

- Huffman encoding
 - 1. Order the symbols according to their probabilities

alphabet set : $S_1, S_2, ..., S_N$

prob. of occurrence : P_1 , P_2 , ..., P_N

 \Rightarrow the symbols are rearranged so that

$$P_1 \ge P_2 \ge \dots \ge P_N$$

2. Apply a contraction process to the two symbols with the smallest probabilities

replace symbols $S_{N\text{-}1}$ and S_N by a "hypothetical" symbol, say $H_{N\text{-}1},$ that has a prob. of occurrence $P_{N\text{-}1}\text{+}P_N$

the new set of symbols has N-1 members:

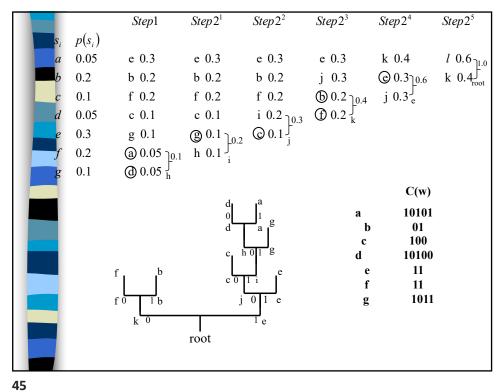
3. Repeat the step 2 until the final set has only one member.

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The recursive procedure in step 2 can be viewed as the construction of a binary tree, since at each step we are merging two symbols.

At the end of the recursion, all the symbols $\ S_1,\,S_2,\,...,\,S_N$ will be leaf nodes of the tree.

The codeword for each symbol S_i is obtained by traversing the binary tree from its root to the leaf node corresponding to S_i



Average codeword length $lave = \sum_{i}^{\Delta} l_{i} p_{i}$

$$lave = \sum_{i}^{\Delta} l_{i} p$$

lave is a measure of the compression ratio.

In the above example,

7 symbols \Rightarrow 3 bits fixed length code representation

lave(Huffman) = 2.6 bits

Compression ration = 3/2.6 = 1.15

Properties of Huffman codes

- Fixed-length symbols → variable-length codewords : error propagation
- H(s) ≤ lave < H(s)+1 H(s) ≤ lave < P+0.086 where P is the prob. Of the most frequently accurring symbol. The equality is achieved when all symbol probs. Are inverse powers of two.
- The Huffman code-tree can be constructed both by bottom-up method — in the above example top-down method

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- The code construction process has a complexity of O(Nlog₂N). With presorting of the input symbol probs, code construction method with complexity O(N) has been presented in IEEE trans. ASSP-38, pp. 1619-1626, Sept. 1990.
- Huffman codes satisfy the prefix-condition: uniquely decodable no codeword is a prefix of another codeword.
- If I_i satisfy the Kraft constraint $\sum_{i=1}^{2^{-l_i}} \le 1$ then the corresponding codewords can be constructed as the first I_i bits in the fractional representation of a_i

$$a_i = \sum_{j=1}^{i-1} 2^{-l_j}, \quad i = 1, 2, \dots, N$$

The complement of a binary Huffman code is also a valid Huffman code.

Huffman Decoding

Bit-Serial Decoding : fixed input bit rate — variable output symbol rate

(Assume the binary coding tree is available to the decoder) In practice, Huffman coding tree can be reconstructed from the symbol-to-codeword mapping table that is known to both the encoder and the decoder

Step 1

Read the input compressed stream bit-by-bit and traverse the tree until a leaf node is reached.

Step 2:

As each bit in the input stream is used, it is discarded. When the leaf node is reached, the Huffman decoder outputs the symbol at the leaf node. This completes the decoding for this symbol.

Step 3

Repeat step1 and 2 until all the input is consumed.

Since the codeword is variable in length, the decoding bit rate is not same for all symbols.

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 Lookup-table-Based Decoding : constant decoding rate for all symbols — variable input rate

The look-up table is constructed at the decoder from the symbol-to-codeword mapping table. If the longest codeword in this table is L bits, then a 2^L entry lookup table is needed. : space constraints image/video longest L = 16 to 20.

Look up table construction:

- Let C_i be the codeword that corresponds to symbol S_i. Assume C_i has I_i bits. We form an L-bit address in which the first I_i bits are C_i and the remaining L- I_i bits take on all possible combinations of "0" and "1". Thus, for the symbol s_i, there will be 2^{L-II} addresses.
- At each entry we form the two-tuple (s_i, l_i).



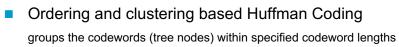
- 1. From the compressed input bit stream, we read in L bits into a buffer.
- 2. We use the L-bit word in the buffer as an address into the lookup table and obtain the corresponding symbol, say s_k . Let the codeword length be l_k . We have now decode one symbol.
- 3. We discard the first I_k bits from the buffer and we append to the buffer, the next I_k bits from the input, so that the buffer has again L bits.
- 4. Repeat step 2 and 3 until all of the symbols have been decoded.

Memory Efficient and High-Speed Search Huffman Coding, by R. Hashemian IEEE trans. Communications, Oct. 1995, pp. 2576-2581

- Due to variable-length coding, the Huffman tree gets progressively sparse as it grow from the root
 - Waste of memory space
 - A lengthy search procedure for locating a symbol

Ex: if K-bit is the longest Huffman code assigned to a set of symbols, the memory size for the symbols may easily reach 2^K words in size.

- → It is desirable to reduce the memory size from typical value of 2^K, to a size proportional to the number of the actual symbols.
 - Reduce memory size
 - Quicker access



- Characteristics of the proposed coding cheme:
 - 1. The search time for more frequent symbols (shorter codes) is substantially reduced compare to less frequent symbols, resulting in an overall faster response.
 - 2. For long codewords the search for the symbol is also speed up. This is achieved through a specific partitioning technique that groups the code bits in a codeword, and the search for a symbol is conducted by jumping over the groups of bits rather than going through the bit individually.
 - 3. The growth of the Huffman tree is directed toward on side of the tree.
 - Single side growing Huffman tree (SGH-tree)

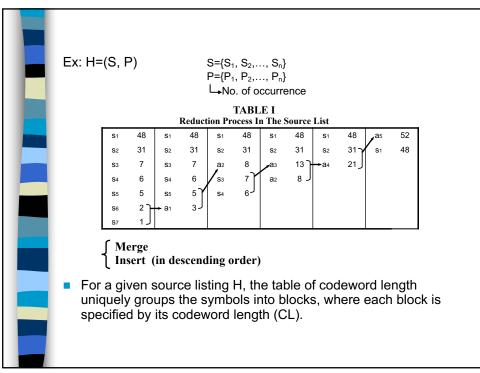


TABLE II
TABLE OF CL-RECORDING

s_i	<i>s</i> _{i-1}	CL
S 7	56	5
a_1	85	4
54	s ₃	4
a_2	a_3	3
a_4	S 2	2
s ₁	a_5	1 1

TABLE III
TABLE OF CODEWORD LENGTHS (TOCL)

CL	symbols	
1	S_1	
2	\boldsymbol{S}_2	
3		
4	S_3 , S_4 , S_5	
5	S_6, S_7	

CL: codeword length

 Each block of symbols, so defined, occupies one level in the associated Huffman tree.

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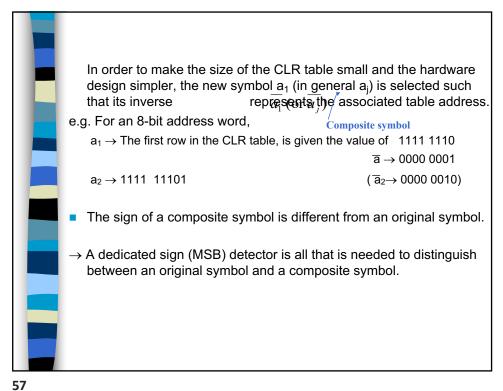


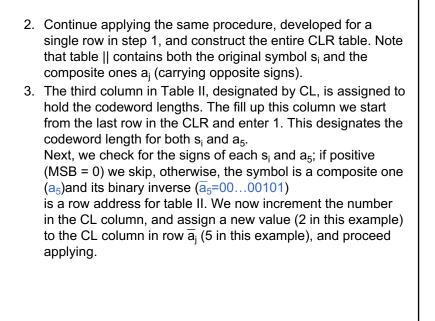
1. The symbols are listed with the probabilities in decending order (the ordering of the symbols with equal probabilities is assumed indifferent).

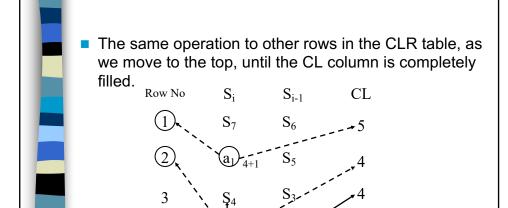
Next, the pair of symbols at the bottom of the ordered list are merged and as a result a new symbol a_1 is created. The symbol a_1 , with probability equal to the sum of the probabilities of the pair, is then inserted at the proper location in the **ordered list**.

To record this operation a codeword length recording (CLR) table is created which consists of three columns:

: Columns 1 and 2 hold the last pair of symbols before being merged, and column 3, initially empty, is identified as the codeword length (CL) column (Table II).







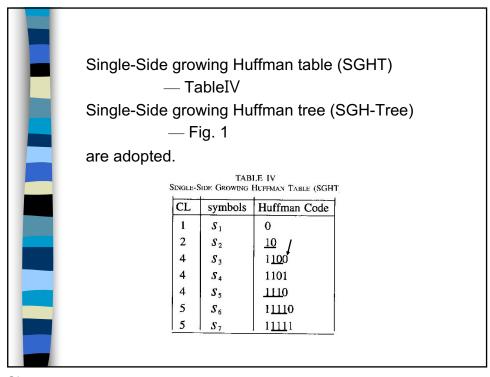
 $6 S_1 (a_5)_{1+1}$

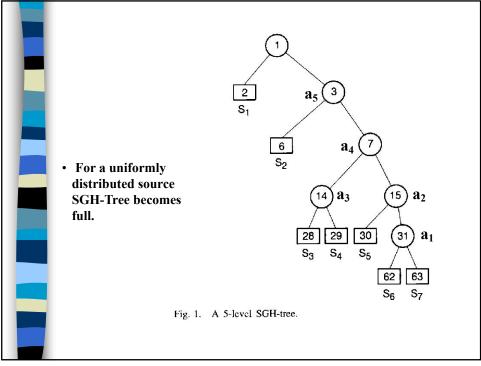
role:
(i) S_i, S_{i-1} 中有一為 composite, 則以 composite 為 CL 增加及 new address 之基礎
(ii) 若S_i, S_{i-1}皆為 original, skip this row and CL 不變

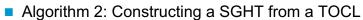
4. Table II indicates that each original symbol in the table has its codeword length (CL) specified.
Ordering the symbols according the their CL values gives Table III.

Associated with the so-obtained TOCL one can actually generate a number of Huffman tables (or Huffman trees), each of which being different in codewords but identical in the codeword lengths.

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- Start from the first row of the table and assign an "all zero" codeword C₁ = 00...0 to the symbol S₁.
- Next we increment this codeword and assign the new value to the next symbol in the table. Similarly, we proceed coreating codewords for the rest of the symbols in the same row of the TOCL.
- When we change rows, we have to expand the last codeword, after being incremented, by placing extra zeros to the right, until the codeword length matches the level (CL).

In general we can write:

$$C_1 = 00 \cdots 0$$

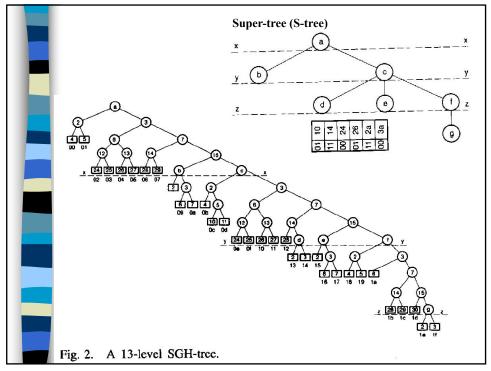
$$C_{i+1} = (C_i + 1) * 2^{p-q}$$

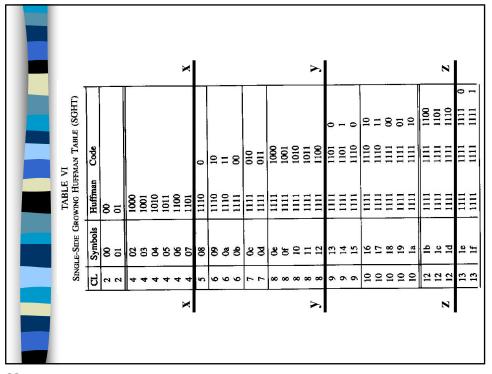
and

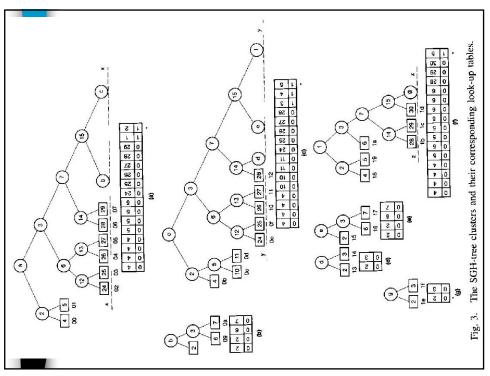
$$C_n = 11 \cdots 1$$

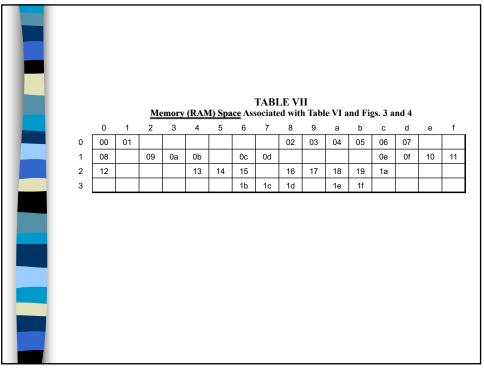
where p and q are the codeword lengths for s_i and s_{i+1} , respectively, and S_n (associated with C_n) denotes the terminal symbol.

 C_1 and C_n have unique forms \rightarrow easy to verify

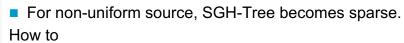












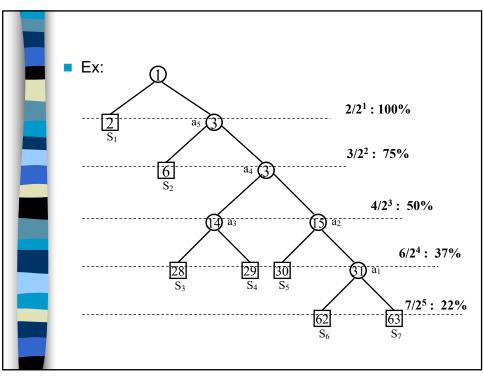
- i) optimize the storage space
- ii) provide quick access to the symbol (data) key Idea:

Break down the SGH-Tree into smaller clusters (subtrees) such that the memory efficiency increases.

The memory efficiency B for a K-level binary Huffman tree

$$B_K = \frac{\text{No. of effective leaf nodes}}{2^K} \times 100\%$$

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Remarks:

- 1. The efficiency changes only when we switch to a new level (or equivalently to a new CL), and it decreases as we proceed to the higher levels.
- 2. Memory efficiency can be interpreted as a measure of the performance of the system in terms of memory space requirement; and it is directly related to the sparsity of the Humman tree.
- Higher memory efficiency for the top levels (with smaller CL) is a clear indication that partitioning the tree into smaller and less sparse clusters will reduce the memory size. In addition, clustering also helps to reduce the search time for a symbol.

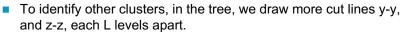
Definition:

A cluster (subtree) T_i with minimum memory efficiency (MME) B_i , if there is no level in T_i memory efficiency less than B_i .

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SGH-Tree Clustering

- Given a SGH-tree, as shown in Fig. 2, depending on the MME (or CL) assigned, the tree is partitioned by a cut line, x-x, at the Lth level (L=4 for the choice of MME=50%, in this example).
- The first cluster (subtree), as shown in Fig.3(a), is formed by removing the remainder of the tree beyond the cut-line x-x.
- The <u>cluster length</u> is defined to be the maximum path length from the root to a node within the cluster → the cluster length for the first cluster is 4.
- Associated with each cluster a <u>look up table</u> (LUT) is assigned, as shown at the bottom of Fig.3(a), to provide the <u>addressing information</u> for the corresponding <u>terminal node</u> (symbol) within the cluster, or beyond.



- More clusters are generated, each of which starting from a single node (<u>root of the cluster</u>) and expanded until it is terminated either by terminal nodes, or nodes being intercepted by the next cute line.
- Next, we construct a <u>super-tree (s-tree)</u> corresponding to a SGH-Tree. In a s-tree <u>each cluster</u> is represented by a <u>node</u>, and the links connecting these nodes, representing the branching nodes in the SGH-tree, shared between two clusters.
- The super-table (ST) associated with the s-tree is shown at the bottom of the tree.
 - Note that the s-tree has 7 nodes, one for each cluster, while its ST has 6 entries. This is because the root cluster a is left out and the table starts from cluster b.

Entries in the ST and the LUT's

- There are two numbers in each location in the ST the first number identifies the cluster length
- the 2nd number is the offset memory address for that cluster Ex:

binary Hexa

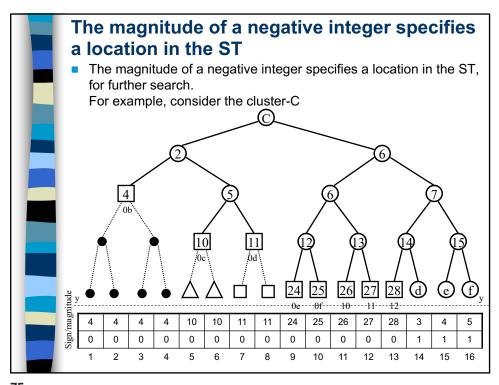
f 11 2a_H

cluster length: 11+1 = 100, or 4

 $2a_{H}$: the starting address of the corresponding LUT, in the memory (see table $\mbox{\rm VII})$

- \Rightarrow the cluster f start at address $2a_H$ in the memory table VII. (i.e. symbol 18)
- Each entry in a LUT is an integer in sign/magnitude format.

Positive integer, with 0 sign, correspond to the nodes existed in the cluster, while negative numbers, with 1 sign, represent the nodes being cut by the next cut line.



In the 15th entry of the above LUT we find ½ as sign/magnitude at that location.

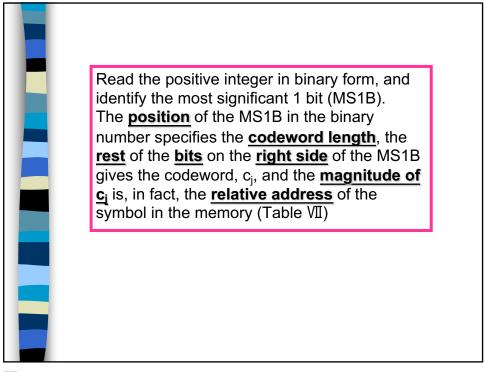
the negative sign (1) indicates that we have to move to other cluster, and 4 refers to the 4th entry in the ST, which corresponds to cluster e, and contains 01 and 26H numbers.

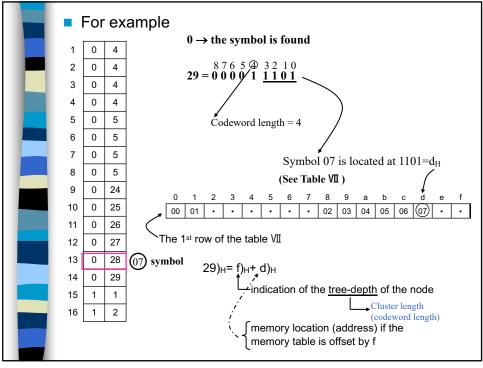
The first number, 01, indicates the cluster length is 01+1=10(or

2), and the 2nd number, shows the starting address of cluster e in

the memory.
 A positive entry in a LUC, indicates that the symbol is already found, and the magnitude comprises three pieces of information

- i) location of the symbol in the memory
- ii) the pertinent codeword
- iii) the codeword length





Huffman Decoding

The decoding procedure basically starts by receiving an L-bit code c_j, where L is the length of the top cluster in the associated SGH-Tree (or the SGHT). This L-bit code c_j is then used as the address to the associated look-up table [fig.3(a)]

Example:

1. Received codeword: 01100 1011... first L=4 bit 0110=6 as an address to the LUT given in Fig.3(a).

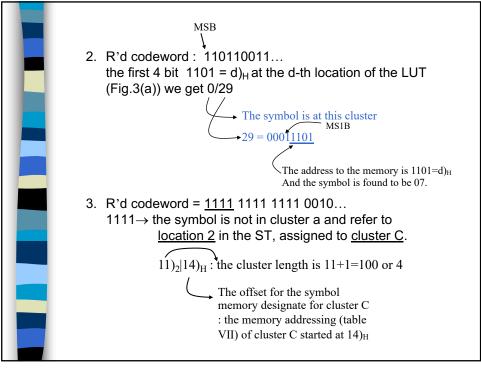
The content of the table at this location is 0/5, as the sign/magnitude.

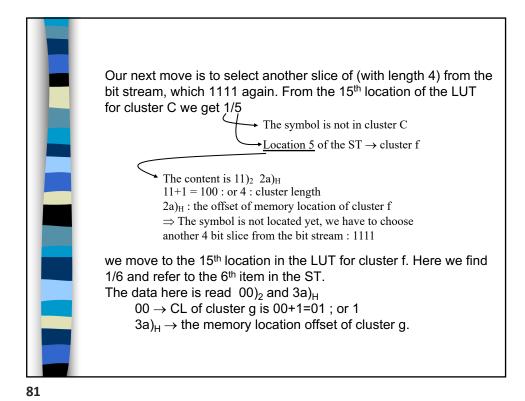
MS1B

0→the symbol is located in this cluster 5=00000101, the MS1B is at location 2

- ⇒ CL=2. Next to the MS1B we have 01, which represents the codeword
- \Rightarrow the symbol (01H) is found at the address 01 in the memory (Table VII)

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Take 1 bit from the bit stream which is "0" referring the LUT of cluster g, location "0" gives 0/2
So the symbol is in cluster g, and 2 = 00000010 ⇒
(i) MS1B is located at location 1 ⇒ CL = 1
(ii) to the right of MS1B is a single 0 identifying the codeword, and (iii) the symbol (1e) is at location 3a+0 = 3a in the memory (tableVII)

⇒ The overall codeword is given as 1111, 1111, 1111, 0, with CL = 4+4+4+1 = 13

Remarks:

1. For high probable symbols with short codewords (4 bits or less) the search for the symbol is very fast, and is completed in the first try. For longer codewords, however, the search time grows almost proportional to the codeword length.

If CL is the codeword length
L is the maximum level selected for each cluster
(L = 4 in the above example) then the search time is
closely proportional to 1+CL/L

2. Increasing L:

- i. Decreasing search time \rightarrow speed up decoding
- ii. Growing
- \Rightarrow Trade-off