



## 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.)

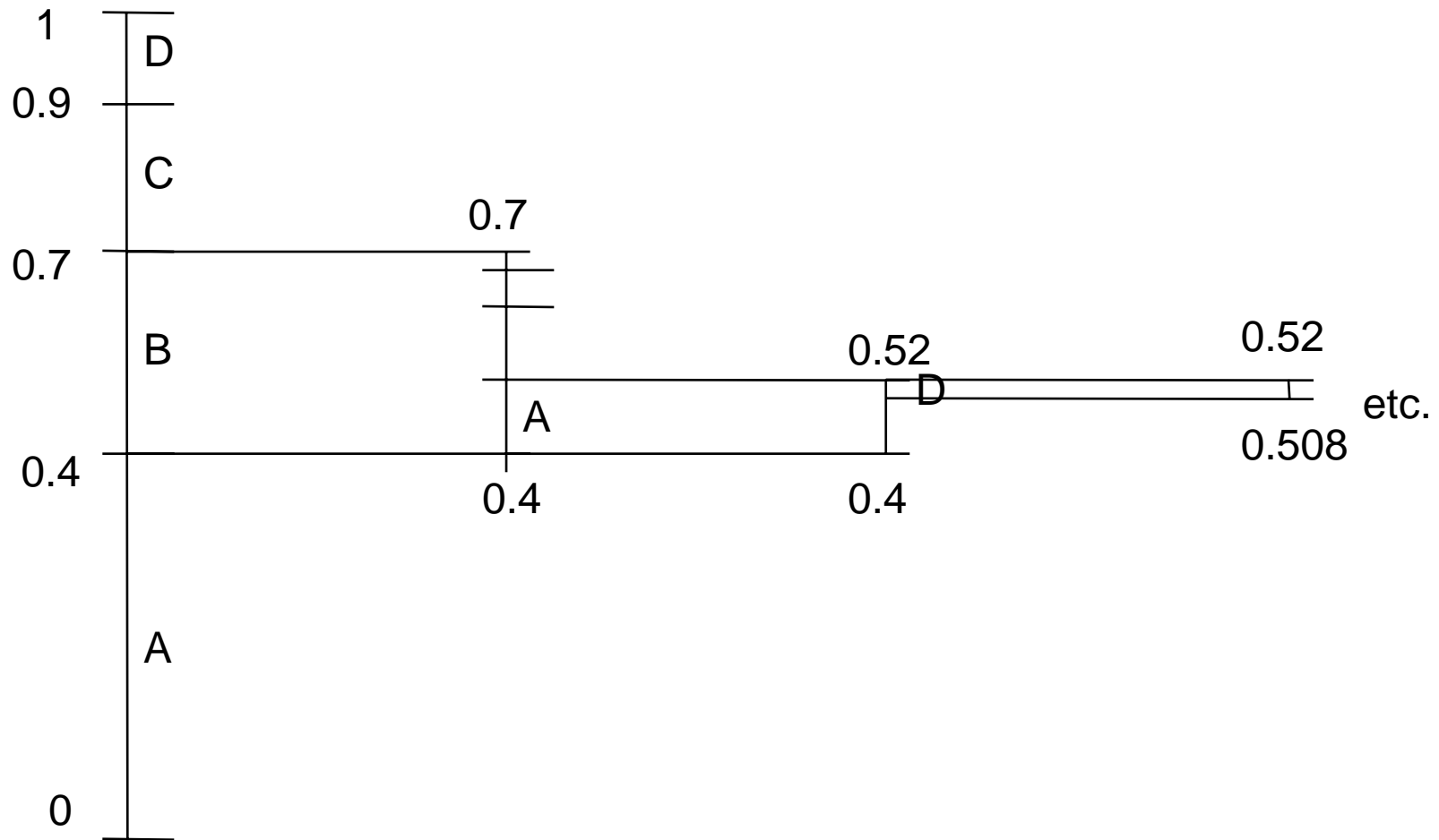
### Characterization:

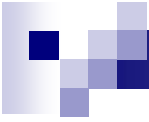
- One codeword for the whole message
- A kind of extreme case of extended Huffman (or Tunstall) coding
- No codebook required
- No clear correspondence between source symbols and code bits

### Basic ideas:

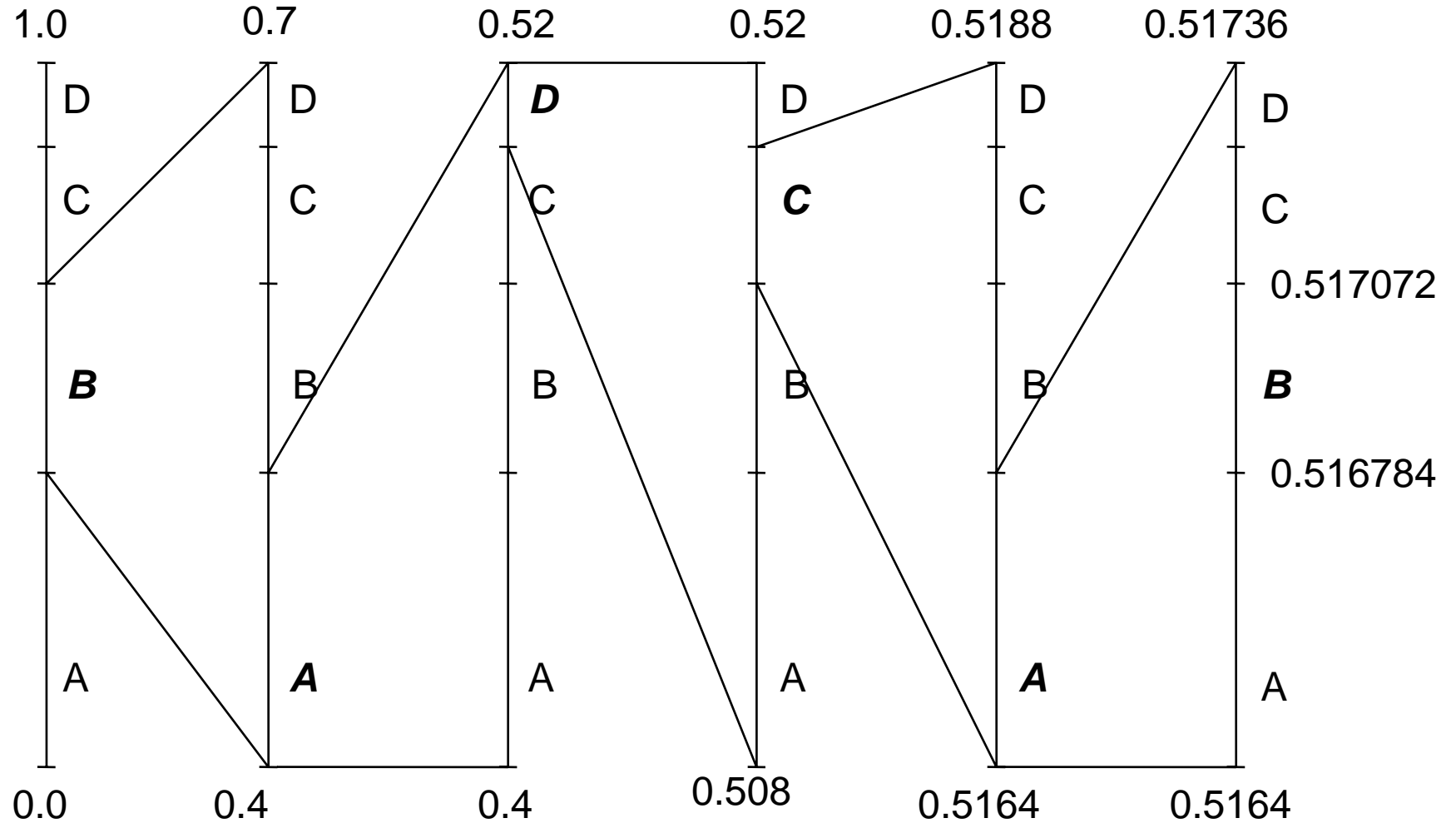
- Message is represented by a (small) interval in  $[0, 1)$
- Each successive symbol reduces the interval size
- Interval size = product of symbol probabilities
- Prefix-free messages result in disjoint intervals
- Final code = any value from the interval
- Decoding computes the same sequence of intervals

# Arithmetic coding: Encoding of "BADCAB"





# Encoding of "BADCAB" with rescaled intervals



# 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:***

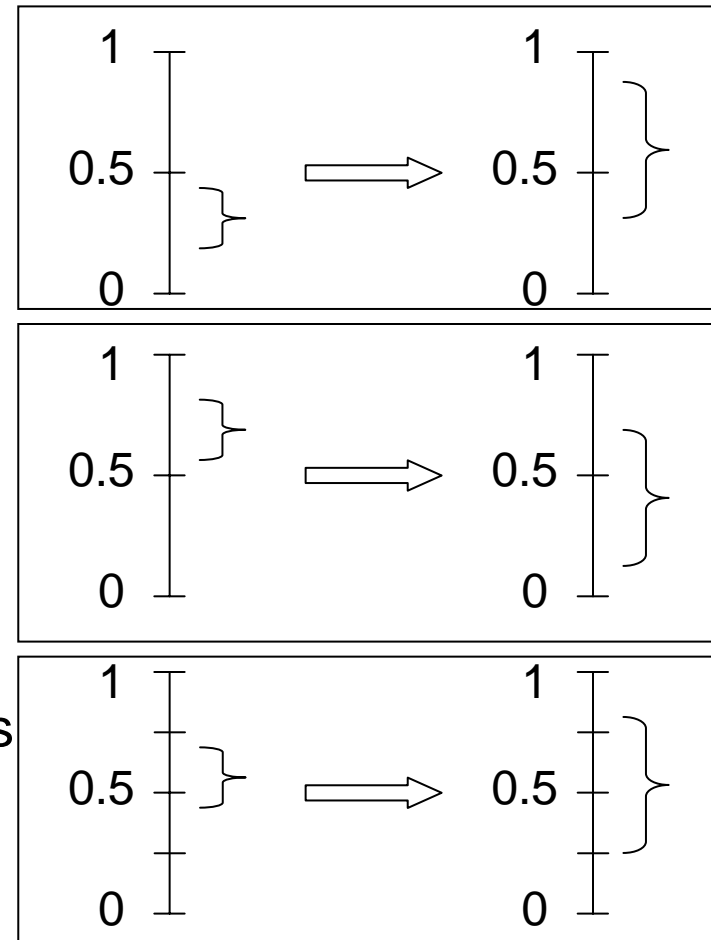
- transmit bit 0 (plus  $p$  pending 1's)
- $lower := 2 \cdot lower$
- $upper := 2 \cdot upper$

## ***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:**

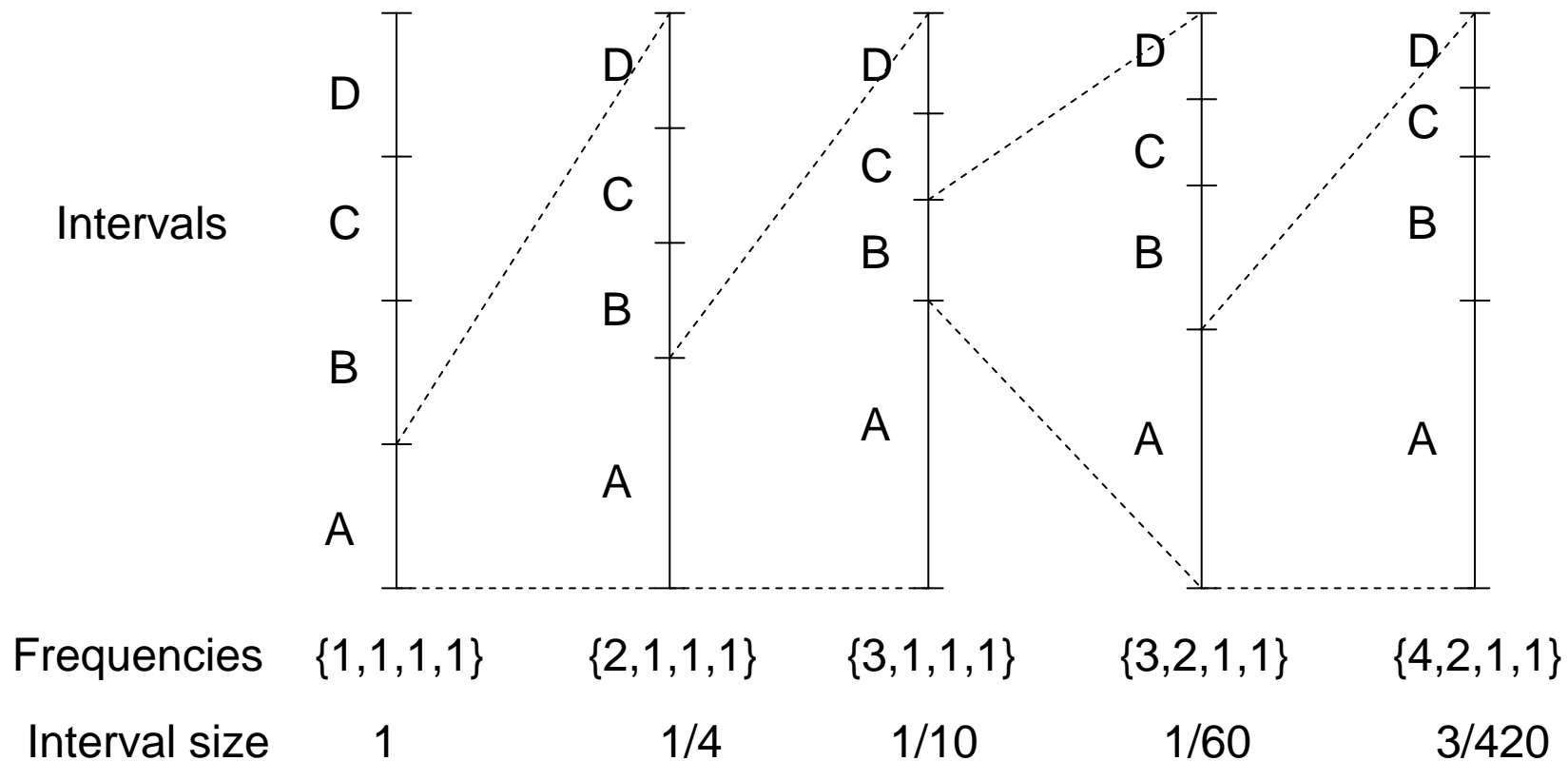
- Maintain frequencies of symbols during the coding
- Use the current frequencies in reducing the interval

### **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:

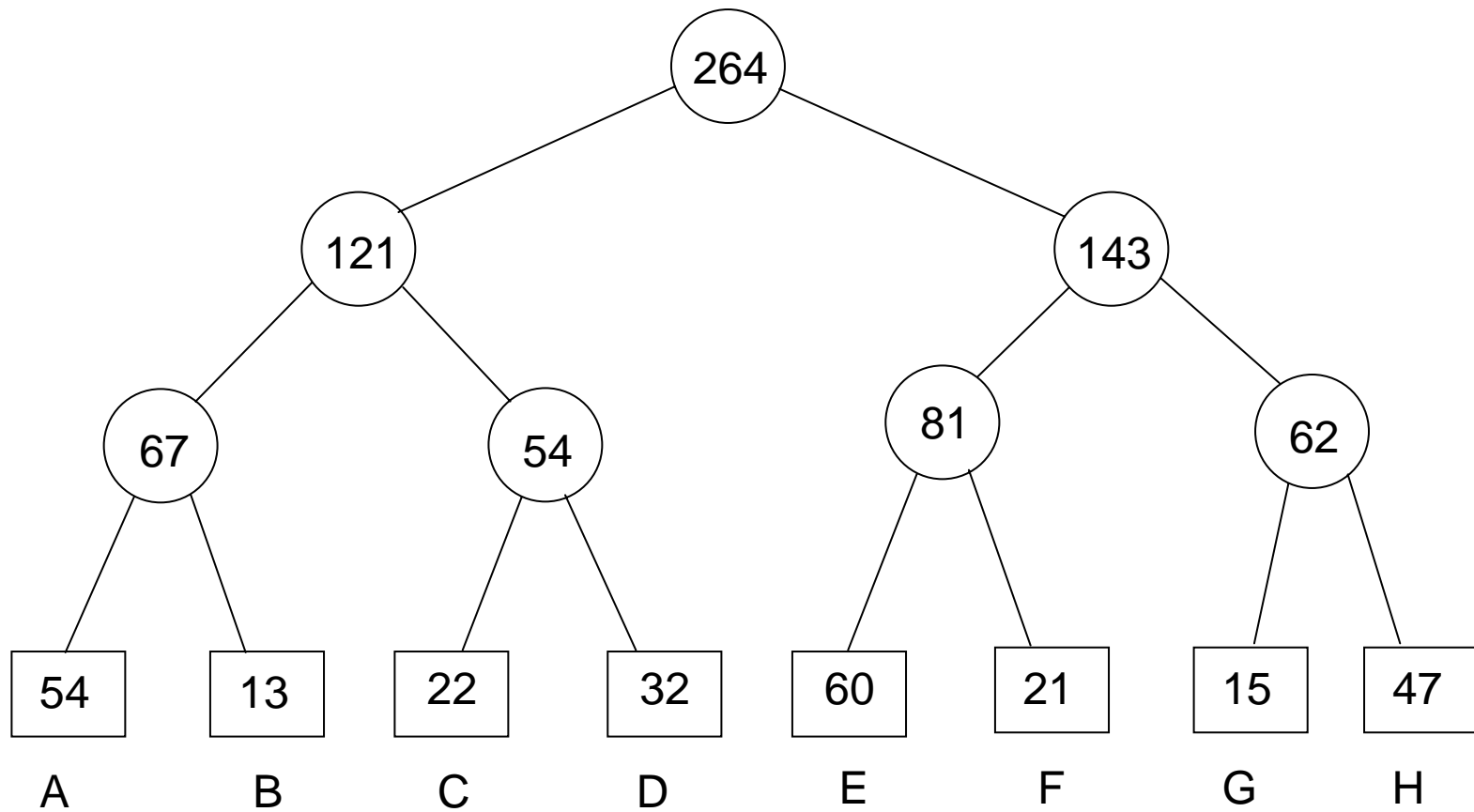
- Maintenance of cumulative frequencies; simple vector implementation has complexity  $O(q)$  for  $q$  symbols.

### 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:

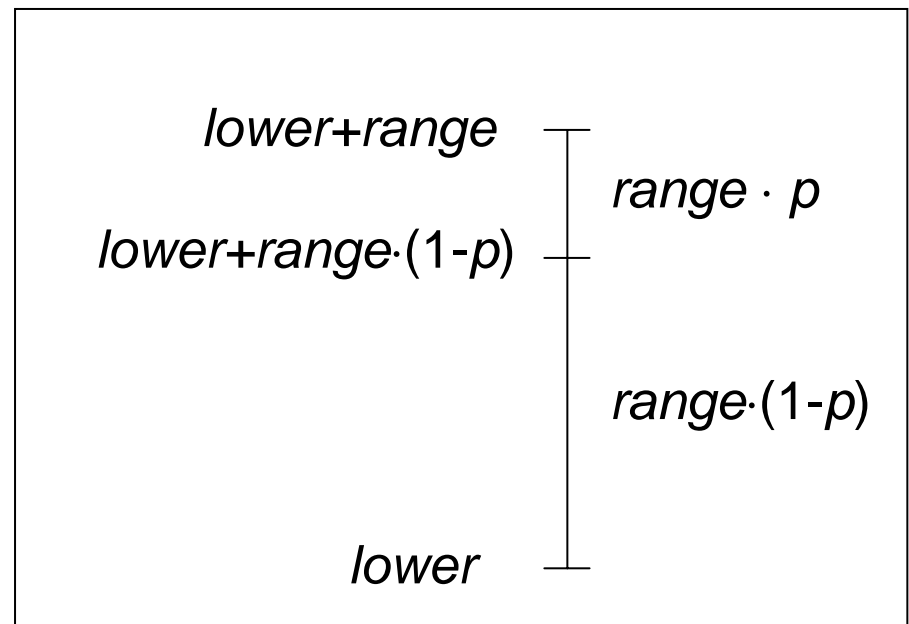
$$\frac{\textit{entropy}}{\textit{averageLength}} = \frac{-p \log p - (1-p) \log(1-p)}{-pQ - (1-p) \log(1-1/2^Q)}$$

## 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