Computer Science 112 Data Structures

Lecture 16: Hashing

Review: Huffman Encoding

Data Compression:

- In most data some symbols appear more often than others
 - Eg English text 'e' appears more often than 'q'
- We can use this fact to represent data in fewer bits total
 - More frequent symbols: shorter codes
 - Less frequent symbols: longer codes

Variable Length Takes Away Some Codes

Suppose codes were

$$1 = a, 11 = b$$

- Decode 111 as 'ab', as 'ba', or as 'aaa'?
- No character's code can be prefix of another

Variable Length Code

• Eg 4-symbols alphabet: {a, b, c, d} with frequencies:

a: 50%, b: 30%, c: 10%, d:10%

- Variable length code: 1, 2, or 3 bits / character
 - e.g: 0 = a, 10 = b, 110 = c, 111 = d
 - aabcbaa = 00101101000
 - 11 bits / 7 characters = 1.6 bits/character
- Decode: 0100110 = 'abac'

Variable Length Code

• Eg 4-symbols alphabet: {a, b, c, d} with frequencies:

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• Variable length code: 1, 2, or 3 bits / character

$$-$$
 e.g: $0 = a$, $10 = b$, $110 = c$, $111 = d$

Average bits per character:

$$0.5*1 + 0.3*2 + 0.1*3 + 0.1*3 = 1.7$$

Frequency of **a**

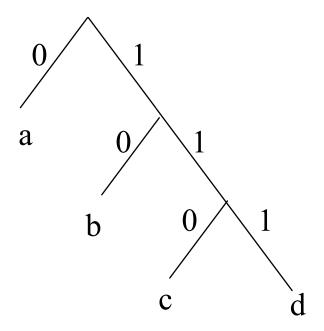
Bits per a

Frequency of **b**

Bits per **b**

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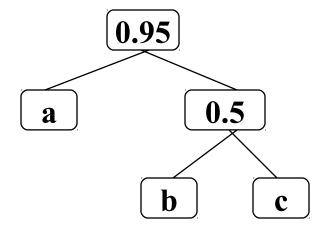
Huffman Code as a Tree



Symbols only at leaves

Algorithm to build tree

- 2 queues: S, T
- Contents of each queue: Tree
 - A leaf node stores (an index of) a symbol
 - A non-leaf mode stores total frequency of all symbols at leaves under this node
- E.g., for frequencies a: 0.45, b: 0.3, c: 0.2:



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Algorithm to build tree

- 2 queues:
 - S initially holds 1-node trees for all symbols, least likely first
 - T empty

while not (S empty and T length == 1)

find two least-weight trees in S, T and dequeue them

make a tree with these two as subtrees enqueue this tree on T

read out codes from final tree

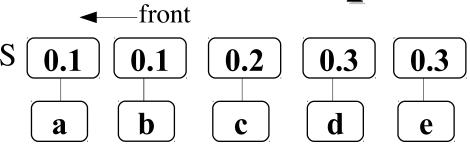
a .1

b .1

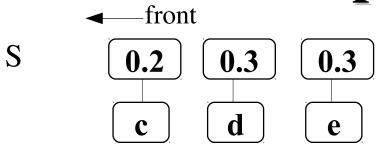
 \mathbf{c} .2

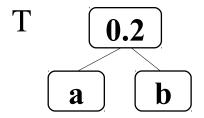
d .3

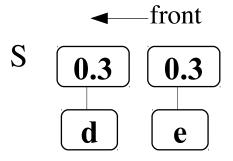
e .3

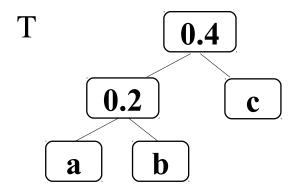


Τ



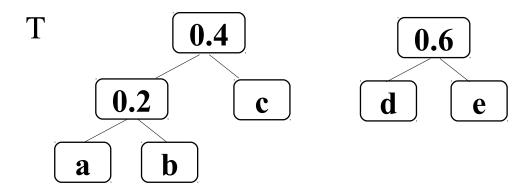






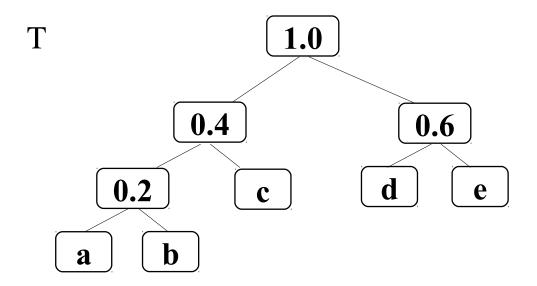
←front

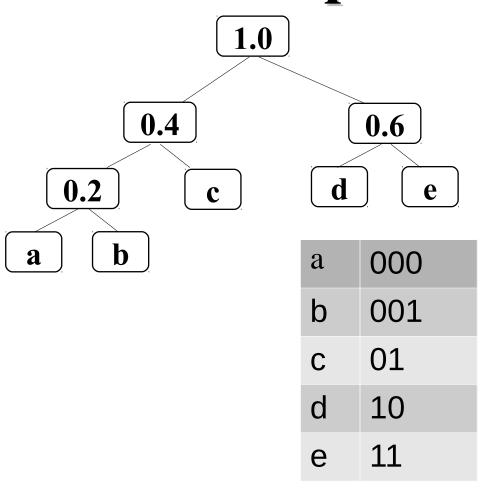
S



←front

S





Big-O to build tree

- Create and enqueue one-symbol trees
 - O(n) where n is number of symbols in symbol set
- While >1 tree in queues

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dequeue 2 trees: 2*O(1) = O(1)
```

merge into 1: create root, attach subtrees:

- enqueue 1 tree
- O(n) iterations, O(1) work each \rightarrow O(n) together
- grand total O(n)

Algorithm to encode

- For each symbol c in original string
 - find table entry for c
 - copy bits from table to encoded version of string
- let
 - n be size of symbol set
 - s be length of original version
 - b be length of encoded version
- O(s*n+b) or (s*log(n)+b)

Algorithm to decode

- Start at:
 - root of tree
 - beginning of encoded version
- For each bit in encoded version:
 - go left if bit is 0, right if 1
 when reach a leaf, output symbol at leaf and go back to root of tree
- O(b)

New: Hashing

- Suppose we want to store a set of numbers
 - add number to set, delete from set, test if in set should all be O(1)
- If range of numbers is small, e.g. 0 .. 9, we can use a boolean array, eg:

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0 1 2 3 4 5 6 7 8 9
t f f t t f f f
```

means $\{0, 4, 5, 7\}$

- What if range of numbers is much larger than set size? eg
 - range 0 ... 499,999
 - set size about 50
- If we use array of 500,000 elements, they will nearly all be false.

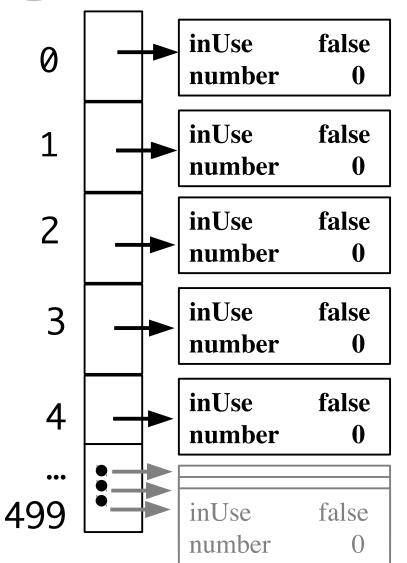
- Use an array of 500 objects
 - each array element <=> 1000 numbers

index	corresponds to numbers	from	to
0		0	999
1		1,000	1,999
2		2,000	2,999
j		floor(j/1000)	floor(j/1000)+999
499		499,000	499,999

- The objects have instance variables:
 - boolean inUse
 - int number

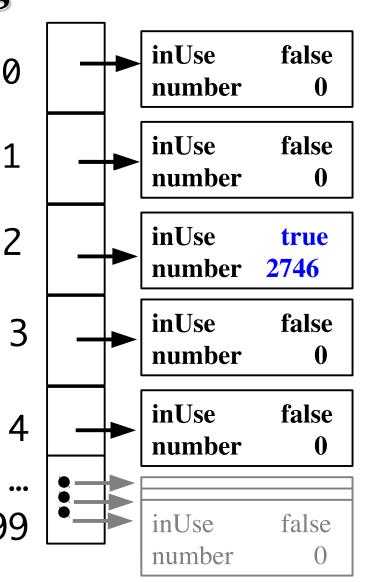
inUse false number 0

The empty set



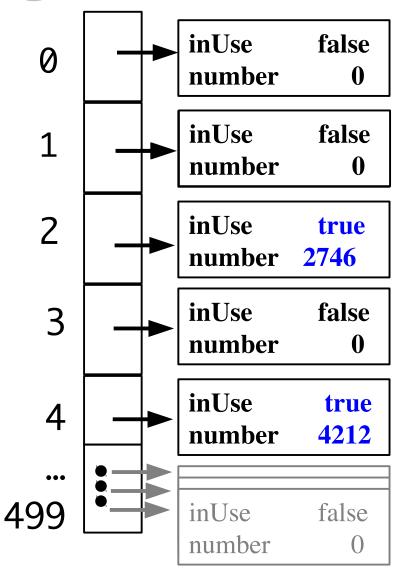
Add 2746 to the set:

- Corresponding index is 2
- Find object at that index
- Set inUse and numberAll O(1)



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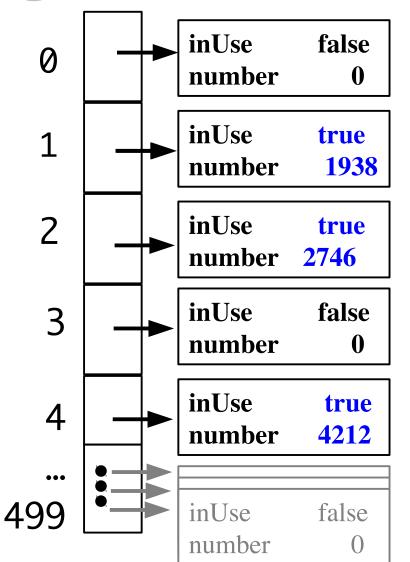
Add 4212 to the set:



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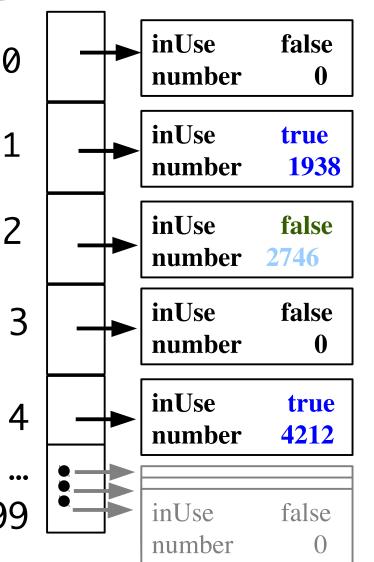
Add 1938 to the set:



Remove 2746 from the set:

Corresponding index is 2

- Find object at that index
- Set inUse to falseAll O(1)

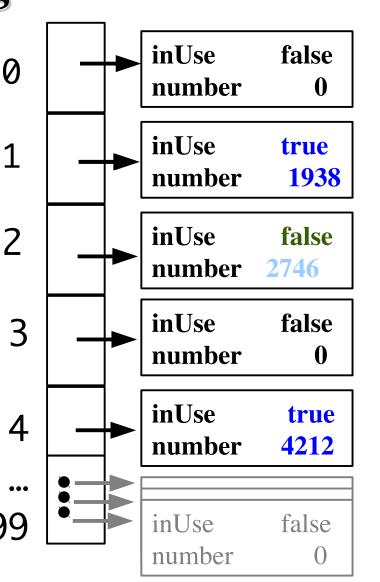


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Is 4352 in the set?

- Corresponding index is 4
- Find object at that index
- inUse is true but number ≠ 4352 → 4352 is not in the set All O(1)

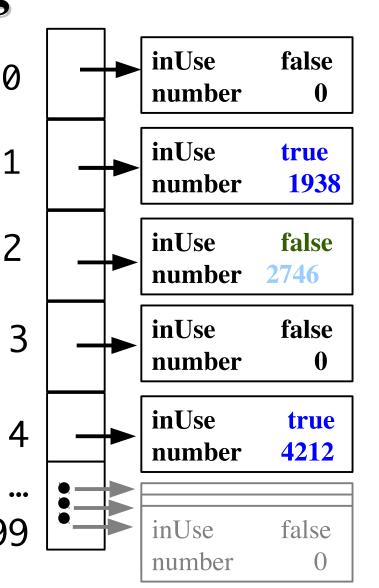


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Is 3314 in the set?

- Corresponding index is 3
- Find object at that index
- not inUse → 3314 is not in the set

All O(1)



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Hash Function

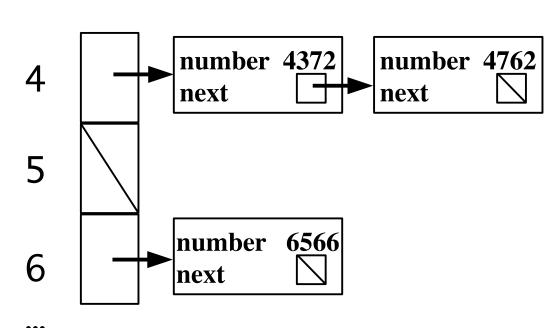
- What if numbers not random, eg likely to be near each other?
 - convert n to index in some other way, e.g. index = n mod 500
 - In general, function that makes each index equally likely: "makes hash out of any pattern in the numbers" -
- Hash function: converts data to hash code
- Mapping function: converts hash code to array index. (Why separate this?)

Collisions

- Even with 500 indices for 10 numbers, it is possible that more than one number will hash to same index
- As we reduce number of indices probability of collision grows
- => must be some way to handle collisions

Chaining

• Instead of an array index referring to a single object, have it refer to a linked list of objects.



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Complexity

- Worst case: O(n)
 - all items hash to same index
- Average: depends on load factor alpha = n / size

alpha	Average compares
.1	1.05
.5	1.3
.8	1.4
.9	1.45
. 99	1.5

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Built-in Hashing in Java

- The class java.util.HashMap<K, V>
 - Mapping from (unique) key to a value
 - Note: generic with two class parameters:
 - K: class of keys
 - V: class of values
 - E.g. NetID => Student
 java.util.HashMap<NetID, Student>
 - See JDK API
 - See Driver.java, UseDriverMap.java