Lecture Notes week 1

$\ddot{\mathrm{O}}\mathrm{mer}$ Şakar

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1 Lecture 1

1.1 Bayes' Rule

Given events A and B:

Definition 1. The Conditional Probability $P(A|B) = \frac{P(A \cap B)}{P(B)}$, with P(B) > 0.

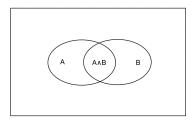


Figure 1: Visual representation of Bayes' Rule

Example: There are 40 Math students of which 15 are girls and there are 50 Computer Science students of which 10 are girls.

Let A = randomly chosen student is a girl and <math>B = randomly chosen student is a math student.

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = -\frac{\frac{15}{90}}{\frac{40}{90}} = \frac{3}{8}$$

And from Bayes' Rule follows $\Rightarrow P(A \cap B) = P(A|B) \cdot P(B) = P(B|A) \cdot P(A), P(A) > 0.$ Thus we can rewrite it as:

Definition 2. The Conditional Probability $P(B|A) = \frac{P(A|B) \cdot P(B)}{P(A)}$.

1.1.1 Full Probability Formula

Definition 3. $P(A) = P(B) \cdot P(A|B) + P(\bar{B}) \cdot P(A|\bar{B})$

Example:
$$P(B|A) = \frac{P(A|B) \cdot P(B)}{P(A)} = \frac{\frac{40}{90} \cdot \frac{3}{8}}{\frac{25}{90}} = \frac{3}{8}$$

1.2 A Herding Experiment

Envelope 1 contains 8 red and 4 blue domino pieces (R) and envelope 1 contains 4 red and 8 blue domino pieces (B).

$$P(R) = P(B) = \frac{1}{2}$$

The first person that draws either a red or blue piece.

$$P(R|(saw)blue) = \frac{P(blue|R) \cdot P(R)}{P(blue)}$$

$$P(blue) = P(R) \cdot P(blue|R) + P(B) \cdot P(blue|B) = \frac{1}{2} \cdot \frac{4}{12} + \frac{1}{2} \cdot \frac{8}{12} = \frac{1}{2}$$

$$P(blue) = P(R) \cdot P(blue|R) + P(B) \cdot P(blue|B) = \frac{1}{2} \cdot \frac{4}{12} + \frac{1}{2} \cdot \frac{8}{12} = \frac{1}{2}$$

$$P(R|blue) = \frac{\frac{4}{12} \cdot \frac{1}{2}}{\frac{1}{2}} = \frac{1}{3} < \frac{1}{2} \text{ and } P(B|blue) = \frac{\frac{8}{12} \cdot \frac{1}{2}}{\frac{1}{2}} = \frac{2}{3} > \frac{1}{2}$$

Now lets look at when a second person draws

Definition 4. $D_i = \{blue\}$ or $\{red\}$ – what person i says they saw

Definition 5. $E_i = \{blue\}\ or\ \{red\}\ -\ what\ person\ i\ saw$

Lets say that the first person says what he sees $(D_1 = E_1)$

$$P(B|blue, blue) = \frac{P(B) \cdot P(blue, blue|B)}{P(blue, blue)}$$

$$P(blue, blue) = \frac{1}{2} \cdot (\frac{2}{3})^2 + \frac{1}{2} \cdot (\frac{2}{3})^2 = \frac{5}{18}$$

Thus
$$P(B|blue, blue) = \frac{\frac{1}{2} \cdot (\frac{2}{3})^2}{\frac{5}{5}} = \frac{4}{5} > \frac{1}{2}$$

Lets say that the first person says what he sees $(D_1 - E_1)$ $P(B|blue, blue) = \frac{P(B) \cdot P(blue, blue|B)}{P(blue, blue)}$ $P(blue, blue) = \frac{1}{2} \cdot (\frac{2}{3})^2 + \frac{1}{2} \cdot (\frac{2}{3})^2 = \frac{5}{18}$ Thus $P(B|blue, blue) = \frac{\frac{1}{2} \cdot (\frac{2}{3})^2}{\frac{5}{18}} = \frac{4}{5} > \frac{1}{2}$ Conclusion: If $D_1 = \{blue\}$ and $E_2 = \{blue\} \implies D_2 = \{blue\}$

$$P(R) = P(B) = \frac{1}{2}$$

$$P(B|blue, red) = \frac{P(B) \cdot P(blue, red|B)}{P(blue, red)} = \frac{\frac{1}{2} \cdot \frac{2}{3} \cdot \frac{1}{2}}{\frac{1}{3} \cdot \frac{2}{5} \cdot \frac{1}{5} + \frac{1}{5} \cdot \frac{2}{5} \cdot \frac{1}{3}} = \frac{1}{2}$$

$$D_1 = \{blue\}, E_2 = \{red\} \implies D_2 = \{red\}$$
 If your opinions differ then

$$P(B|blue, red) = \frac{P(B) \cdot P(blue, red|B)}{P(blue, red)} = \frac{\frac{1}{2} \cdot \frac{2}{3} \cdot \frac{1}{2}}{\frac{1}{2} \cdot \frac{2}{3} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{2}{3} \cdot \frac{1}{2}} = \frac{1}{2}$$

$$D_1 = \{blue\}, E_2 = \{red\} \implies D_2 = \{red\} \text{ If your opinions differ then}$$

$$P(Bcascade|R) = \frac{1}{3} \cdot \frac{1}{3} + (\frac{1}{3} \cdot \frac{2}{3}) \cdot 2 \cdot \frac{1}{3} \cdot \frac{1}{3} + (\frac{1}{3} \cdot \frac{2}{3} \cdot 2)^2 \cdot \frac{1}{3} \cdot \frac{1}{3} + \cdots = \frac{1}{9} \cdot \frac{1}{1 - \frac{4}{9}} = \frac{1}{5} \equiv 20\%$$

1.3 Genral Cascade Model

Example: Adopting a product

- 1) Let B be bad and G be good. We estimate P(B) = 1 p and P(G) = p.
- 2) And let v_g be if we guess G right and v_b if we guess B right.

 $v_g \cdot p + v_b \cdot (1-p) = 0$ - Expected Reward. Rewritten it looks like $v_b + (v_g - v_b) \cdot p$.

It is accepted if P(G|signal) > p and rejected if P(G|signal) < p.

3) low(L) and high(H)

	В	G
L	q	1-q
Н	1-q	q

Table 1: With $q > \frac{1}{2}$

$$P(L|B) = q > \frac{1}{2}, \ P(H|B) = 1 - q. \ P(L|G) = 1 - q, \ P(H|G) = q > \frac{1}{2}.$$

First person
$$P(G|H) = \frac{P(H|G) \cdot P(G)}{P(H)} = \frac{p \cdot q}{p \cdot q + (1-p) \cdot (1-q)} > \frac{p \cdot q}{p \cdot q + (1-p) \cdot q} = \frac{p \cdot q}{q} = p$$

Thus $E_1 = \{H\} \implies D_1 = \{G\}$

Let a + b - person and S - signal a times H and b times L.

$$P(G|S) = \frac{P(S|G) \cdot P(G)}{P(S)} = \frac{p \cdot q^a \cdot (1-q)^b}{p \cdot q^a \cdot (1-q)^b + (1-p) \cdot (1-q)^a \cdot q^b} > \frac{p \cdot q^a \cdot (1-q)^b}{p \cdot q^a \cdot (1-q)^b + (1-p) \cdot q^a \cdot (1-q)^b} = p$$
This is when $a < b \iff (1-q)^a \cdot q^b > q^a \cdot (1-q)^b$ (because $q > 1-q$).

Thus when $a < b, D = \{B\}$

If $a > b \iff (1-q)^a \cdot q^b < q^a \cdot (1-q)^b$. This would result in $P(G|S) > p \implies D = \{G\}$ //TODO insert picture of graph

2 Lecture 2