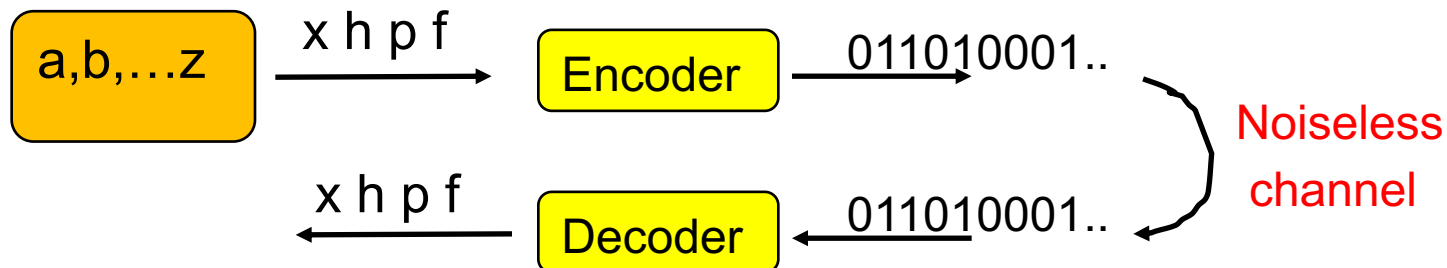


# Measure of efficiency for codes

- Expected length of the code

$$L_{\text{exp}} = \sum_{i=1}^n p_i l_i$$

- Communication efficiency
  - not computational efficiency



# Example

alph	code	prob
a	101	0.125
b	100	0.125
c	11	0.25
d	0	0.5

- Source  $\mathcal{X} = \{a, b, c, d\}$
- $p_a = p_b = 0.125$ ,  $p_c = 0.25$ ,  $p_d = 0.5$
- Code I:  $d \rightarrow 0$ ,  $c \rightarrow 11$   $a \rightarrow 101$   $b \rightarrow 100$
- $L_{\text{exp}} = 1 \times 0.5 + 2 \times 0.25 + 2 \times 3 \times 0.125$
- $L_{\text{exp}}(\text{Code I}) = 1.75$  binary digits
- Code II:  $d \rightarrow 00$   $c \rightarrow 11$   $a \rightarrow 10$   $b \rightarrow 01$
- $L_{\text{exp}}(\text{Code II}) = 2$  binary digits

# Noiseless coding theorem

Fundamental  
Theorem of  
source coding

- **Theorem:**  $L_{\text{exp}} \geq H(X)$

The expected length of any prefix code for a DMC source is lower bounded by the entropy of the source.

*(proof omitted – can be found in CT, or reference below)*

- That is, the expected number of binary digits to encode a source will be at least equal to the average information of the source.

# Entropy of the source

$$H(X) = -\sum_i p(x_i) \log_2 p(x_i)$$

$$- \log_2 p(a) = - \log_2 p(b) = 3 \text{ bits}$$

$$- \log_2 p(c) = 2 \text{ bits}$$

$$- \log_2 p(d) = 1 \text{ bits}$$

$$H(S) = \sum_i p(s_i) [-\log_2 p(s_i)] =$$

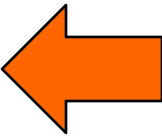
$$0.5 \times 1 + 0.25 \times 2 + 0.125 \times 3 \times 2 = 1.75 \text{ bits (information)}$$

$$H(S) = L_{\text{exp}} (\text{Code I})$$

Code I has **optimal expected length**.

# Summary

- Entropy is a measure of
  - Information
  - uncertainty
- Entropy is used for measuring password strength
- Probabilistic modeling of information source
- Encoding source output using binary digits
  - Efficient codes
- Expected length of the best code
- **A code with the shortest expected length**



# Codes with shortest expected length

## Theorem

- Let  $X$  be a DMS with symbol probabilities  $p_1, \dots, p_m$ .
- Let  $L_{\min}$  be the **smallest expected codeword length over all prefix-free codes for  $X$** .
- Then

$$H(X) \leq L_{\min} < H(X) + 1 \quad \text{bit/symbol}$$

•

# Optimal codes

- A source code with shortest expected length is called **optimal**.
- **Huffman code** is a prefix-free optimal source code  
→ Lossless compression  
**Assumes source distribution is known.**
- Universal source coding algorithms do not assume the source distribution is known.
  - ZIP is based on Ziv-Lempel algorithm
    - Optimal
    - Universal

# Huffman code

*(David Huffman 1952)*

- A source with alphabet  $\{1,2,3,4,5\}$  and probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$
1	0.25
2	0.25
3	0.2
4	0.15
5	0.15

- In 1951, David A. Huffman and his MIT information theory classmates were given the choice of a term paper or a final exam. The professor, Robert M. Fano, assigned a term paper on the problem of finding the most efficient binary code. Huffman, unable to prove any codes were the most efficient, was about to give up and start studying for the final when he hit upon the idea of using a frequency-sorted binary tree and quickly proved this method the most efficient.[3].

wikipedia



(1925-1999)



# Huffman code

*(David Huffman 1952)*

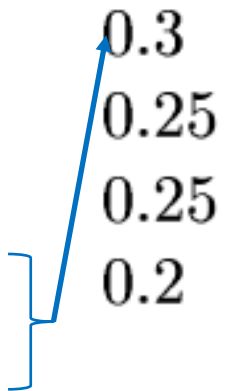
- A source with alphabet  $\{1,2,3,4,5\}$  and probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$
1	0.25
2	0.25
3	0.2
4	0.15
5	0.15

# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  and probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

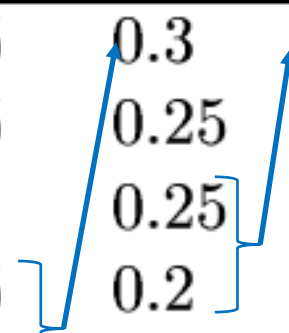
$X$	$P(X)$	
1	0.25	0.3
2	0.25	0.25
3	0.2	0.25
4	0.15	0.2
5	0.15	



# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  and probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$		
1	0.25	0.3	0.45
2	0.25	0.25	0.3
3	0.2	0.25	0.25
4	0.15	0.2	
5	0.15		



# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  and probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$				
1	0.25	0.3	0.45	0.55	
2	0.25	0.25	0.3	0.45	
3	0.2	0.25	0.25		
4	0.15	0.2			
5	0.15				

# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  with probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$					
1	0.25	0.3	0.45	0.55	0	
2	0.25	0.25	0.3	0.45	1	
3	0.2	0.25	0.25			
4	0.15	0.2				
5	0.15					

The diagram illustrates the Huffman coding process. Blue arrows and brackets show the merging of probabilities from the bottom row upwards. A black arrow points to the final node for symbol 1.

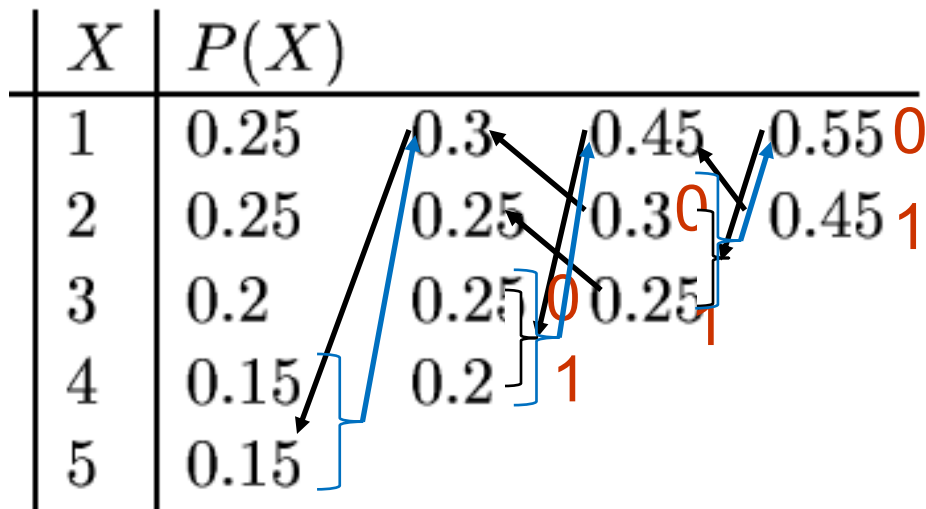
# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  with probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$					
1	0.25	0.3	0.45	0.55	0	
2	0.25	0.25	0.3	0.45	1	
3	0.2	0.25	0.25	1		
4	0.15	0.2				
5	0.15					

# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  with probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$



# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  with probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$
1	0.25
2	0.25
3	0.2
4	0.15
5	0.15

The diagram illustrates the construction of a Huffman tree for the given probabilities. The process starts with five leaf nodes representing symbols 1 through 5. Symbols 4 and 5 (probabilities 0.15 each) are merged into an internal node with probability 0.2. This node is then merged with symbol 3 (probability 0.2) to form a node with probability 0.4. Next, symbols 1 and 2 (probabilities 0.25 each) are merged into a node with probability 0.5. Finally, the 0.4 node and the 0.5 node are merged into the root node with probability 1.0. The tree structure is shown with black arrows indicating the merge process. The resulting binary codes are shown in red next to each symbol: 1 is 0, 2 is 1, 3 is 00, 4 is 01, and 5 is 11. Blue lines and arrows highlight the specific merge steps and the final code assignment.



# Huffman code

- A source with alphabet  $\{1,2,3,4,5\}$  with probabilities  $\{0.25, 0.25, 0.2, 0.15, 0.15\}$

$X$	$P(X)$		
1	0.25	0.3	0.45
2	0.25	0.25	0.3
3	0.2	0.25	0.25
4	0.15	0.2	0.15
5	0.15		

01
10
11
000
001

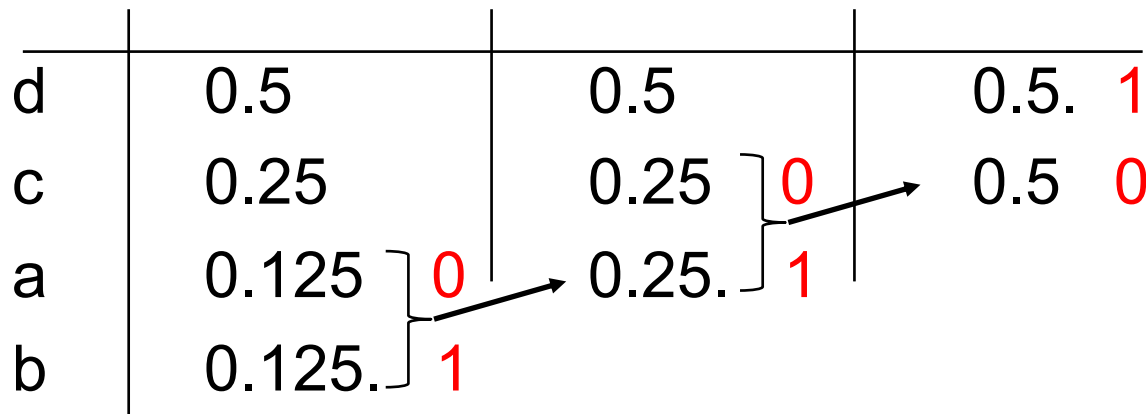
- Theorem
- Huffman code is an optimal code.

$$H(X) \leq L_{\text{exp}} \leq H(X)+1$$

# Example

alph	code	prob
a	101	0.125
b	100	0.125
c	11	0.25
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- $p_a = p_b = 0.125$ ,  $p_c = 0.25$ ,  $p_d = 0.5$

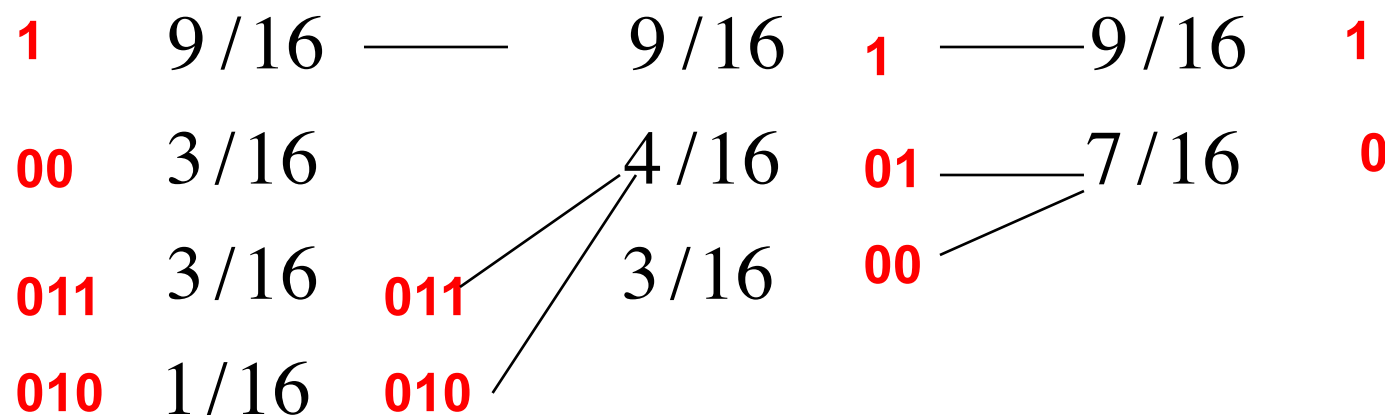


# Reducing the length

- $X=\{0,1\}$  - DMS
- $p_0=1/4,$        $p_1=3/4$
- For the optimal code,       $L_{\text{exp}} = 1$
- $H(X) = 0.19$  bit  
 $0.19 < L_{\text{exp}} < 1.19$

# Encoding blocks of symbols

- $X'=\{00,01,10,11\}$
- $P00=1/16$ ,  $p10=p11=3/16$ ,  $p11=9/16$



$$L_{\text{exp}} = (9/16) + (3/16) \times 2 + (3/16) \times 3 + (1/16) \times 3 = 1.69 \text{ bits/2 sym}$$

$$\rightarrow L_{\text{exp}} = 0.845 \text{ bit/symbol}$$

# DMC with block coding

- Huffman code when applied to blocks of length  $n$ :

$$H(X) \leq L_{\text{exp}} \leq H(X) + 1/n$$

# Lossy and Lossless

- ZIP is an archive file format with **lossless data compression**.
- JPEG, MPEG, MP3 are lossy compression
  - Trade-off between quality and size of the encoder output

100% fidelity  
Image is 725kB



90%  
250kB



10%  
37kB



1%  
20kB



# Summary

- Source coding
- Efficiency of source codes
- Optimal codes
- Huffman code

