

**A. Conditional Probability.** For example, let  $\Omega = \{\text{population}\}$ .

$H = [\text{event of heart disease}] \rightarrow P(H) = 5\%$

$O = [\text{event of over 75 years old}] \rightarrow P(O) = 7\%$

$O$  stands for old.

$HO = [\text{heart disease and over 75 years old}] \rightarrow P(HO) = 1.8\%$

Given that your patient is over 75 years old, what is the probability they have heart disease?

Yes, you are a doctor now.  
Congratulations.

$P(\text{heart disease given over 75 years old}) = P(H | O)$

**Conditional Probability.** If  $E$  and  $F$  are events, and  $F$  occurs with positive probability, then the **conditional probability** of  $E$  given  $F$  is:

$P(E | F) =$

If the probability of event  $F$  is zero, the conditional probability is treated as finite but undefined, basically meaning in practice that, if you multiply the conditional probability by  $0$ , even if the conditional probability is undefined, you may treat the result as  $0$ .

You can think of the conditional probability as what happens if you shrink your sample space  $\Omega$  to be the event  $F$ . In the example above, we shrunk from the general population, to those people who are really old.

The ordinary probability space utilizes the probability function  $P(\cdot)$ . However, if we have observed event  $F$ , then the way we compute probability changes: our new probability function is  $P(\cdot | F)$  which we call a **conditioned probability**. Importantly, this new probability function defines a probability space, meaning all the usual axioms are satisfied.

**Conditioned Probability Function.** If  $F$  is an event that occurs with positive probability, then the **conditioned probability** given  $F$  is  $P(\cdot | F)$ . This defines the **conditioned probability space** given  $F$ , which is in fact a probability space.

See the textbook for a proof that it is a probability space, i.e. that it satisfies all the axioms.

**Example 1.** Joe lost his key. He thinks the key is in his pants with probability  $\frac{1}{3}$ , in his jacket with probability  $\frac{1}{2}$ , and elsewhere with probability  $\frac{1}{6}$ . Joe just checked his jacket, and did not find the key. What is the probability that the key is in his pants?

**B. Multiplication Rule.** For events  $E$  and  $F$  we have:

$$\mathbb{P}(E | F) = \frac{\mathbb{P}(EF)}{\mathbb{P}(F)} \implies \mathbb{P}(EF) =$$

We can iterate this equality to obtain an identity for the probability of the intersection of multiple events.

**Multiplication Rule.** For events  $E_1, E_2, \dots, E_n$ , we have:

$$\mathbb{P}(E_1 E_2 \cdots E_n) =$$

**Example 2.** Recall the matching hats problem: Suppose that  $n$  people throw their hats into the center of a room, then the hats are mixed and returned to people uniformly at random.

We had calculated the probability that no one gets their hat back is:

$$\mathbb{P}(n\text{-derangement}) = \sum_{r=0}^n \frac{(-1)^r}{r!}$$

What is the probability that exactly  $k$  people get their hat back?

Recall: a **derangement** is a permutation where no object ends up back in its original place.

C. **Law of Total Probability.** For events  $E$  and  $F$ , since  $\Omega = F \sqcup F^c$ , we have:

$E =$

Therefore, if the event  $F$  has positive probability:

$P(E) =$

**Law of Total Probability.**

If  $\Omega = F_1 \sqcup F_2 \sqcup \dots$  expresses the sample space as a countable **disjoint** union, or, in other words, if the events  $F_i$  are mutually exclusive and their total probability is 1, then:

$P(E) =$

As an important special case, for an event  $F$  with positive probability:

$P(E) = P(E | F)P(F) + P(E | F^c)P(F^c)$

**Example 3.** A blood test detects a certain disease, when present, with probability 95%. However, the test gives a “false positive” with probability 1%. In the population, 0.5% of people have the disease.

What is the probability a randomly selected person from the population tests positive?