

Question 1 MIXED SHORT QUESTIONS

Part 1 (Bagging and Random Forests)

- (a) FALSE (the order of examples doesn't matter)
- (b) FALSE (sample per split, not per tree)
- (c) TRUE
- (d) TRUE

Part 2 (Frequent Pattern Mining)

- (a) TRUE
- (b) FALSE (3 times)
- (c) FALSE (candidates are generated by rightmost extension)
- (d) TRUE

Part 3 (Classification Trees)

If we predict class  $c$ , the probability of making a wrong prediction is  $1-p(c|t)$ . We predict class  $c$  with probability  $p(c|t)$ , so the probability of making an error is  $\sum_{c=1}^C p(c|t)(1-p(c|t))$ , which is the formula for the gini-index.

Part 4 (Graphical Models)

The correct answer is (b). Model (b) fits the independence model, and  $x_1$  and  $x_2$  are exactly independent in the data, so the independence model gives a perfect fit of the observed counts. The saturated model (c) fits equally well, but uses one more parameter, so has a worse BIC score. Model (a) fits the uniform table of counts 25,25,25,25. It has two less parameters than the independence model, but a much worse fit.

Part 5 (Logistic Regression)

The correct answer is (d). The odds are multiplied by  $\exp(0.14)$  which is approximately 1.15. Multiplication by 1.15 is the same as an increase of 15%.

Question 2 CLASSIFICATION TREES

- (a)  $i(t_1) = 1/2 * 1/2 = 1/4$ .  $i(t_2) = i(t_3) = 1/5 * 4/5 = 4/25$ . Reduction is  $1/4-(1/2*4/25+1/2*4/25)=9/100$ .
- (b) The smallest minimizing subtree (SMS) for  $a_1=0$  is obtained by pruning in  $t_2$ . Next we compute  $g(t_1) = 21/100$  and  $g(t_3)=2/100$ . We prune in  $t_3$  and set  $a_2=2/100$ . Next we recompute  $g(t_1)=3/10$ , and set  $a_2=3/10$ . Summarizing:  $T_1$  is obtained by pruning in  $t_2$ , and it is the SMS for  $a$  in  $[0,2/100)$ .  $T_2$  is obtained by pruning  $T_1$  in node  $t_3$ , and it is the SMS for  $a$  in  $[2/100,3/10)$ . The root node is the SMS for  $a \geq 3/10$ .
- (c)  $\sqrt{2/100 * 3/10} = 0.077$  (approximately)

Question 3 CLOSED FREQUENT ITEM SET MINING

(a)

LEVEL 1:			LEVEL 2:		
	sup	gen?		sup	gen?
A	2	v	AB	2	x
B	4	v	AC	2	x
C	5	v	AD	0	x
D	3	v	BC	4	x
E	1	x	BD	2	v
			CD	3	x

E and AD are pruned due to insufficient support, all other pruning because the itemset has a subset with the same support.

(b)

gen	closure	sup
A	ABC	2
B	BC	4
C	C	5
D	CD	3
BD	BCD	2

Question 4 BAYESIAN NETWORKS

- (a) Yes, the resulting model has the same skeleton and v-structures.
- (b) The current score of lipo is -1208. After deleting the edge mental --> lipo, lipo has one parent left, which is smoke. The score of lipo then becomes:  
  
 $598 \log 598/961 + 363 \log 363/961 + 463 \log 463/880 + 417 \log 417/880 = -1245.855$ , which we will round to -1246.  
  
The change in log-likelihood score is  $-1246 + 1208 = -38$ , so a decrease of 38.
- (c) The penalty per parameter is  $\log(1841)/2 = 3.76$ . Deleting the edge mental --> lipo reduces the number of parameters by 2, so the change in BIC score is  $-38 + 2 * 3.76 = -30.48$ , or -30 after rounding. The BIC score decreases by 30.

Question 5 MULTINOMIAL NAIVE BAYES

(a)  $|V|=10$

$P(\text{good}|\text{Pos}) = (2+1)/(8+10) = 1/6$   
 $P(\text{good}|\text{Neg}) = (0+1)/(8+10) = 1/18$   
 $P(\text{teacher}|\text{Pos}) = (1+1)/(8+10) = 1/9$   
 $P(\text{teacher}|\text{Neg}) = (2+1)/(8+10) = 1/6$

(b) The class priors are  $P(\text{Pos}) = P(\text{Neg}) = 1/2$   
We ignore "very" in the test document, because it didn't occur in the training set

$P(\text{Pos})P(\text{good teacher}|\text{Pos}) = 1/2 * 1/6 * 1/9 = 1/108$   
 $P(\text{Neg})P(\text{good teacher}|\text{Neg}) = 1/2 * 1/18 * 1/6 = 1/216$

$P(\text{Pos}|\text{good teacher}) = (1/108) / (1/108 + 1/216) = 2/3$

(c) One training document of class A containing the word "a", and one training document of class B containing the word "b"  
Test document containing the words "a" and "b".