

# Assignment 2

1.1)

The box is in the **first box**.

Proof by brute force:

1. If #1 was true, then box 2 shouldn't be empty according to #2, but the money shouldn't be in box 2 according to #3. #2 and #3 contradict themselves.
2. If #2 was true, then box 2 is empty. According to #1, box 1 should have the money, and according to the #3, box 2 doesn't have the money. This works out.

1.2)

Let's define  $A, B, C$  to represent whether the first, second or third box hold the money respectively.

$A$ : The first box has the money

$B$ : The second box has the money

$C$ : The third box has the money

1.3)

Label 1 implies  $\neg A$

Label 2 implies  $\neg B$

Label 3 implies  $B$

We also know that only one label is true:

$(\neg A \wedge \neg(\neg B) \wedge \neg(B)) \vee$  // label 1 is true

$(\neg(\neg A) \wedge \neg B \wedge \neg(B)) \vee$  // label 2 is true

$(\neg(\neg A) \wedge \neg(\neg B) \wedge B)$  // label 3 is true

We also know that only one box has the money:

$(A \wedge \neg B \wedge \neg C) \vee$  // box 1 has the money

$(\neg A \wedge B \wedge \neg C) \vee$  // box 2 has the money

$(\neg A \wedge \neg B \wedge C)$  // box 3 has the money

**CNF Form:**

$((\neg A \wedge \neg(\neg B) \wedge \neg(B)) \vee$

$(\neg(\neg A) \wedge \neg B \wedge \neg(B)) \vee$

$(\neg(\neg A) \wedge \neg(\neg B) \wedge B))$

$\wedge$

$((A \wedge \neg B \wedge \neg C) \vee$

$(\neg A \wedge B \wedge \neg C) \vee$

$(\neg A \wedge \neg B \wedge C))$

$\vdash A$

1.4)

**Prove by contradiction**

Reduce all  $\neg\neg$

$$\begin{aligned} & (((\neg A \wedge B \wedge \neg B) \vee \\ & (A \wedge \neg B \wedge \neg B) \vee \\ & (A \wedge B \wedge B)) \\ & \wedge \\ & ((A \wedge \neg B \wedge \neg C) \vee \\ & (\neg A \wedge B \wedge \neg C) \vee \\ & (\neg A \wedge \neg B \wedge C))) \\ & \wedge \neg A \end{aligned}$$

Reduce redundant  $\wedge$  terms

$$\begin{aligned} & (((\neg A \wedge B \wedge \neg B) \vee \\ & (A \wedge \neg B) \vee \\ & (A \wedge B)) \\ & \wedge \\ & ((A \wedge \neg B \wedge \neg C) \vee \\ & (\neg A \wedge B \wedge \neg C) \vee \\ & (\neg A \wedge \neg B \wedge C))) \\ & \wedge \neg A \end{aligned}$$

Remove contradicting clauses

$$\begin{aligned} & (((A \wedge \neg B) \vee \\ & (A \wedge B)) \\ & \wedge \\ & ((A \wedge \neg B \wedge \neg C) \vee \\ & (\neg A \wedge B \wedge \neg C) \vee \\ & (\neg A \wedge \neg B \wedge C))) \\ & \wedge \neg A \end{aligned}$$

Demorgans law

$$\begin{aligned} & (((A \wedge \neg B) \vee A) \wedge ((A \wedge \neg B) \vee B)) \\ & \wedge \\ & ((A \wedge \neg B \wedge \neg C) \vee \\ & (\neg A \wedge B \wedge \neg C) \vee \\ & (\neg A \wedge \neg B \wedge C))) \\ & \wedge \neg A \end{aligned}$$

Demorgans law again

$$\begin{aligned} & (((A \vee A) \wedge (\neg B \vee A) \wedge (A \vee B) \wedge (\neg B \vee B)) \\ & \wedge \\ & ((A \wedge \neg B \wedge \neg C) \vee \\ & (\neg A \wedge B \wedge \neg C) \vee \\ & (\neg A \wedge \neg B \wedge C))) \\ & \wedge \neg A \end{aligned}$$

Reduce contradictions and redundant terms

$$(A \wedge (\neg B \vee A) \wedge (A \vee B))$$
$$\wedge$$
$$((A \wedge \neg B \wedge \neg C) \vee$$
$$(\neg A \wedge B \wedge \neg C) \vee$$
$$(\neg A \wedge \neg B \wedge C))$$
$$\wedge \neg A$$

Break first parenthesis

$$A \wedge (\neg B \vee A) \wedge (A \vee B)$$
$$\wedge$$
$$((A \wedge \neg B \wedge \neg C) \vee$$
$$(\neg A \wedge B \wedge \neg C) \vee$$
$$(\neg A \wedge \neg B \wedge C))$$
$$\wedge \neg A$$

$A$  cannot be both true and false, so we have a contradiction. As a result, box 1 has the money.

2)

**Accuracy:**

Average: 85.431%

Weighted: 81.188%

**Code:**

```
1  from collections import defaultdict
2  from Queue import PriorityQueue as queue
3  import math
4
5  AVERAGE_INFO_GAIN = 1
6  WEIGHTED_INFO_GAIN = 2
7  APPROACH = WEIGHTED_INFO_GAIN
8
9  RENDER = True
10
11
12  class Node:
13
14      def __init__(self, word_id, docs, word_ids, is_terminal=False, contains=
None):
```

```

15     self.word_id = word_id
16     self.entropy = entropy(docs)
17     self.children = {}
18     self.docs = docs
19     self.word_ids = word_ids
20     if is_terminal:
21         self.info_gain = 0.0
22     else:
23         self.info_gain = information_gain(self.docs, self.word_id)
24     self.split = self.get_split()
25     self.contains = contains
26     self.is_terminal = is_terminal
27
28     def add_child(self, node, val):
29         self.children[val] = node
30
31     def has_child(self, val):
32         return val in self.children
33
34     def child_count(self):
35         return len(self.children)
36
37     def is_end(self):
38         # Checks if there are any empty
39         labels = self.split
40         if labels[True][1] + labels[True][2] == 0 or labels[False][1] + labels[False][2] == 0:
41             return True
42         return False
43
44     def get_split(self):
45         labels = {}
46         labels[True] = defaultdict(int)
47         labels[False] = defaultdict(int)
48
49         for d in self.docs:
50             if d.has_word(self.word_id):
51                 labels[True][d.label] += 1

```

```

52         else:
53             labels[False][d.label] += 1
54
55         return labels
56
57     def format_split(self):
58         labels = self.split
59         return 'True-1: {} True-2: {} False-1: {} False-2: {}'.format(labels
[True][1], labels[True][2], labels[False][1], labels[False][2])
60
61     def __cmp__(self, node):
62         if self.entropy < node.entropy:
63             return -1
64         elif self.entropy > node.entropy:
65             return 1
66         return 0
67
68     def __str__(self):
69         return str(self.word_id) + ' ' + str(self.info_gain) + ' ' + 'docs:
' + str(len(self.docs)) + ' ' + 'word_ids: ' + str(self.word_ids)
70
71     def get_id(self):
72         word_ids = map(str, self.word_ids)
73         if self.is_terminal:
74             return str(self.contains) + '-' + '-'.join(word_ids)
75         return '-'.join(word_ids)
76
77
78 class Option:
79
80     def __init__(self, n1, n2, word_id, contains):
81         self.n1 = n1
82         self.n2 = n2
83         self.gain = information_gain(n2.docs, n2.word_id)
84         self.contains = contains
85
86     def __cmp__(self, option):
87         if self.gain < option.gain:

```

```
88         return 1
89     elif self.gain > option.gain:
90         return -1
91     return 0
92
93     def __str__(self):
94         return str(self.n1.word_id) + ' ' + str(self.n2.word_id) + ' ' + str
95         (self.gain) + ' ' + str(self.contains)
96
97 class Doc:
98
99     def __init__(self, id):
100         self.id = id
101         self.label = None
102         self.words = defaultdict(bool)
103
104     def add_word(self, index):
105         self.words[index] = True
106
107     def set_label(self, label):
108         self.label = label
109
110     def has_word(self, word_id):
111         return self.words[word_id]
112
113     def get_label(self, node):
114         has_word = self.has_word(node.word_id)
115         if node.has_child(has_word):
116             return self.get_label(node.children[has_word])
117         if node.split[has_word][1] > node.split[has_word][2]:
118             return 1
119         else:
120             return 2
121
122
123 def entropy(docs):
124     count = defaultdict(int)
```

```
125     for doc in docs:
126         count[doc.label] += 1
127
128     s = 0.0
129     for i in count:
130         p = 1.0 * count[i] / len(docs)
131         s += -1.0 * p * math.log(p, 2)
132
133     return s
134
135
136 def information_gain(docs, word_id):
137     has_word = []
138     missing_word = []
139     for doc in docs:
140         if doc.has_word(word_id):
141             has_word.append(doc)
142         else:
143             missing_word.append(doc)
144
145     has_word_entropy = entropy(has_word)
146     missing_word_entropy = entropy(missing_word)
147
148     if APPROACH == 2:
149         return entropy(docs) - \
150             (1.0 * len(has_word) / len(docs) * has_word_entropy +
151              1.0 * len(missing_word) / len(docs) * missing_word_entropy)
152     else:
153         return entropy(docs) - (0.5 * has_word_entropy + 0.5 * missing_word_
entropy)
154
155
156 def load_data(filename):
157     docs = {}
158     last_index = 0
159     with open(filename, 'r') as f:
160         for line in f.readlines():
161             doc_id, word_id = line.rstrip().split('\t')
```

```
162         doc_id = int(doc_id)
163         word_id = int(word_id)
164
165         if doc_id not in docs:
166             docs[doc_id] = Doc(doc_id)
167             if doc_id - last_index > 1:
168                 for i in range(last_index+1, doc_id):
169                     docs[i] = Doc(i)
170
171             docs[doc_id].add_word(word_id)
172             last_index = doc_id
173
174     return docs
175
176
177 def load_labels(docs, filename):
178     with open(filename, 'r') as f:
179         for i, line in enumerate(f.readlines()):
180             docs[i].set_label(int(line.strip()))
181
182
183 def load_words(filename):
184     words = []
185     with open(filename, 'r') as f:
186         for line in f.readlines():
187             words.append(line.strip())
188
189     return words
190
191
192 def render_tree(node, words):
193     global RENDER
194     if not RENDER:
195         return
196     from graphviz import Digraph
197     dot = Digraph()
198     register_node(dot, node, words)
```



```
199     dot.render('decision-tree.gv', view=True)
200
201
202 def register_node(dot, node, words):
203     if node.is_terminal:
204         label = 1
205         if node.split[node.contains][2] > node.split[node.contains][1]:
206             label = 2
207         dot.node(node.get_id(), str(label))
208     else:
209         dot.node(node.get_id(), str(words[node.word_id]) + ', info gain: ' +
210 str(node.info_gain)[0:6])
211
212     if True in node.children:
213         dot.edge(node.get_id(), node.children[True].get_id(), label='Contains')
214         register_node(dot, node.children[True], words)
215
216     if False in node.children:
217         dot.edge(node.get_id(), node.children[False].get_id(), label='Not contains')
218         register_node(dot, node.children[False], words)
219
220 def fill_single_childs(node):
221     print(node.get_id())
222     if node.has_child(True):
223         fill_single_childs(node.children[True])
224     if node.has_child(False):
225         fill_single_childs(node.children[False])
226
227     print(node.child_count())
228     if node.child_count() == 2:
229         return
230
231     is_true = node.has_child(True)
232
233     has_word = []
234     missing_word = []
```

```
234     for d in node.docs:
235         if d.has_word(node.word_id):
236             has_word.append(d)
237         else:
238             missing_word.append(d)
239
240     if node.child_count() == 1:
241         if is_true:
242             node.add_child(Node(node.word_id, missing_word, node.word_ids, T
rue, False), False)
243         else:
244             node.add_child(Node(node.word_id, has_word, node.word_ids, True,
True), True)
245         else:
246             node.add_child(Node(node.word_id, missing_word, node.word_ids, True,
False), False)
247             node.add_child(Node(node.word_id, has_word, node.word_ids, True, Tru
e), True)
248
249
250 def create_tree():
251     docs = load_data('trainData.txt')
252     docs = [docs[i] for i in docs]
253     load_labels(docs, 'trainLabel.txt')
254     words = load_words('words.txt')
255     pq = queue()
256
257     root = Node(None, docs, [])
258     for i in range(1, len(words)+1):
259         n = Node(i, docs, [i])
260         pq.put(Option(root, n, n.word_id, True))
261
262     start = pq.get()
263     while not pq.empty():
264         try:
265             pq.get(False)
266         except:
267             continue
```

```
268         pq.task_done()
269
270     pq.put(start)
271     node_count = 1
272
273     while node_count < 200:
274         o = pq.get()
275         parent_parent_node = o.n1
276         parent_node = o.n2
277         if parent_parent_node.has_child(o.contains) or parent_node.is_end():
278             continue
279
280         print(str(parent_node), parent_node.format_split())
281         print(str(o))
282         parent_parent_node.add_child(parent_node, o.contains)
283         node_count += 2
284
285         has_word = []
286         missing_word = []
287         print('Parent node word id', parent_node.word_id)
288         for d in parent_node.docs:
289             if d.has_word(parent_node.word_id):
290                 has_word.append(d)
291             else:
292                 missing_word.append(d)
293
294         for word in range(1, len(words)+1):
295             if word not in parent_node.word_ids:
296                 n1 = Node(word, has_word, parent_node.word_ids + [word])
297                 o1 = Option(parent_node, n1, word, True)
298                 pq.put(o1)
299
300                 n2 = Node(word, missing_word, parent_node.word_ids + [word])
301                 o2 = Option(parent_node, n2, word, False)
302                 pq.put(o2)
303
304         print('')
```

```

305     fill_single_childs(root.children[True])
306
307
308     render_tree(root.children[True], words)
309
310     return root.children[True]
311
312
313 def load_test_label(filename):
314     label = []
315     with open(filename, 'r') as f:
316         for i, line in enumerate(f.readlines()):
317             label.append(int(line.strip()))
318     return label
319
320
321 def test_tree(root):
322     correct = 0
323     data = load_data('testData.txt')
324     label = load_test_label('testLabel.txt')
325     for i in range(1, len(label)+1):
326         predicted_label = data[i].get_label(root)
327         if predicted_label == label[i-1]:
328             correct += 1
329     print('Accuracy: ' + str(correct * 1.0 / len(label)))
330
331 if __name__ == '__main__':
332     tree = create_tree()
333     test_tree(tree)

```