



# Lecture 6

## Arithmetic coding

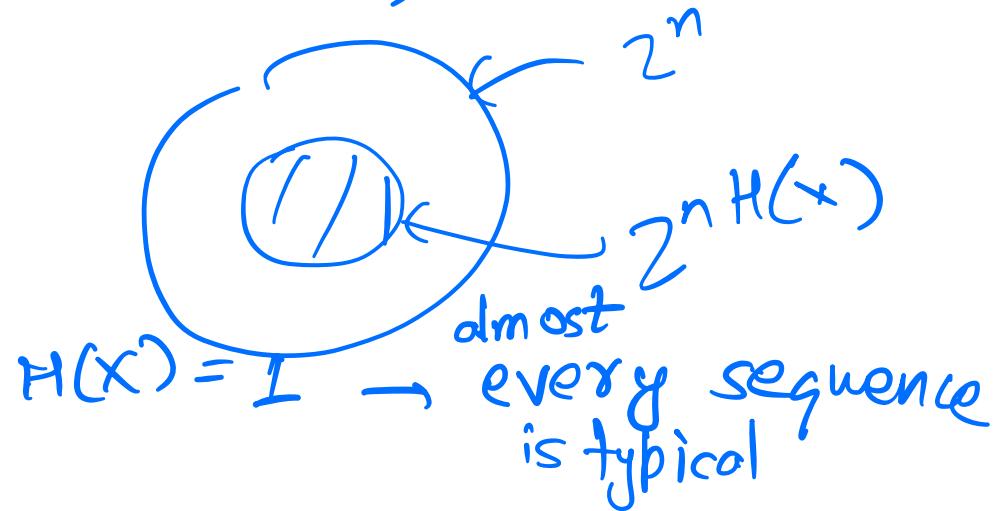
# Announcements

- Hw1 due on Wednesday
- Clarifications
  - Q1 - ~~nb.assert(almost-eq~~
  - Q5 - hints: floats  $\rightarrow$  bytes
- OH → Shubham after lecture today

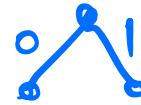
## Quiz Q1 Typical Set Size

Consider a binary source with  $P(0) = P(1) = 0.5$ . What is the size of the typical set  $A^{(n)}$ , in terms of  $n$ ?

$$\underline{\underline{2^n}}$$



## Quiz Q2 KL Divergence



Consider a source  $P = \text{Ber}(0.5)$ . Zoro knows the distribution of this source and designs a per-symbol Huffman code for  $P = \text{Ber}(0.5)$  to encode a sequence of symbols obtained using this source. However, Luffy doesn't know the distribution of this source and encodes it using a per-symbol Huffman code assuming that the sequence of symbols came from  $Q = \text{Ber}(0.25)$ .



2.1 How many extra number of bits in expectation (per-symbol) does Luffy need over Zoro to encode a sequence from the above source  $P$ ?

Some code

## Quiz Q2 KL Divergence

Consider a source  $P = \text{Ber}(0.5)$ . Zoro knows the distribution of this source and designs a per-symbol Huffman code for  $P = \text{Ber}(0.5)$  to encode a sequence of symbols obtained using this source. However, Luffy doesn't know the distribution of this source and encodes it using a per-symbol Huffman code assuming that the sequence of symbols came from  $Q = \text{Ber}(0.25)$ .

2.2 What is the KL divergence  $D(P||Q)$  between distributions  $P$  and  $Q$  specified above?

0.2075

$$0.5 \log_2 \frac{0.5}{0.25} + 0.5 \log_2 \frac{0.5}{0.75}$$

## Quiz Q2 KL Divergence

Consider a source  $P = \text{Ber}(0.5)$ . Zoro knows the distribution of this source and designs a per-symbol Huffman code for  $P = \text{Ber}(0.5)$  to encode a sequence of symbols obtained using this source. However, Luffy doesn't know the distribution of this source and encodes it using a per-symbol Huffman code assuming that the sequence of symbols came from  $Q = \text{Ber}(0.25)$ .

2.3 In the class we learnt that KL divergence is an indicator of the excess code-length for mismatched codes. How do you explain that the two answers above do not match?

$\% \rightarrow$  does not achieve entropy for  $Q$ .  
For a large block size  $\rightarrow$  you would get  $D(P||Q)$ .

**Slides credit - Kedar Tatwawadi**

## RECAP

$$H(X) \leq \underset{\text{P}}{\mathbb{E}} l_{\text{Huff}}(x) \leq \mathbb{E} l_{\text{Shann}}(x) < H(X) + 1$$

### Issues with symbol codes:

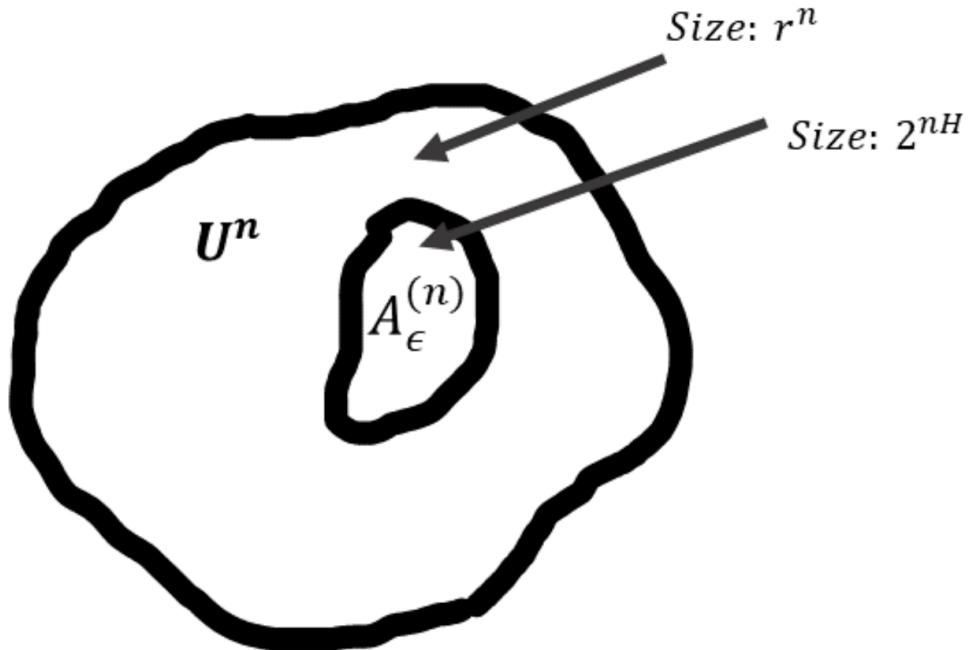
1.  $P = \{A: 0.1, B: 0.9\}$ ,  $H(P) = 0.47$

Huffman code can only compress this to 1 bit.

2. For any symbol  $s$ , ideally we want to use  $l(s) = \log_2 \frac{1}{P(s)}$  bits. But, as we are using a symbol code, we can't use fractional bits.

Thus, there is always going to be ~1 bit overhead per symbol with symbol codes.

## RECAP - AEP



$U^n$  = set of all possible n-tuples

$A_\epsilon^{(n)}$  = set of all typical n-tuples

$$P\left(U^n \in A_\epsilon^{(n)}\right) \approx 1$$

$$\forall u^n \in A_\epsilon^{(n)} : P(u^n) \approx 2^{-nH} *$$

\* where  $u^n$  represents a particular n – tuple

# RECAP

## Solution -> use block codes

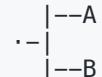
We can do better by considering blocks of size 2:  $P = \{\text{AA: } 0.01, \text{ AB: } 0.09, \text{ BA: } 0.09, \text{ BB: } 0.81\}$ , this way the overhead is ~1 bit per 2 symbol!

(or in case of blocks of size B, the overhead is ~1 bit per symbol)

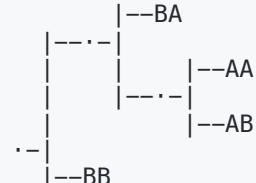
```
## Huffman code for blocks
block_size: 1, entropy: 0.47, avg_codelen: 1.00 bits/symbol
block_size: 2, entropy: 0.47, avg_codelen: 0.65 bits/symbol
block_size: 3, entropy: 0.47, avg_codelen: 0.53 bits/symbol
block_size: 4, entropy: 0.47, avg_codelen: 0.49 bits/symbol
block_size: 5, entropy: 0.47, avg_codelen: 0.48 bits/symbol
```

# Huffman codes on blocks

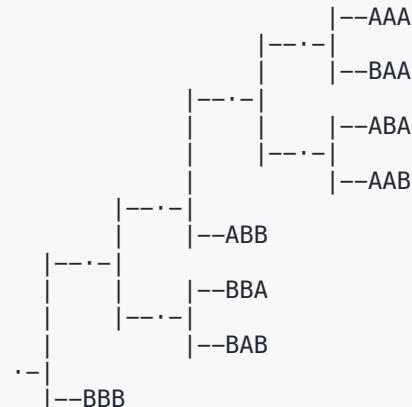
```
block_size: 1, entropy: 0.47, avg_codelen: 1.00 bits/symbol
```



```
block_size: 2, entropy: 0.47, avg_codelen: 0.65 bits/symbol
```

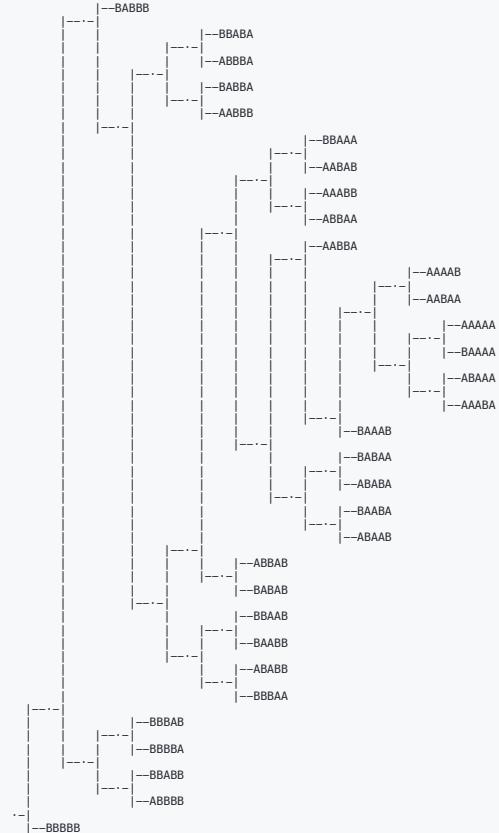


```
block_size: 3, entropy: 0.47, avg_codelen: 0.53 bits/symbol
```



# Huffman codes on blocks

block\_size: 5, entropy: 0.47, avg\_codelen: 0.48 bits/symbol



# Huffman codes on blocks

1. Huffman codes

$$H(X) \leq \mathbb{E}[l(X)] \leq H(X) + 1$$

2. Huffman codes on blocks of size B

$$H(X) \leq \frac{\mathbb{E}[l(X_1^B)]}{B} \leq H(X) + \frac{1}{B}$$

if  $H(X)$  is small  
 $\frac{1}{B} \gg H(X)$   
 $H(X) = 0.01$   
 $B \approx 100$   
for  $2 \times$  entropy

## Huffman codes on blocks

1. Convergence to entropy  $H(X)$  is quite fast ->  $1/B$
2. But, not very practical, as the codebook size needed is large (exponential):

$$\text{size} = |\mathcal{X}|^B$$

3. Larger codebook -> difficult to handle,

block size limited by device memory,

higher latency, ... → not streaming

boundary conditions (data is not multiple of  $B$ )

# Arithmetic coding

1. For data  $x_1^n$ , the block\_size = n

i.e. the entire data is a single block!

2. Codeword is computed on *on the fly*

No need to pre-compute the codebook beforehand

3. Very Efficient! -> *theoretically* the performance can be shown to be:

$$H(X) \leq \frac{\mathbb{E}[l(X_1^n)]}{n} \leq H(X) + \frac{2}{n}$$

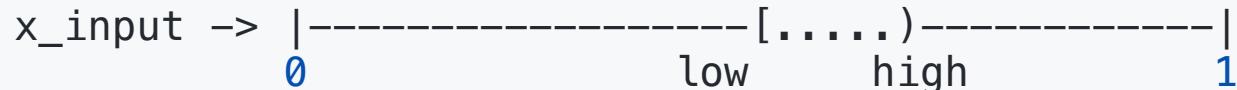
i.e. ~2 bits of overhead for the *entire sequence*

| Block coding         | Arithmetic coding    |
|----------------------|----------------------|
| $H(x) + \frac{2}{n}$ | $H(x) + \frac{2}{n}$ |
| $O(2^n)$             | $O(n)$               |

# How does Arithmetic coding work?

# Arithmetic Encoding

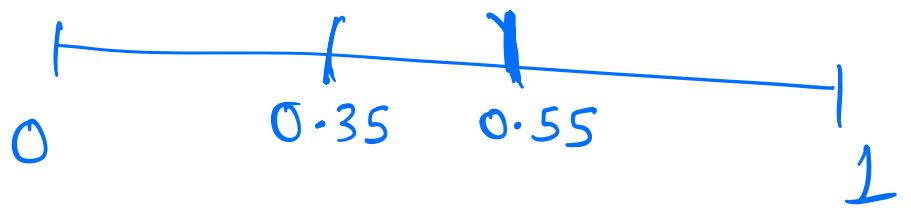
1. **STEP I:** Find an *interval* (or a *range*)  $[L, H)$ , corresponding to the *entire sequence*  $x_1^n$



2. **STEP II:** Communicate the *interval*  $[L, H)$  *efficiently*

(i.e. using less number of bits)

$x_{\text{input}} \rightarrow [L, H) \rightarrow 011010$



0.355555...

0.41786...

0.5  
0.4

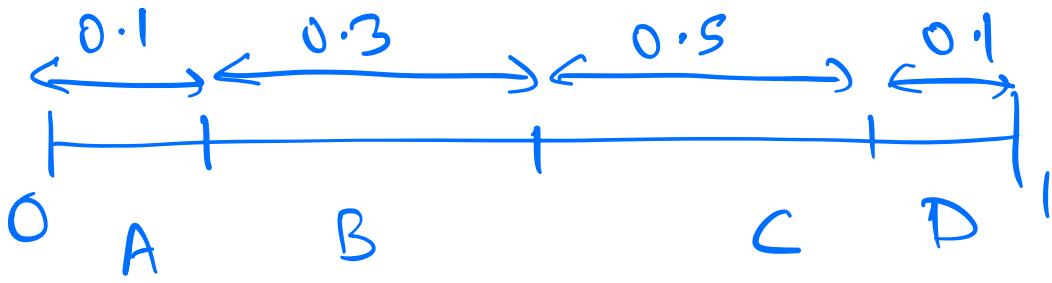
→ easy to describe

(0.3567, 0.3568)

0.35675

① Shorter intervals

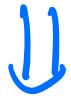
↓  
take more bits  
to describe



$$P = \{A: 0.1, B: 0.3, C: 0.5, D: 0.1\}$$

② → Map sequence to interval  
with length =  $P(\text{sequence})$

① + ②  $\Rightarrow$  More probable sequence



Bigger interval



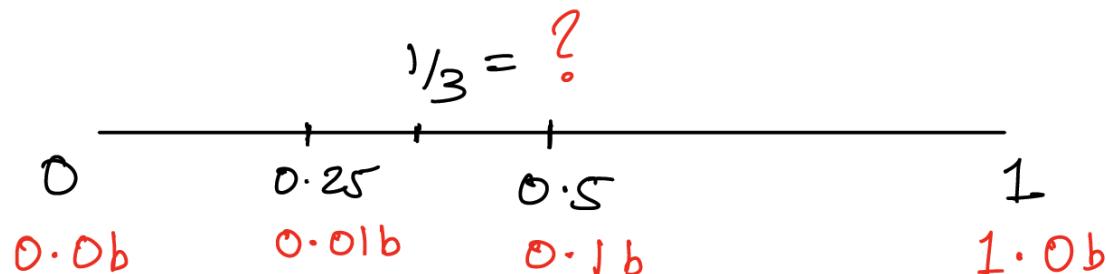
Shorter bit length

$$0.1b = \frac{1}{2}$$

$$0.01b = \frac{1}{2^2}$$

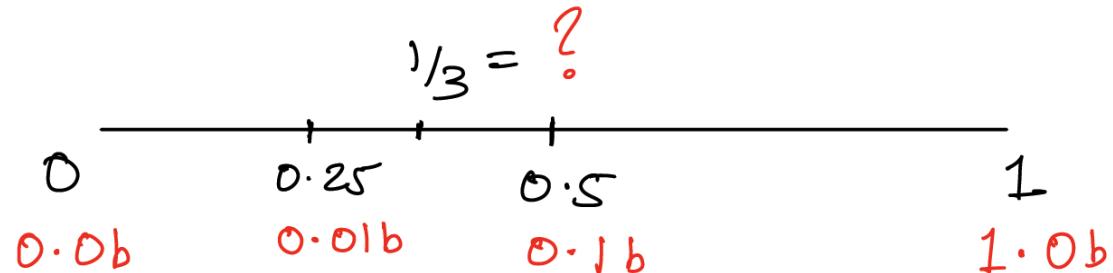
⋮

$$\begin{aligned}0.11b &= \frac{1}{2} + \frac{1}{2^2} \\&= \frac{3}{4}\end{aligned}$$



# floating point values in binary  
 $0.\underline{3333} = \# b0.... ? \rightarrow b0.0101\cdots$   
 $\underline{0.6666} = \# ?$

# Primer: the number line (in binary)



```
# floating point values in binary
from utils.bitarray_utils import float_to_bitarrays
_, binary_exp = float_to_bitarrays(0.3333333, 20)
```

$0.3333 = b0.010101\dots$

$0.6666 = b0.101010\dots$

# Arithmetic Encoding

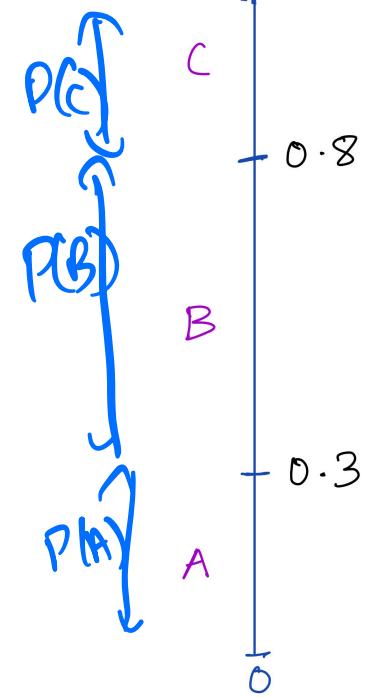
We will consider the following running example:

```
P = ProbabilityDist({A: 0.3, B: 0.5, C: 0.2})  
x_input = BACB
```

We want to encode the sequence  $x_1^n = BACB$  sampled from the distribution  $P$ .

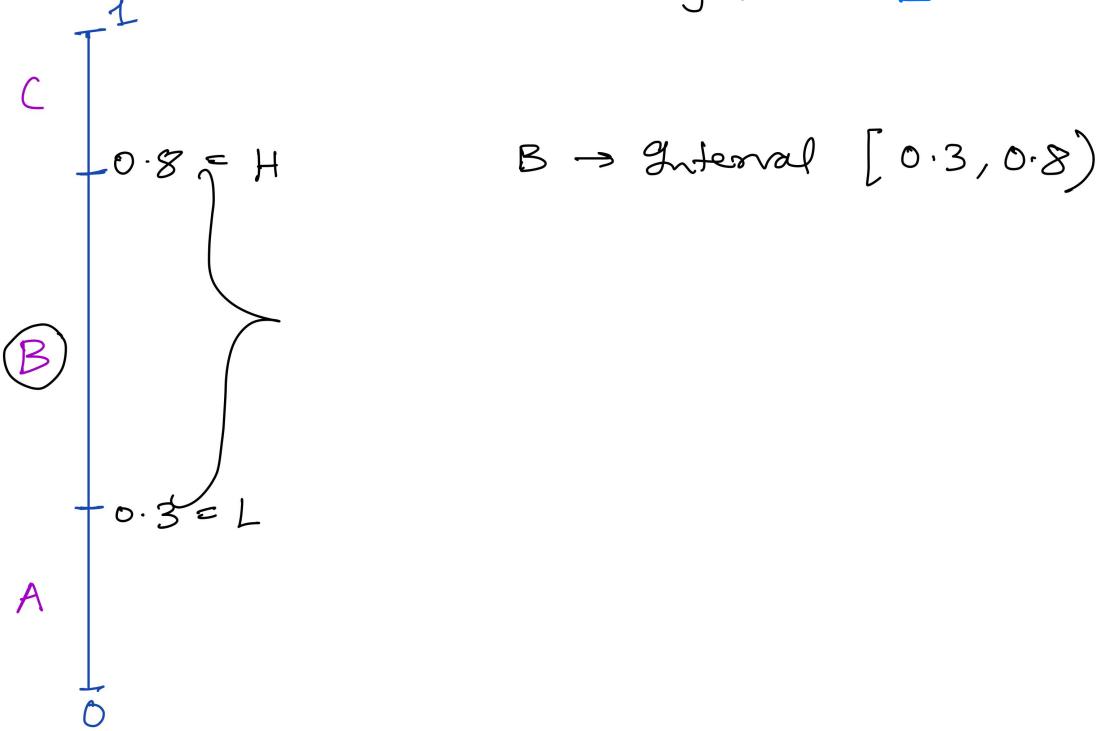
# Arithmetic coding example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, X_1^n = BACB$$



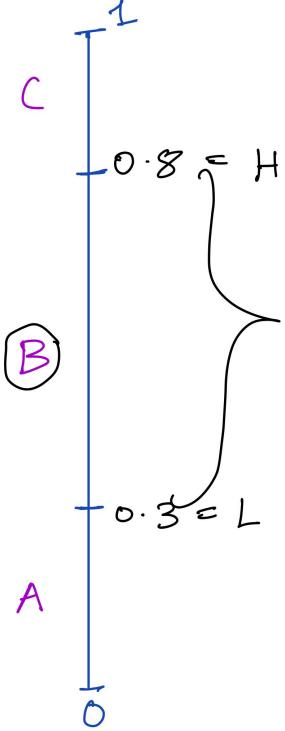
# Arithmetic coding example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, X_1^k = \underline{BACB}$$

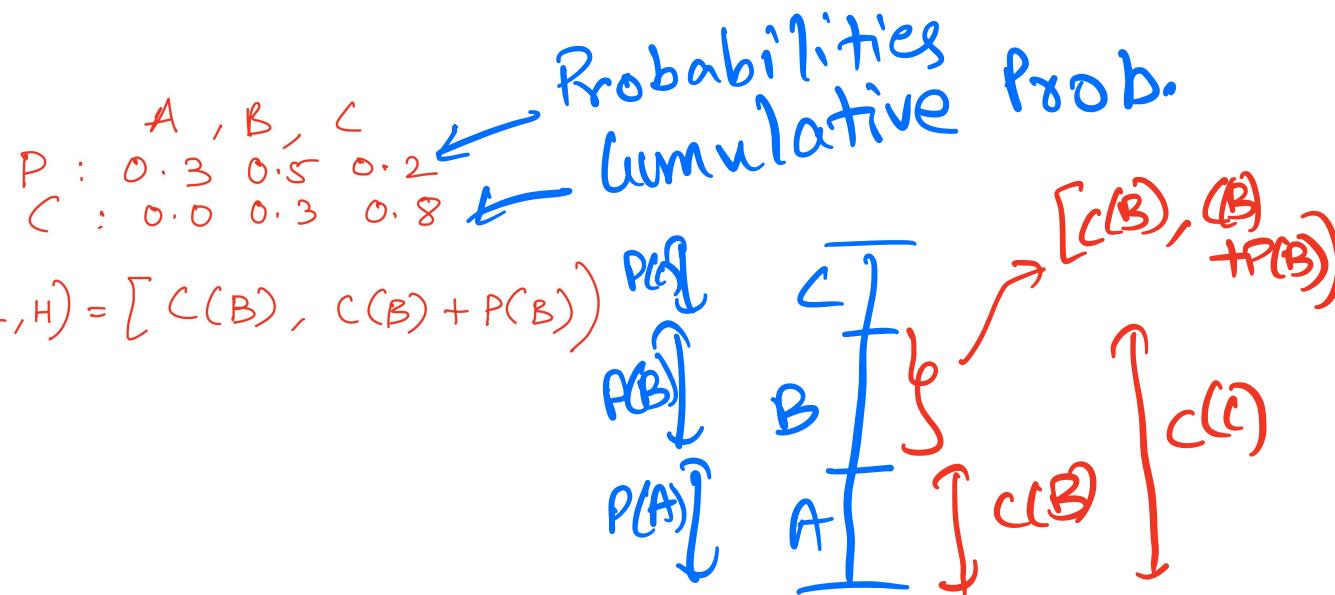


# Arithmetic coding example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, X = BACB$$

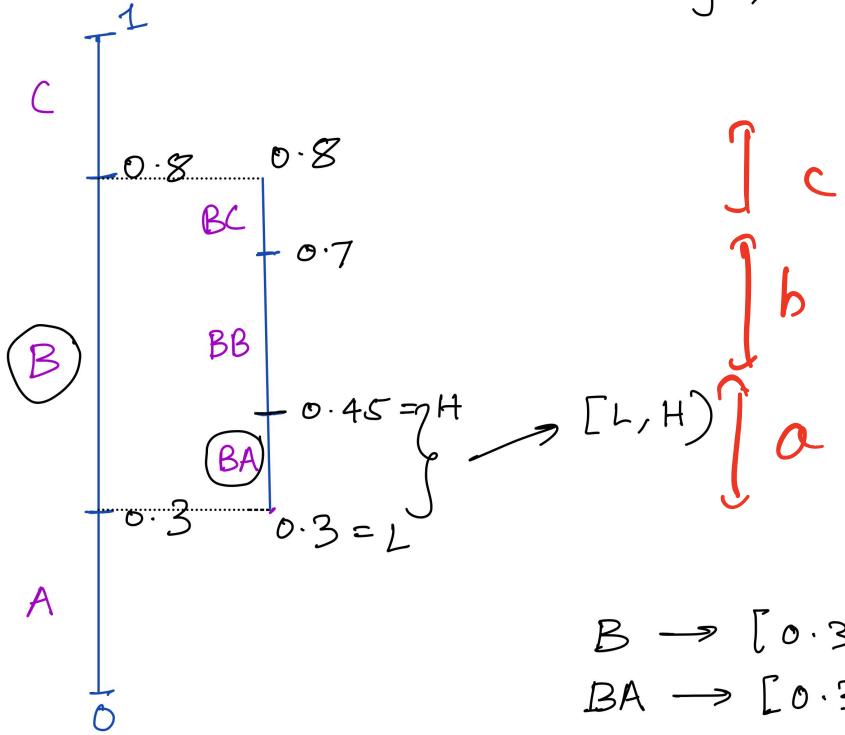


$B \rightarrow \text{Interval } [0.3, 0.8)$



# Arithmetic coding example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, X_1^4 = BACB$$



$$a: b: c = 0.3: 0.5: 0.2$$

$$B: [0.3, 0.8]$$

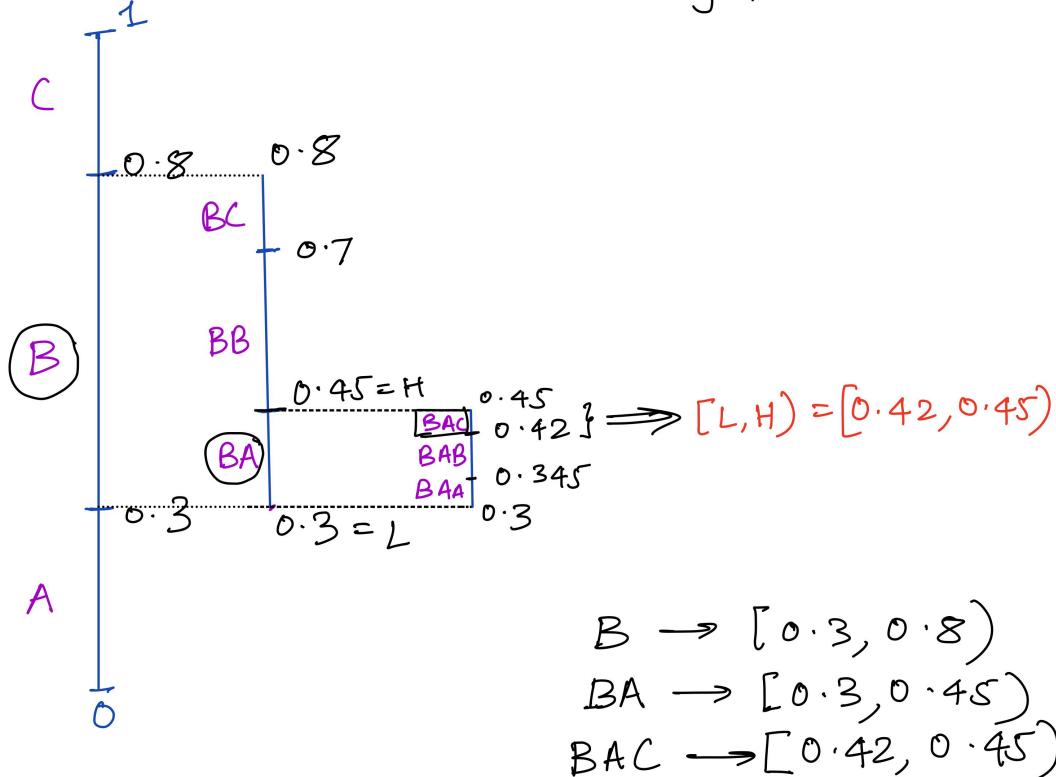
$$\text{length} = 0.5$$

$$\begin{aligned} BA &\rightarrow 0.5 \times 0.3 \\ &= 0.15 \end{aligned}$$

$$BA : [0.3, 0.3+0.15]$$

# Arithmetic coding example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, X = BACCB$$



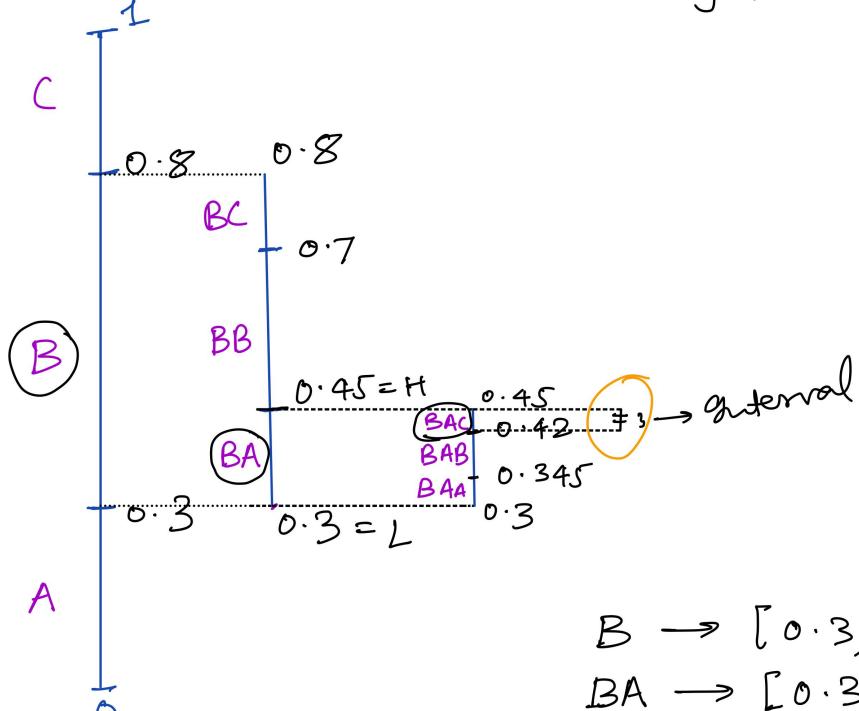
$$B \rightarrow [0.3, 0.8)$$

$$BA \rightarrow [0.3, 0.45)$$

$$BAC \rightarrow [0.42, 0.45)$$

# Arithmetic coding example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, X_1^k = BACCB$$



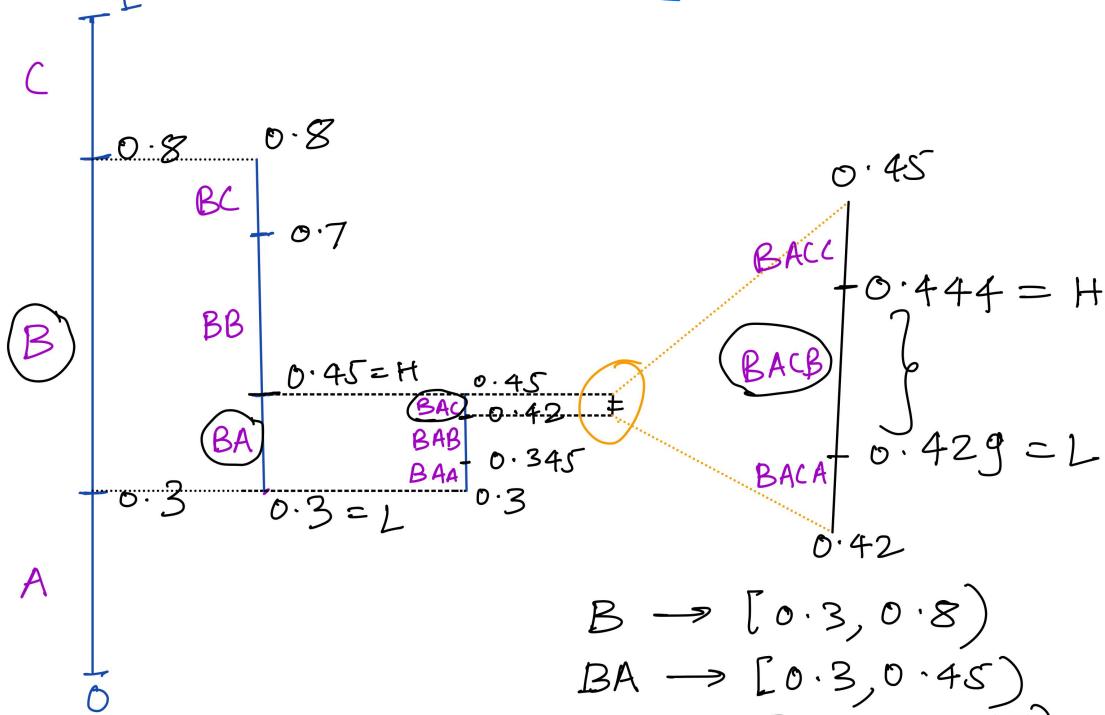
$$B \rightarrow [0.3, 0.8)$$

$$BA \rightarrow [0.3, 0.45)$$

$$BAC \rightarrow [0.42, 0.45)$$

# Arithmetic coding example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, X^* = BACB$$



$$B \rightarrow [0.3, 0.8)$$

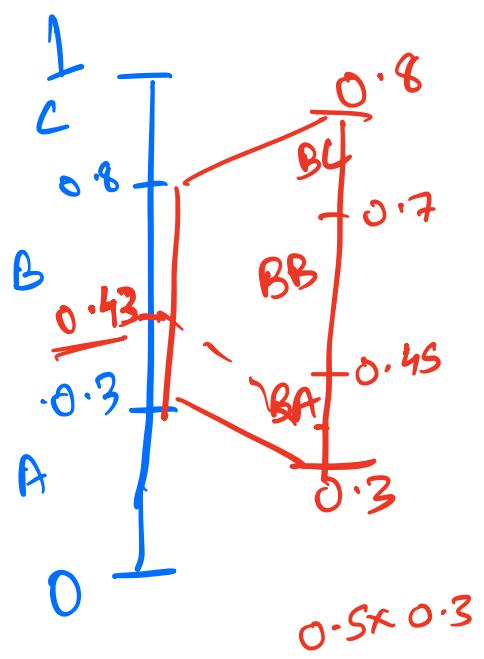
$$BA \rightarrow [0.3, 0.45)$$

$$BAC \rightarrow [0.42, 0.45)$$

$$BACB \rightarrow [0.429, 0.444)$$

0.43

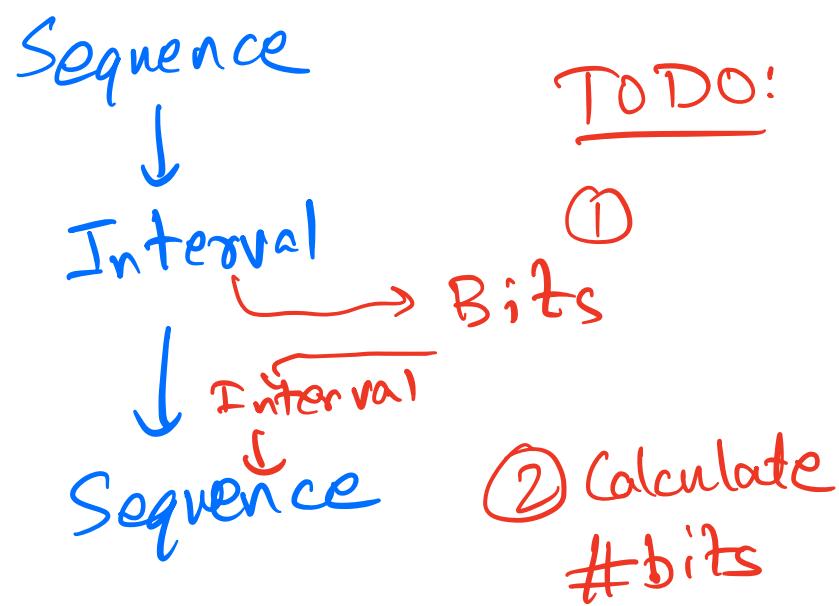
Decoding A: 0.3, B: 0.5, C: 0.2



B A

When to stop decoding?

- Prepend the length "n" at the start
- Create a special symbol called EOF (end of file), assign a probability, encode {sequence <EOF>}



# Arithmetic coding example

1. STEP I: Find an *interval* (or a *range*)  $[L, H)$ , corresponding to the *entire sequence*  $x_1^n$

```
prob = ProbabilityDist({A: 0.3, B: 0.5, C: 0.2})
x_input = BACB

# find interval corresp to BACB
ENCODE: B -> [L,H) = [0.30000,0.80000) → P(B)=0.5
ENCODE: A -> [L,H) = [0.30000,0.45000) → P(B)P(A)=0.5×0.3
ENCODE: C -> [L,H) = [0.42000,0.45000) → 0.5×0.3×P(C)
ENCODE: B -> [L,H) = [0.42900,0.44400) →
```

Thus, the final interval is:  $x_{\text{input}} \rightarrow [0.429, 0.444)$

# Arithmetic coding pseudo-code

```
class ArithmeticEncoder:  
    ...  
  
    def shrink_range(self, L, H, s):  
        rng = H - L  
        new_L = L + (rng * self.P.cumul[s])  
        new_H = new_L + (rng * self.P.probs(s))  
        return new_L, new_H  
  
    def find_interval(self, x_input):  
        L,H = 0.0, 1.0  
        for s in x_input:  
            L,H = self.shrink_range(L,H,s)  
        return L,H  
  
    def encode_block(self, x_input):  
        # STEP1  
        L,H = self.find_interval(x_input)  
  
        # STEP-II  
        ...
```

prob. of current symbol  
previous range  
\*  
next range

## Arithmetic coding example-2

```
P = {A: 0.2, B: 0.4, C: 0.4}
```

```
x_input = BAAB
```

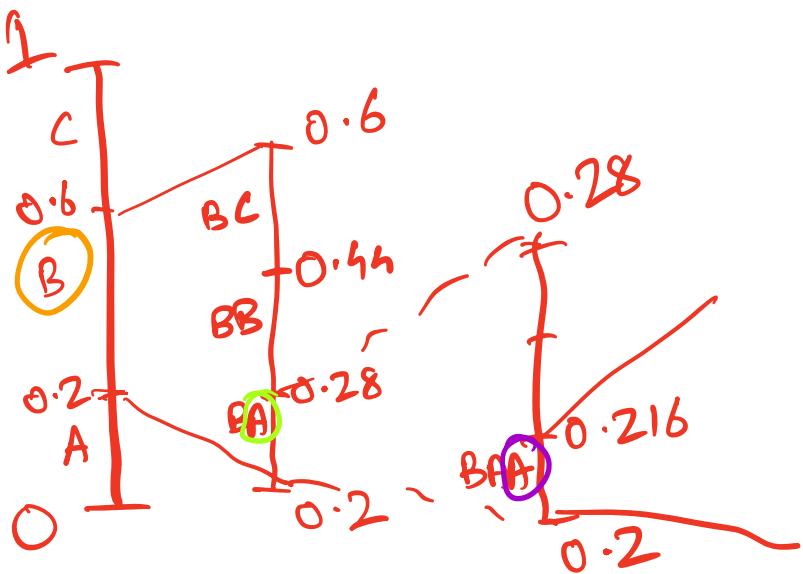
$$P = \{A: 0.2, B=0.4, C=0.4\}$$

seq: (BAAB)

$$\begin{aligned} BA - \text{length} \\ = 0.4 \times 0.2 \\ = 0.08 \end{aligned}$$

$$\begin{aligned} BB = 0.4 \times 0.4 \\ = 0.16 \\ BC = 0.16 \end{aligned}$$

$$\begin{aligned} BAA \rightarrow \text{length} \\ = \text{length}(BA) \times P(A) \\ = 0.08 \times 0.2 \\ = 0.016 \end{aligned}$$



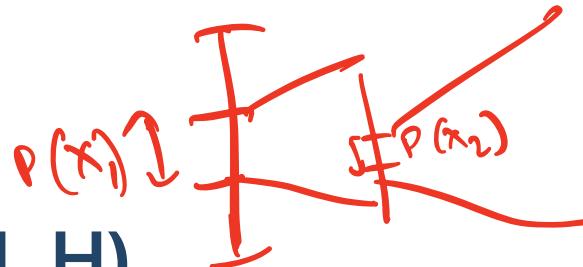
## Arithmetic coding example:2

```
P = {A: 0.4, B: 0.4, C: 0.2}  
x_input = BACA
```

```
ENCODE: B -> [L,H) = [0.40000,0.80000)  
ENCODE: A -> [L,H) = [0.40000,0.56000)  
ENCODE: C -> [L,H) = [0.52800,0.56000)  
ENCODE: A -> [L,H) = [0.52800,0.54080)
```

Probability distribution changed out of nowhere





## STEP-I: Find the interval $[L, H]$

```
P = {A: 0.3, B: 0.5, C: 0.2}
x_input = BACB
L = [0.429, 0.444)
```

1. **Observation:** Interval size reduces as we encode more symbols
2. **QUIZ-1:** What is the size of the interval ( $H-L$ ) for the input  $X_1^n$ ?

$$P(X_1^n) = P(x_1)P(x_2)\dots P(x_n)$$

# STEP-I: Find the interval [L,H)

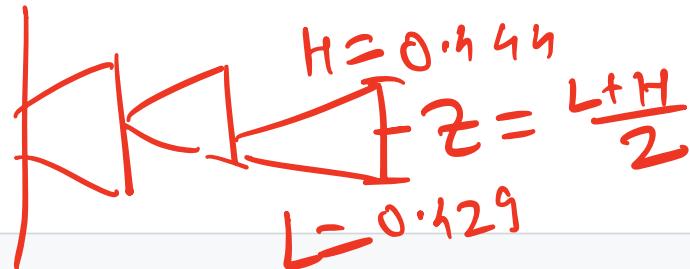
```
P = {A: 0.3, B: 0.5, C: 0.2}  
x_input = BACB  
L = [0.429, 0.444)
```

1. **Observation:** Interval size reduces as we encode more symbols
2. **QUIZ-1:** What is the size of the interval ( H-L ) for the input  $X_1^n$ ?

$$\begin{aligned}(H - L) &= p(x_1) * p(x_2) \dots p(x_n) \\ &= \prod_{i=1}^n p(x_i) \\ &= p(x_1^n)\end{aligned}$$

# Arithmetic Encoding

```
P = {A: 0.3, B: 0.5, C: 0.2}  
x_input = BACB  
L = [0.429, 0.444)
```



1. **STEP-I:** Find an *interval* (or a *range*)  $[L, H]$

corresponding to the *entire sequence*  $x_1^n$

2. **STEP-II:** Communicate the interval  $[L, H]$  using a value  $Z \in [L, H]$

For example:  $Z = \frac{(L+H)}{2}$ , i.e. the midpoint of the range.  
(in our example  $Z = 0.4365$  )

# Arithmetic decoding

**Quiz-2:** If the decoder knows:

- $n=4$
- $P = \{A: 0.3, B: 0.5, C: 0.2\}$
- $Z = 0.4365$

How can it decode the entire input sequence?  $X_1^n$ .

# Arithmetic decoding - example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, Z = 0.4365$$

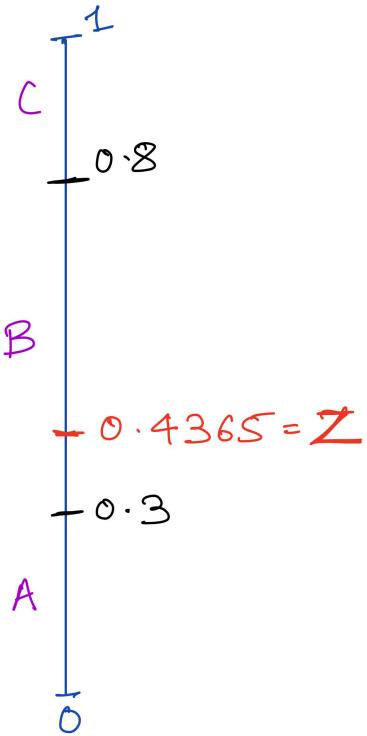
1

$$0.4365 = Z$$

0

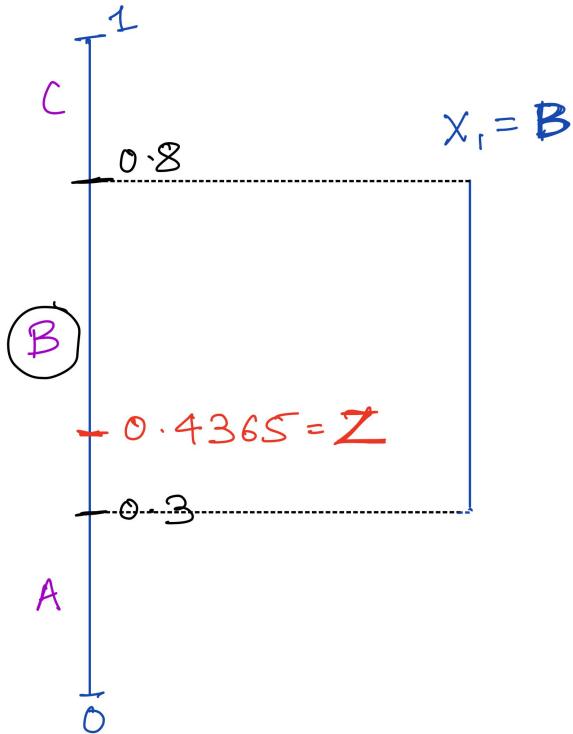
# Arithmetic decoding - example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, Z = 0.4365$$



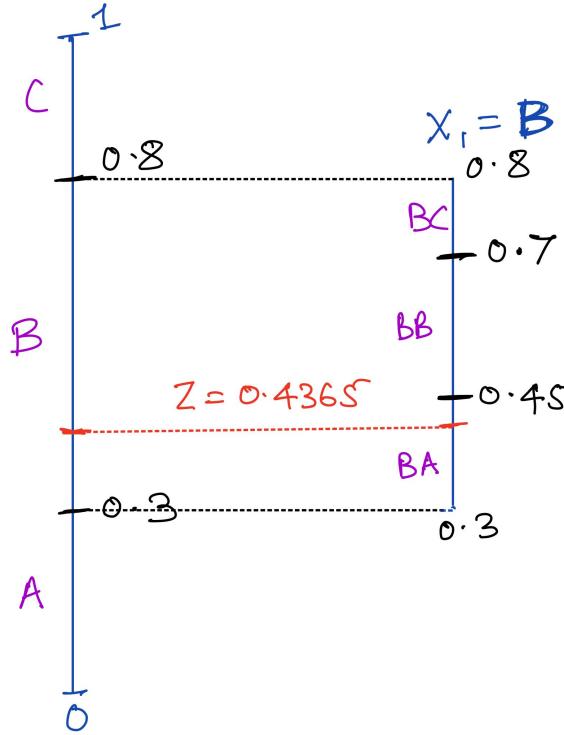
# Arithmetic decoding - example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, Z_{n=4} = 0.4365$$



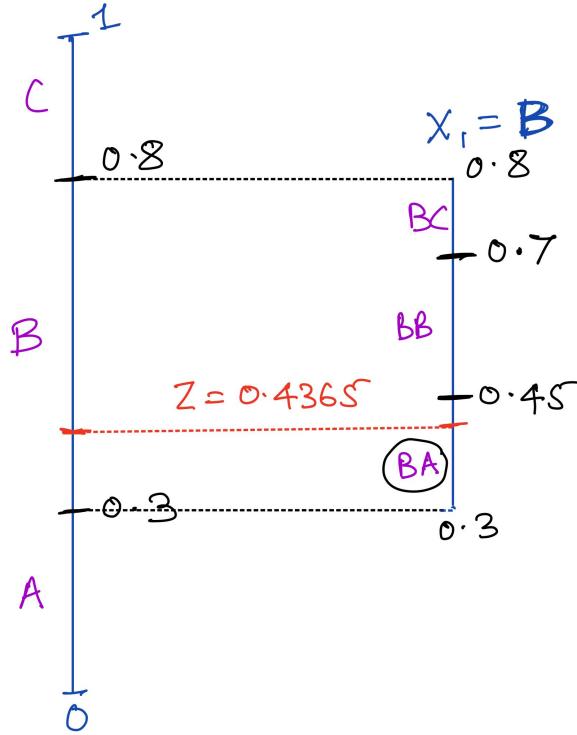
# Arithmetic decoding - example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, \sum_{n=4} = 0.4365$$



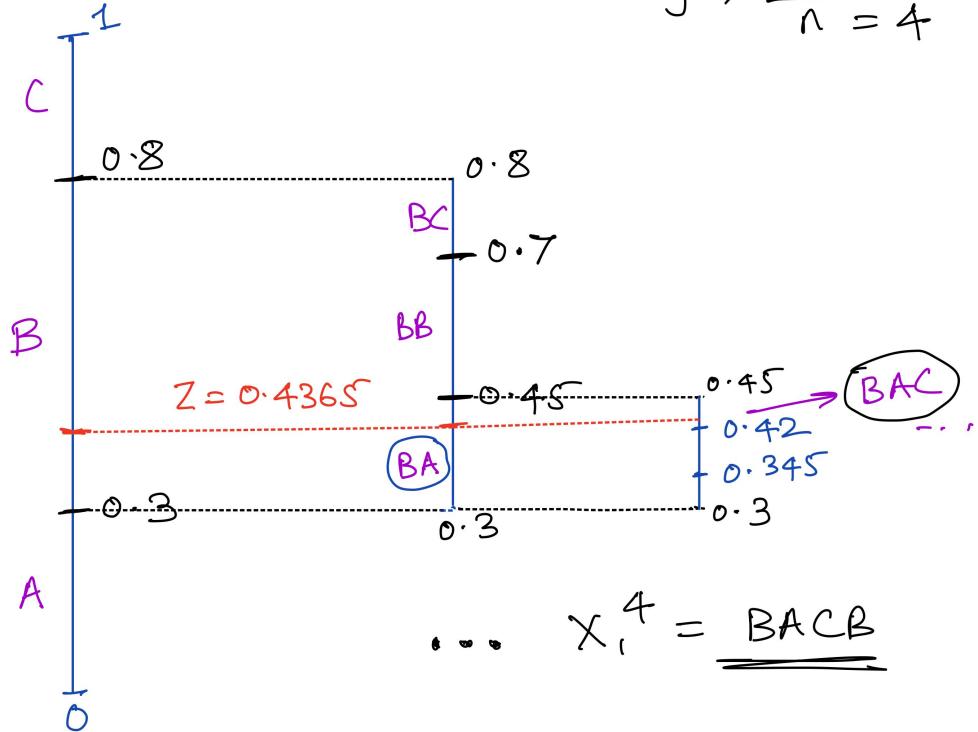
# Arithmetic decoding - example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, \sum_{n=4} = 0.4365$$



# Arithmetic decoding - example

$$P = \{A: 0.3, B: 0.5, C: 0.2\}, \sum_{n=1}^N = 0.4365$$



# Arithmetic decoding

Z = **0.4365**

ENCODE: B → [L,H) = **[0.30000,0.80000)**

ENCODE: A → [L,H) = **[0.30000,0.45000)**

ENCODE: C → [L,H) = **[0.42000,0.45000)**

ENCODE: B → [L,H) = **[0.42900,0.44400)**

---

DECODE: B → [L,H) = **[0.30000,0.80000)**

DECODE: A → [L,H) = **[0.30000,0.45000)**

DECODE: C → [L,H) = **[0.42000,0.45000)**

DECODE: B → [L,H) = **[0.42900,0.44400)**

# Arithmetic decoding-pseudocode

```
class ArithmeticDecoder:  
    ...  
    def shrink_range(self, L, H, s):  
        ...  
        return new_L, new_H  
  
    def decode_symbol(self, L, H, Z):  
        rng = H - L  
        search_list = L + (self.P.cumul * rng)  
        symbol_ind = np.searchsorted(search_list, Z)  
        return self.P.alphabet[symbol_ind]  
  
    def decode_block(self, Z, n):  
        L,H = 0.0, 1.0  
        for _ in range(n): #main decoding loop  
            s = self.decode_symbol(L, H, Z)  
            L,H = self.shrink_range(L,H,s)
```

## Arithmetic decoding:

**Quiz-2:** If the decoder knows:

- $n=4$
- $P = \{A: 0.3, B: 0.5, C: 0.2\}$
- $Z = 0.4365$

**Ans ->** The decoder creates intervals same as the ones encoder creates, and find which symbol corresponds to the interval in which  $Z$  lies.

$$\frac{1}{3} = 0.3333\ldots$$

## Arithmetic encoding

1. **STEP-I:** Find an *interval* (or a *range*)  $[L, H)$   
corresponding to the *entire sequence*  $x_1^n$  (  $[0.429, 0.444]$  )
2. **STEP-II:** Find the midpoint of the interval  $[L, H)$ ,  $Z = \frac{(L+H)}{2}$  (  $Z = 0.4365$  )
3. **STEP-III:** Write the binary expansion of  $Z$  to the bitstream ->  
eg:  $Z = 0.4365 = b0.0110111101\ldots$   
then the final **encoded\_bitstream = 0110111101...**

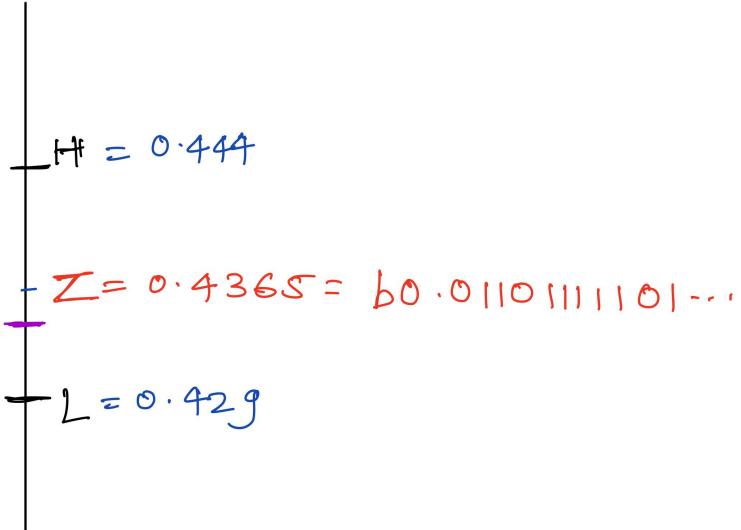
# Arithmetic encoding

1. **STEP-I:** Find an *interval* (or a *range*)  $[L, H]$   
corresponding to the *entire sequence*  $x_1^n$  ( $[0.429, 0.444]$  )
2. **STEP-II:** Find the midpoint of the interval  $[L, H]$ ,  $Z = \frac{(L+H)}{2}$ . ( $Z = 0.4365$  )
3. **STEP-III:** Write the binary expansion of  $Z$  to the bitstream ->  
eg:  $Z = 0.4365 = b0.0110111101\dots$   
then the final **encoded\_bitstream = 0110111101...**

**Quiz-4:**  $Z$ 's binary representation can be long, can also have infinite bits.  
How can we fix this?

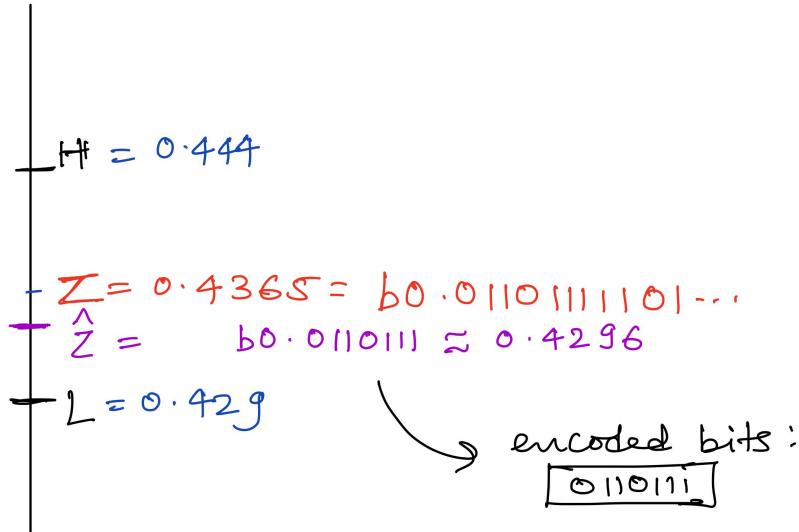
# Communicating the interval $[L, H)$

Communicating  $Z$



# Communicating the interval $[L, H)$

Communicating  $Z$



We just need a  $\hat{Z}$  such the  $\hat{Z} \in [L, H)$   
 $\hat{Z}$  should have a short binary representation.

# Arithmetic coding example:

1. **STEP-I:** Find an *interval* (or a *range*)  $[L, H]$

corresponding to the *entire sequence*  $x_1^n$  (  $[0.429, 0.444]$  )

2. **STEP-II:** Find the midpoint of the interval  $[L, H]$ ,  $Z = \frac{(L+H)}{2}$ . (  $Z = 0.4365$  )

3. **STEP-III:** Truncate  $Z$  to  $k$  bits ( $\hat{Z}$ )

e.g:

$$L, H = 0.429, 0.444$$

$$Z = 0.4365 = b0.0110111101\dots$$

$$Z_{\text{hat}} = b0.01101111 \sim 0.4296$$

Final Encoding = **encoded\_bitstream = 01101111**

## Communicating the interval $[L, H)$

1. Cond 1: Truncate  $Z$  to  $\hat{Z}$  with  $k$  bits, so that  $\hat{Z} \in [L, H)$
2. Cond 2: If  $\hat{Z}$  has binary representation:  $Z_{\text{hat}} = b0.011011111$  for example, then we also need, any extension of it  $Z_{\text{ext}} \in [L, H)$ .

For eg:

$$\hat{Z} \in [L, H)$$

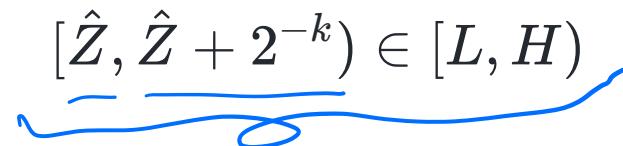
$$Z_{\text{hat}} = b0.\underline{011011111}$$
$$Z_{\text{ext}} = b0.\underline{01101111111011110101\dots}$$

Quiz-5: Why so?

## Communicating the interval $[L, H)$

1. **Cond 1:** Truncate  $Z$  to  $\hat{Z}$  with  $k$  bits, so that  $\hat{Z} \in [L, H)$
2. **Cond 2:** If  $\hat{Z}$  has binary representation:  $z_{\text{hat}} = b_0.011011111$  for example, then we also need, any extension of it  $Z_{ext} \in [L, H)$ .

The two conditions can be written together as:

$$[\hat{Z}, \hat{Z} + 2^{-k}) \in [L, H)$$


## Communicating the interval $[L, H)$

Given the interval  $[L, H)$ , and  $Z = \frac{(L+H)}{2}$ , truncate  $Z$  to  $k$  bits so that:

$$[\hat{Z}, \hat{Z} + 2^{-k}) \in [L, H)$$

**Quiz-6:** What should the  $k$  be?

```
## Examples
# Ex1: L=0.429, H=0.444, Z = 0.4365.. how many digits we can truncate from Z?

## Ex2: L=0.552398714, H=0.5524123
Z = 0.5524058..., how many digits we can truncate Z from?
```

## Communicating the interval $[L, H]$

Given the interval  $[L, H]$ , and  $Z = \frac{(L+H)}{2}$ , truncate  $Z$  to  $k$  bits so that:

$$[\hat{Z}, \hat{Z} + 2^{-k}) \in [L, H)$$

**Quiz-6:** What should the  $k$  be?

- Shorter the interval,  $|H - L|$ , the more the number of bits we need to use.

## Communicating the interval $[L, H)$

Given the interval  $[L, H)$ , and  $Z = \frac{(L+H)}{2}$ , truncate  $Z$  to  $k$  bits so that:

$$[\hat{Z}, \hat{Z} + 2^{-k}) \in [L, H)$$

### Quiz-6: What should the $k$ be?

- Shorter the interval,  $|H - L|$ , the more the number of bits we need to use.
- the numbers of bits we need to truncate  $Z$  by is:

$$k \leq \left\lceil \log_2 \frac{1}{(H - L)} \right\rceil + 1$$

$$[0.35, 0.45] \rightarrow L = 0.1$$
$$\underline{\underline{0.4}} \in [0.35, 0.45]$$

$$[0.355, 0.365] \rightarrow L = 0.01$$
$$0.36 \in [0.355, 0.365]$$

$$[0.3555, 0.3565] \rightarrow \text{len} = 0.001$$
$$0.356 \in [0.3 \dots]$$

# bits needed  $\approx \log_2 \frac{1}{|H-L|}$

# Arithmetic Encoding pseudo-code

```
class ArithmeticEncoder:  
    def shrink_range(self, L, H, s):  
        ...  
    def find_interval(self, x_input):  
        L,H = 0.0, 1.0  
        for s in x_input:  
            L,H = self.shrink_range(L,H,s)  
        return L,H  
  
    def encode_block(self, x_input):  
        # STEP-1 find interval  
        L,H = self.find_interval(x_input)  
  
        # STEP-II, III communicate interval  
        Z = (L+H)/2  
        num_bits = ceil(log2((H-L))) + 1  
        _, code = float_to_bitarray(Z, num_bits)  
        return code
```

# Arithmetic decoding-pseudocode

```
class ArithmeticDecoder:  
    ...  
    def shrink_range(self, L, H, s):  
        ...  
  
    def decode_symbol(self, L, H, Z):  
        ...  
  
    def decode_block(self, code, n):  
        Z = bitarray_to_float(code)  
  
        # start decoding  
        L,H = 0.0, 1.0  
        for _ in range(n): #main decoding loop  
            s = self.decode_symbol(L, H, Z)  
            L,H = self.shrink_range(L,H,s)  
  
            # add code to remove additional bits read
```

## Arithmetic coding compression performance:

- Size of interval  $H - L = \log_2 1p(x_1^n)$
- $k \leq \log_2 \frac{1}{H-L} + 2$

**Quiz-7:** What is the codelength for arithmetic coding?

## Arithmetic coding compression performance:

- Size of interval  $H - L = \log_2 1p(x_1^n)$
- $k \leq \log_2 \frac{1}{H-L} + 2$

Quiz-7: What is the codelength for arithmetic coding?

$$codelen = k \leq \log_2 \frac{1}{p(x_1^n)} + 2$$

## Arithmetic coding compression performance:

- Size of interval  $H - L = \log_2 1/p(x_1^n)$
- $k \leq \log_2 \frac{1}{H-L} + 2$

Quiz-7: What is the codelength for arithmetic coding?

$$codelen = k \leq \log_2 \frac{1}{p(x_1^n)} + 2$$

Thus, Arithmetic coding is within 2 bits of the optimal on the ENTIRE sequence!

$$\begin{aligned} \text{IE}(codelen) &\leq \text{IE} \log_2 \frac{1}{p(x_1^n)} + 2 \\ &= n \text{IE} \log_2 \frac{1}{p(x_1^n)} + 2 \\ &= n H(x) + 2 \end{aligned}$$

## Arithmetic coding compression performance:

**THEOREM:** Arithmetic coding achieves average codelength:

$$H(X) \leq \frac{\mathbb{E}[l(X_1^n)]}{n} \leq H(X) + \frac{2}{n}$$

# Arithmetic coding Summary

1. Given *any* distribution  $P$ , achieves *optimal* compression. Thus, Arithmetic coding allows for model and entropy coding separation.
2. Encoding, decoding is linear time and quite efficient!
3. As we are not saving a large codebook, memory requirements are not very high
4. Can work very well with changing distribution  $P$ .  
i.e. Adaptive algorithms work well with Arithmetic coding

lecture 8

Don't even  
need to compute  $P(x^n)$

# Arithmetic coding in practice

Quiz-8: What are the practical issues with our Arithmetic encoding/decoding?

Hint ->

```
prob = ProbabilityDist({A: 0.3, B: 0.5, C: 0.2})  
x_input = BACBBCCBA  
  
# find interval corresp to BACB  
ENCODE: B -> [L,H) = [0.30000,0.80000)  
ENCODE: A -> [L,H) = [0.30000,0.45000)  
ENCODE: C -> [L,H) = [0.42000,0.45000)  
ENCODE: B -> [L,H) = [0.42900,0.44400)  
ENCODE: C -> [L,H) = [0.44100,0.44400)  
ENCODE: C -> [L,H) = [0.44340,0.44400)  
ENCODE: B -> [L,H) = [0.44358,0.44388)  
ENCODE: A -> [L,H) = [0.44358,0.44367)
```

# Arithmetic coding in practice

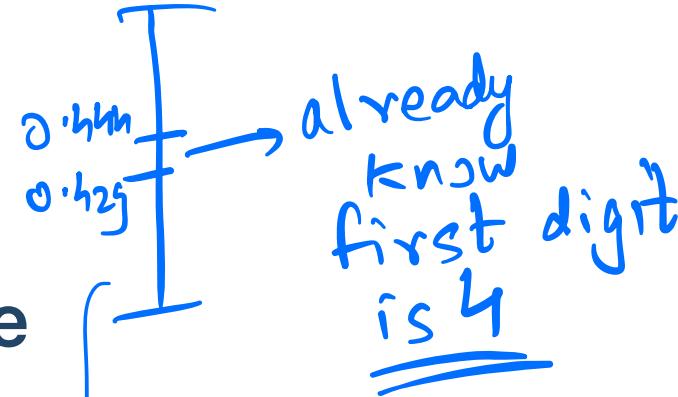
**Quiz-8:** What are the practical issues with our Arithmetic encoding/decoding?

Ans -> The interval becomes too small very quickly and we run out of bits to represent L, H .

```
prob = ProbabilityDist({A: 0.3, B: 0.5, C: 0.2})
x_input = BACBBCCBA

# find interval corresp to BACB
ENCODE: B -> [L,H) = [0.30000,0.80000)
ENCODE: A -> [L,H) = [0.30000,0.45000)
ENCODE: C -> [L,H) = [0.42000,0.45000)
ENCODE: B -> [L,H) = [0.42900,0.44400)
...
...
```

## Arithmetic coding in practice

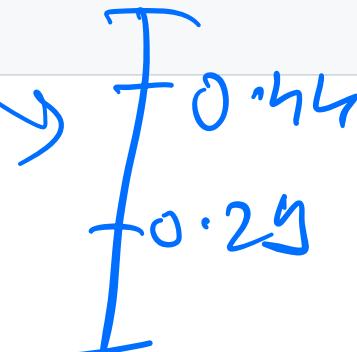


Quiz-9: What can we do to avoid the interval  $[L, H)$  from getting too small?

Hint ->

$$L = 0.429 = b0.0110110\dots$$

$$H = 0.444 = b0.01110001\dots$$



# Arithmetic coding in practice

Quiz-9: What can we do to avoid the interval  $[L, H)$  from getting too small?

Idea: If  $L, H$  start with  $011$  then any value lying inside the interval  $[L, H)$  also will start with  $011$  !

$$L = 0.429 = b0.0110110\dots$$

$$H = 0.444 = b0.01110001\dots$$

$$\hat{Z} = b0.011\dots$$

# Arithmetic coding in practice

**Quiz-9:** What can we do to avoid the interval  $[L, H)$  from getting too small?

**Idea:** If  $L, H$  start with  $011$  then any value lying inside the interval  $[L, H)$  also will start with  $011$ !

**Rescale::** Already output bits  $011$ , and rescale  $L, H$

$$L = 0.429 = b0.0110110\dots$$

$$H = 0.444 = b0.01110001\dots$$

Rescaled:  $L=0.8580, H=0.8880, \text{bitarray}='0'$

Rescaled:  $L=0.7160, H=0.7760, \text{bitarray}='01'$

Rescaled:  $L=0.4320, H=0.5520, \text{bitarray}='011'$

ENCODE:  $B \rightarrow [L, H) = [0.42900, 0.44400)$

# Arithmetic Encoding with rescaling

```
class ArithmeticEncoder:  
    def shrink_range(self, L, H, s):  
        ...  
    def rescale_range(self, L, H):  
        ...  
    def find_interval(self, x_input):  
        L,H, bitarray = 0.0, 1.0, Bitarray("")  
        for s in x_input:  
            L,H = self.shrink_range(L,H,s)  
            L,H, bits = self.rescale_range(L,H)  
            bitarray += bits  
        return L,H, bitarray
```

# Arithmetic Encoding with rescaling

```
def rescale_range(self, L, H):
    bitarray = ""
    while (L >= 0.5) or (H < 0.5):
        if (L < 0.5) and (H < 0.5):
            bitarray+= "0"
            L,H = L*2, H*2
        elif ((L >= 0.5) and (H >= 0.5)):
            bitarray += "1"
            L,H = (L - 0.5)*2, (H - 0.5)*2
    return L, H, bitarray
```

# Arithmetic Encoding with rescaling

Rescale:: Already output bits which are same between  $L, H$  ( 011 ), and rescale  $L, H$ .

$L = 0.429 = b0.0110110\dots$

$H = 0.444 = b0.01110001\dots$

Rescaled:  $L=0.8580=b0.11011\dots, H=0.8880$ , bitarray='0'

Rescaled:  $L=0.7160=b0.1011\dots, H=0.7760$ , bitarray='01'

Rescaled:  $L=0.4320=b0.011\dots, H=0.5520$ , bitarray='011'

ENCODE:  $B \rightarrow [L,H] = [0.42900, 0.44400)$

Quiz-10: There is one case in which our algorithm can still have  $L, H$  being really close.

What is that?

# Arithmetic Encoding with rescaling

Lots of Variants of Arithmetic coding; mainly come from how they implement the rescaling.

1. **Arithmetic coding:** Bit-based rescaling -> keeping a count of the mid-ranges etc.  
[SCL arithmetic coder](#)
2. **Range Coding** Byte (8-bit based rescaling), word-based rescaling  
[SCL range coder](#)
3. Variants on the above based on how compressors handle the edge case ( $L$  starts with `b0.0` and  $H$  starts with `b0.1..`, but the interval is very small)

# Arithmetic/Range coders in practice

Used almost everywhere! (either as Range coder or Arithmetic coding)

- 1. JPEG2000, BPG, H265, H266, VP8 → *second half*
- 2. CMIX, tensorflow-compress, NNCP etc. → *lec. 9*

# What are the problems with Arithmetic coding

Although Arithmetic coding algorithms are very fast, they are not fast enough!  
(especially when compared with Huffman coding)

| Codec             | Encode speed | Decode speed | Compression |
|-------------------|--------------|--------------|-------------|
| Huffman coding    | 252 MB/s     | 300 MB/s     | 1.66        |
| Arithmetic coding | 120 MB/s     | 69 MB/s      | 1.24        |

NOTE -> Speed numbers from: [Charles Bloom's blog](#)

# Beyond Arithmetic coding

| Codec             | Encode speed | Decode speed | Compression |
|-------------------|--------------|--------------|-------------|
| Huffman coding    | 252 MB/s     | 300 MB/s     | 1.66        |
| Arithmetic coding | 120 MB/s     | 69 MB/s      | 1.24        |
| rANS              | 76 MB/s      | 140 MB/s     | 1.24        |
| tANS              | 163 MB/s     | 284 MB/s     | 1.25        |

NOTE -> Speed numbers from: [Charles Bloom's blog](#)

Next Class -> ANS: Asymmetric Numeral Systems