channels $(\mathfrak{A}^{(n)}, \mathfrak{B}^{(n)}, P_\gamma(v|u))$ with $Q(u) = Q(x_1) \cdots Q(x_n)$ and $Q(x)$ a distribution for which $C = \inf_{\gamma \in \mathfrak{C}} E_\gamma J_\gamma$ and $G = e^{Rn}$, $\alpha = (R + \theta/2)n$, $\delta = \theta n/2$ we see that there is an (e^{Rn}, ϵ_n, n) code for \mathfrak{C} with

$$\epsilon_n = (L + L^2)e^{-\theta^2 n} + \sum_1^L P_\gamma \left(\frac{1}{n} J_\gamma \leq R + \theta \right).$$

Let $\theta = (C - R)^2$ and note that $R + (C - R)^2 \leq R + (C - R)/2 \leq R + (E_\gamma J_\gamma - R)/2 = (E_\gamma J_\gamma + R)/2$. Thus,

$$\epsilon_n \leq (L + L^2)e^{-\frac{(C-R)^2}{16ab}n} + \sum_1^L P_\gamma \left(\frac{1}{n} J_\gamma \leq \frac{1}{2}(R + E_\gamma J_\gamma) \right).$$

Now

$$P_\gamma \left(\frac{1}{n} J_\gamma \leq \frac{1}{2}(R + E_\gamma J_\gamma) \right) \leq P_\gamma \left(\frac{1}{n} J_\gamma \leq \frac{1}{2}(R' + E_\gamma J_\gamma) \right)$$

where $R' = E_\gamma J_\gamma - (C - R) \geq R$ and $0 \leq E_\gamma J_\gamma - R' \leq 1/2$. Therefore, we can apply the result obtained in the proof of Theorem 3 and get

$$P_\gamma \left(\frac{1}{n} J_\gamma \leq \frac{1}{2}(R' + E_\gamma J_\gamma) \right) \leq e^{-\frac{(E_\gamma J_\gamma - R')^2}{16ab}n} = e^{-\frac{(C-R)^2}{16ab}n}.$$

Now $L \geq 2$ so that $2L + L^2 = L(L + 2) \leq 2L^2$ and since Theorem 4 reduces to Theorem 3 for $L = 1$, the proof is completed.

6. The direct half of Theorem 1. Lemmas 4 and 5 are needed for the proof of part (a) of Theorem 1.

LEMMA 4: *Let $\mathfrak{A}, \mathfrak{B}$ be given. For every integer $M \geq 2b^2$ there is a class of channels $(\mathfrak{A}, \mathfrak{B}, P_j(y|x))$ with $\epsilon \in \mathfrak{D}_M$, where \mathfrak{D}_M has at most $(M + 1)^{ab}$ elements, such that for any channel $(\mathfrak{A}, \mathfrak{B}, P(y|x))$ there is a channel $(\mathfrak{A}, \mathfrak{B}, P'(y|x))$ in \mathfrak{D}_M such that:*

- (a) $|P(y|x) - P'(y|x)| \leq b/M$ for all x, y .
- (b) $P(y|x) \leq e^{2b^2/M} P'(y|x)$ for all x, y .
- (c) For any distribution $Q(x)$ on \mathfrak{A} let $P(x, y) = P(y|x)Q(x)$, $P'(x, y) = P'(y|x)Q(x)$, then

$$|EJ - E'J'| \leq 2b \left(\frac{b}{M} \right)^{1/2}.$$

PROOF. Let \mathfrak{D}_M be the class of channels $(\mathfrak{A}, \mathfrak{B}, P(y|x))$ such that for all x, y we have $MP(y|x) =$ an integer. Clearly \mathfrak{D}_M has at most $(M + 1)^{ab}$ elements. Given the distributions $P(y|x)$ we will first construct $P'(y|x)$ and prove (a),

(b). For this purpose it is enough to carry out the construction for one x_0 . Arrange the “b” numbers $P(y | x_0)$ in ascending order and designate them by $p_1 \leq p_2 \leq \dots \leq p_b$. For $i = 1, \dots, (b - 1)$ select p'_i uniquely by $p_i \leq p'_i < p_i + 1/M$, $Mp'_i =$ an integer. p'_i will be $P'(y | x_0)$ with the y being the one corresponding to p_i . Clearly

$$p_i \leq e^{\frac{2b^2}{M}} p'_i \quad \text{and} \quad |p_i - p'_i| \leq \frac{b}{M}$$

for $i = 1, \dots, (b - 1)$. It remains to show that if $p'_b = 1 - \sum_{i=1}^{b-1} p'_i$ then $p'_b \geq 0$ and p'_b, p'_b satisfy the same relations. Now

$$p'_b \geq 1 - \sum_1^{b-1} \left(p_i + \frac{1}{M} \right) \geq p_b - \frac{b}{M} \geq \frac{1}{b} - \frac{b}{M} \geq \frac{1}{b} - \frac{1}{2b} = \frac{1}{2b}.$$

Thus p'_1, \dots, p'_b form a distribution and $p_b \geq p'_b \geq p_b - b/M$ so that

$$|p_b - p'_b| \leq b/M.$$

Also

$$p_b \leq p'_b + \frac{b}{M} \leq p'_b + \frac{2b^2}{M} \frac{1}{2b} \leq p'_b \left(1 + \frac{2b^2}{M} \right) \leq e^{\frac{2b^2}{M}} p'_b$$

completing the proof of parts (a) and (b).

In the proof of part (c) we will use part (a) and Lemma 1. In order to use Lemma 1 we observe that $b/M \leq 1/2b \leq 1/4 < 1/e$. We also note that

$$|P(y) - P'(y)| \leq \sum_x |P(y | x) - P'(y | x)| Q(x) \leq b/M.$$

Now

$$\begin{aligned} |EJ - E'J'| &\leq \left| \left[-\sum_y P(y) \log P(y) \right] - \left[-\sum_y P'(y) \log P'(y) \right] \right| \\ &+ \left| \left[-\sum_{x,y} P(x, y) \log P(x, y) \right] - \left[-\sum_{x,y} P'(x, y) \log P'(x, y) \right] \right| \leq b \left(\frac{b}{M} \right)^{1/2} \\ &+ \sum_x Q(x) \left| \left[-\sum_y P(y | x) \log P(y | x) \right] - \left[-\sum_y P'(y | x) \log P'(y | x) \right] \right| \\ &\leq b \left(\frac{b}{M} \right)^{1/2} + b \left(\frac{b}{M} \right)^{1/2} \end{aligned}$$

and the lemma is proved.

LEMMA 5: Let $(\mathfrak{A}, \mathfrak{B}, P'(y | x))$, $(\mathfrak{A}, \mathfrak{B}, P(y | x))$ be two channels and A a non-negative number such that $P(y | x) \leq e^A P'(y | x)$ for all x, y . Any (e^{Rn}, ϵ_n, n) code for $(\mathfrak{A}, \mathfrak{B}, P'(y | x))$ is an $(e^{Rn}, \epsilon_n e^{An}, n)$ code for $(\mathfrak{A}, \mathfrak{B}, P(y | x))$.

PROOF: Let $u = (x_1, \dots, x_n) \in \mathfrak{A}^{(n)}$, $v = (y_1, \dots, y_n) \in \mathfrak{B}^{(n)}$. Then

$$P(v | u) = \prod_1^n P(y_i | x_i) \leq e^{An} \prod_1^n P'(y_i | x_i) = e^{An} P'(v | u).$$

Thus for any subset D of $\mathfrak{B}^{(n)}$ and any $u \in \mathfrak{A}^{(n)}$ we have

$$P(D | u) \leq e^{An} P'(D | u).$$

Let $u_i \in \mathcal{A}^{(n)}$ be an input message and B_i the corresponding decoding set of an (e^{Rn}, ϵ_n, n) code for $(\mathcal{A}, \mathcal{B}, P'(y|x))$. Then

$$P(B_i^c | u_i) \leq e^{An} P'(B_i^c | u_i) \leq e^{An} \epsilon_n$$

and the proof is completed.

We turn now to the proof of part (a) of Theorem 1. For each $P(y|x) \in \mathcal{C}$ select a $P'(y|x) \in \mathcal{D}_M$ according to Lemma 4 and let \mathcal{C}' denote this set of channels. Let $C' = C(\mathcal{C}')$. Since \mathcal{C}' has at most $(M+1)^{ab}$ elements we know from Theorem 4 that if $R' > 0, 0 \leq C' - R' \leq 1/2$ then there is an $(e^{R'n}, \epsilon'_n, n)$ code for \mathcal{C}' with

$$\epsilon'_n = 2(M+1)^{2ab} e^{-\frac{(C'-R')^2}{16ab}n}$$

For each $P(y|x) \in \mathcal{C}$ there is a $P'(y|x) \in \mathcal{C}'$ such that

$$P(y|x) \leq e^{\frac{2b^2}{M}} P'(y|x)$$

so that from Lemma 5 the code which we have for \mathcal{C}' is an $(e^{R'n}, \epsilon_n, n)$ code for \mathcal{C} with

$$\epsilon_n = 2(M+1)^{2ab} \exp \left\{ -\left[\frac{(C'-R')^2}{16ab} - \frac{2b^2}{M} \right] n \right\}$$

Let $C = C(\mathcal{C})$ and let $Q(x)$ be a maximizing distribution for \mathcal{C} . We wish to show that C' cannot be very much smaller than C . For every $P'(y|x) \in \mathcal{C}'$ there is a $P(y|x) \in \mathcal{C}$ such that $EJ \leq E'J' + 2b(b/M)^{1/2}$ where we use $Q(x)$ in both cases. Thus for every $P'(y|x) \in \mathcal{C}'$

$$C = \inf_{\mathcal{C}} EJ \leq E'J' + 2b \left(\frac{b}{M} \right)^{1/2}$$

so that

$$C \leq \inf_{\mathcal{C}'} E'J' + 2b \left(\frac{b}{M} \right)^{1/2} \leq C' + 2b \left(\frac{b}{M} \right)^{1/2}$$

Let $R > 0$ be given such that $0 < C - R \leq 1/2$. We must show how to select R' and M to get our result into the final form.

We select an integer M such that

$$\frac{2^8 ab^3}{(C-R)^2} \leq M \quad \text{and} \quad (M+1) \leq \frac{2^9 ab^3}{(C-R)^2}$$

so that

$$2b \left(\frac{b}{M} \right)^{1/2} \leq \frac{C-R}{2} \quad \text{and} \quad \frac{2b^2}{M} \leq \frac{(C-R)^2}{2^7 ab}$$

We define R' by

$$C' - R' = C - R - 2b \left(\frac{b}{M} \right)^{1/2} \geq \frac{C-R}{2} > 0.$$

Clearly $C' - R' \leq 1/2$ so that we have an $(e^{R'n}, \epsilon_n, n)$ code for \mathfrak{C} with

$$\begin{aligned} \epsilon_n &\leq 2(M + 1)^{2ab} \exp - \left\{ \frac{(C - R)^2}{4(16ab)} - \frac{(C - R)^2}{2^7 ab} \right\} \\ &\leq 2 \left[\frac{2^9 ab^3}{(C - R)^2} \right]^{2ab} \exp - \left\{ \frac{(C - R)^2}{2^7 ab} \right\}. \end{aligned}$$

The inequality $C \leq C' + 2b(b/M)^{1/2}$ shows that $R' \geq R$ and an $(e^{R'n}, \epsilon_n, n)$ code for \mathfrak{C} can easily be reduced to an (e^{Rn}, ϵ_n, n) code for \mathfrak{C} so that part (a) of Theorem 1 is proved.

7. Converse half of Theorem 1. The proof is based on Lemma 6.

LEMMA 6: Let G be an integer, \mathfrak{A} a finite set and let u_1, \dots, u_G be distinct elements of $\mathfrak{A}^{(n)}$. Define $Q(x)$ on \mathfrak{A} by

$$Q(x) = \frac{1}{nG} \sum_{j=1}^G (\text{the number of times that } x \text{ appears in } u_j).$$

Then any (G, ϵ, n) code, for a channel $(\mathfrak{A}, \mathfrak{B}, P(y | x))$ which uses these u_1, \dots, u_G for inputs must satisfy

$$(1 - \epsilon) \log G - \log 2 \leq nEJ$$

where $Q(x)$ is used to define $P(x, y)$ and $J(x, y)$.

PROOF: Define a distribution $\nu(u)$ on $\mathfrak{A}^{(n)}$ by $\nu(u) = 1/G$ if u is one of u_1, \dots, u_G and $\nu(u) = 0$ otherwise. Define a distribution $P(u, v)$ on $\mathfrak{A}^{(n)} \times \mathfrak{B}^{(n)}$ by $P(u, v) = P(v | u)\nu(u)$ where $P(v | u)$ is obtained from the n -extension of $(\mathfrak{A}, \mathfrak{B}, P(y | x))$. Now define n distributions on $\mathfrak{A} \times \mathfrak{B}$ by

$$P^{(i)}(x, y) = P(y | x)\nu^{(i)}(x)$$

for $i = 1, \dots, n$ where

$$\nu^{(i)}(x) = \sum_{x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n} \nu(x_1, \dots, x_{i-1}, x, x_{i+1}, \dots, x_n)$$

and observe that $Q(x) = (1/n) \sum_{i=1}^n \nu^{(i)}(x)$. Thus, the lemma will be proved if the following chain of inequalities is proved.

$$\begin{aligned} (1 - \epsilon) \log G - \log 2 &\leq \sum_{u,v} P(u, v) \log \frac{P(u, v)}{P(u)P(v)} \\ &\leq \sum_{i=1}^n \sum_{x,y} P^{(i)}(x, y) \log \frac{P^{(i)}(x, y)}{P^{(i)}(x)P^{(i)}(y)} \leq n \sum_{x,y} P(x, y) \log \frac{P(x, y)}{P(x)P(y)}. \end{aligned}$$

Using $\log x = (\log 2) \log_2 x$ to convert a result from Feinstein [8], pp. 29, 39, 44; which is due to Fano [5]; we obtain the first inequality. The second inequality follows from page 30 of Feinstein [8]. We proceed to prove the third inequality. Now

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n \sum_{x,y} P^{(i)}(x, y) [\log P(y | x) - \log P^{(i)}(y)] \\ = \sum_{x,y} P(x, y) \log P(y | x) - \frac{1}{n} \sum_{i=1}^n \sum_y P^{(i)}(y) \log P^{(i)}(y) \end{aligned}$$

but

$$\begin{aligned} -\frac{1}{n} \sum_{i=1}^n \sum_y P^{(i)}(y) \log P^{(i)}(y) &\leq -\sum_y \left(\frac{1}{n} \sum_{i=1}^n P^{(i)}(y) \right) \log \left(\frac{1}{n} \sum_{i=1}^n P^{(i)}(y) \right) \\ &= -\sum_y P(y) \log P(y) = -\sum_{x,y} P(x, y) \log P(y) \end{aligned}$$

where this last inequality follows from Lemma 4 on page 16 of Feinstein [8]. Combining the above, we complete the proof of the third inequality and hence of the lemma.

From Lemma 6 we immediately obtain that if G is an integer then for any (G, ϵ, n) code for a class \mathcal{C} of channels there is a $Q(x)$ on \mathcal{A} such that

$$(1 - \epsilon) \log G - \log 2 \leq n \inf_{\gamma \in \mathcal{C}} E_\gamma J_\gamma \leq nC.$$

Now e^{Rn} may not be an integer but

$$\log [e^{Rn}] \geq \log (e^{Rn} - 1) \geq nR + \log(1 - e^{-Rn}) \geq nR - \log 2$$

so that

$$(1 - \epsilon) (nR - \log 2) \leq nC + \log 2$$

which completes the proof of part (b) of Theorem 1.

REFERENCES

- [1] DAVID BLACKWELL, LEO BREIMAN, A. J. THOMASIAN, "Proof of Shannon's transmission theorem for finite-state indecomposable channels," *Ann. Math. Stat.*, Vol. 29 (1958), pp. 1209-1220.
- [2] HERMAN CHERNOFF, "A measure of asymptotic efficiency for tests of a hypothesis based on the sum of observations," *Ann. Math. Stat.*, Vol. 23 (1952), pp. 493-507.
- [3] PETER ELIAS, "Coding for noisy channels," *I.R.E. Convention Record* (1955), part 4, pp. 37-44.
- [4] PETER ELIAS, "Coding for two noisy channels," *Proceedings of the London Symposium on Information Theory*, Butterworth Scientific Publications, London, 1955.
- [5] R. M. FANO, "Statistical theory of communication," notes on a course given at the Massachusetts Institute of Technology, 1952, 1954.
- [6] AMIEL FEINSTEIN, "A new basic theorem of information theory," *I.R.E. Trans. P.G.I.T.*, September, 1954, pp. 2-22.
- [7] AMIEL FEINSTEIN, "Error bounds in noisy channels without memory," *I.R.E. Trans. P.G.I.T.*, September, 1955, pp. 13-14.
- [8] AMIEL FEINSTEIN, *Foundations of Information Theory*, McGraw-Hill, New York, 1958.
- [9] A. I. KHINCHIN, "On the fundamental theorems of information theory," *Uspekhi Matematicheskikh Nauk.*, Vol. 21 (1956), pp. 17-75.
- [10] A. I. KHINCHIN, *Mathematical Foundations of Information Theory*, Dover Publications, Inc., 1957.
- [11] BROCKWAY McMILLAN, "The basic theorems of information theory," *Ann. Math. Stat.*, Vol. 24(1953), pp. 196-219.
- [12] C. E. SHANNON, "A mathematical theory of communication," *Bell System Technical Journal*, Vol. 27 (1948), pp. 379-423, and 623-656.
- [13] CLAUDE E. SHANNON, "Certain results in coding theory for noisy channels," *Information and Control*, Vol. 1 (1957), pp. 6-25.
- [14] J. WOLFOWITZ, "The coding of messages subject to chance errors," *Illinois J. Math.*, Vol. 1 (1957), pp. 591-606.