

STAT 25100 Lecture 6
3.3 Bayes Rule & Law of Total Probability
3.4 Independence

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Law of Total Probability and Bayes' Rule

Example – A Nervous Job Applicant

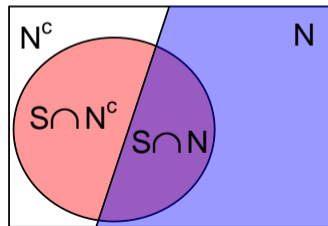
An job applicant has been invited for an interview.

Suppose The probability that

- ▶ he is nervous is $P(N) = 0.7$,
- ▶ the interview is successful given he is nervous is $P(S | N) = 0.2$,
- ▶ the interview is successful given he is not nervous is $P(S | N^c) = 0.9$.

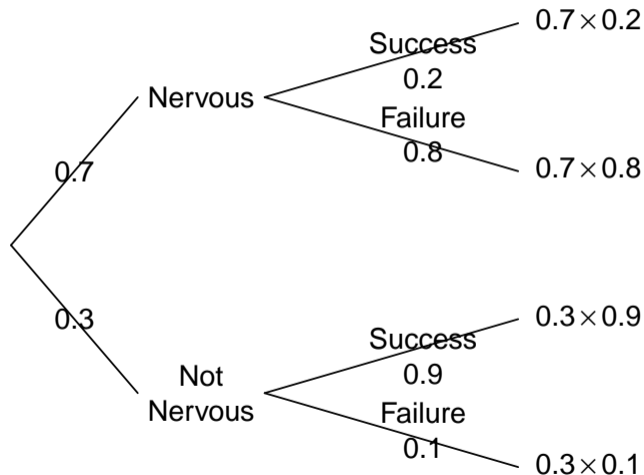
What is the probability that he succeeds in the interview?

$$\begin{aligned}P(S) &= P(S \cap N) + P(S \cap N^c) \\ &= P(N)P(S | N) + P(N^c)P(S | N^c) \\ &= 0.7 \times 0.2 + 0.3 \times 0.9 = 0.41.\end{aligned}$$



Tree Diagram for the Nervous Job Applicant Example

Another look of the computation:



Nervous Job Applicant Example Continued

Conversely, given the interview is successful, what is the probability that the job applicant is nervous during the interview?

$$\begin{aligned}P(N | S) &= \frac{P(N \cap S)}{P(S)} \\&= \frac{P(N \cap S)}{0.41} && P(S) = 0.41 \text{ from previous page} \\&= \frac{P(N)P(S | N)}{0.41} && \text{since } P(N \cap S) = P(N)P(S | N) \\&= \frac{0.7 \times 0.2}{0.41} = \frac{14}{41} \approx 0.34.\end{aligned}$$

Bayes' Rule = Bayes' Theorem = Bayes' Formula

The problem in the previous slide is an example of **Bayes' Rule**.

Knowing $P(B | A)$, $P(B | A^c)$, and $P(A)$, is there a way to know $P(A | B)$?

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

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Medical Testing

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- ▶ $P(T+ | D)$ is called the *sensitivity* of the test
- ▶ $P(T- | D^c)$ is called the *specificity* of the test
- ▶ Ideally, we hope $P(T+ | D)$ and $P(T- | D^c)$ both equal 1. However, medical tests are not perfect. They may give **false positives** and **false negatives**.

Enzyme-Linked Immunosorbent Assay (ELISA) Test for HIV

- ▶ $P(T+ | D) = 0.98$ (sensitivity - positive for infected)
- ▶ $P(T- | D^c) = 0.995$ (specificity - negative for not infected)
- ▶ $P(D) = 1/300$ (prevalence of HIV in USA)

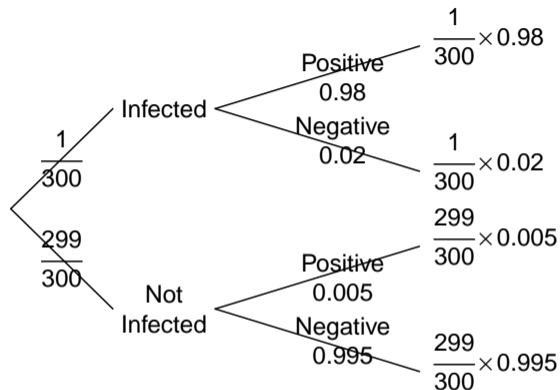
Suppose we apply ELISA test to the US population to screen for HIV.

- What percentage of people will be tested positive?
- What percentage of people that were tested positive actually are actually infected?

$$\begin{aligned}P(T+) &= P(D)P(T+ | D) + P(D^c)P(T+ | D^c) \\ &= \frac{1}{300} \times 0.98 + \frac{299}{300} \times 0.005 = 0.00825 \\ P(D | T+) &= \frac{P(D)P(T+ | D)}{P(T+)} = \frac{1/300 \times 0.98}{0.00825} = 39.6\%\end{aligned}$$

This test is not confirmatory. Need to confirm by a second test.

Tree Diagram for the HIV Test

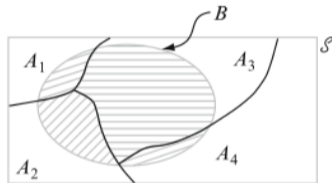


$$P(D | T+) = \frac{(1/300) \times 0.98}{(1/300) \times 0.98 + (299/300) \times 0.005}$$

Bayes' Rule for 3 or More Cases

- ▶ The 2 examples above both split the sample space into 2 parts A or A^c (nervous or not nervous, infected or not infected)
- ▶ In many cases, we need to calculate $P(B)$ by splitting it into several parts, using the **Law of Total Probability**:

Suppose A_1, A_2, \dots, A_k are disjoint and $A_1 \cup A_2 \cup \dots \cup A_k = S$ and $A_i \cap A_j = \emptyset$ for all $i \neq j$, then



$$\begin{aligned}P(B) &= P(B \cap A_1) + P(B \cap A_2) + \dots + P(B \cap A_k) \\ &= P(A_1)P(B | A_1) + P(A_2)P(B | A_2) + \dots + P(A_k)P(B | A_k).\end{aligned}$$

Using the **Law of Total Probability**, Bayes Rule becomes

$$P(A_i | B) = \frac{P(A_i)P(B | A_i)}{P(A_1)P(B | A_1) + P(A_2)P(B | A_2) + \dots + P(A_k)P(B | A_k)}$$

Example (Bayes' Rule for 3 Cases)

At a gas station,

- ▶ 40% of the customers use regular gas (A_1),
- ▶ 35% use mid-grade gas (A_2), and
- ▶ 25% use premium gas (A_3).

Moreover,

- ▶ of those customers using regular gas, only 30% fill their tanks;
- ▶ of those using mid-grade, 60% fill their tanks;
- ▶ of those using premium, 50% fill their tanks.

Let $B = \{\text{the next customer fills the tank.}\}$

Example (Bayes' Rule for 3 Cases)

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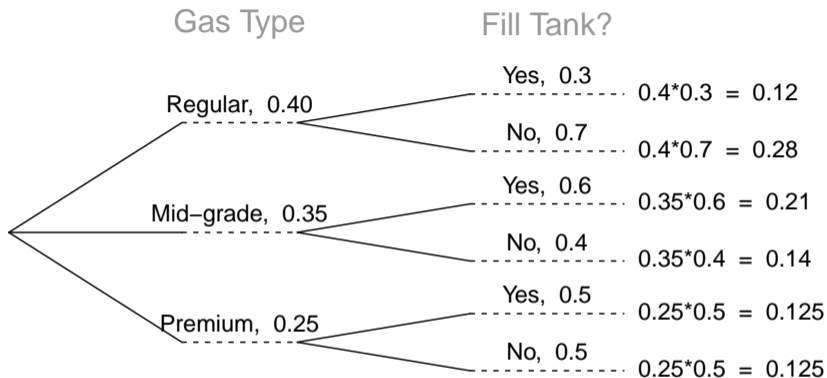
- ▶ 40% of the customers use regular gas (A_1),
- ▶ 35% use mid-grade gas (A_2), and
- ▶ 25% use premium gas (A_3).

Moreover,

- ▶ of those customers using regular gas, only 30% fill their tanks; $P(B | A_1) = 0.3$
- ▶ of those using mid-grade, 60% fill their tanks; $P(B | A_2) = 0.6$
- ▶ of those using premium, 50% fill their tanks. $P(B | A_3) = 0.5$

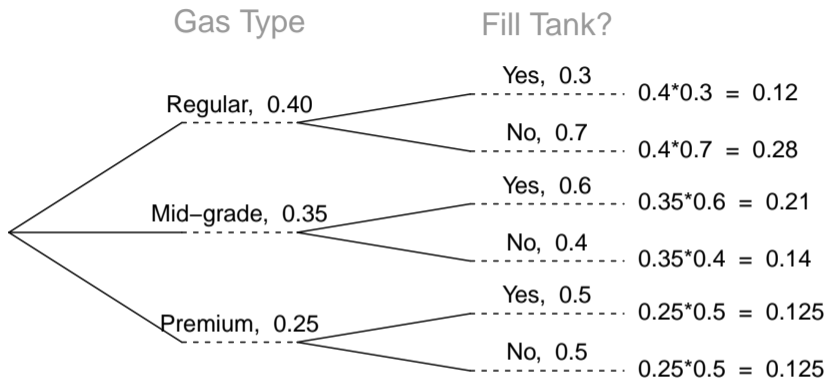
Let $B = \{\text{the next customer fills the tank.}\}$

Gas Station Example — Tree Diagram



Q1: What is the probability that the next customer requests premium gas and fills the tank?

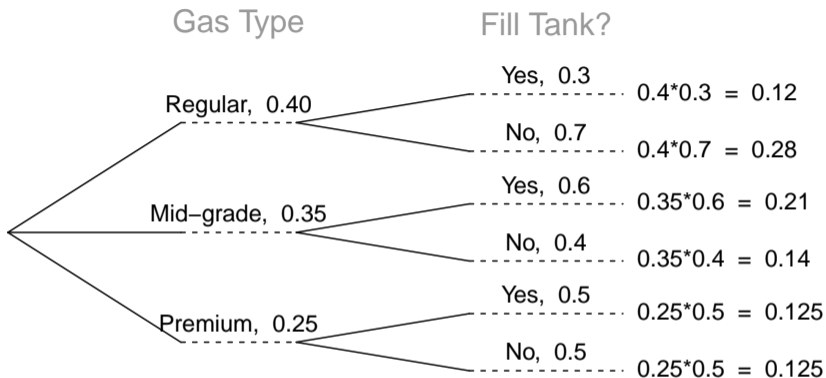
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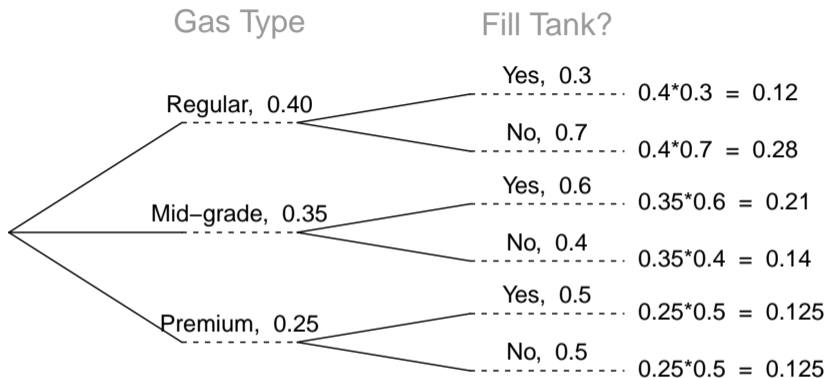
$$P(A_3 \cap B) = P(A_3)P(B | A_3) = 0.25 \times 0.5 = 0.125.$$

Gas Station Example — Tree Diagram



Q2: What is the probability that the next customer fills the tank.

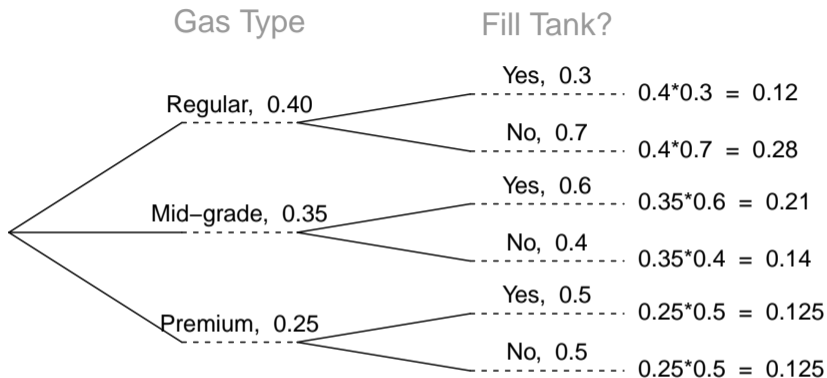
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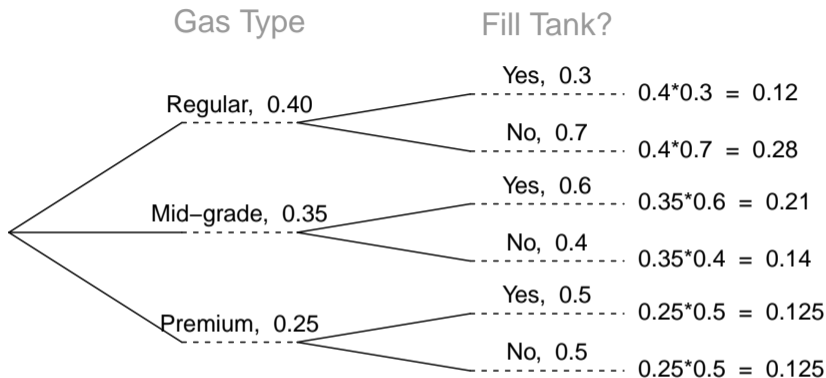
$$\begin{aligned} P(B) &= P(A_1)P(B | A_1) + P(A_2)P(B | A_2) + P(A_3)P(B | A_3) \\ &= 0.4 \times 0.3 + 0.35 \times 0.6 + 0.25 \times 0.5 = 0.455 \end{aligned}$$

Gas Station Example — Tree Diagram



Q3: If the next customer fills the tank, what is the probability that premium gas is requested?

Gas Station Example — Tree Diagram



Q3: If the next customer fills the tank, what is the probability that premium gas is requested?

$$P(A_3 | B) = \frac{P(A_3 \cap B)}{P(B)} = \frac{0.125}{0.455} \approx 0.275.$$

Example — Law of Total Probability

If one rolls a pair of 6-face fair dice continuously until the sum of the two dice is a 5 or a 7, what is the probability that a sum of 5 appears before a sum of 7?

Sol: Let $A = \{\text{a sum of 5 appears before a sum of 7}\}$ and

$$S_5 = \{\text{first roll} = 5\}, \quad S_7 = \{\text{first roll} = 7\}, \quad S_o = \{\text{first roll is not 5 or 7}\}.$$

Note S_5 , S_7 and S_o are disjoint and $S_5 \cup S_7 \cup S_o = S$ is the sample space. Moreover,

$$P(S_5) = \frac{4}{36}, \quad P(S_7) = \frac{6}{36}, \quad P(S_o) = 1 - P(S_5) - P(S_7) = \frac{26}{36}.$$

By the Law of Total Probability,

$$\begin{aligned} P(A) &= P(S_5)P(A | S_5) + P(S_7)P(A | S_7) + P(S_o)P(A | S_o) \\ &= \end{aligned}$$

Independence

Independence

Two events A and B are said to be **independent** if any of the following is true

- ▶ $P(A | B) = P(A)$ B happens doesn't affect how likely A happens
- ▶ $P(A | B) = P(A | B^c)$
..... How likely A happens is not affected by B happens or not
- ▶ $P(B | A) = P(B)$ A happens doesn't affect how likely B happens
- ▶ $P(A \cap B) = P(A) \times P(B)$

If any of the identities above is true, then all remaining identities will also be true.

Proof of $P(A | B) = P(A)$ implies $P(B | A) = P(B)$

$$\begin{aligned}P(B | A) &= \frac{P(A \cap B)}{P(A)} && \text{definition of conditional prob.} \\ &= \frac{P(B)P(A | B)}{P(A)} && \text{Multiplication Rule} \\ &= \frac{P(B)P(A)}{P(A)} && \text{since } P(A | B) = P(A) \\ &= P(B)\end{aligned}$$

Thus, $P(A | B) = P(A)$ implies $P(B | A) = P(B)$.

Proof of $P(B | A) = P(B)$ implies $P(A \cap B) = P(A)P(B)$

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

(by Multiplication Rule)
(since $P(B | A) = P(B)$)

Independent Events vs Disjoint Events

- ▶ If A and B are independent, $P(A \cap B) = P(A) \times P(B)$.
- ▶ If A and B are disjoint: $A \cap B = \emptyset \Rightarrow P(A \cap B) = 0$.
- ▶ If $P(A) > 0$ and $P(B) > 0$,
 - ▶ Independent events **cannot** be disjoint.
 - ▶ Disjoint events **cannot** be independent.
- ▶ Conceptually, A and B are disjoint means that one happens prevents the other from happening, so one's occurrence definitely affects the other's.

Examples (Independent or not)

Rolling a black die and the white die, both 6-face and fair. Define

$$B_6 = \{\text{black die shows 6}\} \rightarrow P(B_6) = 1/6$$

$$W_6 = \{\text{white die shows 6}\} \rightarrow P(W_6) = 1/6$$

$$T_6 = \{\text{total is 6}\} \rightarrow P(T_6) = 5/36$$

$$T_7 = \{\text{total is 7}\} \rightarrow P(T_7) = 6/36 = 1/6$$

- Are B_6 and W_6 independent?
- Are B_6 and T_6 independent?
- Are B_6 and T_7 independent?

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- Are B_6 and W_6 independent? **Yes, as they involve different dice.**
- Are B_6 and T_6 independent?
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- Are B_6 and W_6 independent? **Yes, as they involve different dice.**
- Are B_6 and T_6 independent?
No. As $B_6 \cap T_6 = \emptyset$, $P(B_6 \cap T_6) = 0 \neq P(B_6)P(T_6)$
- Are B_6 and T_7 independent?

Examples (Independent or not)

Rolling a black die and the white die, both 6-face and fair. Define

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a. Are B_6 and W_6 independent? **Yes, as they involve different dice.**

b. Are B_6 and T_6 independent?

No. As $B_6 \cap T_6 = \emptyset$, $P(B_6 \cap T_6) = 0 \neq P(B_6)P(T_6)$

c. Are B_6 and T_7 independent?

Yes, since

$$P(B_6 \cap T_7) = P(\text{black die} = 6, \text{white die} = 1) = \frac{1}{36} = P(B_6)P(T_7) = \frac{1}{6} \cdot \frac{1}{6}.$$

Independence of 3 Events

Three events E , F , and G are said to be *independent* if **ALL** of the following are true

$$P(EFG) = P(E)P(F)P(G)$$

$$P(EF) = P(E)P(F)$$

$$P(EG) = P(E)P(G)$$

$$P(FG) = P(F)P(G)$$

Pairwise Independence \neq Independence

Rolling a black die and the white die, both 6-face and fair. Define

$$B_6 = \{\text{black die shows 6}\}$$

$$W_6 = \{\text{white die shows 6}\}$$

$$T_7 = \{\text{total is 7}\}$$

Two pages ago, we just showed

- ▶ the independence of B_6 and W_6
- ▶ the independence of B_6 and T_7
- ▶ the independence of W_6 and T_7 can be shown similarly as B_6 and T_7

However, B_6, W_6 and T_7 are NOT independence since $B_6 \cap W_6 \cap T_7 = \emptyset$,

$$P(B_6 \cap W_6 \cap T_7) = 0 \neq P(B_6)P(W_6)P(T_7) = \frac{1}{6} \cdot \frac{1}{6} \cdot \frac{1}{6}.$$

Properties of Independence

If E , F , and G are independent, then E will be independent of **any event formed by F and G** , like

$$F \cup G, \quad F^c \cap G, \quad F \cup G^c, \quad \text{etc.}$$

For example, let's prove the independence of E and $F \cup G$.

$$\begin{aligned} P[E(F \cup G)] &= P(EF \cup EG) && \text{(Distributive law)} \\ &= P(EF) + P(EG) - P(EFG) && \text{(Inclusion-Exclusion formula)} \\ &= P(E)P(F) + P(E)P(G) - P(E)P(FG) && \text{(Definition of Independence)} \\ &= P(E)[P(F) + P(G) - P(FG)] \\ &= P(E)P(F \cup G) && \text{(Inclusion-Exclusion formula)} \end{aligned}$$

Independence of n Events

The events E_1, E_2, \dots, E_n are said to be *independent* if for every subset $E_{i_1}, E_{i_2}, \dots, E_{i_r}$, for $1 \leq i_1 < i_2 < \dots < i_r \leq n$ of these events, we have

$$P(E_{i_1} E_{i_2} \dots E_{i_r}) = P(E_{i_1})P(E_{i_2}) \dots P(E_{i_r}).$$

Finally, we define an **infinite set** of events to be *independent* if every finite subset of those events is independent.

Example — 5 or 7 First (Revisit)

If one rolls a pair of 6-face fair dice continuously until the sum of the two dice is a 5 or a 7, what is the probability that a sum of 5 appears before a sum of 7?

Sol. On p.13, we solved using the Law of Total Prob. Here is another approach. Let

$$E_n = \{\text{no 5 or 7 appears on the first } n - 1 \text{ rolls, } n\text{th roll is 5}\}$$

$$A_k = \{k\text{th roll is neither 5 nor 7}\}$$

$$B_n = \{n\text{th roll is 5}\}$$

- ▶ $\{\text{5 appears before 7}\} = \cup_{n=1}^{\infty} E_n$. Are E_n 's disjoint?
- ▶ $E_n = A_1 A_2 \dots A_{n-1} B_n$. Are $A_1, A_2, \dots, A_{n-1}, B_n$ independent?

$$P(A_i) = \quad , P(B_j) =$$

$$P(E_n) = P(A_1)P(A_2) \dots P(A_{n-1})P(B_n) = \left(\frac{26}{36}\right)^{n-1} \frac{4}{36}$$

$$P(\text{5 before 7}) = P(\cup_{n=1}^{\infty} E_n) = \sum_{n=1}^{\infty} P(E_n) = \sum_{n=1}^{\infty} \left(\frac{26}{36}\right)^{n-1} \frac{4}{36} = \frac{4}{36} \frac{1}{1 - \frac{26}{36}} = \frac{2}{5}.$$