CS578/STAT590: Introduction to Machine Learning

Fall 2014

Problem Set 4

Gen Nishida Handed In: December 3, 2014

Questions

1. Boosting

(1) The weak hypothesis used at each round, and its error **1st round** $h_1(x) = \text{sign}(x_1 > 6), \ \epsilon_1 = 0.2 \ (\alpha_1 = 0.6931)$

index	x_1	x_2	y	prediction
1	1	10	-	-
2	4	4	-	-
3	8	7	+	+
4	5	6	-	-
5	3	16	-	-
6	7	7	+	+
7	10	14	+	+
8	4	2	-	-
9	4	10	+	-
10	8	8	-	+

Table 1: The prediction by h_1

2nd round $h_2(x) = sign(x_2 > 9), \ \epsilon_1 = 0.25 \ (\alpha_2 = 0.5493)$

index	x_1	x_2	y	prediction
1	1	10	-	+
2	4	4	-	-
3	8	7	+	-
4	5	6	-	-
5	3	16	-	+
6	7	7	+	-
7	10	14	+	+
8	4	2	-	-
9	4	10	+	+
10	8	8	-	-

Table 2: The prediction by h_1

(2) The distribution D_i over the examples for each round The distribution D_i evolves over the rounds as shown in Table 3. Gen Nishida 2

index	D_1	D_2	D_3
1	0.1	0.0625	0.125
2	0.1	0.0625	0.0417
3	0.1	0.0625	0.125
4	0.1	0.0625	0.0417
5	0.1	0.0625	0.125
6	0.1	0.0625	0.125
7	0.1	0.0625	0.0417
8	0.1	0.0625	0.0417
9	0.1	0.25	0.1666
10	0.1	0.25	0.1666

Table 3: The distribution D_i

(3) The final hypothesis after running two rounds. $H_{\text{final}}(x) = \text{sign} \left[0.6931 \times \text{sign}(x_1 > 6) + 0.5493 \times \text{sign}(x_2 > 9) \right]$

2. Naive Bayes

(1) Write the joint log-likelihood of a document and labels (i.e., $\log P(D_i, y_i)$). First, the joint probability $P(D_i, y_i)$ is

$$\left[\frac{n!}{a_i!b_i!c_i!d_i!}\alpha_1^{a_i}\beta_1^{b_i}\gamma_1^{c_i}\delta_1^{d_i}\times P(y_i=1)\right]^{y_i}\times \left[\frac{n!}{a_i!b_i!c_i!d_i!}\alpha_0^{a_i}\beta_0^{b_i}\gamma_0^{c_i}\delta_0^{d_i}\times P(y_i=0)\right]^{(1-y_i)}$$

Since we do not know a prior probability of P(y), we just use the uniform distribution, which is P(y) = 1/2. Thus,

$$\frac{1}{2} \times \left[\frac{n!}{a_i!b_i!c_i!d_i!} \alpha_1^{a_i} \beta_1^{b_i} \gamma_1^{c_i} \delta_1^{d_i} \right]^{y_i} \times \left[\frac{n!}{a_i!b_i!c_i!d_i!} \alpha_0^{a_i} \beta_0^{b_i} \gamma_0^{c_i} \delta_0^{d_i} \right]^{(1-y_i)}$$

Then, its logarithm becomes

$$\log \frac{1}{2} + y_i \log \left[\frac{n!}{a_i!b_i!c_i!d_i!} \alpha_1^{a_i} \beta_1^{b_i} \gamma_1^{c_i} \delta_1^{d_i} \right] + (1 - y_i) \log \left[\frac{n!}{a_i!b_i!c_i!d_i!} \alpha_0^{a_i} \beta_0^{b_i} \gamma_0^{c_i} \delta_0^{d_i} \right]$$

$$= \log \frac{1}{2} + \log \frac{n!}{a_i!b_i!c_i!d_i!} + y_i \left[a_i \log \alpha_1 + b_i \log \beta_1 + c_i \log \gamma_1 + d_i \log \delta_1 \right]$$

$$+ (1 - y_i) \left[a_i \log \alpha_0 + b_i \log \beta_0 + c_i \log \gamma_0 + d_i \log \delta_0 \right]$$

3. Bayesian Network

(1) How many parameters are needed to define the network?

21 parameters

A requires 1 parameter, and B requires 2 parameters. For C, there are $2 \times 3 = 6$ combinations of A and B and for each A and B, it requires 2 parameters, so $6 \times 2 = 12$ parameters. Finally, D requires $3 \times 2 = 6$ parameters. Thus, 1 + 2 + 12 + 6 = 21 parameters are required in total.

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(2) Write the expression calculating

(a)
$$P(A = 1, D = 2)$$

$$= \sum_{B,C} P(A = 1, B, C, D = 2)$$

$$= \sum_{B,C} P(A = 1)P(B)P(C|A = 1, B)P(D = 2|C)$$

(b)
$$P(A = 1, D = 2|C = 1)$$

$$= \frac{\sum_{B} P(A=1,B,C=1,D=2)}{P(C=1)}$$

$$= \frac{\sum_{B} P(A=1)P(B)P(C=1|A=1,B)P(D=2|C=1)}{P(C=1)}$$

4. Variable Elimination

(1) Write the expression calculating P(Marry calls), according to the network below. How many operations are needed to compute it?

79 operations in total.

$$\sum_{E,B,A,J} P(E)P(B)P(A|B,E)P(J|A)P(M|A)$$

The product consists of 5 terms, which requires 4 multiplications. Also, since each summation with regard to E, B, A, and J requires two iterations, the number of multiplications becomes $4 \times 2^4 = 64$ in addition to $2^4 - 1 = 15$ additions. Thus, 64 + 15 = 79 operations are required in total.

(2) Run the variable elimination algorithm. How many operations are needed now? (Write down each step)

23 operations in total.

At each step, we eliminate a variable which has the fewst degree in the interaction graph (Figure 1).

1) Eliminate J

$$= \sum_{E.B.A} P(E)P(B)P(A|B,E)P(M|A)f_1(A)$$

where $f_1(A) = \sum_J P(J|A)$. Apparently, $f_1(A) = 1$, so no operation is required for this elimination.

2) Eliminate E

$$= \sum_{B,A} P(B)P(M|A)f_1(A)f_2(A,B)$$

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where $f_2(A, B) = \sum_E P(E)P(A|B, E)$. For each pair of A and B, one multiplication is required for each E, so the number of multiplications is 2. Also, the summation requires 1 addition. Therefore, 3 operations are necessary for each pair of A and B. Since we have $2^2 = 4$ combinations of A and B, $3 \times 4 = 12$ operations are required for $f_2(A, B)$.

3) Eliminate B

$$= \sum_{A} P(M|A)f_1(A)f_3(A)$$

where $f_3(A) = \sum_B P(B) f_2(A, B)$. For a given A, one addition and two multiplications are necessary. Thus, $3 \times 2 = 6$ operations are required for $f_3(A)$.

4) Eliminate A

$$f_4(M) = \sum_{A} P(M|A)f_1(A)f_3(A)$$

We need two multiplications for each A, so $2 \times 2 = 4$ multiplications are required. Also, we need one addition for summation. Thus, 4+1=5 operations are required for $f_4(M)$. To sumarize, we need 12+6+5=23 operations are required in total. Note that the number of operations are much less than the one without variable elimination.

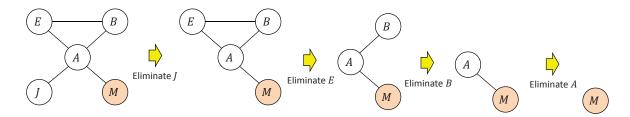


Figure 1: This figure shows how the interaction graph changes over the variable eliminations. Note that at each step the node which has the fewest degrees in the graph is chosen to eliminate.