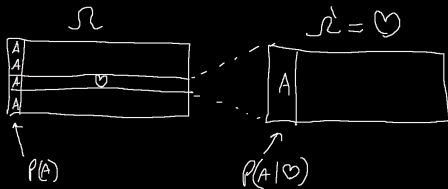


Return to the universe of the 52-card deck and draw one card.

$$P(A|B) = \frac{4}{52} = \frac{1}{13}, \quad P(A|\heartsuit) = \frac{1}{13} = \frac{P(A, \heartsuit)}{P(\heartsuit)} = \frac{\frac{1}{52}}{\frac{13}{52}}$$

Did providing the information that the card was a heart change the probability? NO. Knowing it was a heart was "informationally irrelevant" or "probabilistic independence".



The heart event and the ace event are "independent" events.

$$P(\text{IBM stock goes up today} \mid \text{raining in DC}) = P(\text{IBM stock goes up today})$$

If two events A, B are independent then: $P(A \mid B) = P(A)$, $P(B \mid A) = P(B)$

$$\Rightarrow P(A|B) = \frac{P(A, B)}{P(B)} \stackrel{!}{=} P(A) \Rightarrow P(A, B) = P(A)P(B)$$

the "multiplication rule"

If you flip 5 coins. What is the probability of all heads? old

$$P(H_1, H_2, H_3, H_4, H_5) = P(H_1) \cdot \dots \cdot P(H_5) = \left(\frac{1}{2}\right)^5 = \frac{1}{\sqrt{2^5}} = \frac{1}{101^5} = \frac{1}{2^5}$$

We will now see if Chevalier de Mere's gambling conjecture is correct. He claimed that $P(\text{one or more double 6 in 24 dice rolls}) < 1/2$.

$$\begin{aligned}
 &= P(1 \text{ 6-6}) + P(2 \text{ 6-6's}) + \dots + P(24 \text{ 6-6's}) && \text{due to [A3]} \\
 &= 1 - P(\text{no 6-6's}) && \text{complement rule} \\
 &= 1 - P(\text{roll 1 is not 6-6, roll 2 is not 6-6, ..., roll 24 is not 6-6}) \\
 &= 1 - P(\text{roll 1 is not 6-6}) P(\text{roll 2 is not 6-6}) * \dots * P(\text{roll 24 is not 6-6}) && \text{mult rule} \\
 &= 1 - (1 - P(6,6))^{24} && \text{complement rule, same event} \\
 &= 1 - (1 - P(6)P(6))^{24} && \text{mult rule} \\
 &= 1 - (1 - 1/6 * 1/6)^{24} = 1 - (1 - 1/36)^{24} = 1 - (35/36)^{24} = 0.4914 < 0.5
 \end{aligned}$$

If $P(A \mid B) \neq P(A)$ then A, B and "not independent" i.e. "dependent".

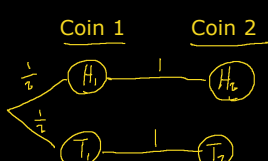
$$P(\text{dying of SARS-Cov-2} \mid \text{vaccine}) < P(\text{dying of SARS-Cov-2})$$

If you can prove this... then the two events are dependent AKA "informationally relevant".

$$P(\text{Q64 is late} \mid \text{snowstorm}) > P(\text{Q64 is late})$$

$$P(\text{lung cancer} \mid \text{smoking}) > P(\text{lung cancer})$$

Consider two magic coins. You flip them together and they always come up the same.



Prove that heads in coin 1 and heads in coin 2 are dependent.

$$1 = P(H_2 \mid H_1) \neq P(H_2) = \frac{1}{2}$$

Let's do some famous fun problems. First is the "birthday problem". What is the probability that at least one pair of people share a birthday in this class? Intuitively, there are 365 days and only 22 people so it's probably low.

$$\begin{aligned}
 &= P(\text{at least one birthday pair in 22}) \\
 &= P(1 \text{ pair}) + P(2 \text{ pairs}) + \dots + P(22 \text{ pairs}) \quad [\text{A3}] \\
 &= 1 - P(\text{no pairs}) \\
 &= 1 - 365/365 * 364/365 * 363/365 * \dots * 344/365 \\
 &= 1 - \frac{365 P_{22}}{365^{22}} = 1 - \frac{365!}{365^{22} (365-22)!} = 1 - .524 = \boxed{.476}
 \end{aligned}$$

$$P(\text{at least one birthday pair in 60}) = 1 - \frac{365 P_{60}}{365^{60}} = 99.5\%$$

Another famous problem is the "hat problem". This is difficult. n people walk into a room and toss their hats in a pile. They each then grab a random hat. What is the probability at least one person gets their own hat?

$$= P(1 \text{ person gets hat}) + P(2 \text{ people get hat}) + \dots + P(n \text{ people get hat})$$

Let $A_i :=$ event i'th person gets their hat

$$= P(A_1 \cup A_2 \cup \dots \cup A_n)$$

Then using the inclusion-exclusion formula,

$$= \sum P(A_i) - \sum P(A_i \cap A_j) + \sum P(A_i \cap A_j \cap A_k) - \dots + \dots + \dots$$

$$P(A_1) = \frac{1}{1} \cdot \frac{n-1}{2} \cdot \frac{n-2}{3} \cdot \dots \cdot \frac{1}{n} \Rightarrow \frac{(n-1)!}{n!} = \frac{1}{n}$$

$$P(A_2) = \frac{n-1}{1} \cdot \frac{1}{2} \cdot \frac{n-2}{3} \cdot \dots \cdot \frac{1}{n} \Rightarrow \frac{(n-1)!}{n!} = \frac{1}{n}$$

$$\Rightarrow P(A_i) = \frac{1}{n}$$

$$P(A_1 \cap A_2) = \frac{1}{1} \cdot \frac{1}{2} \cdot \frac{n-2}{3} \cdot \dots \cdot \frac{1}{n} \Rightarrow \frac{(n-2)!}{n!} = P(A_i \cap A_j)$$

$$P(A_i \cap A_j \cap A_k) = \frac{(n-3)!}{n!}$$

$$= \sum P(A_i) - \sum P(A_i \cap A_j) + \sum P(A_i \cap A_j \cap A_k)$$

$$\sum \frac{1}{n} - \sum \frac{(n-2)!}{n!} + \sum \frac{(n-3)!}{n!}$$

$$\frac{n \cdot \frac{1}{n}}{1} - \frac{\binom{n}{2} \frac{(n-2)!}{n!}}{1} + \frac{\binom{n}{3} \frac{(n-3)!}{n!}}{1} = \frac{1}{1!} - \frac{1}{2!} + \frac{1}{3!} - \frac{1}{4!} + \dots$$

$$= \frac{1}{1!} - \frac{1}{2!} + \frac{1}{3!} - \frac{1}{4!} + \dots = \sum_{i=1}^n \frac{(-1)^{i+1}}{i!} \approx \sum_{i=1}^{\infty} \frac{(-1)^{i+1}}{i!}$$

Taylor series

$$= 1 - \frac{1}{e} \approx \boxed{67\%}$$