

# Chapter 5

## Weak Typicality

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Department of Information Engineering  
The Chinese University of Hong Kong

# The Notion of Typical Sequences

- Toss a fair coin  $n$  times.
- If the outcome is “head” approximately half of the time, the sequence of outcomes is “normal”, or “typical”.
- How to measure the typicality of a sequence w.r.t. to a generic distribution of an i.i.d. process?
- Two common such measures in information theory: weak typicality and strong typicality.
- The main theorems are weak and strong *Asymptotic Equipartition Properties* (AEP), which are consequences of WLLN.

# 5.1 The Weak AEP

Setup

- $\{X_k, k \geq 1\}$ ,  $X_k$  i.i.d.  $\sim p(x)$ .
- $X$  denotes generic r.v. with entropy  $H(X) < \infty$ .
- $\mathbf{X} = (X_1, X_2, \dots, X_n)$ . Then

$$p(\mathbf{X}) = p(X_1)p(X_2) \cdots p(X_n).$$

- $\mathcal{X}$  may be countably infinite.

### Theorem 5.1 (Weak AEP I)

$$-\frac{1}{n} \log p(\mathbf{X}) \rightarrow H(X)$$

in probability as  $n \rightarrow \infty$ , i.e., for any  $\epsilon > 0$ , for  $n$  sufficiently large,

$$\Pr \left\{ \left| -\frac{1}{n} \log p(\mathbf{X}) - H(X) \right| \leq \epsilon \right\} > 1 - \epsilon.$$

**Note:**  $X_n \rightarrow X$  in probability means that

$$\lim_{n \rightarrow \infty} \Pr\{|X_n - X| \geq \epsilon\} = 0$$

for all  $\epsilon > 0$ .

**Definition 5.2** The weakly typical set  $W_{[X]_\epsilon}^n$  with respect to  $p(x)$  is the set of sequences  $\mathbf{x} = (x_1, x_2, \dots, x_n) \in \mathcal{X}^n$  such that

$$\left| -\frac{1}{n} \log p(\mathbf{x}) - H(X) \right| \leq \epsilon,$$

or equivalently,

$$H(X) - \epsilon \leq -\frac{1}{n} \log p(\mathbf{x}) \leq H(X) + \epsilon,$$

where  $\epsilon$  is an arbitrarily small positive real number. The sequences in  $W_{[X]_\epsilon}^n$  are called weakly  $\epsilon$ -typical sequences.

# Empirical Entropy

- $$-\frac{1}{n} \log p(\mathbf{x}) = -\frac{1}{n} \sum_{k=1}^n \log p(x_k)$$

is called the *empirical entropy* of the sequence  $\mathbf{x}$ .
- The empirical entropy of a weakly typical sequence is close to the true entropy  $H(X)$ .

**Theorem 5.2 (Weak AEP II)** The following hold for any  $\epsilon > 0$ :

1) If  $\mathbf{x} \in W_{[X]\epsilon}^n$ , then

$$2^{-n(H(X)+\epsilon)} \leq p(\mathbf{x}) \leq 2^{-n(H(X)-\epsilon)}.$$

2) For  $n$  sufficiently large,

$$\Pr\{\mathbf{X} \in W_{[X]\epsilon}^n\} > 1 - \epsilon.$$

3) For  $n$  sufficiently large,

$$(1 - \epsilon)2^{n(H(X)-\epsilon)} \leq |W_{[X]\epsilon}^n| \leq 2^{n(H(X)+\epsilon)}.$$

WAEP says that for large  $n$ ,

- the probability of occurrence of the sequence drawn is close to  $2^{-nH(X)}$  with very high probability;
- the total number of weakly typical sequences is approximately equal to  $2^{nH(X)}$ .

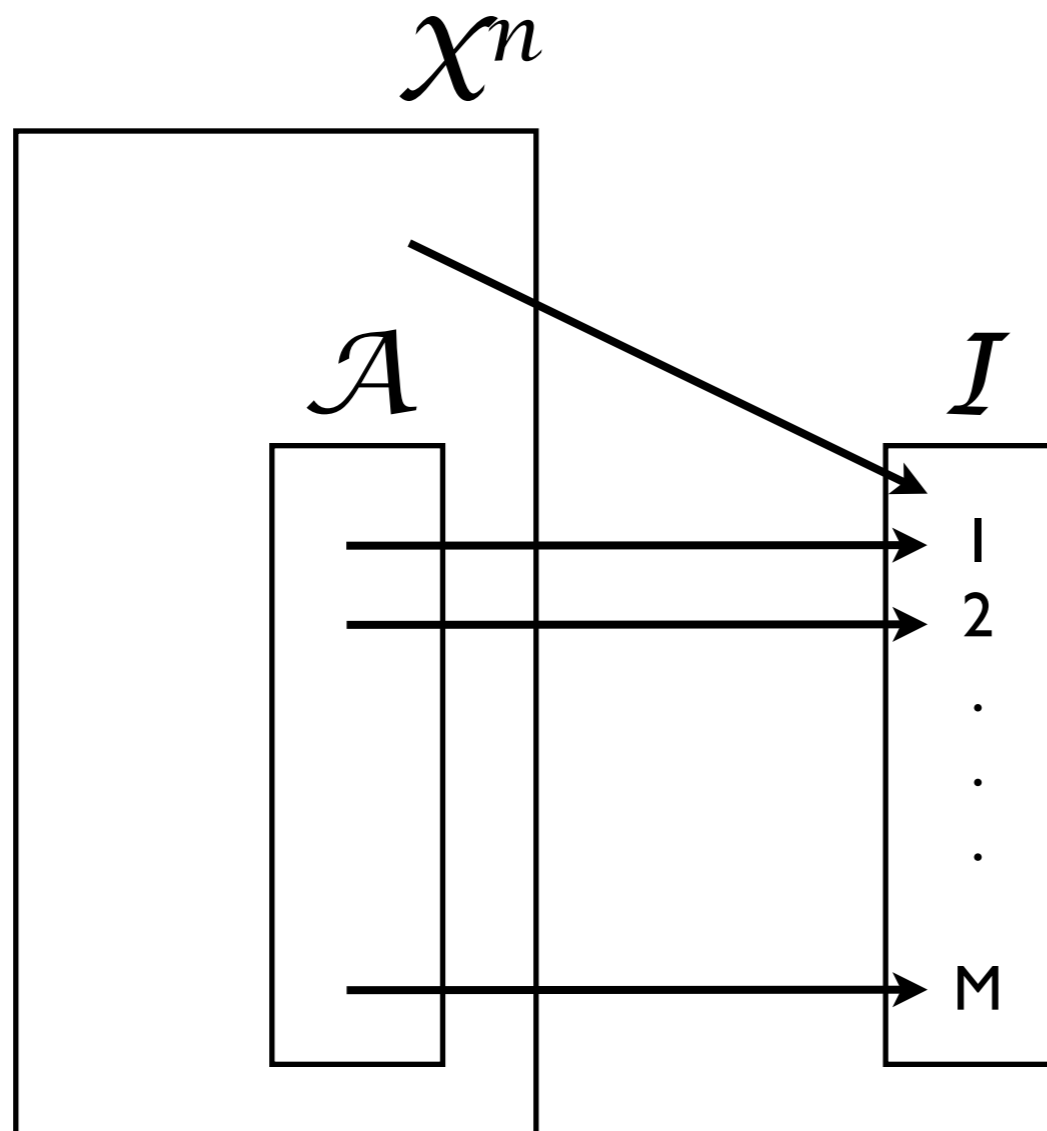
WAEP DOES NOT say that

- most of the sequences in  $\mathcal{X}^n$  are weakly typical;
- the most likely sequence is weakly typical.

When  $n$  is large, one can almost think of the sequence  $\mathbf{X}$  as being obtained by choosing a sequence from the weakly typical set according to the uniform distribution.



# 5.2 The Source Coding Theorem



A block code:  $\mathcal{X}^n \rightarrow \mathcal{I}$

- $\mathcal{I} = \{1, 2, \dots, M\}$
- blocklength =  $n$
- coding rate =  $n^{-1} \log M$
- $P_e = \Pr\{\mathbf{X} \notin \mathcal{A}\}$ .

# Direct Part & Converse

- Direct part: For arbitrarily small  $P_e$ , there exists a block code whose coding rate is arbitrarily close to  $H(X)$  when  $n$  is sufficiently large.
- Converse: For any block code with block length  $n$  and coding rate less than  $H(X) - \zeta$ , where  $\zeta > 0$  does not change with  $n$ , then  $P_e \rightarrow 1$  as  $n \rightarrow \infty$ .

# Direct Part

- Fix  $\epsilon > 0$  and take  $\mathcal{A} = W_{[X]\epsilon}^n$  and hence  $M = |\mathcal{A}|$ .
- For sufficiently large  $n$ , by WAEP,

$$(1 - \epsilon)2^{n(H(X) - \epsilon)} \leq M = |\mathcal{A}| = |W_{[X]\epsilon}^n| \leq 2^{n(H(X) + \epsilon)}.$$

- Coding rate satisfies

$$\frac{1}{n} \log(1 - \epsilon) + H(X) - \epsilon \leq \frac{1}{n} \log M \leq H(X) + \epsilon.$$

- By WAEP,

$$P_e = \Pr\{\mathbf{X} \notin \mathcal{A}\} = \Pr\{\mathbf{X} \notin W_{[X]\epsilon}^n\} < \epsilon.$$

- Letting  $\epsilon \rightarrow 0$ , the coding rate tends to  $H(X)$ , while  $P_e$  tends to 0.

# Converse

- Consider any block code with rate less than  $H(X) - \zeta$ , where  $\zeta > 0$  does not change with  $n$ . Then total number of codewords  $\leq 2^{n(H(X)-\zeta)}$ .
- Use some indices to cover  $\mathbf{x} \in W_{[X]_\epsilon}^n$ , and others to cover  $\mathbf{x} \notin W_{[X]_\epsilon}^n$ .
- Total probability of typical sequences covered is upper bounded by

$$2^{n(H(X)-\zeta)} 2^{-n(H(X)-\epsilon)} = 2^{-n(\zeta-\epsilon)}.$$

- Total probability covered is upper bounded by

$$2^{-n(\zeta-\epsilon)} + \Pr\{\mathbf{X} \notin W_{[X]_\epsilon}^n\} < 2^{-n(\zeta-\epsilon)} + \epsilon.$$

- Then  $P_e > 1 - (2^{-n(\zeta-\epsilon)} + \epsilon)$  holds for any  $\epsilon > 0$  and  $n$  sufficiently large.
- Take  $\epsilon < \zeta$ . Then  $P_e > 1 - 2\epsilon$  for  $n$  sufficiently large.
- Finally, let  $\epsilon \rightarrow 0$ .