

# Digital Communication Systems

## EES 452

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## **2. Source Coding**

### **2.3 Source Extension:**

Introduction to Source Extension,  
Memoryless Source and Independent Symbols

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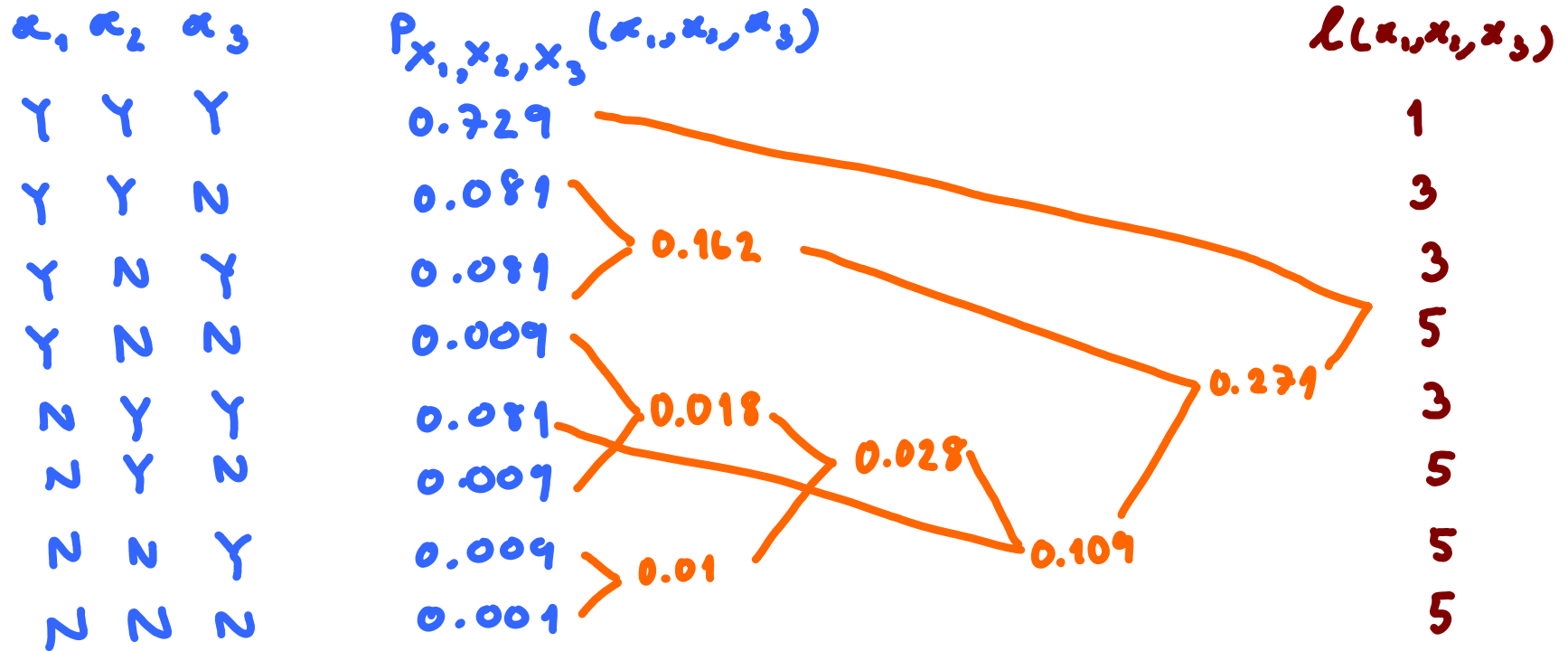
## **2. Source Coding**

### **2.3 Source Extension:**

Calculating the probabilities for Blocks of Source Symbols,  
Second- and Third-Order Extensions,  $L_n$

[Ex. 2.40]

# Huffman Coding: Source Extension





# Kronecker Product

- An operation on two matrices of arbitrary size
- Named after German mathematician Leopold Kronecker.
- If  $\mathbf{A}$  is an  $m$ -by- $n$  matrix and  $\mathbf{B}$  is a  $p$ -by- $q$  matrix, then the **Kronecker product**  $\mathbf{A} \otimes \mathbf{B}$  is the  $mp$ -by- $nq$  matrix

Use  
`kron(A, B)`  
in MATLAB.

$$\mathbf{A} \otimes \mathbf{B} = \begin{bmatrix} a_{11}B & \cdots & a_{1n}B \\ \vdots & \ddots & \vdots \\ a_{m1}B & \cdots & a_{mn}B \end{bmatrix}.$$

- Example

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \otimes \begin{bmatrix} 0 & 5 \\ 6 & 7 \end{bmatrix} = \begin{bmatrix} 1 \cdot 0 & 1 \cdot 5 & 2 \cdot 0 & 2 \cdot 5 \\ 1 \cdot 6 & 1 \cdot 7 & 2 \cdot 6 & 2 \cdot 7 \\ 3 \cdot 0 & 3 \cdot 5 & 4 \cdot 0 & 4 \cdot 5 \\ 3 \cdot 6 & 3 \cdot 7 & 4 \cdot 6 & 4 \cdot 7 \end{bmatrix} = \begin{bmatrix} 0 & 5 & 0 & 10 \\ 6 & 7 & 12 & 14 \\ 0 & 15 & 0 & 20 \\ 18 & 21 & 24 & 28 \end{bmatrix}.$$



# Kronecker Product

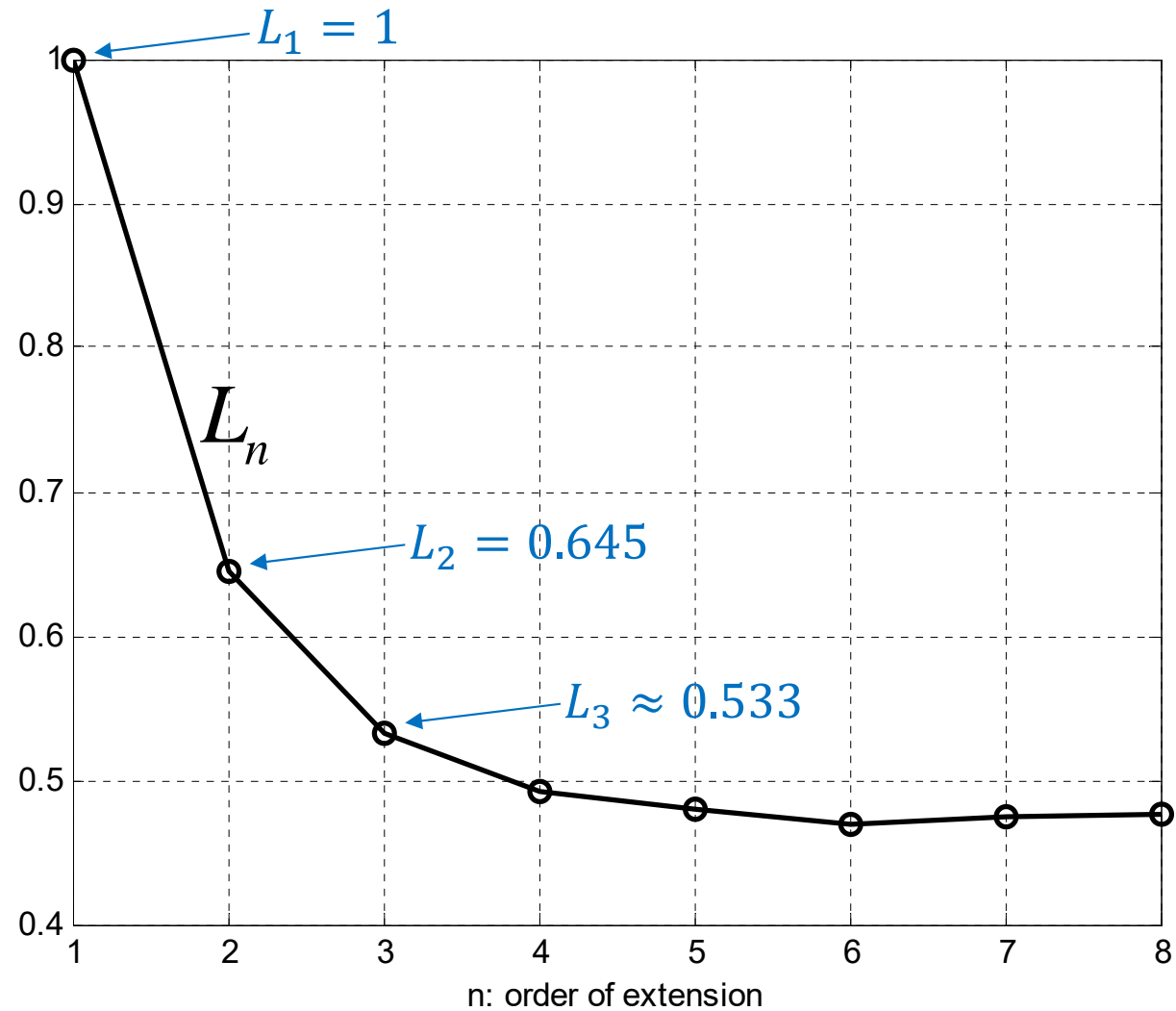
```
>> p = [0.9 0.1]
p =
    0.9000    0.1000
>> p2 = kron(p,p)
p2 =
    0.8100    0.0900    0.0900    0.0100
>> p3 = kron(p2,p)
p3 =
Columns 1 through 7
    0.7290    0.0810    0.0810    0.0090    0.0810    0.0090    0.0090
Column 8
    0.0010
```



[Ex.2.40]

# Huffman Coding: Source Extension

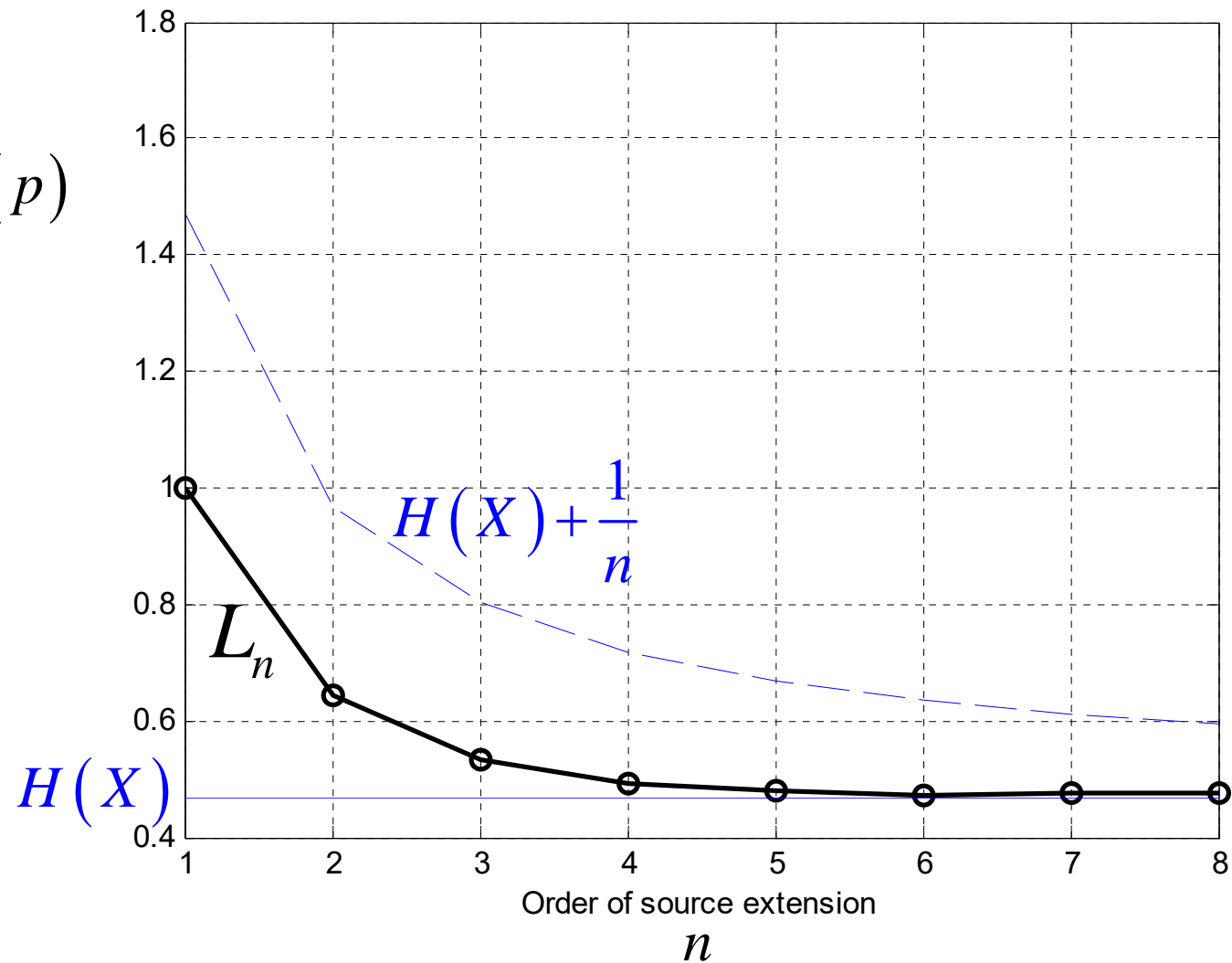
$$X_k \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(p)$$
$$p = 0.1$$





[Ex.2.40]

# Huffman Coding: Source Extension

$$X_k \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(p)$$
$$p = 0.1$$



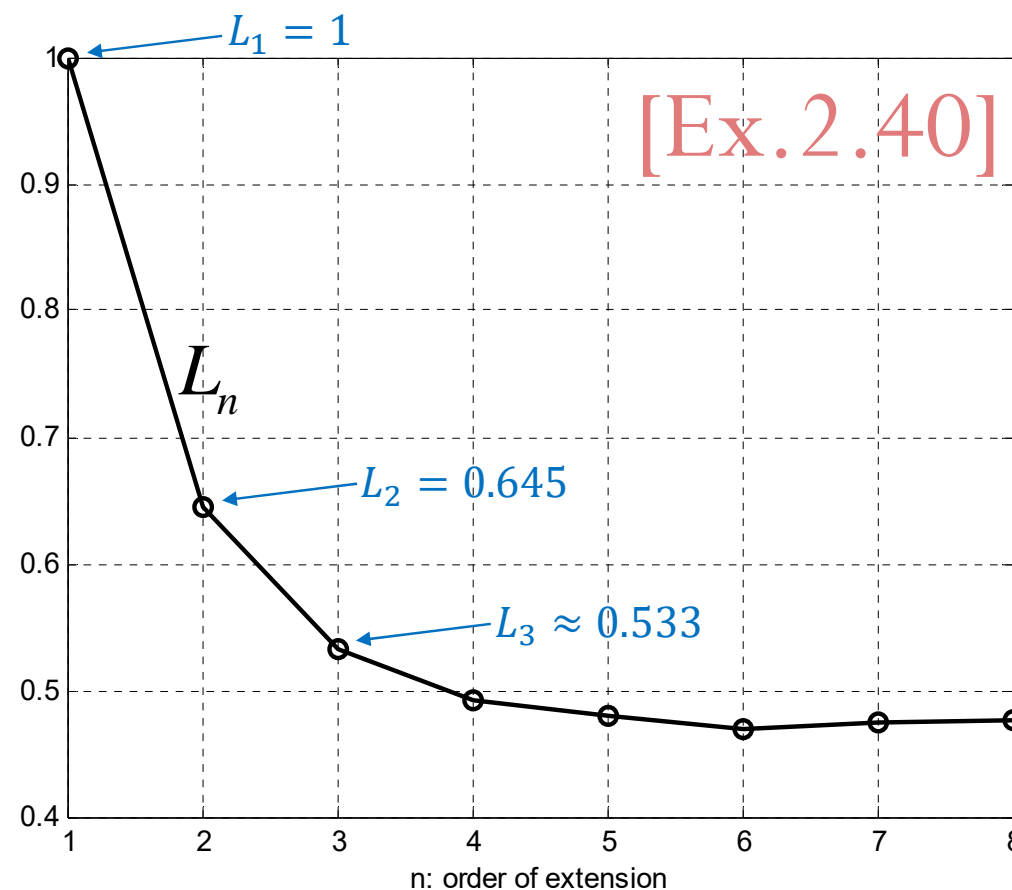
# Summary: Source Extension

Source Coding w/o Extension	Source Coding with Extension
The encoder operates on individual source symbols.	The encoder operates on blocks of consecutive source symbols. [Defn 2.39] <b><math>n</math>-th extension coding:</b> 1 block = $n$ successive source symbols
Source Alphabet 	Cartesian Product 
$c: \mathcal{A} \rightarrow \{0,1\}^*$	$c: (\mathcal{A})^n \rightarrow \{0,1\}^*$
Discussed in Sections 2.1-2.2.	Discussed in Section 2.3.
Assuming that the pmf $p(\mathbf{x})$ of the source symbols is known, the optimal encoding technique was discovered by Huffman.	The probabilities for the blocks $p(\underline{\mathbf{x}})$ are calculated from the memoryless property of the source. Huffman's technique can then be applied to the blocks.



# Summary: Source Extension

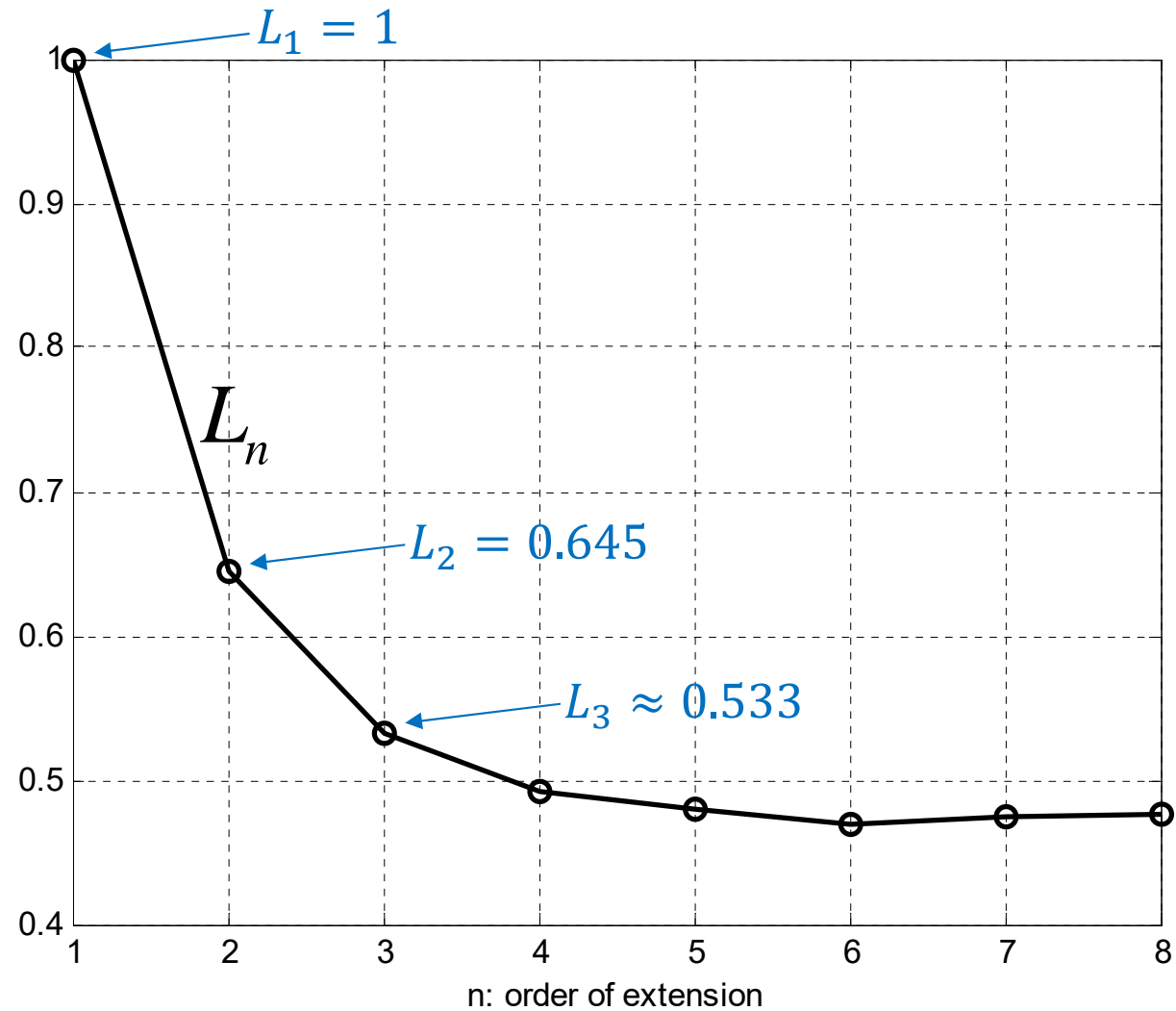
- $L_n$  = expected (average) codeword length per source symbol when Huffman coding is used with  $n$ -th extension
- $\lim_{n \rightarrow \infty} L_n = H(X)$



[Ex.2.40]

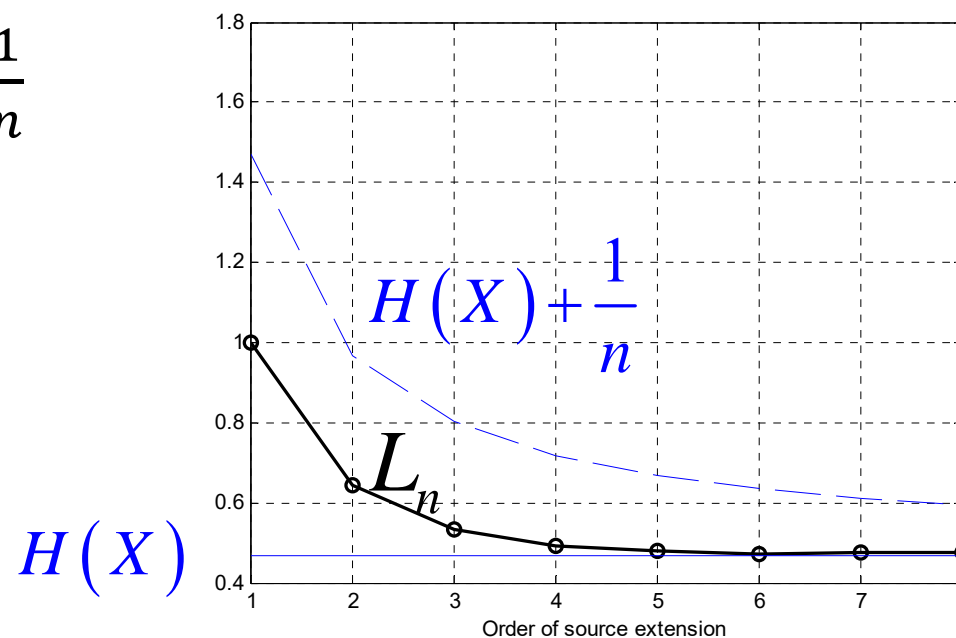
# Huffman Coding: Source Extension

$$X_k \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(p)$$
$$p = 0.1$$



# Summary: Source Extension

- The encoder operates on the blocks rather than on individual symbols.
- [Defn 2.39]  **$n$ -th extension coding**:  
1 block =  $n$  successive source symbols
- $L_n$  = expected (average) codeword length per source symbol when Huffman coding is used with  $n$ -th extension
  - $H(X) \leq L_n < H(X) + \frac{1}{n}$
  - $\lim_{n \rightarrow \infty} L_n = H(X)$



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## **2. Source Coding**

# Random Vectors

# Multiple Random Variables

- Consider a collection of  $n$  random variables

$$X_1, X_2, \dots, X_n$$

- Compact notations: **Random Vectors**

- $\mathbf{X}_1^n$

- $\vec{\mathbf{X}} = \begin{pmatrix} X_1 \\ X_2 \\ \vdots \\ X_i \\ \vdots \\ X_n \end{pmatrix}$

- $\underline{\mathbf{X}} = (X_1, X_2, \dots, X_i, \dots, X_n)$

# Vector Notation

$$\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_i \\ \vdots \\ v_n \end{pmatrix}$$

- $\vec{\mathbf{v}}$ : column vector

- $\underline{\mathbf{r}}$ : row vector

$$(r_1, r_2, \dots, r_i, \dots, r_n)$$

$\vec{\mathbf{0}}, \underline{\mathbf{0}}$ : the zero vector  
(the all-zero vector)

$\vec{\mathbf{1}}, \underline{\mathbf{1}}$ : the one vector  
(the all-one vector)

- **Subscripts** represent element indices inside individual vectors.

- $v_i$  and  $r_i$  refer to the  $i^{\text{th}}$  elements inside the vectors  $\vec{\mathbf{v}}$  and  $\underline{\mathbf{r}}$ , respectively.

- When we have a list of vectors, we use **superscripts** in parentheses as indices of vectors.

- $\vec{\mathbf{v}}^{(1)}, \vec{\mathbf{v}}^{(2)}, \dots, \vec{\mathbf{v}}^{(M)}$  is a list of  $M$  column vectors

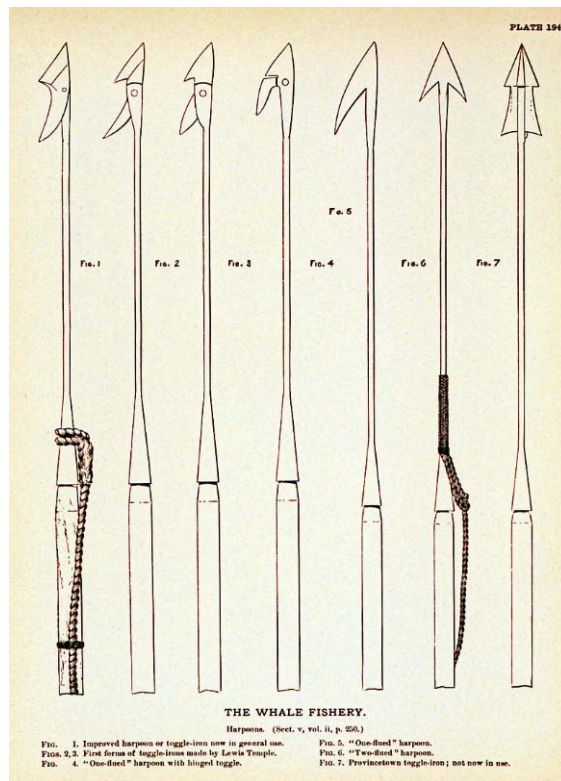
- $\underline{\mathbf{r}}^{(1)}, \underline{\mathbf{r}}^{(2)}, \dots, \underline{\mathbf{r}}^{(M)}$  is a list of  $M$  row vectors

- $\vec{\mathbf{v}}^{(i)}$  and  $\underline{\mathbf{r}}^{(i)}$  refer to the  $i^{\text{th}}$  vectors in the corresponding lists.



# Harpoon

- a long, heavy spear attached to a rope, used for killing large fish or whales



# Harpoon

Word

Prapun Suksompong

What you want to do

$\sqrt{x}$   $\int_{-x}^x$   $\sum_{i=0}^n$   $\{()\}$   $\sin \theta$   $\ddot{a}$   $\lim_{n \rightarrow \infty}$   $\triangle$   $\begin{bmatrix} 10 \\ 01 \end{bmatrix}$

Radical Integral Large Bracket Function Accent Limit and Operator Matrix

Log

Structures

5 6

Accents

$\acute{a}$	$\grave{a}$	$\tilde{a}$	$\hat{a}$
$\grave{a}$	$\acute{a}$	$\grave{a}$	$\grave{a}$
$\grave{a}$	$\grave{a}$	$\tilde{a}$	$\hat{a}$
$\grave{a}$	$\grave{a}$	$\tilde{a}$	$\hat{a}$
$\grave{a}$	$\grave{a}$	$\tilde{a}$	$\hat{a}$

Boxed Formulas

Rightwards Harpoon Above

$P(\mathcal{E})$ ; then so is the error



# Joint pmf

$$\underbrace{p_{X_1, X_2, \dots, X_n}(x_1, x_2, \dots, x_n)}_{p_{\underline{X}}(\underline{x})} = P[\underbrace{X_1 = x_1, X_2 = x_2, \dots, X_n = x_n}_{P[\underline{X} = \underline{x}]}]$$

$$p_{\underline{X}}(\underline{x}) = P[\underline{X} = \underline{x}]$$

# Independence

$n$  random variables  $X_1, X_2, \dots, X_n$

are **independent** if for any  $x_1, x_2, \dots, x_n$ ,

$$\underbrace{p_{X_1, X_2, \dots, X_n}(x_1, x_2, \dots, x_n)}_{p_{\underline{X}}(\underline{x})} = \underbrace{p_{X_1}(x_1)p_{X_2}(x_2) \cdots p_{X_n}(x_n)}_{\prod_i p_{X_i}(x_i)}$$

$$p_{\underline{X}}(\underline{x}) = \prod_i p_{X_i}(x_i)$$

# Independence

$n$  random variables  $X_1, X_2, \dots, X_n$

are **independent** if for any  $x_1, x_2, \dots, x_n$ ,

$$p_{X_1, X_2, \dots, X_n}(x_1, x_2, \dots, x_n) = p_{X_1}(x_1) p_{X_2}(x_2) \cdots p_{X_n}(x_n)$$

$$p_{X_1^n}(\mathbf{x}_1^n)$$

$$\prod_{i=1}^n p_{X_i}(x_i)$$

$$p_{X_1^n}(\mathbf{x}_1^n) = \prod_{i=1}^n p_{X_i}(x_i)$$

# Extension Coding

List Notation	Vector Notation
$p_{X_1, X_2, \dots, X_n}(x_1, x_2, \dots, x_n)$	$p_{\underline{X}}(\underline{x})$ or $p_{X_1^n}(x_1^n)$
$c(x_1, x_2, \dots, x_n)$	$c(\underline{x})$ or $c(x_1^n)$
$\ell(x_1, x_2, \dots, x_n)$	$\ell(\underline{x})$ or $\ell(x_1^n)$
$\mathbb{E}[\ell(X_1, X_2, \dots, X_n)]$	$\mathbb{E}[\ell(\underline{X})]$ or $\mathbb{E}[\ell(\mathbf{X}_1^n)]$
$L_n = \frac{\mathbb{E}[\ell(X_1, X_2, \dots, X_n)]}{n}$	$L_n = \frac{1}{n} \mathbb{E}[\ell(\underline{X})]$ or $\frac{1}{n} \mathbb{E}[\ell(\mathbf{X}_1^n)]$