## COMP2610 / COMP6261 - Information Theory

Lecture 10: Typicality and Asymptotic Equipartition Property

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#### Last time

Markov's inequality

Chebyshev's inequality

Law of large numbers

## Law of Large Numbers

#### Theorem

Let  $X_1, \ldots, X_n$  be a sequence of iid random variables, with

$$\mathbb{E}[X_i] = \mu$$

and  $\mathbb{V}[X_i] < \infty$ . Define

$$\bar{X}_n = \frac{X_1 + \ldots + X_n}{n}.$$

Then, for any  $\beta > 0$ ,

$$\lim_{n\to\infty} p(|\bar{X}_n - \mu| < \beta) = 1.$$

This is also called  $\bar{X}_n \to \mu$  in probability.

**Definition**: For random variables  $v_1, v_2, ...$ , we say  $v_n \to v$  in probability if for all  $\beta > 0$   $\lim_{n \to \infty} P(|v_n - v| > \beta) = 0$ .

 $\beta$  is fixed (not shrinking like  $\frac{1}{n}$ ). Not max/min. Reduction in variability.

#### This time

• Ensembles and sequences

Typical sets

Asymptotic Equipartition Property (AEP)

- Ensembles and sequences
  - Counting Types of Sequences

- 2 Typical sets
- 3 Asymptotic Equipartition Property (AEP)
- Wrapping Up

#### **Ensembles**

#### Ensemble

An ensemble X is a triple  $(x, \mathcal{A}_X, \mathcal{P}_X)$ ; x is a random variable taking values in  $\mathcal{A}_X = \{a_1, a_2, \dots, a_l\}$  with probabilities  $\mathcal{P}_X = \{p_1, p_2, \dots, p_l\}$ .

We will call  $A_X$  the alphabet of the ensemble

#### Ensembles

Example: Bent Coin



Let X be an ensemble with outcomes h for heads with probability 0.9 and t for tails with probability 0.1.

- The outcome set is  $A_X = \{h, t\}$
- The probabilities are

$$\mathcal{P}_X = \{ p_h = 0.9, p_t = 0.1 \}$$

We can also consider blocks of outcomes, which will be useful to describe sequences:

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#### Example (Coin Flips):

hhhhthhthh  o hh hh th ht ht	$(6 \times 2 \text{ outcome blocks})$
------------------------------	---------------------------------------

ightarrow hhh hth hth thh (4 imes 3 outcome blocks)

ightarrow hhhh thht hthh (3 imes 4 outcome blocks)

#### **Extended Ensemble**

Let X be a single ensemble. The **extended ensemble** of blocks of size N is denoted  $X^N$ . Outcomes from  $X^N$  are denoted  $\mathbf{x} = (x_1, x_2, \dots, x_N)$ . The **probability** of  $\mathbf{x}$  is defined to be  $P(\mathbf{x}) = P(x_1)P(x_2)\dots P(x_N)$ .

Example: Bent Coin



Let X be an ensemble with outcomes  $A_X = \{h, t\}$  with  $p_h = 0.9$  and  $p_t = 0.1$ .

Consider  $X^4$  – i.e., 4 flips of the coin.

 $\mathcal{A}_{X^4} = \{\mathtt{hhhh},\mathtt{hhht},\mathtt{hhth},\ldots,\mathtt{tttt}\}$ 

Example: Bent Coin



Let X be an ensemble with outcomes 
$$\mathcal{A}_X = \{ \mathtt{h}, \mathtt{t} \}$$
 with  $p_\mathtt{h} = 0.9$  and  $p_\mathtt{t} = 0.1$ .

Consider  $X^4$  – i.e., 4 flips of the coin.

$$\mathcal{A}_{X^4} = \{\text{hhhh}, \text{hhht}, \text{hhth}, \dots, \text{tttt}\}$$

$$\begin{split} P(\text{hhhh}) &= (0.9)^4 \approx 0.6561 \\ P(\text{tttt}) &= (0.1)^4 = 0.0001 \\ P(\text{hthh}) &= 0.9 \cdot 0.1 \cdot 0.9 \cdot 0.9 = (0.9)^3 (0.1) \approx 0.0729 \\ P(\text{htht}) &= 0.9 \cdot 0.1 \cdot 0.9 \cdot 0.1 = (0.9)^2 (0.1)^2 \approx 0.0081. \end{split}$$

Example: Bent Coin

#### Entropy of extended ensembles

We can view  $X^4$  as comprising 4 independent random variables, based on the ensemble X

Entropy is additive for independent random variables

Thus,

$$H(X^4) = 4H(X) = 4.(-0.9 \log_2 0.9 - 0.1 \log_2 0.1) = 1.88$$
bits.

More generally,

$$H(X^N) = NH(X).$$

## Criteria for dividing $2^N$ sequences into **types**

In the bent coin example,

$$(0.9)^{2}(0.1)^{2} = P(hhtt)$$

$$= P(htht)$$

$$= P(htth)$$

$$= P(thht)$$

$$= P(thht)$$

$$= P(thhh).$$

The order of outcomes in the sequence is irrelevant

Let *X* be an ensemble with alphabet  $A_X = \{a_1, \dots, a_l\}$ .

Let 
$$p(X = a_i) = p_i$$
.

For a sequence  $\mathbf{x} = x_1, x_2, \dots, x_N$ , how to compute  $p(\mathbf{x})$ ?

let  $n_i$  = # of times symbol  $a_i$  appears in **x** (symbol count)

Given the  $n_i$ 's, we can compute the probability of seeing **x**:

$$P(\mathbf{x}) = P(x_1) \cdot P(x_2) \cdot \ldots \cdot P(x_N)$$
  
=  $P(a_1)^{n_1} \cdot P(a_2)^{n_2} \cdot \ldots \cdot P(a_l)^{n_l}$   
=  $p_1^{n_1} \cdot p_2^{n_2} \dots p_l^{n_l}$ 

Sufficient statistics:  $\{n_1, n_2, \dots, n_l\}$ . Use it as a criteria of partitioning.

#### Sequence Types

Each unique choice of  $(n_1, n_2, \dots, n_l)$  gives a different type of sequence

- 4 heads, (3 heads, 1 tail), (2 heads, 2 tails), ...
- Sequences in each type are equiprobable.

For a given type of sequence how many sequences are there with these symbol counts?

# of sequences with 
$$n_i$$
 copies of  $a_i = \frac{N!}{n_1! n_2! \dots n_l!}$ 

$$\binom{N}{n_1} \binom{N-n_1}{n_2} \binom{N-n_1-n_2}{n_3} \dots$$

$$= \frac{N!}{n_1!(N-n_1)!} \cdot \frac{(N-n_1)!}{n_2!(N-n_1-n_2)!} \cdot \frac{(N-n_1-n_2)!}{n_3!(N-n_1-n_2-n_3)!} \dots$$

Example

#### Probability of types

Let 
$$A = \{a, b, c\}$$
 with  $P(a) = 0.2$ ,  $P(b) = 0.3$ ,  $P(c) = 0.5$ .

Example

#### Probability of **types**

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Each sequence of type  $(n_a, n_b, n_c) = (2, 1, 3)$  has length 6 and probability  $(0.2)^2(0.3)^1(0.5)^3 = 0.0015$ .

Example

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The probability **x** is of type (2, 1, 3) is  $(0.0015) \cdot 60 = 0.09$ .

Study probabilities at the level of types (most likely, average/typical)

- Ensembles and sequences
  - Counting Types of Sequences

- 2 Typical sets
- 3 Asymptotic Equipartition Property (AEP)
- Wrapping Up

#### Example

With  $p_{\rm h}=0.75$ , what are the probabilities for  $X^N$ ?

$$N = 2$$

X	$P(\mathbf{x})$
hh	0.5625
ht	0.1875
th	0.1875
tt	0.0625

#### Example

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$$N = 2$$

X	$P(\mathbf{x})$
hh	0.5625
ht	0.1875
th	0.1875
tt	0.0625

$$N=2$$
  $N=3$ 

X	$P(\mathbf{x})$
hhh	0.4219
hht	0.1406
hth	0.1406
thh	0.1406
htt	0.0469
tht	0.0469
tth	0.0469
ttt	0.0156

#### Example

With  $p_h = 0.75$ , what are the probabilities for  $X^N$ ?

N = 2

N = 3

N = 4

X	$P(\mathbf{x})$
hh	0.5625
ht	0.1875
th	0.1875
tt	0.0625
-	

X	$P(\mathbf{x})$
hhh	0.4219
hht	0.1406
hth	0.1406
thh	0.1406
htt	0.0469
tht	0.0469
tth	0.0469
ttt	0.0156

N=4			
×	$P(\mathbf{x})$	х	$P(\mathbf{x})$
hhhh	0.3164	thht	0.0352
hhht	0.1055	thth	0.0352
hhth	0.1055	tthh	0.0352
hthh	0.1055	httt	0.0117
thhh	0.1055	thtt	0.0117
htht	0.0352	ttht	0.0117
htth	0.0352	ttth	0.0117
hhtt	0.0352	tttt	0.0039

#### Observations

As *N* increases, there is an increasing spread of probabilities

The most likely single sequence will always be the all h's

However, for N = 4, the most likely sequence type is 3 h's and 1 t

Not surprising because  $3 = N \cdot p_h$ , pretty much average case.

## Symbol Frequency in Long Sequences

To judge if a sequence is typical/average, a natural question to ask is:

How often does each symbol appear in a sequence  $\mathbf{x}$  from  $X^N$ ?

Intuitively, in a sequence of length N, let  $a_i$  appear for  $n_i$  times. Then **in expectation** 

$$n_i \approx N \cdot P(a_i)$$

Note  $p_i = P(a_i)$ , and

$$P(\mathbf{x}) = P(a_1)^{n_1} P(a_2)^{n_2} \dots P(a_l)^{n_l} \approx p_1^{Np_1} p_2^{Np_2} \dots p_l^{Np_l}$$

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So the *information content*  $-\log_2 P(\mathbf{x})$  of that sequence is approximately

$$-p_1 N \log_2 p_1 - \ldots - p_l N \log_2 p_l = -N \sum_{i=1}^l p_i \log_2 p_i = NH(X)$$

We want to consider elements  $\mathbf{x}$  that have  $-\log_2 P(\mathbf{x})$  "close" to NH(X)

### Typical Set

For "closeness"  $\beta > 0$  the typical set  $T_{N\beta}$  for  $X^N$  is

$$T_{N\beta} \stackrel{\text{def}}{=} \left\{ \mathbf{x} : \left| -\log_2 P(\mathbf{x}) - NH(X) \right| < N\beta \right\}$$
$$= \left\{ \mathbf{x} : \left| -\frac{1}{N} \log_2 P(\mathbf{x}) - H(X) \right| < \beta \right\}$$

Union of types

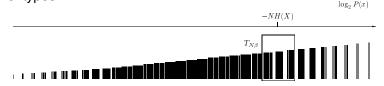
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Union of types



What when  $\beta = 0$  (and replace < by  $\le$ )?

Criterion based on information content. Other criterion (KL divergence)?

The name "typical" is used since  $\mathbf{x} \in T_{N\beta}$  will have roughly  $p_1 N$  occurrences of symbol  $a_1, p_2 N$  of  $a_2, \ldots, p_K N$  of  $a_K$ .

x lo	$g_2(P(\mathbf{x}))$
11111111	-50.1
	-37.3
1111111111	-65.9
1.11	-56.4
11	-53.2
	-43.7
111	-46.8
$\ldots 1 \ldots 1 \ldots 1 \ldots \ldots 1$	-56.4
1111111	-37.3
1	-43.7
111	-56.4
	-37.3
.111.111111	-56.4
1111111	-59.5
11.111111	-46.8
	-15.2
	-332.1

Randomly drawn sequences for P(1) = 0.1. Note:  $H(X) \approx 0.47$ 

**Properties** 

Typical sequences are nearly equiprobable: Every  $\mathbf{x} \in T_{N\beta}$  has

$$2^{-N(H(X)+\beta)} \le P(\mathbf{x}) \le 2^{-N(H(X)-\beta)}$$
.

Variation is small when  $\beta$  is small

Number of sequences in the typical set: For any N,  $\beta$ ,

$$|T_{N\beta}| \leq 2^{N(H(X)+\beta)}.$$

#### Proof of Cardinality Bound

For every  $\mathbf{x} \in T_{N\beta}$ ,

$$p(\mathbf{x}) \geq 2^{-N(H(X)-\beta)}.$$

Thus,

$$1 = \sum_{\mathbf{x}} p(\mathbf{x})$$

$$\geq \sum_{\mathbf{x} \in T_{N\beta}} p(\mathbf{x})$$

$$\geq \sum_{\mathbf{x} \in T_{N\beta}} 2^{-N(H(X) - \beta)}$$

$$= 2^{-N(H(X) - \beta)} \cdot |T_{N\beta}|.$$

Thus

$$|T_{N\beta}| \leq 2^{N(H(X)+\beta)}$$

Most Likely Sequence

The most likely sequence may not belong to the typical set

e.g. with  $p_h = 0.75$ , we have

$$-\frac{1}{4}\log_2 P(\text{hhhh}) = 0.4150$$

whereas H(X) = 0.8113

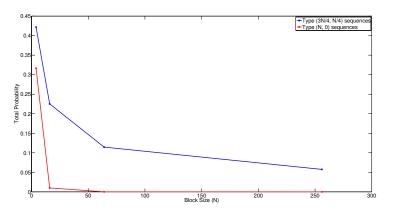
The most likely single sequence  $\rightarrow hhhh$ 

The most likely single sequence type  $\rightarrow$  {hhht, hthh, ...}

Most Likely Sequence

Probability of most likely sequence decays like  $(p_h)^N$   $(p_h = 0.75)$ 

Sequences with  $N \cdot p_h$  heads contain much more total probability mass



Blue curve corresponds to typical set with  $\beta = 0$ . What if  $\beta > 0$ ?

- Ensembles and sequences
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# Asymptotic Equipartition Property Eventually Equally Divided Informally

## Asymptotic Equipartition Property (Informal)

As  $N \to \infty$ ,  $\log_2 P(x_1, \dots, x_N)$  is close to -NH(X) with high probability.

For large block sizes "almost all sequences are typical" (i.e., in  $T_{N\beta}$ )

$$n(r)P(\mathbf{x}) = \binom{N}{r}p_1^r(1-p_1)^{N-r} \\ \begin{pmatrix} 0.14 \\ 0.12 \\ 0.04 \\ 0.08 \\ 0.06 \\ 0.04 \\ 0.02 \\ 0 \\ 10 \ 20 \ 30 \ 40 \ 50 \ 60 \ 70 \ 80 \ 90 \ 100 \end{pmatrix} \\ \begin{pmatrix} 0.044 \\ 0.035 \\ 0.035 \\ 0.025 \\ 0.01$$

Probability sequence **x** has *r* heads for N = 100 (left) and N = 1000 (right). Here P(X = head) = 0.1.

Formally

## Asymptotic Equipartition Property

If  $x_1, x_2, \ldots$  are i.i.d. with distribution P then, in probability,

$$-\frac{1}{N}\log_2 P(x_1,\ldots,x_N)\to H(X).$$

In precise language:

$$(\forall \beta > 0) \lim_{N \to \infty} p\left(\left|-\frac{1}{N}\log_2 P(x_1,\ldots,x_N) - H(X)\right| < \beta\right) = 1.$$

Exactly the probability of  $\mathbf{x} \in T_{N\beta}$ .

Formally

## Asymptotic Equipartition Property

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Exactly the probability of  $\mathbf{x} \in T_{N\beta}$ .

Recall definition: for random variables  $v_1, v_2, ...,$  we say  $v_N \to v$  in **probability** if for all  $\beta > 0$   $\lim_{N \to \infty} P(|v_N - v| > \beta) = 0$ 

Here  $v_N$  corresponds to  $-\frac{1}{N}\log_2 P(x_1,\ldots,x_N)$ .

Comments

#### Why is it surprising/significant?

For an ensemble with binary outcomes, and low entropy,

$$|T_{N\beta}| \leq 2^{NH(X)+\beta} \ll 2^N$$

i.e. the typical set is a small fraction of all possible sequences

AEP says that for N sufficiently large, we are virtually guaranteed to draw a sequence from this small set

Significance in information theory

Proof

Since  $x_1, \ldots, x_N$  are independent,

$$-\frac{1}{N}\log p(x_1,\ldots,x_N) = -\frac{1}{n}\log\prod_{n=1}^N p(x_n)$$
$$= -\frac{1}{N}\sum_{n=1}^N \log p(x_n).$$

Let 
$$Y=-\log p(X)$$
 and  $y_n=-\log p(x_n)$ . Then,  $y_n\sim Y$ , and  $\mathbb{E}[Y]=H(X)$ .

But then by the law of large numbers,

$$(\forall \beta > 0) \lim_{N \to \infty} \rho \left( \left| \frac{1}{N} \sum_{n=1}^{N} y_n - H(X) \right| > \beta \right) = 0.$$

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## Summary & Conclusions

Ensembles and sequences

Typical sets

Asymptotic Equipartition Property (AEP)

Next: Source Coding.