

## **Step 1: Reason and Decide**

What do we know initially?

(H<sub>1</sub>) The car is behind door No. 1.

(H<sub>2</sub>) The car is behind door No. 2.

(H<sub>3</sub>) The car is behind door No. 3.

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The three hypotheses are **mutually exclusive and exhaustive**:

$$P(H_1 \vee H_2 \vee H_3) = P(H_1) + P(H_2) + P(H_3) = 1$$

## **Step 2: Learn**

What is it, precisely, that we learn?

### **Step 3: Reason and Decide**

Shall we switch doors?

**Bayes Theorem:**

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Note: We use the notation  $P(H, E) = P(H \wedge E)$  for the probability of the conjunction of two propositions.



## Bayes Rule:

$$P(H|E) = \frac{P(E|H) P(H)}{P(E)}$$

$P(H)$ : prior probability of H

$P(E|H)$ : likelihood of E

$P(E)$ : expectedness of E

$P(H|E)$ : posterior probability of H

From the **Rule of Total Probability** and the definition of conditional probability, we obtain for the expectedness of E:

$$\begin{aligned} P(E) &= P(E, H) + P(E, \neg H) \\ &= P(E|H) P(H) + P(E|\neg H) P(\neg H) \end{aligned}$$

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If  $H_1, H_2, \dots, H_n$  are mutually exclusive and exhaustive, then

$$\begin{aligned} P(E) &= P(E|H_1) P(H_1) + \dots + P(E|H_n) P(H_n) \\ &= \sum_{i=1}^n P(E|H_i) P(H_i). \end{aligned}$$

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Remember:  $P(H_1) = P(H_2) = P(H_3) = 1/3$

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$$P(E) = 1 \cdot (1/3) + 1 \cdot (1/3) = 2/3$$



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$$P(E|H_1) = 1$$

$$P(E|H_2) = 1$$

$$P(E|H_3) = 0$$

$$P(E) = 2/3$$

$$P(H_1) = P(H_2) = P(H_3) = 1/3$$

We obtain:

$$\begin{aligned} P_E(H_1) = P_E(H_2) &= \frac{1 \cdot 1/3}{2/3} \\ &= 1/2 \end{aligned}$$

## **Reasoning 2:**

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$$\begin{aligned} P(F) &= P(F|H_1)P(H_1) + P(F|H_2)P(H_2) + P(F|H_3)P(H_3) \\ &= (1/2) \cdot (1/3) + 1 \cdot (1/3) \\ &= 1/2 \end{aligned}$$

## Confirmation – The Bayesian Way

E confirms H iff  $P_E(H) > P(H)$ .

E disconfirms H, iff  $P_E(H) < P(H)$ .

E is irrelevant for H, iff  $P_E(H) = P(H)$ .

**Irrelevance:**

$$P(H|E) = P(H) \Leftrightarrow P(H, E) = P(H)P(E)$$

We assume: E is a deductive consequence of H.

Hence,  $P(E|H) = 1$ .

Using Bayes Theorem and Bayes Rule:

$$P_E(H) = \frac{P(H)}{P(E)}$$

We conclude: The smaller  $P(E)$ , the the greater  $P_E(H)$  for a fixed  $P(H)$ .



Let  $P(H_1) = h_1$ ,  $P(H_2) = h_2$ , and  $P(H_3) = h_3$ .

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Let  $h_1 > h_2, h_3$ .

$$P_F(H_1) = \frac{1/2 \cdot h_1}{P(F)}$$

$$P_F(H_2) = \frac{h_2}{P(F)}$$

$$P(F) = 1/2 \cdot h_1 + h_2$$

$$P_F(H_2) > P_F(H_1) \text{ iff } h_2 > 1/2 \cdot h_1$$

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$$P_F(H_1) = \frac{a \cdot 1/3}{P(F)}$$

$$P_F(H_2) = \frac{1/3}{P(F)}$$

$$P(F) = (1 + a) \cdot 1/3$$

Hence,  $P_F(H_2) > P_F(H_1)$ . Note that this results holds for all  $a \in (0, 1)$ .