

4.4. Arithmetic coding

Advantages:

- Reaches the entropy (within computing precision)
- Superior to Huffman coding for small alphabets and skewed distributions
- Clean separation of modelling and coding
- Suits well for adaptive one-pass compression
- Computationally efficient

History:

- Original ideas by Shannon and Elias
- Actually discovered in 1976 (Pasco; Rissanen)

Arithmetic coding (cont.)

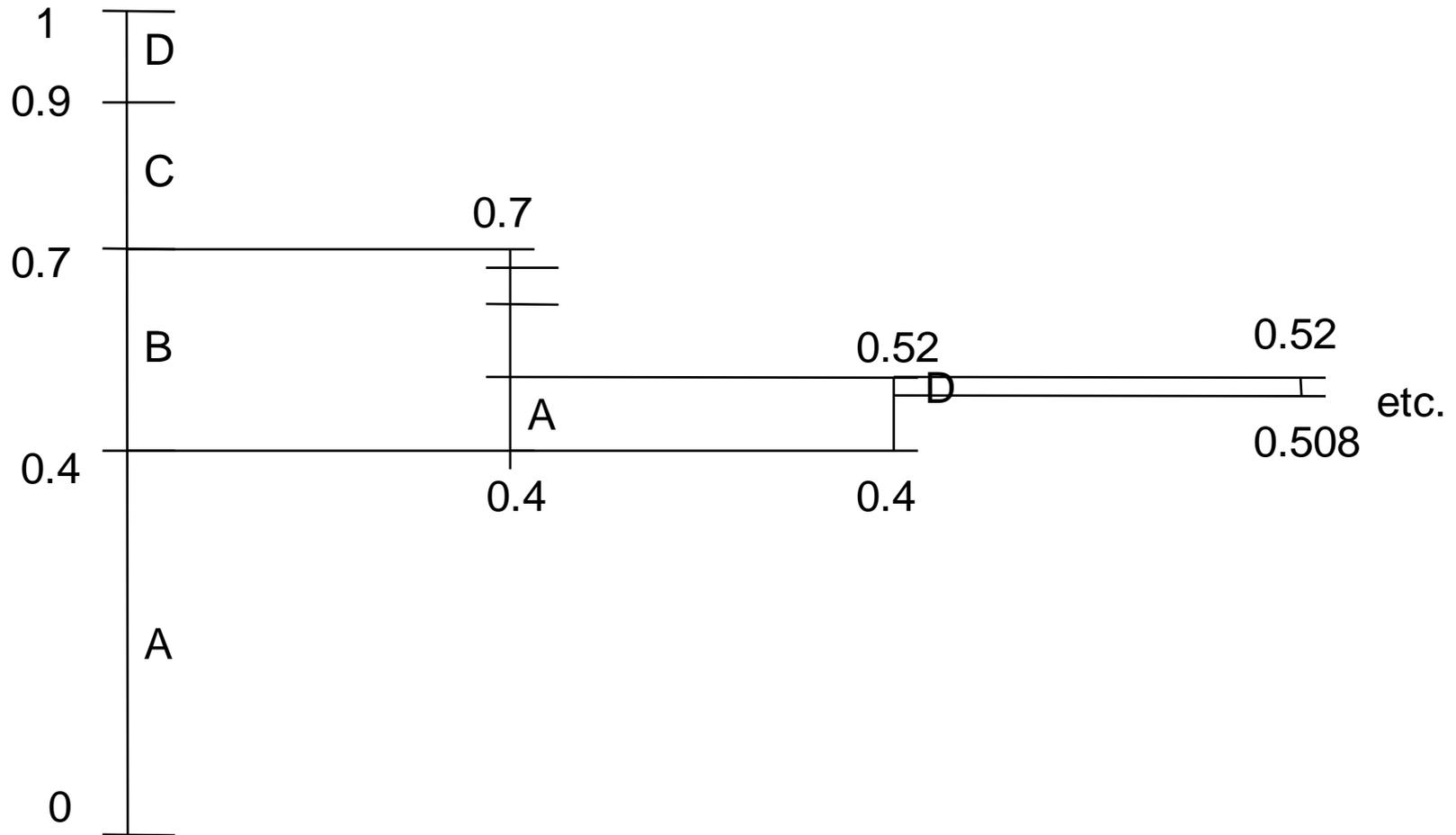
Characterization:

- One codeword for the whole message
- A kind of extreme case of extended Huffman (or Tunstall) coding
- No codebook required
- No clear correspondence between source symbols and code bits

Basic ideas:

- Message is represented by a (small) interval in $[0, 1)$
- Each successive symbol reduces the interval size
- Interval size = product of symbol probabilities
- Prefix-free messages result in disjoint intervals
- Final code = any value from the interval
- Decoder computes the same sequence of intervals

Arithmetic coding: Encoding of "BADCBAB"



Algorithm: Arithmetic encoding

Input: Sequence $x = x_i, i=1, \dots, n$; probabilities p_1, \dots, p_q of symbols $1, \dots, q$.

Output: Real value between $[0, 1)$ that represents X .

begin

$cum[0] := 0$

for $i := 1$ **to** q **do** $cum[i] := cum[i-1] + p_i$

$lower := 0.0$

$upper := 1.0$

for $i := 1$ **to** n **do**

begin $range := upper - lower$

$upper := lower + range * cum[x_i]$

$lower := lower + range * cum[x_i - 1]$

end

return $(lower + upper) / 2$

end

Algorithm: Arithmetic decoding

Input: v : Encoded real value; n : number of symbols to be decoded; probabilities p_1, \dots, p_q of symbols $1, \dots, q$.

Output: Decoded sequence x .

begin

$cum[1] := p_1$

for $i := 2$ **to** q **do** $cum[i] := cum[i-1] + p_i$

$lower := 0.0$

$upper := 1.0$

for $i := 1$ **to** n **do**

begin $range := upper - lower$

$z := (v - lower) / range$

Find j such that $cum[j-1] \leq z < cum[j]$

$x_i := j$

$upper := lower + range * cum[j]$

$lower := lower + range * cum[j-1]$

end

return $x = x_1, \dots, x_n$

end

Arithmetic coding (cont.)

Practical problems to be solved:

- Arbitrary-precision real arithmetic
- The whole message must be processed before the first bit is transferred and decoded.
- The decoder needs the length of the message

Representation of the final binary code:

- Midpoint between *lower* and *upper* ends of the final interval.
- Sufficient number of significant bits, to make a distinction from both *lower* and *upper*.
- The code is prefix-free among prefix-free messages.

Example of code length selection

midpoint \neq lower and upper

- *upper*: 0.517072 = .100001000101 **1**1101...
 - *midpoint*: 0.516928 = .10000100010 **10**1010...
 - *lower*: 0.516784 = .10000100010 **01**0111...
- 13 bits

range = 0.00028

$\log_2(1/\text{range}) \approx 11.76$ bits

Another source message

“ABCDABCABA”

- *Precise probabilities:*

$$P(A) = 0.4, \quad P(B) = 0.3, \quad P(C) = 0.2, \quad P(D) = 0.1$$

- *Final range length:*

$$0.4 \cdot 0.3 \cdot 0.2 \cdot 0.1 \cdot 0.4 \cdot 0.3 \cdot 0.2 \cdot 0.4 \cdot 0.3 \cdot 0.4 = \\ 0.4^4 \cdot 0.3^3 \cdot 0.2^2 \cdot 0.1 = 0.000002764$$

$$-\log_2 0.000002764 \approx 18.46 = \textit{entropy}$$

Arithmetic coding: Basic theorem

Theorem 4.2.

Let $range = upper - lower$ be the final probability interval in Algorithm 4.8. The binary representation of $mid = (upper + lower) / 2$ truncated to $l(x) = \lceil \log_2(1/range) \rceil + 1$ bits is a uniquely decodable code for message x among prefix-free messages.

Proof: Skipped.

Optimality

Expected length of an n -symbol message x :

$$\begin{aligned}L^{(n)} &= \sum P(x)l(x) \\&= \sum P(x) \left[\left\lceil \log_2 \frac{1}{P(x)} \right\rceil + 1 \right] \\&\leq \sum P(x) \left[\log_2 \frac{1}{P(x)} + 2 \right] \\&= \sum P(x) \log_2 \frac{1}{P(x)} + 2 \sum P(x) \\&= H(S^{(n)}) + 2\end{aligned}$$

Bits per symbol:

$$\begin{aligned}\frac{H(x^{(n)})}{n} \leq L \leq \frac{H(x^{(n)})}{n} + \frac{2}{n} \\H(S) \leq L \leq H(S) + \frac{2}{n}\end{aligned}$$

Ending problem

- The above theorem holds only for prefix-free messages.
- The ranges of a message and its prefix overlap, and may result in the same code value.
- How to distinguish between “VIRTA” and “VIRTANEN”?
- Solutions:
 - Transmit the *length* of the message before the message itself: “5VIRTA” and “8VIRTANEN”.
This is not good for online applications.
 - Use a special end-of-message symbol, with prob = $1/n$ where n is an *estimated* length of the message.
Good solution unless n is totally wrong.

Arithmetic coding: Incremental transmission

- Bits are sent as soon as they are known.
- Decoder can start well before the encoder has finished.
- The interval is *scaled (zoomed)* for each output bit: Multiplication by 2 means shifting the binary point one position to the right:

$$\begin{array}{l} \text{upper: } 0.011010\dots \\ \text{lower: } 0.001101\dots \end{array} \quad \longrightarrow \quad \begin{array}{l} 0.11010\dots \\ 0.01101\dots \end{array} \quad \text{and transmit 0}$$

$$\begin{array}{l} \text{upper: } 0.110100\dots \\ \text{lower: } 0.100011\dots \end{array} \quad \longrightarrow \quad \begin{array}{l} 0.10100\dots \\ 0.00011\dots \end{array} \quad \text{and transmit 1}$$

- **Note:** The common bit also in midpoint value.

Arithmetic coding: Scaling situations

// Number p of pending bits initialized to 0

upper < 0.5:

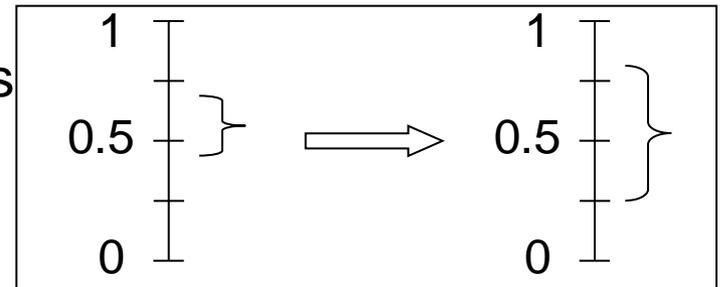
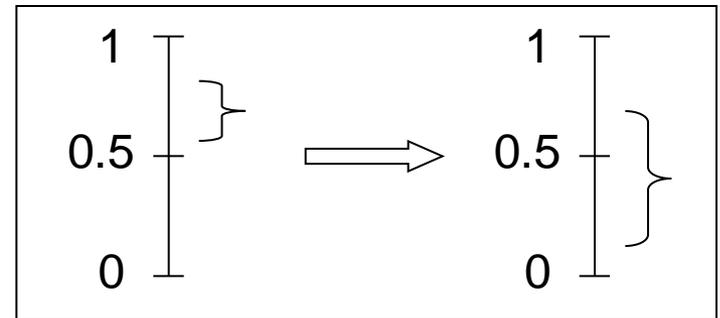
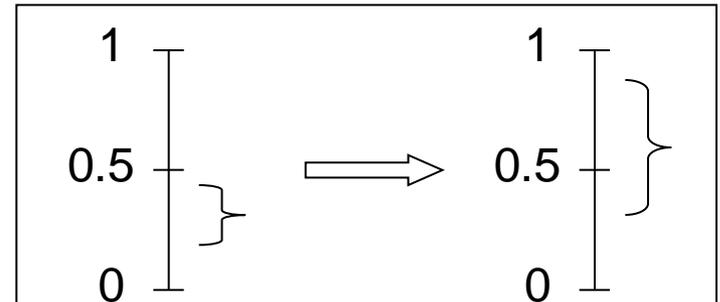
- transmit bit 0 (plus p pending 1's)
- $lower := 2 \cdot lower$
- $upper := 2 \cdot upper$

lower > 0.5

- transmit bit 1 (plus p pending 0's)
- $lower := 2 \cdot (lower - 0.5)$
- $upper := 2 \cdot (upper - 0.5)$

lower > 0.25 and upper < 0.75:

- Add one to the number p of pending bits
- $lower = 2 \cdot (lower - 0.25)$
- $upper = 2 \cdot (upper - 0.25)$



Decoder operation

- Reads a sufficient number of bits to determine the first symbol (unique interval of cumulative probabilities).
- Imitates the encoder: performs the same scalings, after the symbol is determined
- Scalings drop the 'used' bits, and new ones are read in.
- No pending bits.

Implementation with integer arithmetic

- Use symbol frequencies instead of probabilities
- Replace $[0, 1)$ by $[0, 2^k - 1)$
- Replace 0.5 by $2^{k-1} - 1$
- Replace 0.25 by $2^{k-2} - 1$
- Replace 0.75 by $3 \cdot 2^{k-2} - 1$

Formulas for computing the next interval:

- $upper := lower + (range \cdot cum[symbol] / total_freq) - 1$
- $lower := lower + (range \cdot cum[symbol-1] / total_freq)$

Avoidance of overflow: $range \cdot cum() < 2^{wordsize}$

Avoidance of underflow: $range > total_frequency$

Solution to avoiding over-/underflow

- Due to scaling, *range* is always $> 2^{k-2}$
- Both overflow and underflow are avoided, if $total_freq < 2^{k-2}$, and $2k-2 \leq w = \text{machine word}$

Suggestion:

- Present *total_freq* with max 14 bits, *range* with 16 bits

Formula for decoding a symbol x from a k -bit value:

$$cum(x - 1) \leq \left\lfloor \frac{(value - lower + 1) \cdot total_freq - 1}{upper - lower + 1} \right\rfloor < cum(x)$$

4.4.1. Adaptive arithmetic coding

Advantage of arithmetic coding:

- Used probability distribution can be changed at any time, but synchronously in the encoder and decoder.

Adaptation:

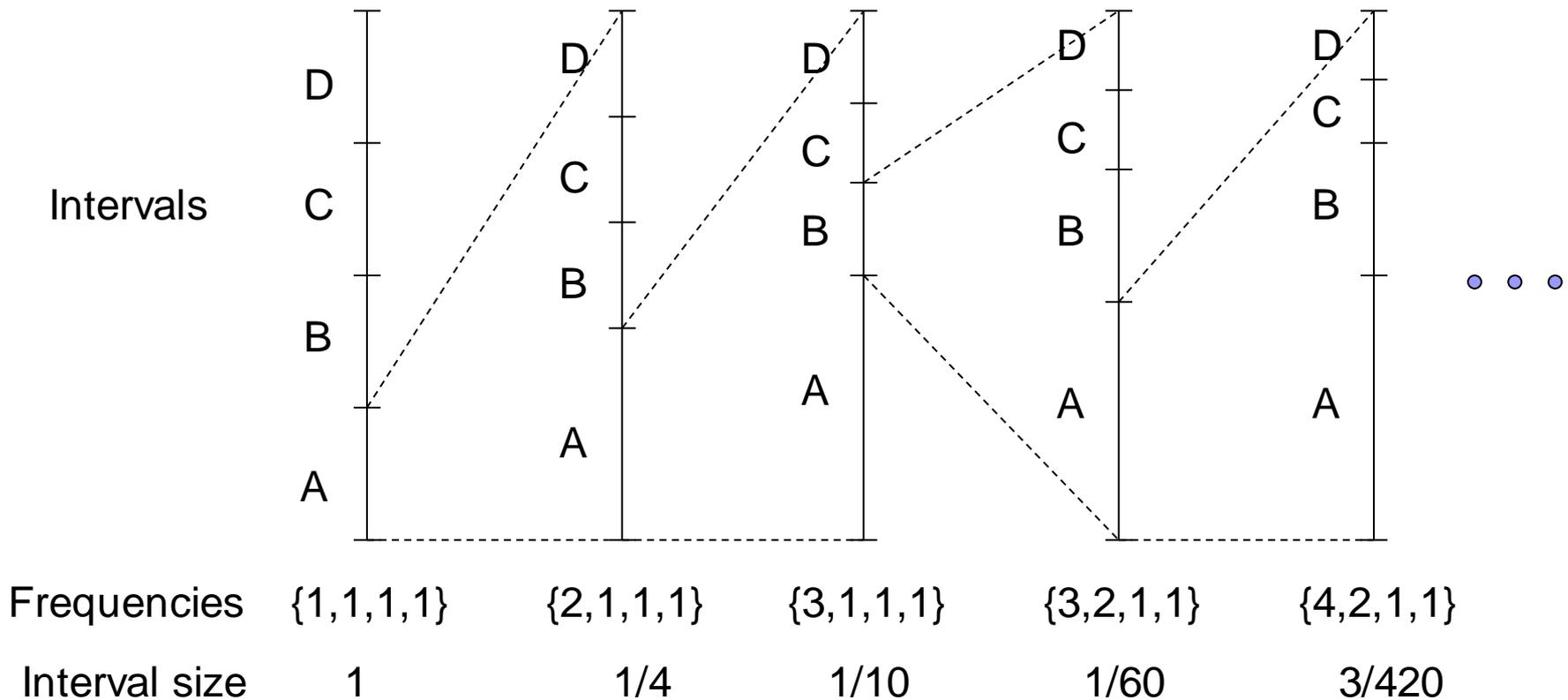
- Maintain frequencies of symbols during the coding
- Use the current frequencies in reducing the interval

Initial model; alternative choices:

- All symbols have an initial frequency = 1.
- Use a placeholder (NYT = Not Yet Transmitted) for the unseen symbols, move symbols to active alphabet at the first occurrence.

Basic idea of adaptive arithmetic coding

- Alphabet: {A, B, C, D}
- Message to be coded: "AABAAB ..."



Adaptive arithmetic coding (cont.)

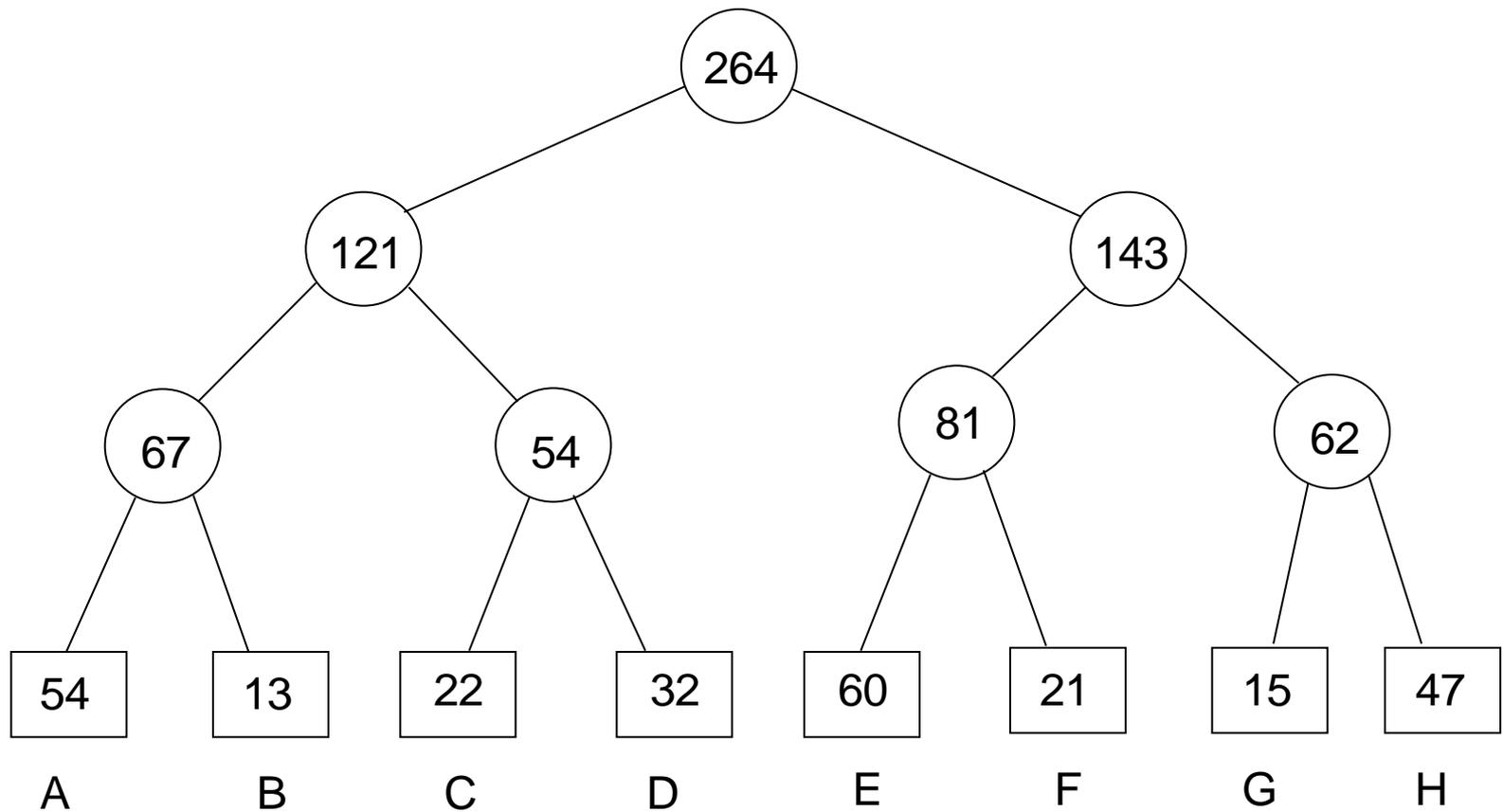
Biggest problem:

- Maintenance of cumulative frequencies; simple vector implementation has complexity $O(q)$ for q symbols.

General solution:

- Maintain partial sums in an explicit or implicit binary tree structure.
- Complexity is $O(\log_2 q)$ for both search and update

Tree of partial sums



Implicit tree of partial sums

1	2	3	4	5	6	7	8
f	f_1+f_2	f_3	$f_1+\dots+f_4$	f_5	f_5+f_6	f_7	$f_1+\dots+f_8$
9	10	11	12	13	14	15	16
f_9	f_9+f_{10}	f_{11}	$f_9+\dots+f_{12}$	f_{13}	$f_{13}+f_{14}$	f_{15}	$f_1+\dots+f_{16}$

Correct indices are obtained by bit-level operations.

4.4.2. Arithmetic coding for a binary alphabet

Observations:

- Arithmetic coding works as well for any size of alphabet, contrary to Huffman coding.
- Binary alphabet is especially easy: *No cumulative probability table.*

Applications:

- Compression of black-and-white images
- Any source, interpreted bitwise

Speed enhancement:

- Avoid multiplications
- Approximations cause additional redundancy

Arithmetic coding for binary alphabet (cont.)

Note:

- Scaling operations need only multiplication by two, implemented as shift-left.
- Multiplications appearing in reducing the intervals are the problem.

Convention:

- **MPS** = More Probable Symbol
- **LPS** = Less Probable Symbol
- The correspondence to actual symbols may change locally during the coding.

Skew coder (Langdon & Rissanen)

- **Idea:** approximate the probability p of LPS by $1/2^Q$ for some integer $Q > 0$.
- Choose LPS to be the first symbol of the alphabet (can be done without restriction)
- Calculating the new *range*:
 - For LPS: $range \leftarrow range \gg Q$;
 - For MPS: $range \leftarrow range - (range \gg Q)$;
- Approximation causes some redundancy
- Average number of bits per symbol ($p =$ exact prob):

$$pQ - (1 - p) \log_2 \left(1 - \frac{1}{2^Q}\right)$$

Solving the ‘breakpoint’ probability \hat{p}

- Choose Q to be either r or $r+1$, where $r = \lfloor -\log_2 p \rfloor$
- Equate the bit counts for rounding down and up:

$$\hat{p}r - (1 - \hat{p}) \log_2 \left(1 - \frac{1}{2^r}\right) = \hat{p}(r+1) - (1 - \hat{p}) \log_2 \left(1 - \frac{1}{2^{r+1}}\right)$$

which gives

$$\hat{p} = \frac{z}{1+z} \quad \text{where} \quad z = \log_2 \frac{1 - 1/2^{r+1}}{1 - 1/2^r}$$

Skew coder (cont.)

Probability approximation table:

Probability range	Q	Effective probability
0.3690 – 0.5000	1	0.5
0.1820 – 0.3690	2	0.25
0.0905 – 0.1820	3	0.125
0.0452 – 0.0905	4	0.0625
0.0226 – 0.0452	5	0.03125
0.0113 – 0.0226	6	0.015625

Proportional compression efficiency:

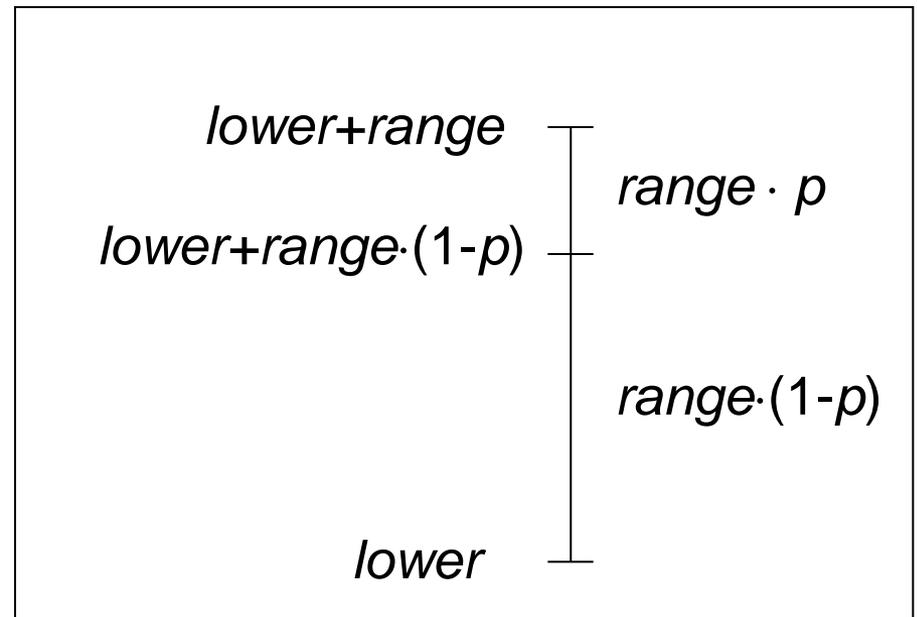
$$\frac{\textit{entropy}}{\textit{averageLength}} = \frac{-p \log p - (1-p) \log(1-p)}{-pQ - (1-p) \log(1-1/2^Q)}$$

QM-coder

- One of the methods for e.g. black-and-white images
- Others:
 - *Q-coder* (predecessor of QM, tailored to hardware impl. / IBM)
 - *MQ-coder* (in JBIG2; Joint Bi-Level Image Compression Group)
 - *M-coder* (in H.264/AVC video compression standard)
- Tuned Markov model (finite-state automaton) for adapting probabilities.

Interval setting:

- MPS is the 'first' symbol
- Maintain *lower* and *range*:



QM-coder (cont.)

Key ideas:

- Operate within interval $[0, 1.5)$
- Rescale when $range < 0.75$
- Approximate $range$ by 1 in multiplications
$$range \cdot p \approx p$$
$$range \cdot (1-p) \approx range - p$$
- No pending bits, but a 'carry' bit can propagate to the output bits, which must be buffered. Unlimited propagation is prevented by 'stuffing' 0-bits after bytes containing only 1's (small redundancy).
- Practical implementation is done using integers within $[0, 65536)$.

4.4.3. Practical problems with arithmetic coding

- *Not partially decodable nor indexable:*
Start decoding always from the beginning even to recover a small section in the middle.
- *Vulnerable:* Bit errors result in a totally scrambled message
- *Not self-synchronizable,* contrary to Huffman code

Solution for static distributions: *Arithmetic Block Coding*

- Applies the idea of arithmetic coding within machine words
- Restarts a new coding loop when the word bits are 'used'.
- Resembles Tunstall code, but no explicit codebook.
- Fast, because avoids the scalings and bit-level operations.
- Non-optimal code length, but rather close