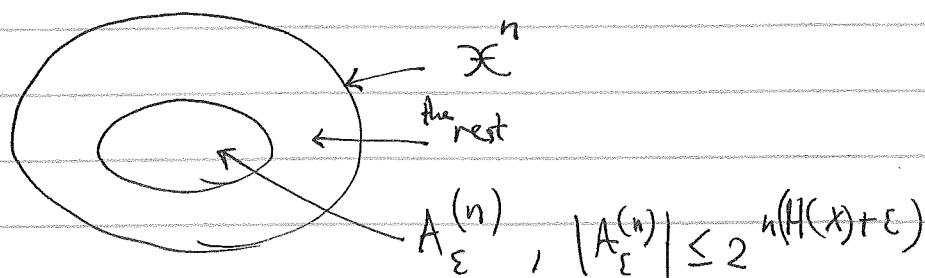


(4)

~~Source Coding by Typicality~~~~Source Coding by Typicality~~Source Coding by typicality: $X_1, \dots, X_n$ iid $\sim p(x)$ over $\mathcal{X}$.

* Encode each element in $A_\epsilon^{(n)}$ by its index

in $A_\epsilon^{(n)}$ following a zero in $\lceil n(H(X)+\epsilon) \rceil + 1$
 $\leq n \lceil H(X)+\epsilon \rceil + 2$ bits.

* Anything else encoded as (1, trivial) in
 $\leq n \log |\mathcal{X}| + 2$ bits.

* Easy to see: uniquely decodable.

* Take n large so that $\Pr(A_\epsilon^{(n)} \geq 1-\epsilon)$

Expected length: $\mathbb{E}(l(X_1^n)) = \sum_{x_1 \rightarrow x_n} p(x_1 \rightarrow x_n) l(x_1 \rightarrow x_n)$

$\approx \sim$

$$= \sum_{x_1, \dots, x_n \in A_\varepsilon^{(n)}} p(x_1, \dots, x_n) l(x_1, \dots, x_n)$$

$$+ \sum_{x_1, \dots, x_n \notin A_\varepsilon^{(n)}} p(x_1, \dots, x_n) l(x_1, \dots, x_n)$$

$$\leq \sum_{x_1, \dots, x_n \in A_\varepsilon^{(n)}} p(x_1, \dots, x_n) (n(H + \varepsilon) + 2)$$

$$+ \sum_{x_1, \dots, x_n \notin A_\varepsilon^{(n)}} p(x_1, \dots, x_n) (n \log |\mathcal{X}| + 2)$$

$$= \Pr((x_1, \dots, x_n) \in A_\varepsilon^{(n)}) (n(H + \varepsilon) + 2)$$

$$+ \Pr((x_1, \dots, x_n) \notin A_\varepsilon^{(n)}) (n \log |\mathcal{X}| + 2)$$

$$\leq n(H + \varepsilon) + \varepsilon n (\log |\mathcal{X}| + 2)$$

$$= n(H + \varepsilon') \longrightarrow nH(x)$$

$\varepsilon + \varepsilon \log |\mathcal{X}| + 2\varepsilon$ arbitrarily small.

Postscript: * Extension to the bounded-entropy
fixed-length universal source coding *
(Csiszár-Körner scheme)

11

Joint Typicality

Suppose $(X_1, Y_1), \dots, (X_n, Y_n) \text{ iid } \leftarrow (X, Y)$ (over $\mathcal{X} \times \mathcal{Y}$).

$A_\varepsilon^{(n)}$ for this joint distribution is defined as:

$$A_\varepsilon^{(n)} \stackrel{\text{def}}{=} \{(x_1^n, y_1^n) \in \mathcal{X}^n \times \mathcal{Y}^n :$$

$$\left| -\frac{\log p(x_1^n)}{n} - H(X) \right| < \varepsilon,$$

$$\left| -\frac{\log p(y_1^n)}{n} - H(Y) \right| < \varepsilon,$$

$$\left| -\frac{\log p(x_1^n, y_1^n)}{n} - H(X, Y) \right| < \varepsilon \}.$$

Joint AEP Theorems

1. $\Pr((X_1^n, Y_1^n) \in A_\varepsilon^{(n)}) \rightarrow 1$ as $n \rightarrow \infty$ (for large n)

2. $|A_\varepsilon^{(n)}| \leq 2^{n(H(X, Y) + \varepsilon)}$, $|A_\varepsilon^{(n)}| \geq (1 - \varepsilon) 2^{n(H(X, Y) - \varepsilon)}$.

3. If $(\tilde{X}_1^n, \tilde{Y}_1^n) \sim p(x_1^n)p(y_1^n)$, then

$$\Pr((\tilde{X}_1^n, \tilde{Y}_1^n) \in A_\varepsilon^{(n)}) \leq 2^{-n(I(X; Y) - 3\varepsilon)}$$

&

$$\geq (1 - \varepsilon) 2^{-n(I(X; Y) + 3\varepsilon)}.$$

proof of lower bound in (2).

for ^{large} $\forall n$, $\Pr(A_\epsilon^{(n)}) \geq 1 - \epsilon \Rightarrow$

$$\begin{aligned} 1 - \epsilon &\leq \sum_{(x^n, y^n) \in A_\epsilon^{(n)}} p(x^n, y^n) \\ &\leq |A_\epsilon^{(n)}| 2^{-n(H(X, Y) - \epsilon)} \Rightarrow \checkmark \end{aligned}$$

upper bound

Proof:

(2) ✓ follows from 1-variable AEP.
(lower bound ←)

Pf (1): By weak law of large numbers:

$\forall \varepsilon > 0, \exists n_0$ s.t. $\forall n > n_0$,

$$\Pr \left(\left| -\frac{\log P(X_1^n)}{n} - H(X) \right| \geq \varepsilon \right) \leq \frac{\varepsilon}{3}$$

and the same for Y and (X, Y) :

$$\Pr \left(\left| -\frac{\log P(Y_1^n)}{n} - H(Y) \right| \geq \varepsilon \right) \leq \frac{\varepsilon}{3},$$

$$\Pr \left(\left| -\frac{\log P(X_1^n, Y_1^n)}{n} - H(X, Y) \right| \geq \varepsilon \right) \leq \frac{\varepsilon}{3}.$$

Take a union bound and done! //

Pf (3):

$$\Pr \left((\tilde{X}_1^n, \tilde{Y}_1^n) \in A_\varepsilon^{(n)} \right) = \sum_{(x_1^n, y_1^n) \in A_\varepsilon^{(n)}} P(x_1^n) P(y_1^n)$$

$$\leq \underbrace{2^{n(H(X, Y) + \varepsilon)}}_{\text{by (2)}} \cdot 2^{-n(H(X) - \varepsilon)} 2^{-n(H(Y) - \varepsilon)}$$

$$= 2^{-n(I(X; Y) - 3\varepsilon)} \quad \checkmark \quad \begin{matrix} \& \\ \text{(lower bound)} \\ \text{similar} \end{matrix}$$

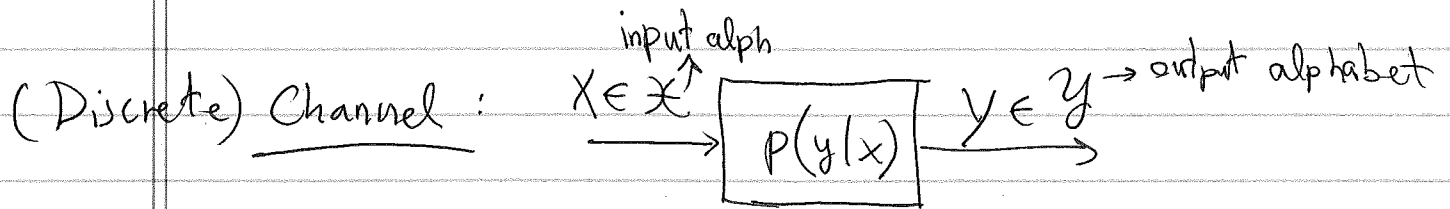
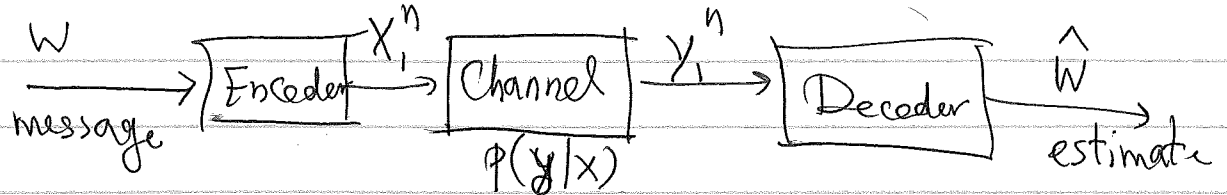
~~Second part:~~

~~For large n , $\Pr(A_\varepsilon^{(n)}) \geq 1 - \varepsilon$~~

$$\checkmark \quad (1 - \varepsilon) \leq \sum_{(x_1^n, y_1^n) \in A_\varepsilon^{(n)}} P(x_1^n, y_1^n) \leq \frac{|A_\varepsilon^{(n)}|}{2^{n(H(X, Y) - \varepsilon)}} \quad \checkmark$$

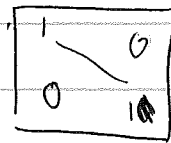
Channel Coding

Basic "point-to-point" Communication System:
"no-feedback"

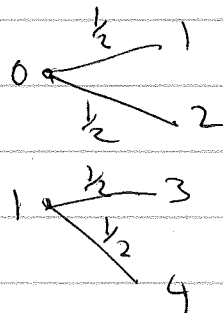


Channel can be described by a "probability transition matrix" $p(y/x)$:

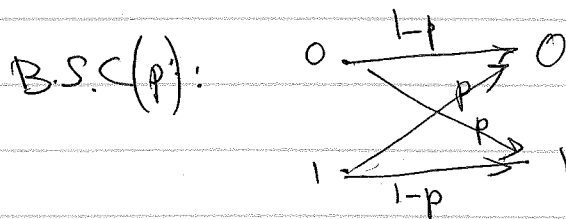
Ex. Noiseless:



"practically noiseless":

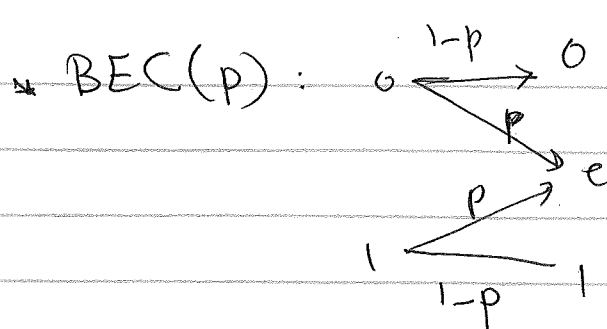


	1	2	3	4	
0	1/2	1/2	0	0	0
1	0	0	1/2	1/2	1



$$p(y/x) = \begin{pmatrix} 1-p & p \\ p & 1-p \end{pmatrix}$$

(14)



$$P(y|x) = \begin{pmatrix} 1-p & p & 0 \\ 0 & p & 1-p \end{pmatrix}$$

and col's

* Symmetric Channel: Rows are permutations of each other.

ex: BSC(p) and $P(y|x) = \begin{pmatrix} .3 & .2 & .5 \\ .5 & .3 & .2 \\ .2 & .5 & .3 \end{pmatrix}$
 (doubly stochastic)

* Capacity

$$C \stackrel{\text{def}}{=} \max_{P(x)} I(X;Y).$$

Ex: * Noiseless: $C=1$ ($P(x) = (\frac{1}{2}, \frac{1}{2})$).

* "practically noiseless": $C=1$ ($P(x) = (\frac{1}{2}, \frac{1}{2})$)

* BSC(p): $I(X;Y) = H(Y) - H(Y|X)$

$$= H(Y) - \sum P(x) \overbrace{H(Y|X=x)}^{h(p)} \rightarrow h(p)$$

$$= H(Y) - h(p) \leq 1 - h(p).$$

equality for $P(x) = (\frac{1}{2}, \frac{1}{2}) \Rightarrow C = 1 - h(p).$

~~* BSC(p):~~

(15)

* BEC (p): same as before,

$$C = \max_{p(x)} H(Y) - h(p)$$

$$\begin{aligned} H(Y) &= H(Y, \text{"erasure"}) = H(\text{erasure}) + H(Y|\text{erasure}) \\ &= h(p) + \underbrace{H(Y|\text{erasure})}_{(1-p)h(p_1), \quad p_1 := \Pr(X=1)} \end{aligned}$$

$$\Rightarrow C = \max_{p_1} (1-p) h(p_1) = 1-p \quad (\text{for } p_1 = 1/2)$$

Ex: Symmetric Channel

$$\begin{aligned} I(X; Y) &= H(Y) - H(Y|X) \\ &= H(Y) - H(\text{row}) \end{aligned}$$

$$\leq \log |Y| - H(\text{row}),$$

equality iff Y is uniform,
achieved by uniform X .