

CSE 312

Foundations of Computing II

Bayes' Theorem

$$P(A|B) = \frac{P(A \cap B)}{Pr(B)} = \frac{Pr(B \cap A)}{Pr(B)}$$

$$\begin{aligned} Pr(B \cap A) &= Pr(B) Pr(A|B) \\ &= Pr(A) Pr(B|A) \end{aligned}$$

Definition (Bayes' Theorem)

$$\Pr(F | E) = \frac{\Pr(F \cap E)}{\Pr(E)} = \frac{\Pr(E \cap F)}{\Pr(E)} = \frac{\Pr(E | F) \Pr(F)}{\Pr(E)}$$

This allows us to use $\Pr(E | F)$ to calculate $\Pr(F | E)$ which turns out to be useful in a variety of applications.

In particular, it allows us swap $\Pr(\text{model} | \text{data})$ for $\Pr(\text{data} | \text{model})$.

It also allows us to “update” our beliefs based on new evidence.

Example

- 60% of email is spam
- 90% of spam has a forged header
- 20% of non-spam has a forged header

What is $\Pr(\text{spam} \mid \text{forged header})$? What happens as we vary how much spam there is?

Let S be the event “the email is spam”.

Let F be the event “there is a forged header”.

$$(0.9) \quad (0.6)$$

↓

$$\Pr(S|F) = \frac{\Pr(S \cap F)}{\Pr(F)} = \frac{\Pr(F|S)\Pr(S)}{\Pr(F)}$$

$$\Pr(F) = \frac{\Pr(F|S)\Pr(S)}{(0.9)(0.6)} + \frac{\Pr(F|\bar{S})\Pr(\bar{S})}{(0.2)(0.4)}$$

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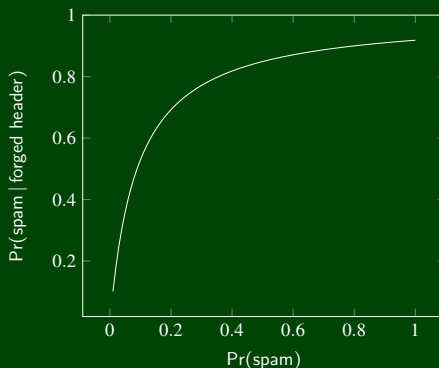
Let F be the event “there is a forged header”.

$$\begin{aligned}\Pr(S \mid F) &= \frac{\Pr(S \cap F)}{\Pr(F)} = \frac{\Pr(F \mid S) \Pr(S)}{\Pr(F)} = \frac{(0.9)(0.6)}{\Pr(F)} \\ &= \frac{(0.9)(0.6)}{\Pr(F \mid S) \Pr(S) + \Pr(F \mid \bar{S}) \Pr(\bar{S})} \\ &= \frac{(0.9)(0.6)}{(0.9)(0.6) + (0.2)(0.4)} \\ &\approx 0.871\end{aligned}$$

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What is $\Pr(\text{spam} \mid \text{forged header})$? What happens as we vary how much spam there is?



Example

- 60% of email is spam
- 10% of spam has the word “Viagra”
- 1% of non-spam has the word “Viagra”

What is $\Pr(\text{spam} \mid \text{viagra})$?

Let V be the event “message contains the word ‘Viagra’ ”

Let S be the event “message is spam”

Example

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- 10% of spam has the word “Viagra”
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Let V be the event “message contains the word ‘Viagra’ ”

Let S be the event “message is spam”

$$\begin{aligned}\Pr(S \mid V) &= \frac{\Pr(V \mid S) \Pr(S)}{\Pr(V \mid S) \Pr(S) + \Pr(V \mid \bar{S}) \Pr(\bar{S})} \\ &= \frac{(0.1)(0.6)}{(0.1)(0.6) + (0.01)(0.4)} \\ &\approx 0.9375\end{aligned}$$

Example

- Child is born with (A,a) gene pair
- Mother has (A,A) gene pair
- Two possible fathers: $M_1 = (a,a)$ and $M_2 = (a,A)$

What is $\Pr(M_1 \text{ is the father} \mid \text{child born with } (A,a))$?

Let Aa be the event “child born with (A,a) ”.

Let F_1 be the event “ M_1 is the father”.

Let F_2 be the event “ M_2 is the father”.

Say $\Pr(F_1) = p$ and $\Pr(F_2) = 1 - p$

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$$\begin{aligned}\Pr(F_1 \mid Aa) &= \frac{\Pr(Aa \mid F_1) \Pr(F_1)}{\Pr(Aa \mid F_1) \Pr(F_1) + \Pr(Aa \mid \bar{F}_1) \Pr(\bar{F}_1)} \\ &= \frac{(1)(p)}{(1)(p) + (0.5)(1 - p)} \\ &= \frac{2p}{1 + p}\end{aligned}$$

Example

- An HIV test's "false negative" rate (negative when present) is 2% and its "false positive" (positive when not present) rate is 1%
- 0.5% of the population has HIV

Let P be the event " X tests positive for HIV".

Let H be the event " X has HIV".

What is $\Pr(H | P)$?

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What is $\Pr(H | P)$?

$$\begin{aligned}\Pr(H | P) &= \frac{\Pr(P | H) \Pr(H)}{\Pr(P | H) \Pr(H) + \Pr(P | \bar{H}) \Pr(\bar{H})} \\ &= \frac{(0.98)(0.005)}{(0.98)(0.005) + (0.01)(0.995)} \\ &\approx 0.330\end{aligned}$$

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What is $\Pr(H | \bar{P})$?

$$\begin{aligned}\Pr(H | \bar{P}) &= \frac{\Pr(\bar{P} | H) \Pr(H)}{\Pr(\bar{P} | H) \Pr(H) + \Pr(\bar{P} | \bar{H}) \Pr(\bar{H})} \\ &= \frac{(0.02)(0.005)}{(0.02)(0.005) + (0.99)(0.995)} \\ &\approx 0.0001\end{aligned}$$