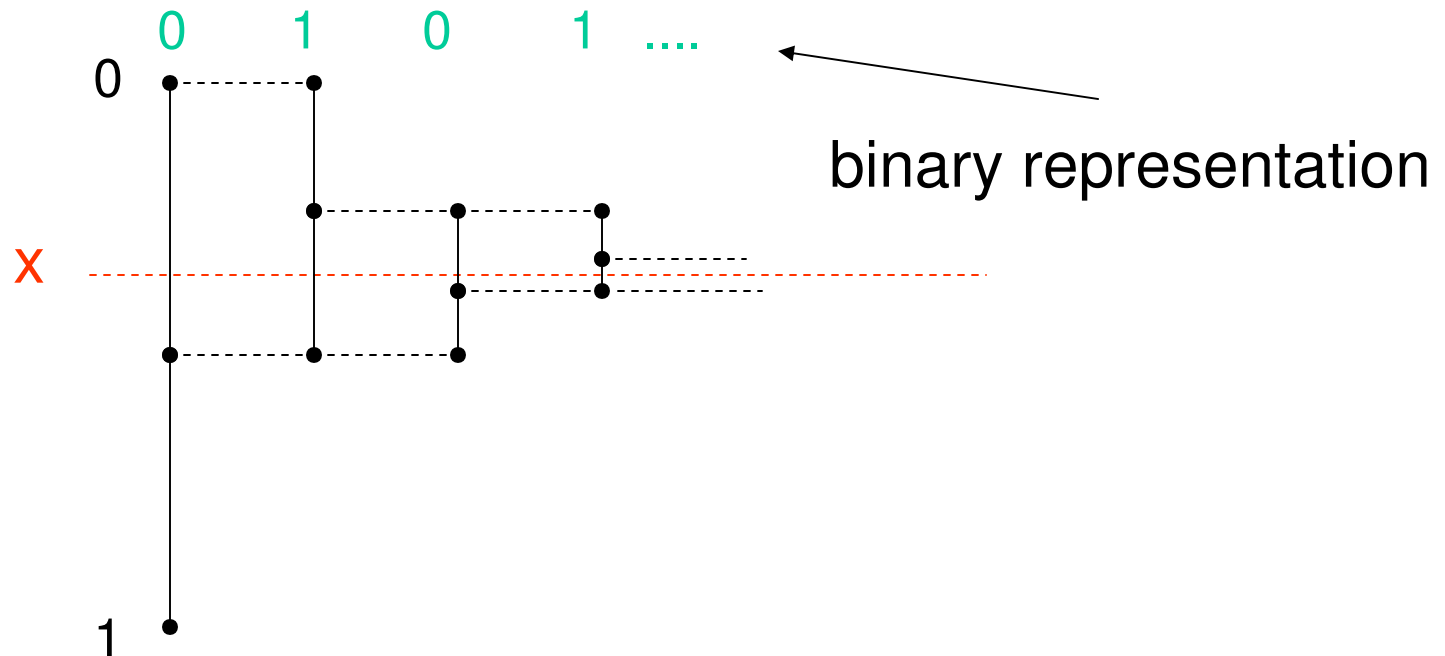


CSEP 590
Data Compression
Autumn 2007

Arithmetic Coding

Reals in Binary

- Any real number x in the interval $[0,1)$ can be represented in binary as $.b_1b_2\dots$ where b_i is a bit.



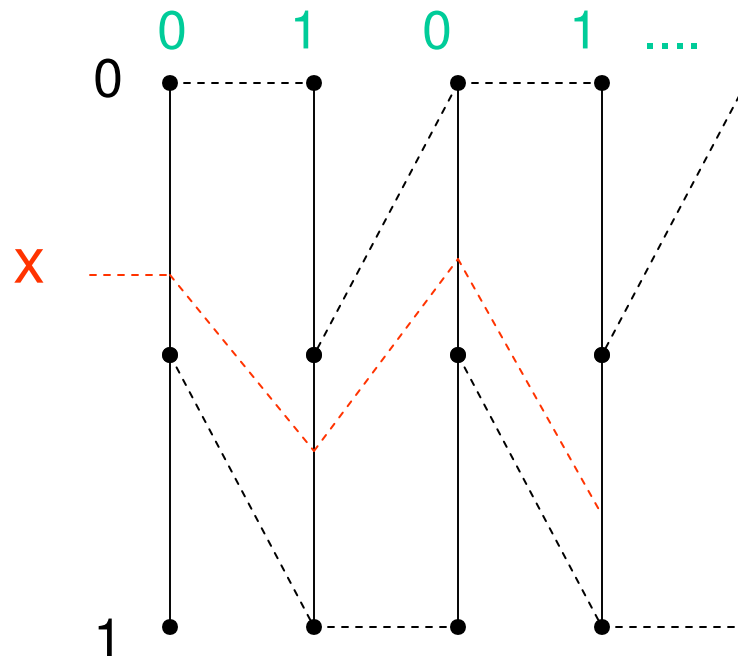
First Conversion

```
L := 0; R := 1; i := 1
while x > L *
  if x < (L+R)/2 then bi := 0 ; R := (L+R)/2;
  if x ≥ (L+R)/2 then bi := 1 ; L := (L+R)/2;
  i := i + 1
end{while}
bj := 0 for all j ≥ i
```

* Invariant: x is always in the interval $[L,R)$

Conversion using Scaling

- Always scale the interval to unit size, but x must be changed as part of the scaling.



Binary Conversion with Scaling

```
y := x; i := 0
while y > 0 *
    i := i + 1;
    if y < 1/2 then bi := 0; y := 2y;
    if y ≥ 1/2 then bi := 1; y := 2y - 1;
end{while}
bj := 0 for all j ≥ i + 1
```

* Invariant: $x = .b_1b_2 \dots b_i + y/2^i$

Proof of the Invariant

- Initially $x = 0 + y/2^0$
- Assume $x = .b_1b_2 \dots b_i + y/2^i$
 - Case 1. $y < 1/2$. $b_{i+1} = 0$ and $y' = 2y$
$$\begin{aligned} .b_1b_2 \dots b_i b_{i+1} + y'/2^{i+1} &= .b_1b_2 \dots b_i 0 + 2y/2^{i+1} \\ &= .b_1b_2 \dots b_i + y/2^i \\ &= x \end{aligned}$$
 - Case 2. $y \geq 1/2$. $b_{i+1} = 1$ and $y' = 2y - 1$
$$\begin{aligned} .b_1b_2 \dots b_i b_{i+1} + y'/2^{i+1} &= .b_1b_2 \dots b_i 1 + (2y-1)/2^{i+1} \\ &= .b_1b_2 \dots b_i + 1/2^{i+1} + 2y/2^{i+1} - 1/2^{i+1} \\ &= .b_1b_2 \dots b_i + y/2^i \\ &= x \end{aligned}$$

Example and Exercise

$$x = 1/3$$

y	i	b
1/3	1	0
2/3	2	1
1/3	3	0
2/3	4	1
...

$$x = 17/27$$

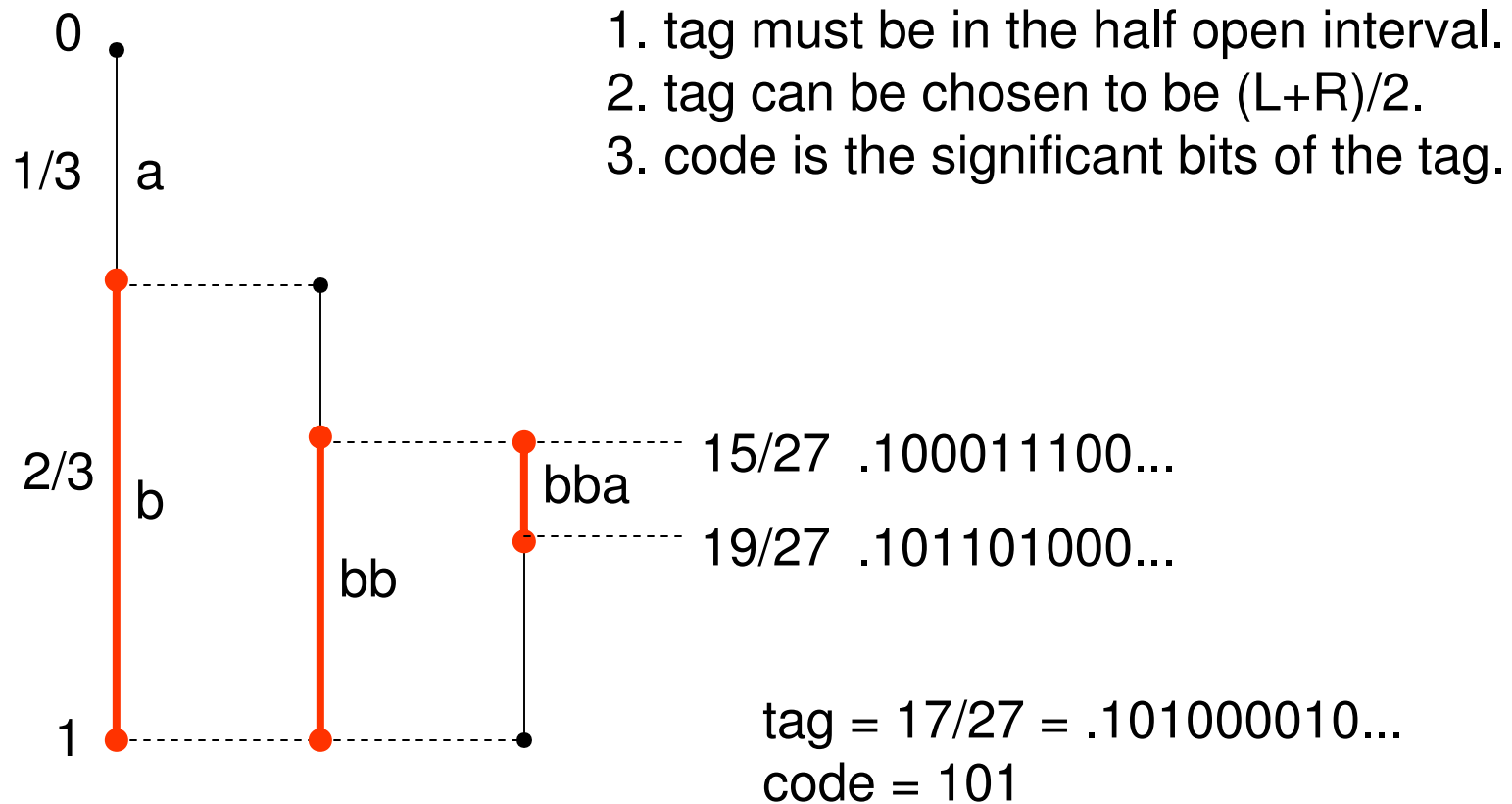
y	i	b
17/27	1	1

Arithmetic Coding

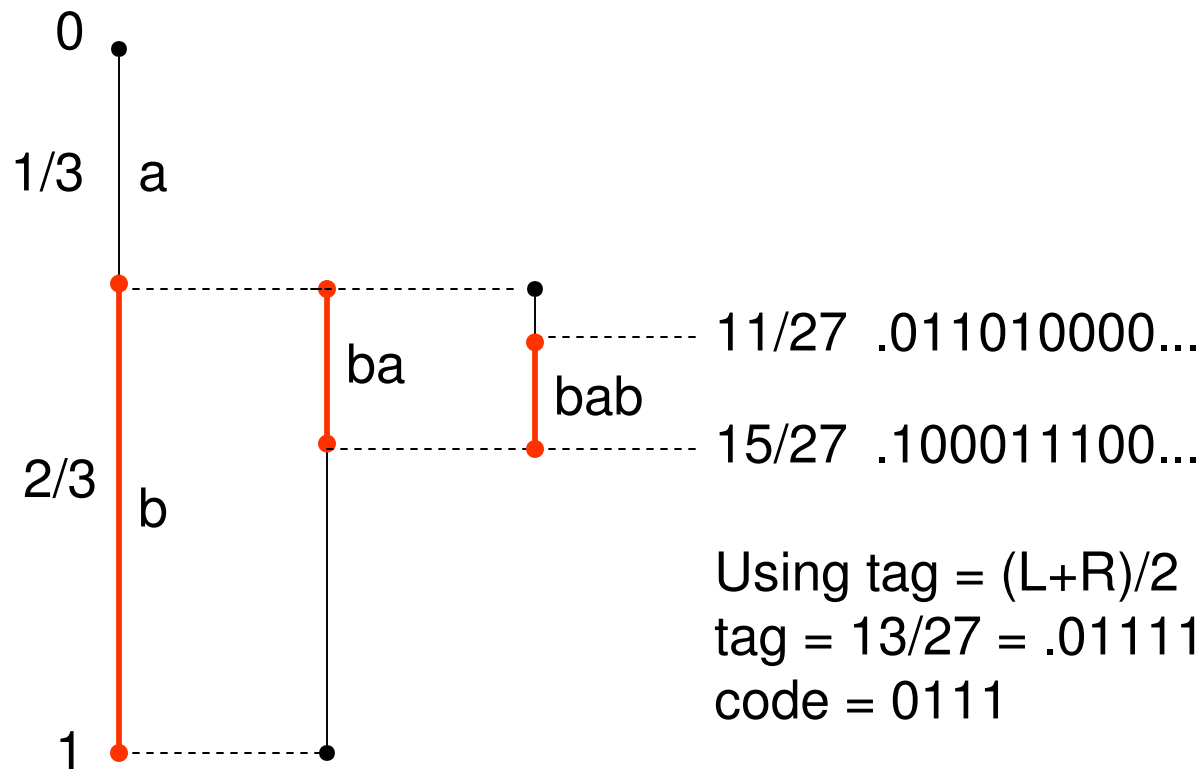
Basic idea in arithmetic coding:

- represent each string x of length n by a unique interval $[L,R)$ in $[0,1)$.
- The width $R-L$ of the interval $[L,R)$ represents the probability of x occurring.
- The interval $[L,R)$ can itself be represented by any number, called a tag, within the half open interval.
- The k significant bits of the tag $.t_1t_2t_3\dots$ is the code of x . That is, $.t_1t_2t_3\dots t_k000\dots$ is in the interval $[L,R)$.
 - It turns out that $k \approx \log_2(1/(R-L))$.

Example of Arithmetic Coding (1)



Some Tags are Better than Others

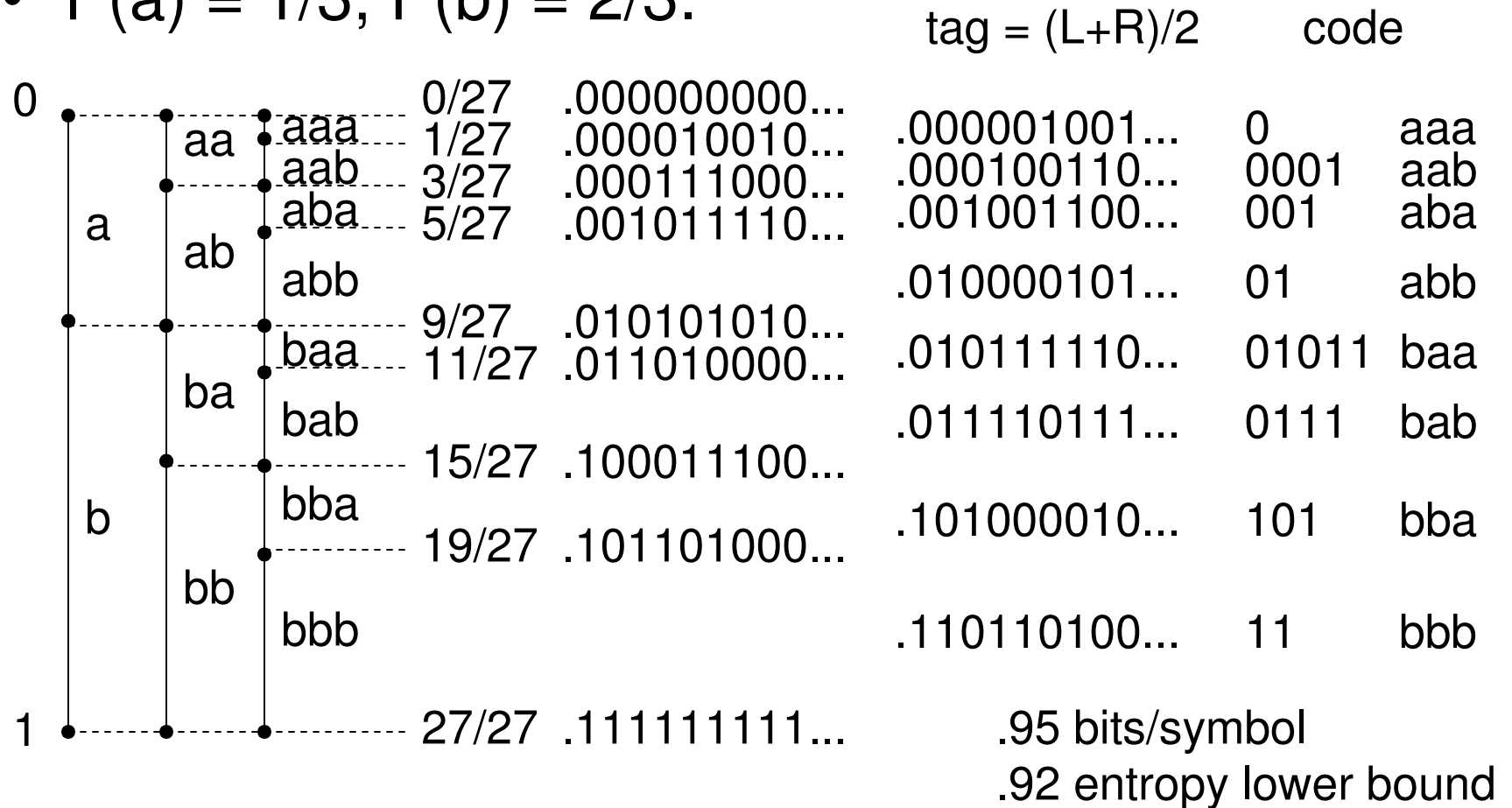


Using tag = $(L+R)/2$
tag = $13/27 = .011110110...$
code = 0111

Alternative tag = $14/27 = .100001001...$
code = 1

Example of Codes

- $P(a) = 1/3, P(b) = 2/3.$

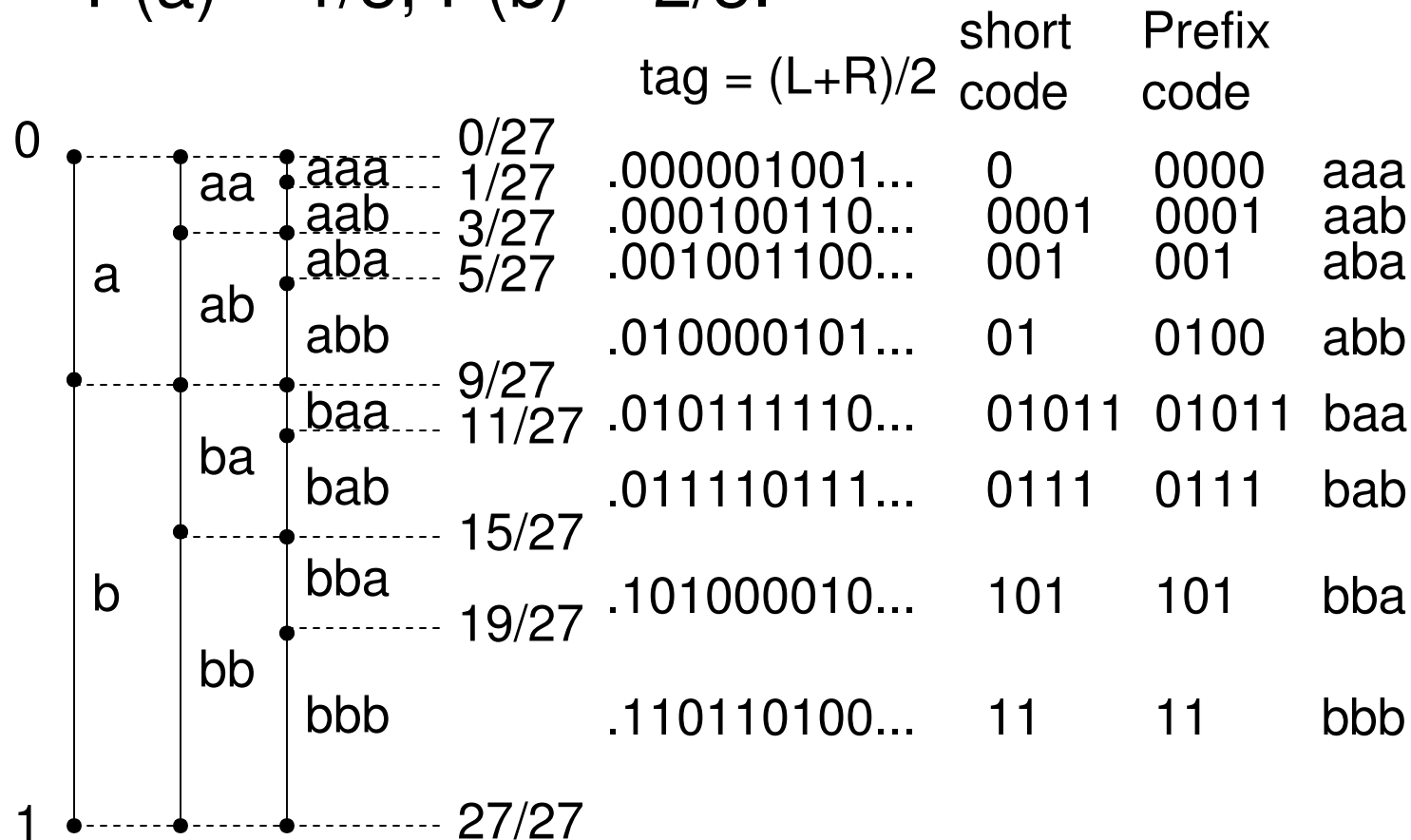


Code Generation from Tag

- If binary tag is $.t_1t_2t_3\dots = (L+R)/2$ in $[L,R)$ then we want to choose k to form the code $t_1t_2\dots t_k$.
- Short code:
 - choose k to be as small as possible so that $L \leq .t_1t_2\dots t_k000\dots < R$.
- Guaranteed code:
 - choose $k = \lceil \log_2 (1/(R-L)) \rceil + 1$
 - $L \leq .t_1t_2\dots t_kb_1b_2b_3\dots < R$ for any bits $b_1b_2b_3\dots$
 - for fixed length strings provides a good prefix code.
 - example: $[.000000000\dots, .000010010\dots)$, tag = $.000001001\dots$
Short code: 0
Guaranteed code: 000001

Guaranteed Code Example

- $P(a) = 1/3, P(b) = 2/3$.



Arithmetic Coding Algorithm

- $P(a_1), P(a_2), \dots, P(a_m)$
- $C(a_i) = P(a_1) + P(a_2) + \dots + P(a_{i-1})$
- Encode $x_1x_2\dots x_n$

```
Initialize L := 0 and R:= 1;  
for i = 1 to n do  
    W := R - L;  
    L := L + W * C(xi);  
    R := L + W * P(xi);  
t := (L+R)/2;  
choose code for the tag
```

Arithmetic Coding Example

- $P(a) = 1/4, P(b) = 1/2, P(c) = 1/4$
- $C(a) = 0, C(b) = 1/4, C(c) = 3/4$
- abca

	symbol	W	L	R
			0	1
$W := R - L;$	a	1	0	1/4
$L := L + W C(x);$	b	1/4	1/16	3/16
$R := L + W P(x)$	c	1/8	5/32	6/32
	a	1/32	5/32	21/128

$$\text{tag} = (5/32 + 21/128)/2 = 41/256 = .001010010\dots$$

$$L = .001010000\dots$$

$$R = .001010100\dots$$

$$\text{code} = 00101$$

$$\text{prefix code} = 00101001$$

Arithmetic Coding Exercise

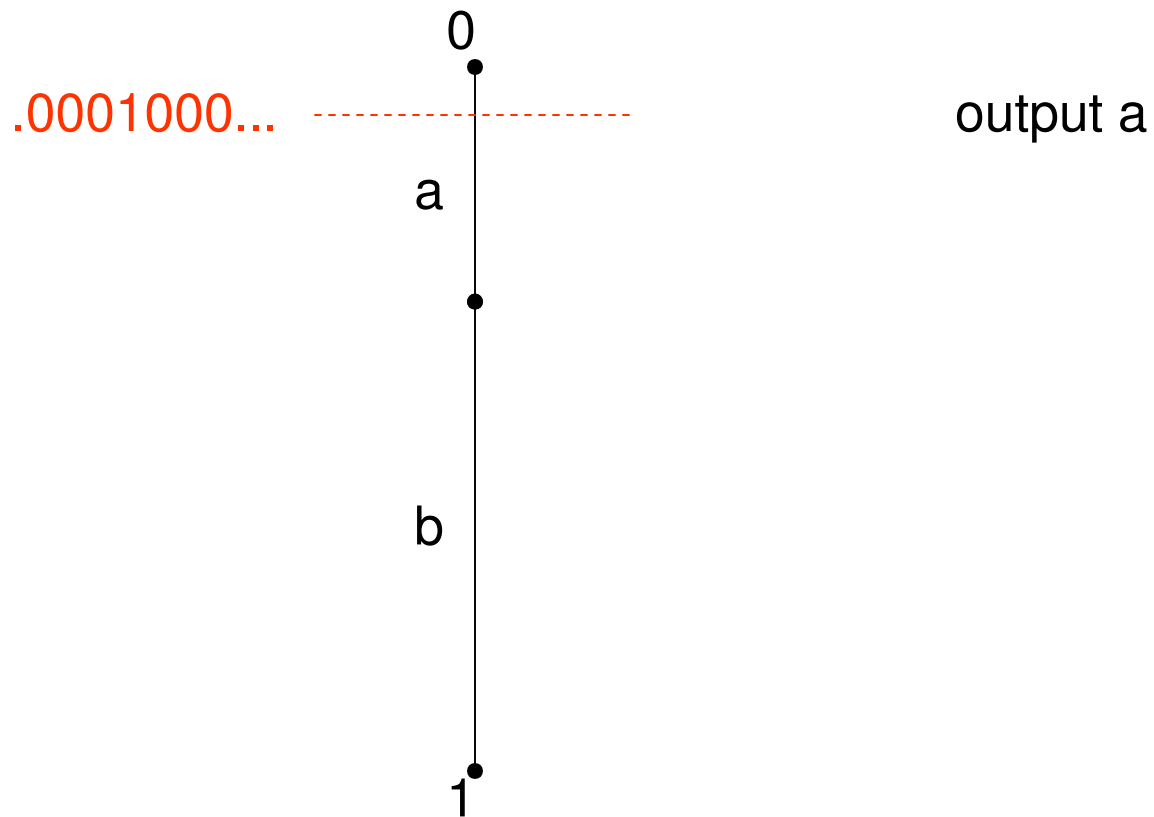
- $P(a) = 1/4, P(b) = 1/2, P(c) = 1/4$
- $C(a) = 0, C(b) = 1/4, C(c) = 3/4$
- bbbb

	symbol	W	L	R
			0	1
$W := R - L;$	b	1		
$L := L + W C(x);$	b			
$R := L + W P(x)$	b			
	b			

tag =
L =
R =
code =
prefix code =

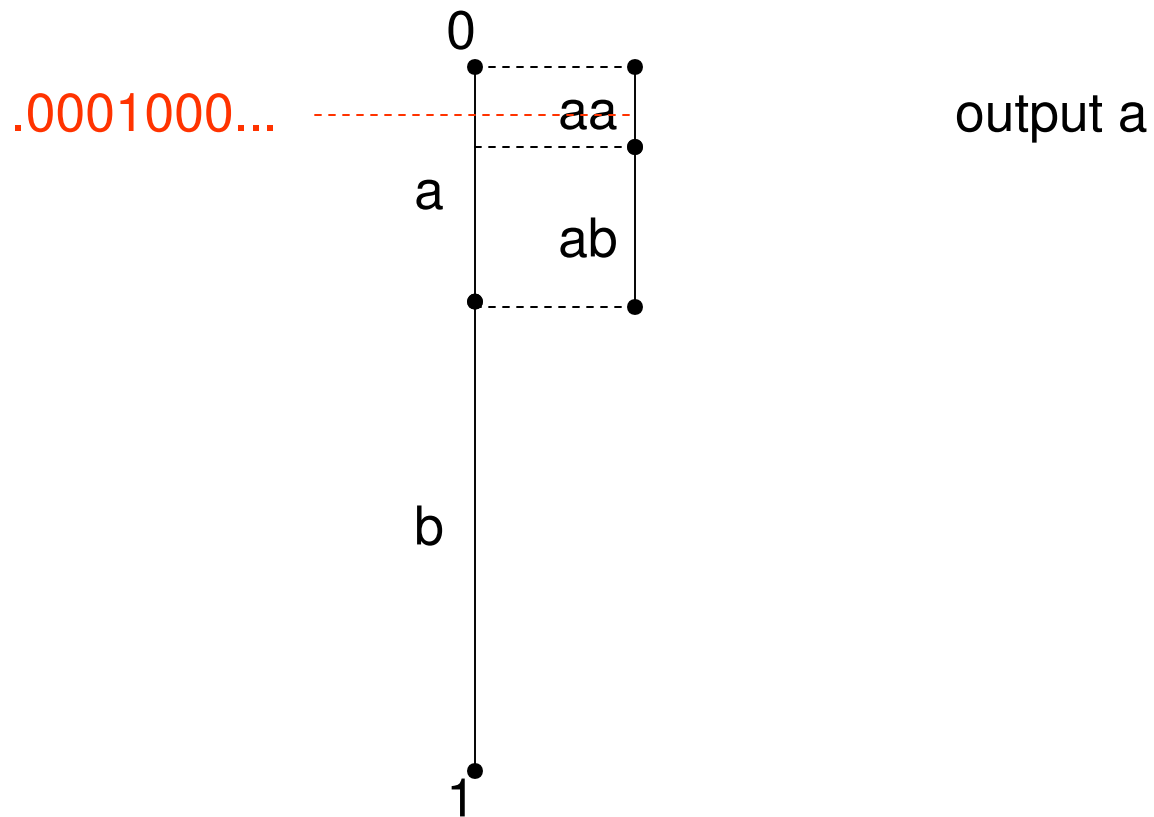
Decoding (1)

- Assume the length is known to be 3.
- 0001 which converts to the tag .0001000...



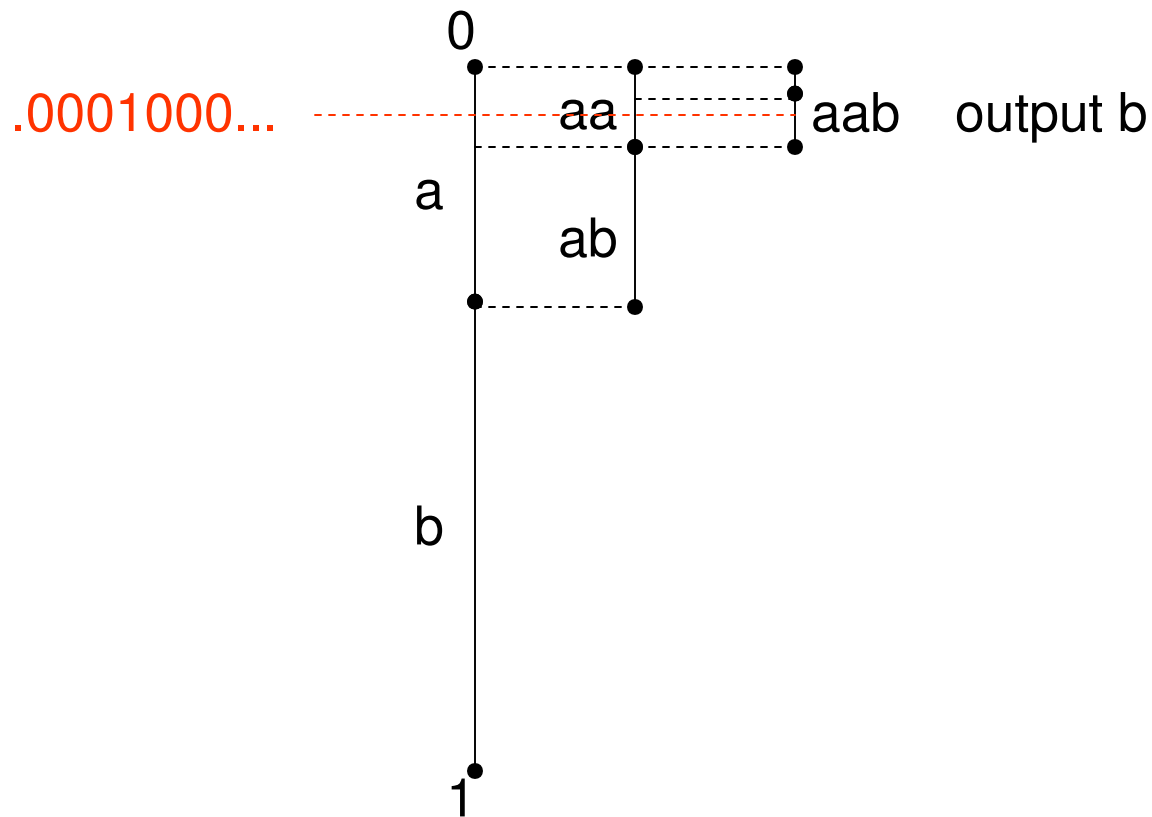
Decoding (2)

- Assume the length is known to be 3.
- 0001 which converts to the tag .0001000...



Decoding (3)

- Assume the length is known to be 3.
- 0001 which converts to the tag .0001000...



Arithmetic Decoding Algorithm

- $P(a_1), P(a_2), \dots, P(a_m)$
- $C(a_i) = P(a_1) + P(a_2) + \dots + P(a_{i-1})$
- Decode $b_1 b_2 \dots b_k$, number of symbols is n .

```
Initialize L := 0 and R := 1;  
t := .b1b2...bk000...  
for i = 1 to n do  
    W := R - L;  
    find j such that  $L + W * C(a_j) \leq t < L + W * (C(a_j) + P(a_j))$   
    output aj;  
    L := L + W * C(aj);  
    R := L + W * P(aj);
```

Decoding Example

- $P(a) = 1/4, P(b) = 1/2, P(c) = 1/4$
- $C(a) = 0, C(b) = 1/4, C(c) = 3/4$
- 00101

tag = .00101000... = 5/32

W	L	R	output
	0	1	
1	0	1/4	a
1/4	1/16	3/16	b
1/8	5/32	6/32	c
1/32	5/32	21/128	a

Decoding Issues

- There are at least two ways for the decoder to know when to stop decoding.
 1. Transmit the length of the string
 2. Transmit a unique end of string symbol

Practical Arithmetic Coding

- Scaling:
 - By scaling we can keep L and R in a reasonable range of values so that $W = R - L$ does not underflow.
 - The code can be produced progressively, not at the end.
 - Complicates decoding some.
- Integer arithmetic coding avoids floating point altogether.

More Issues

- Context
- Adaptive
- Comparison with Huffman coding

Scaling

- Scaling:
 - By scaling we can keep L and R in a reasonable range of values so that $W = R - L$ does not underflow.
 - The code can be produced progressively, not at the end.
 - Complicates decoding some.

Scaling during Encoding

Lower half

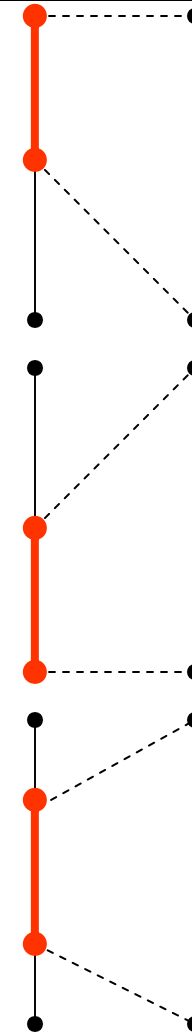
If $[L,R)$ is contained in $[0,.5)$ then
 $L := 2L$; $R := 2R$
output 0, followed by C 1's
 $C := 0$.

Upper half

If $[L,R)$ is contained in $[.5,1)$ then
 $L := 2L - 1$, $R := 2R - 1$
output 1, followed by C 0's
 $C := 0$

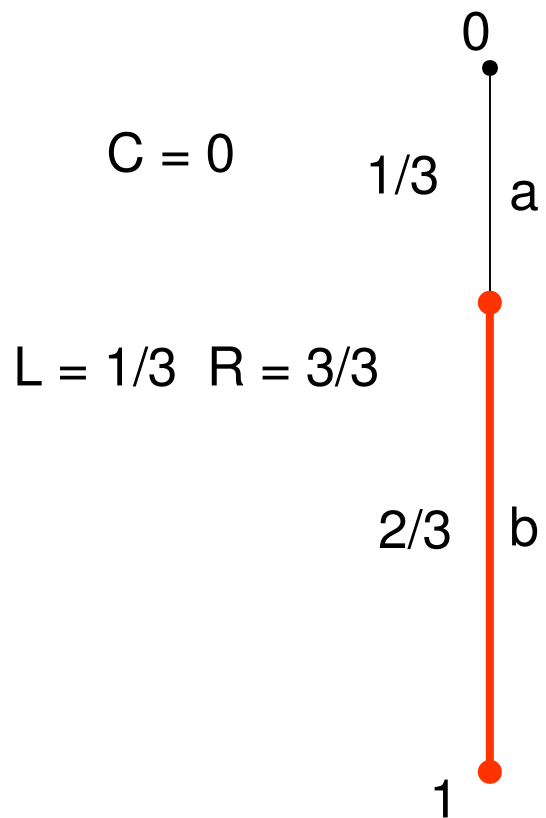
Middle Half

If $[L,R)$ is contained in $[.25,.75)$ then
 $L := 2L - .5$, $R := 2R - .5$
 $C := C + 1$.



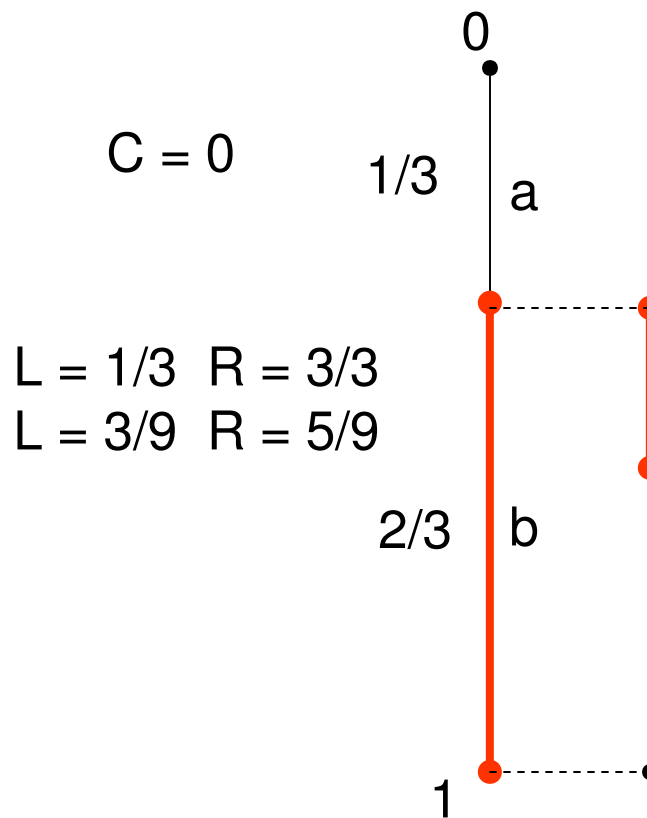
Example

- baa



Example

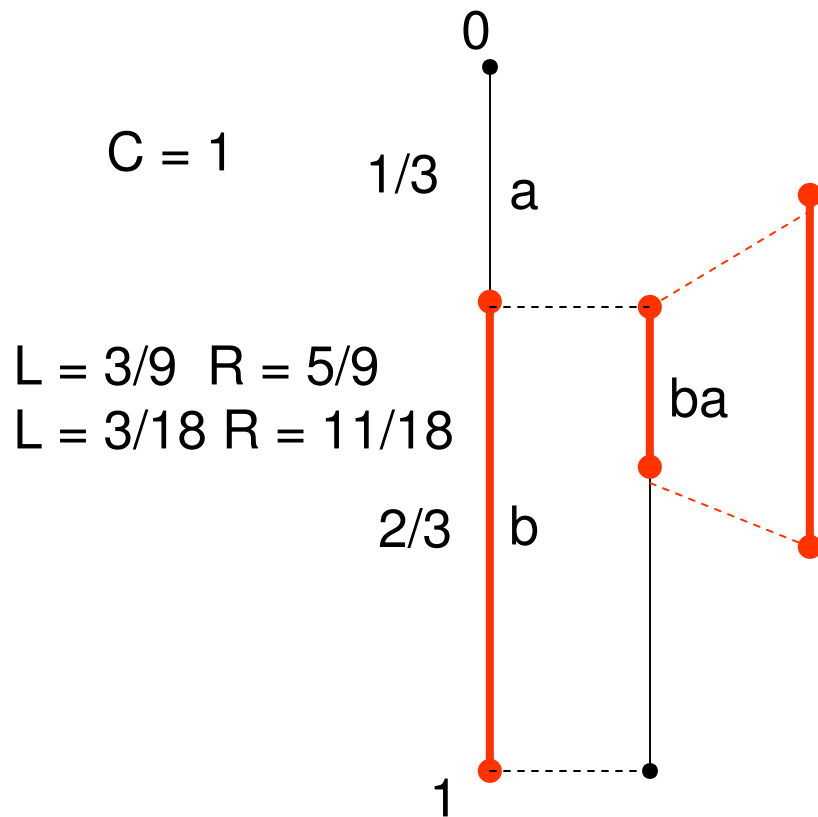
- baa



Scale middle half

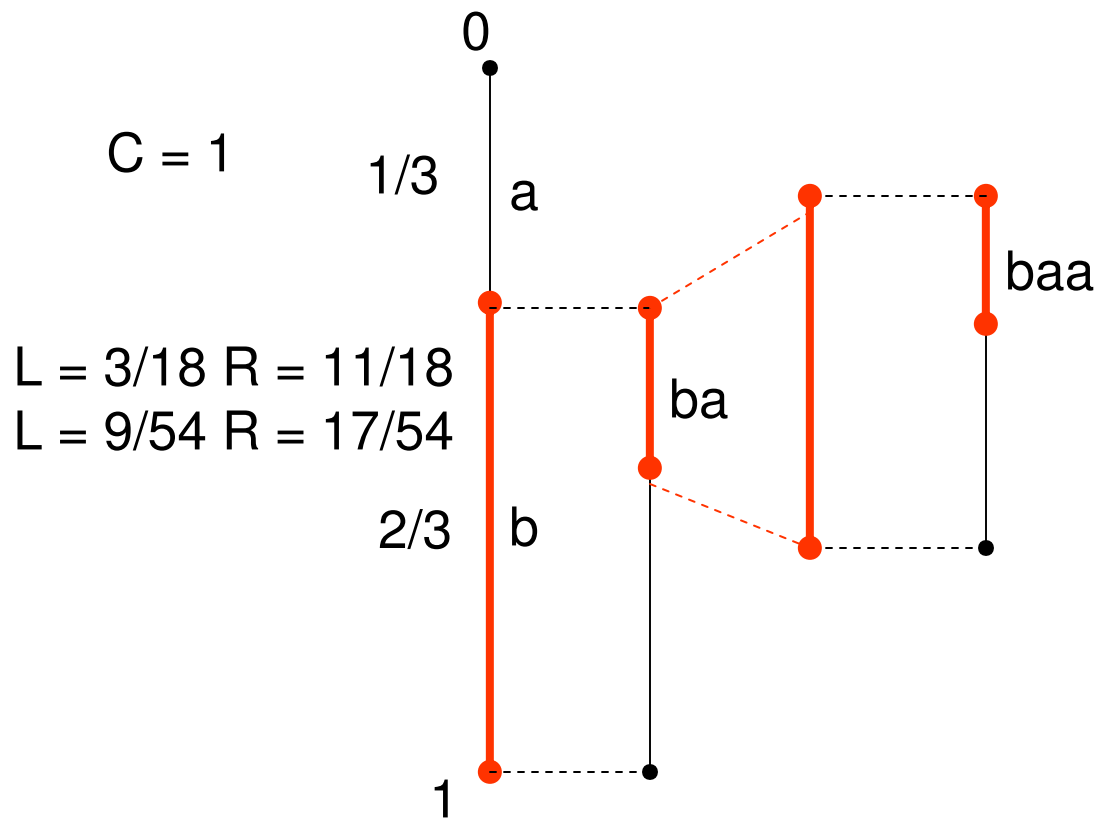
Example

- baa



Example

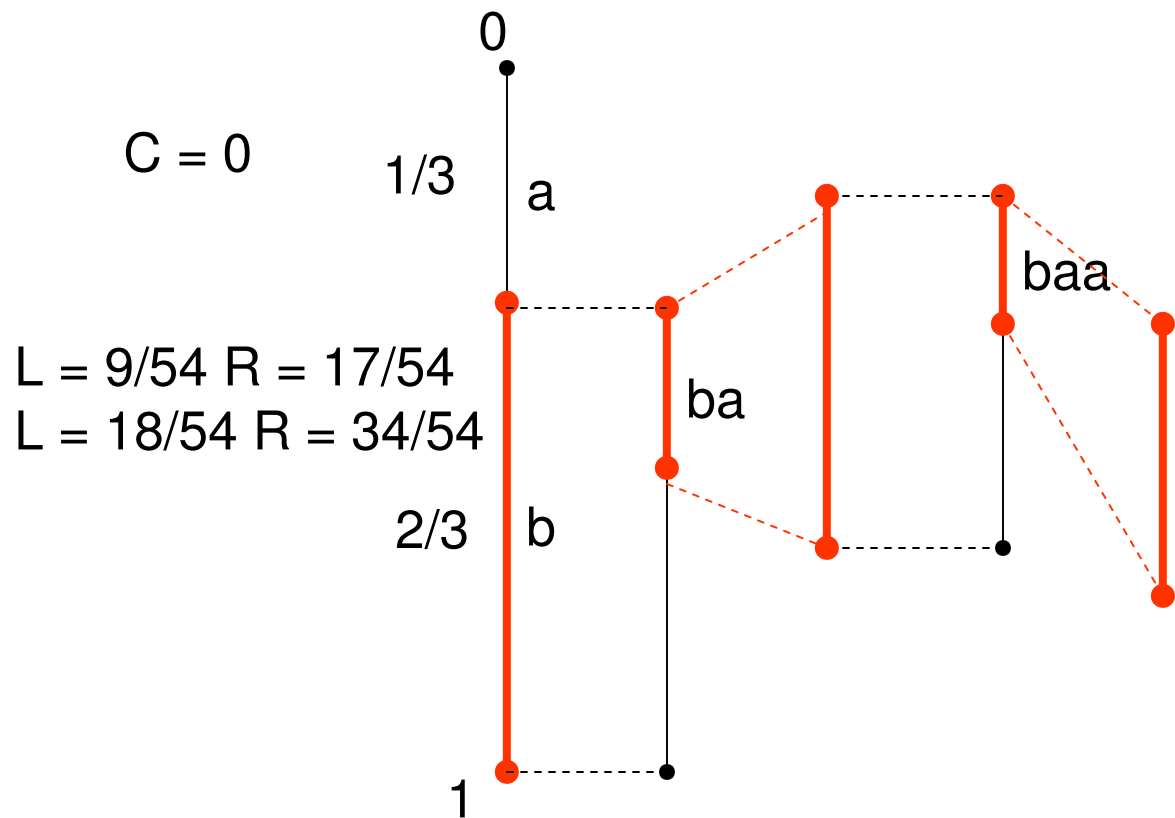
- $ba\bar{a}$



Scale lower half

Example

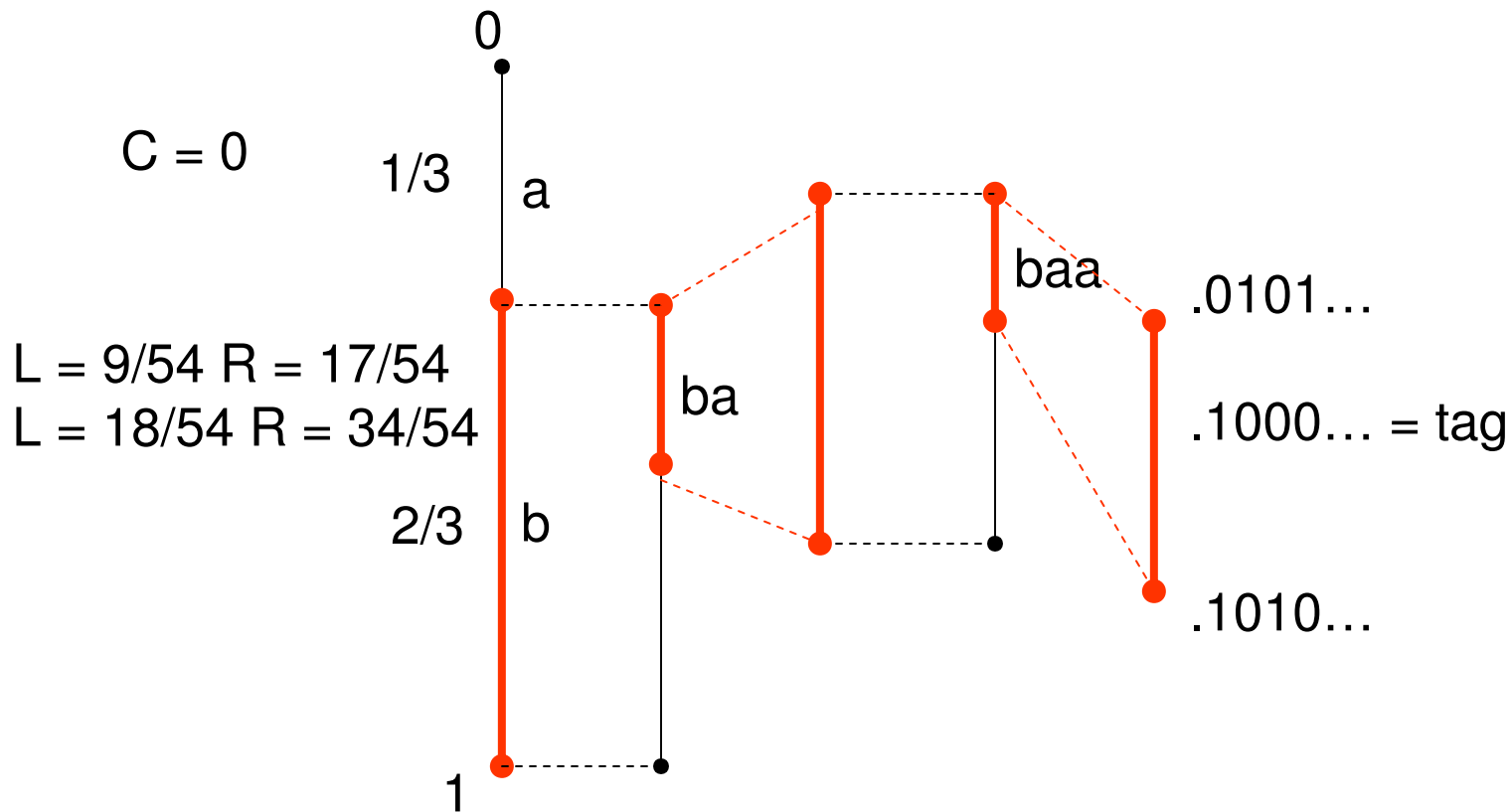
- baa 01



Example

- baa 011

In end $L < \frac{1}{2} < R$, choose tag to be $\frac{1}{2}$



Exercise

Model: a: $1/4$; b: $3/4$

Encode: bba

Decoding

- The decoder behaves just like the encoder except that C does not need to be maintained.
- Instead, the input stream is consumed during scaling.

Scaling during Decoding

Lower half

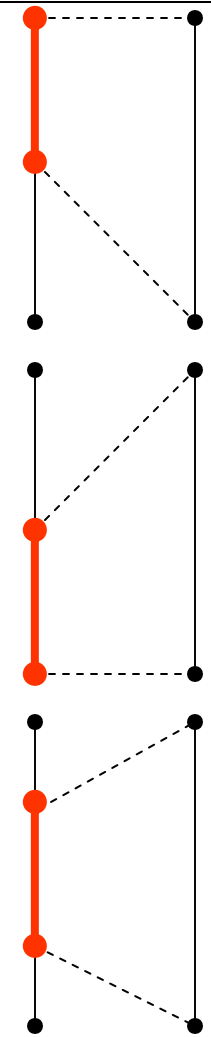
If $[L,R)$ is contained in $[0,.5)$ then
 $L := 2L; R := 2R$
consume 0 from the encoded stream

Upper half

If $[L,R)$ is contained in $[.5,1)$ then
 $L := 2L - 1, R := 2R - 1$
consume 1 from the encoded stream

Middle half

If $[L,R)$ is contained in $[.25,.75)$ then
 $L := 2L - .5, R := 2R - .5$
Replace 01 with 0 on stream
Replace 10 with 1 on stream



Scaling Math for the Tag

- Lower Half
 - $.0b_1b_2\dots \times 10 = .b_1b_2$
- Upper Half
 - $.1b_1b_2\dots \times 10 - 1 = .b_1b_2$
- Middle Half
 - $.01b_2b_3\dots \times 10 - .1 = .0b_2b_3$
 - $.10b_2b_3\dots \times 10 - .1 = .1b_2b_3$

Exercise

Model: a: $1/4$; b: $3/4$

Decode: 001 to 3 symbols

Integer Implementation

- m bit integers
 - Represent 0 with 000...0 (m times)
 - Represent 1 with 111...1 (m times)
- Probabilities represented by frequencies
 - n_i is the number of times that symbol a_i occurs
 - $C_i = n_1 + n_2 + \dots + n_{i-1}$
 - $N = n_1 + n_2 + \dots + n_m$

$$W := R - L + 1$$

$$L' := L + \left\lfloor \frac{W \cdot C_i}{N} \right\rfloor$$

$$R := L + \left\lfloor \frac{W \cdot C_{i+1}}{N} \right\rfloor - 1$$

$$L := L'$$

Coding the i-th symbol using integer calculations.
Must use scaling!

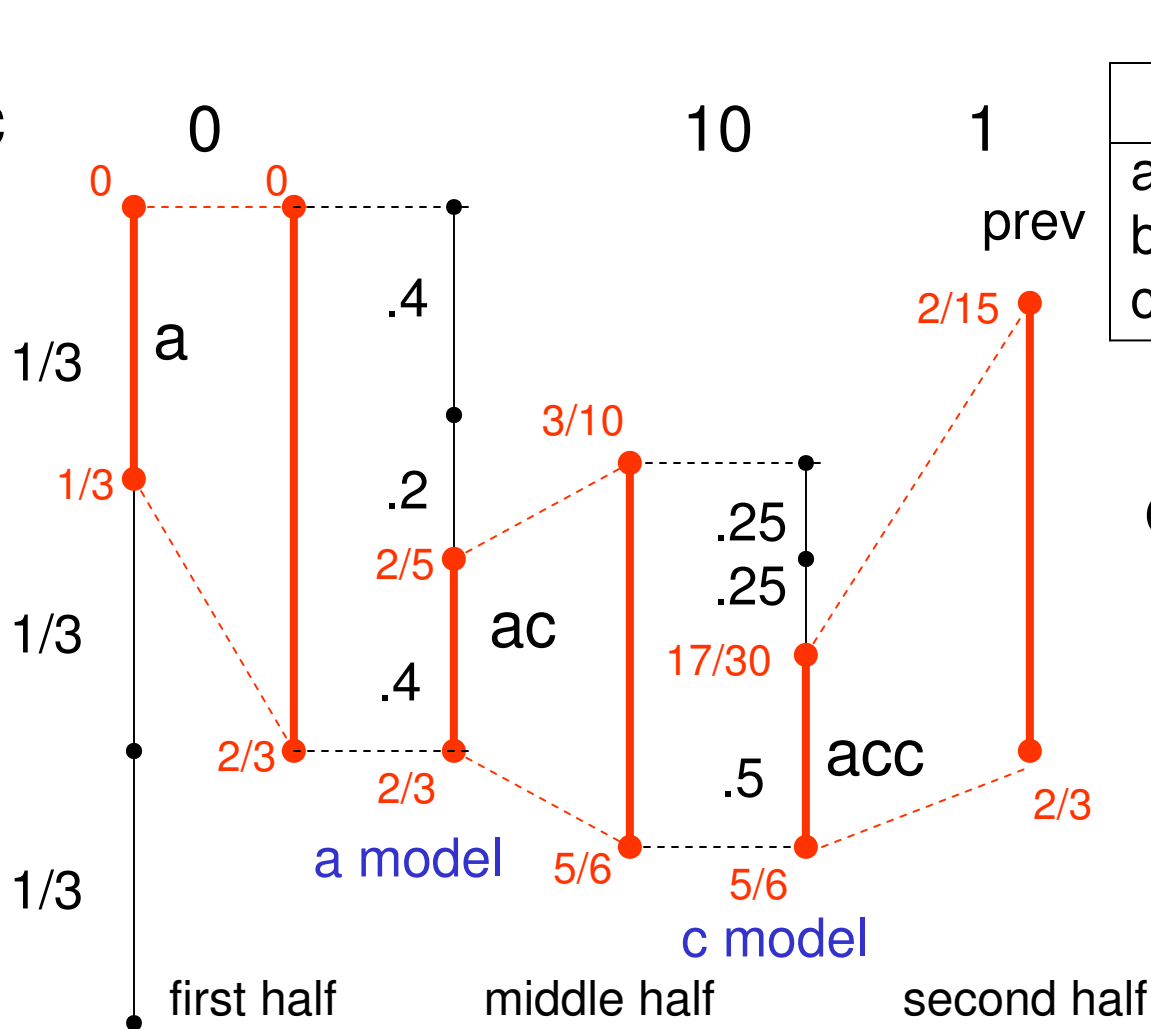
Context

- Consider 1 symbol context.
- Example: 3 contexts.

		next		
		a	b	c
prev	a	.4	.2	.4
	b	.1	.8	.1
	c	.25	.25	.5

Example with Scaling

- acc



	next		
	a	b	c
a	.4	.2	.4
b	.1	.8	.1
c	.25	.25	.5

Code = 0101

Arithmetic Coding with Context

- Maintain the probabilities for each context.
- For the first symbol use the equal probability model
- For each successive symbol use the model for the previous symbol.

Adaptation

- Simple solution – **Equally Probable Model**.
 - Initially all symbols have frequency 1.
 - After symbol x is coded, increment its frequency by 1
 - Use the new model for coding the next symbol
- Example in alphabet a,b,c,d

		a	a	b	a	a	c
a	1	2	3	3	4	5	5
b	1	1	1	2	2	2	2
c	1	1	1	1	1	1	2
d	1	1	1	1	1	1	1

After aabaac is encoded
The probability model is
a 5/10 b 2/10
c 2/10 d 1/10

Zero Frequency Problem

- How do we weight symbols that have not occurred yet.
 - Equal weights? Not so good with many symbols
 - Escape symbol, but what should its weight be?
 - When a new symbol is encountered send the <esc>, followed by the symbol in the equally probable model. (Both encoded arithmetically.)

		a	a	b	a	a	c
a	0	1	2	2	3	4	4
b	0	0	0	1	1	1	1
c	0	0	0	0	0	0	1
d	0	0	0	0	0	0	0
<esc>	1	1	1	1	1	1	1

After aabaac is encoded
The probability model is
a 4/7 b 1/7
c 1/7 d 0
<esc> 1/7

PPM

- Prediction with Partial Matching
 - Cleary and Witten (1984)
- State of the art arithmetic coder
 - Arbitrary order context
 - Adaptive
- Needs good data structures to be efficient.

PPM Example

- abracadabra

0-order context	
a	3
b	1
r	1
c	1
<esc>	1

1st-order context		
a		
	b	1
	c	1
	<esc>	1
b		
	r	1
	<esc>	1
r		
	a	1
	<esc>	1
c		
	a	1
	<esc>	1

2nd-order context		
ab		
	r	1
	<esc>	1
br		
	a	1
	<esc>	1
ra		
	c	1
	<esc>	1
ac		
	a	1
	<esc>	1

PPM Example

- abracadabra

0-order context	
a	3
b	1
r	1
c	1
<esc>	1

Output

1-order <esc>

0-order <esc>

(-1)-order d.

Update tables

1st-order context		
a		
	b	1
	c	1
	<esc>	1
b		
	r	1
	<esc>	1
r		
	a	1
	<esc>	1
c		
	a	1
	<esc>	1

2nd-order context		
ab		
	r	1
	<esc>	1
br		
	a	1
	<esc>	1
ra		
	c	1
	<esc>	1
ac		
	a	1
	<esc>	1

- abracadabra

0-order context	
a	3
b	1
r	1
c	1
d	1
<esc>	1

1st-order context		
a	b	1
	c	1
	d	1
	<esc>	1
b	r	1
	<esc>	1
r	a	1
	<esc>	1
c	a	1
	<esc>	1

2nd-order context		
ab	r	1
	<esc>	1
br	a	1
	<esc>	1
ra	c	1
	<esc>	1
ac	a	1
	<esc>	1
ca	d	1
	<esc>	1

- abracadabra

0-order context	
a	3
b	1
r	1
c	1
d	1
<esc>	1

0-order d
Update tables

1st-order context		
a	b	1
	c	1
	d	1
	<esc>	1
b	r	1
	<esc>	1
r	a	1
	<esc>	1
c	a	1
	<esc>	1

2nd-order context		
ab	r	1
	<esc>	1
br	a	1
	<esc>	1
ra	c	1
	<esc>	1
ac	a	1
	<esc>	1
ca	d	1
	<esc>	1

- abracadabra

0-order context	
a	4
b	1
r	1
c	1
d	1
<esc>	1

1st-order context		
a	b	1
	c	1
	d	1
	<esc>	1
b	r	1
	<esc>	1
r	a	1
	<esc>	1
c	a	1
	<esc>	1
d	a	1
	<esc>	1

2nd-order context		
ab	r	1
	<esc>	1
br	a	1
	<esc>	1
ra	c	1
	<esc>	1
ac	a	1
	<esc>	1
ca	d	1
	<esc>	1
ad	a	1
	<esc>	1

- abracadabra

0-order context	
a	4
b	1
r	1
c	1
d	1
<esc>	1

1st order b in
context a
Update tables

1st-order context		
a	b	1
	c	1
	d	1
	<esc>	1
b	r	1
	<esc>	1
r	a	1
	<esc>	1
c	a	1
	<esc>	1
d	a	1
	<esc>	1

2nd-order context		
ab	r	1
	<esc>	1
br	a	1
	<esc>	1
ra	c	1
	<esc>	1
ac	a	1
	<esc>	1
ca	d	1
	<esc>	1
ad	a	1
	<esc>	1

Arithmetic vs. Huffman

- Both compress very well. For m symbol grouping.
 - Huffman is within $1/m$ of entropy.
 - Arithmetic is within $2/m$ of entropy.
- Context
 - Huffman needs a tree for every context.
 - Arithmetic needs a small table of frequencies for every context.
- Adaptation
 - Huffman has an elaborate adaptive algorithm
 - Arithmetic has a simple adaptive mechanism.
- Bottom Line – Arithmetic is more flexible than Huffman.