EECS 445: Machine Learning

Fall 2016

Discussion 9: Explaining Away

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9.1 Probabilistic influence

Recall the HW5 Problem 3.

Lemma 9.1.

$$P(t_1|d_1) < P(t_1|d_0)$$

Intuitively, this statement is kind of obvious. The person is on good diet (d_1) should be less likely to test for high cholesterol (t_1)

Proof: From the factorization theorem, T is independent of all variables other than C given D. So,

$$P(T|D,u) = P(T|D) \tag{9.1}$$

for all values of u.

Now from the definition, we have that

$$P(c_1|d_1) < P(c_1|d_0) \tag{9.2}$$

meaning people with good diet is less likely to have a high cholesterol than those with bad diet. We were able to eliminate the conditioning on u because of (9.1). Similarly,

$$P(t_1|c_1) > P(t_1|c_0).$$

We write

$$P(T|D) = \sum_{c} P(T,C=c|D) = \sum_{c} P(T|C=c)P(C=c|D)$$

The second equality is because T and D are independent given C.

Now,

$$P(t_1|d_1) = P(t_1|c_1)P(c_1|d_1) + P(t_1|c_0)P(c_0|d_1)$$

and

$$P(t_1|d_0) = P(t_1|c_1)P(c_1|d_0) + P(t_1|c_1)P(c_1|d_0)$$

We compute the difference:

$$P(t_1|d_0) - P(t_1|d_1) = P(t_1|c_1)(P(c_1|d_0) - P(c_1|d_1)) + P(t_1|c_0)(P(c_0|d_0) - P(c_0|d_1))$$

which is less than 0 because both terms in the parentheses are negative.

Corollary 9.2.

$$P(t_1|d_1) < P(t_1)$$

Proof: Note that

$$P(t_1) = P(t_1|d_1)p(d_1) + p(t_1|d_0)p(d_0) < P(t_1|d_1)p(d_1) + p(t_1|d_1)p(d_0) = P(t_1|d_1)$$

Lemma 9.3.

$$P(d_1|e_1, m_1) < P(d_1|m_1)$$

Proof:

$$P(d_1|e_1, m_1) = P(m_1|d_1, e_1)P(d_1)/P(m_1) > P(m_1|d_1, e_0)P(d_1)/P(m_1) = P(d_1|e_0, m_1).$$

So,

$$P(d_1|m_1) = P(d_1|m_1, e_0)P(e_0) + P(d_1|m_1, e_1)P(e_1) < P(d_1|m_1, e_1)P(e_0) + P(d_1|m_1, e_1)P(e_1) = P(d_1|m_1, e_1)P(e_1) < P(d_1|m_1, e_1)P(e_1)P(e_1) < P(d_1|m_1, e_1)P(e_1)P(e_1)P(e_1) < P(e_1|m_1, e_1)P(e_1)P(e_1)P(e_1)P(e_1)P(e_1)$$

9.2 Explaining Away

The explaining away is the following phenomenon:

$$P(h_1|b_0, e_1) < P(h_1|b_1, e_1). (9.3)$$

It follows that

$$P(h_1|b_0, e_1) < P(h_1|e_1) < P(h_1|b_1, e_1).$$

Let's explore the sufficient and necessary condition for this to happen.

By the Bayes rule,

$$P(h_1|e_1, B) = P(e_1|h_1, B) \frac{P(h_1|B)}{P(e_1|B)} = P(e_1|h_1, B) \frac{P(h_1)}{P(e_1|B)}$$

since H and B are independent.

Note that

$$P(e_1|B) = P(e_1|h_1, B)P(h_1) + P(e_1|h_0, B)P(h_0)$$

So,

$$\frac{1}{P(h_1|e_1,B)} = 1 + \frac{P(e_1|h_0,B)P(h_0)}{P(e_1|h_1,B)P(h_1)}$$

Hence, an equivalent statement of (9.3)

$$1/P(h_1|b_0,e_1) > 1/P(h_1|b_1,e_1)$$

is equivalent to

$$1 + \frac{P(e_1|h_0, b_0)P(h_0)}{P(e_1|h_1, b_0)P(h_1)} > 1 + \frac{P(e_1|h_0, b_1)P(h_0)}{P(e_1|h_1, b_1)P(h_1)}$$

which simplifies to

$$\frac{P(e_1|h_0,b_0)}{P(e_1|h_1,b_0)} > \frac{P(e_1|h_0,b_1)}{P(e_1|h_1,b_1)}.$$

Finally, we rearrange this:

$$\frac{P(e_1|h_0, b_0)}{P(e_1|h_0, b_1)} > \frac{P(e_1|h_1, b_0)}{P(e_1|h_1, b_1)}.$$
(9.4)

So, the time constraint has more dramatic effect on the people that are not health-conscious. Health-conscious people are less likely to exercise if they are busy, but this probability drop is much less severe because they try to make time for exercise.

Excercise 9.1. Give a realistic example where (9.4) fails to hold.

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