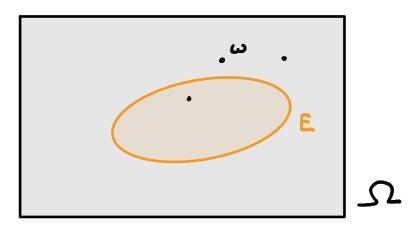
M378K: January 15th 2025.

M378K Introduction to Mathematical Statistics Problem Set #1 Probability spaces.

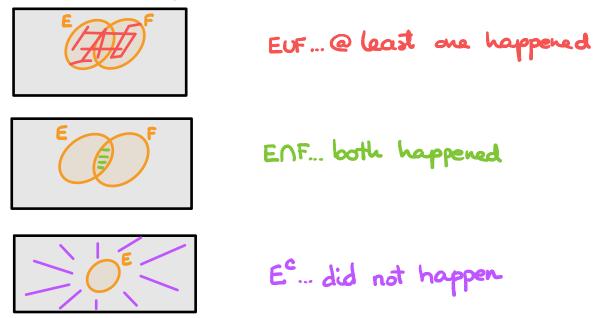
1.1. Probability distributions. Consider an outcome space (also known as a sample space) Ω . In statistics, it's convenient to imagine it as containing your entire target population. The individual elements $\omega \in \Omega$ are known in probability as elementary outcomes; in statistics, they can be understood as individuals in your target population.

We are usually not interested that much in individual ω , but want to consider **events** E in Ω . In full mathematical generality, the set Ω can be complicated so that the family of all of its subsets is also a complicated object. Due to mathematical difficulties, it is sometimes impossible to measure in any meaningful way absolutely all subsets of Ω ¹. For the purposes of this course, we will not have to dwell on these matters. So, we will refer to any (nice) subset of Ω as an **event**.



¹See https://en.wikipedia.org/wiki/Banach\T1\textendashTarski_paradox

We will inherit all the notation and rules from set theory while adding some more vocabulary and context. Most notably, we will consider *intersections*, *unions*, and *complements* of events. These are best understood via Venn diagrams.



Moreover, in a probabilistic setting, we have the following definition:

Definition 1.1. Let E and F be two events on the same Ω such that

$$E \cap F = \emptyset$$
.

Then, we say that E and F are mutually exclusive (or disjoint).



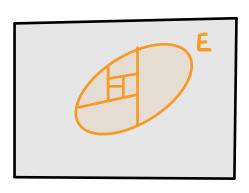
Now, we are ready for the following (crucial!) definition:

Definition 1.2. Consider a mapping \mathbb{P} from the set of all events on Ω to \mathbb{R} . We say that \mathbb{P} is a probability (distribution, measure) if it satisfies the following three conditions:

- $\mathbb{P}[E] \ge 0$ for all events E;
- $\bullet \ \mathbb{P}[\Omega] = 1;$
- ullet for all pairwise disjoint sequences of events $\{E_j: j=1,2,\dots\}$, we have that

$$\mathbb{P}\left[\bigcup_{j=1}^{\infty} E_j\right] = \sum_{j=1}^{\infty} \mathbb{P}[E_j].$$

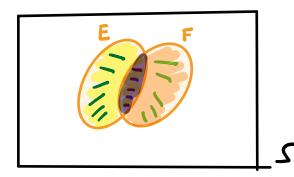




One immediately useful consequence is the **inclusion-exclusion formula**. This is its simplest version.

Proposition 1.3. Let E and F be two events on Ω . Then,

$$\mathbb{P}[E \cup F] = \mathbb{P}[E] + \mathbb{P}[F] - \mathbb{P}[E \cap F].$$

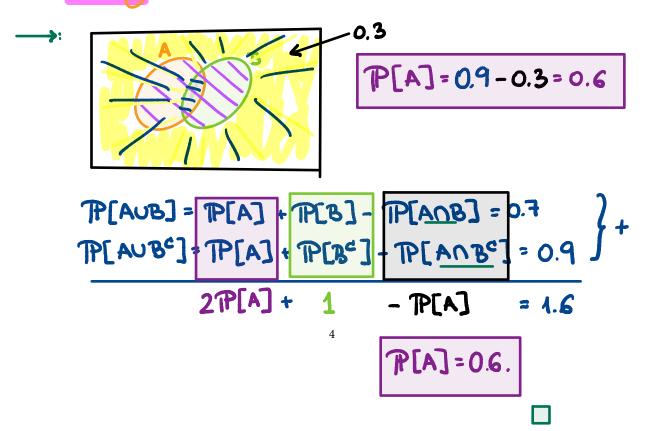


Of course, the above formula can be generalized to arbitrary unions of finitely many events. *Try to figure it out!*

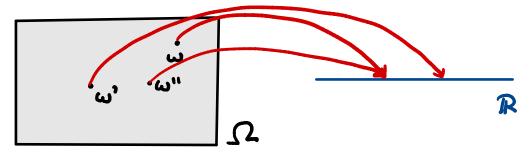
Problem 1.1. Source: An old P exam problem. For two events A and B, you are given that

$$\mathbb{P}[A \cup B] = 0.7 \quad and \quad \mathbb{P}[A \cup B^c] = 0.9$$

Calculate $\mathbb{F}[A]$



1.2. **Random variables.** Informally speaking, any "nice" mapping/function from Ω to a target set S is a *random element* 2 . When S is \mathbb{R} , we like to use the term *random variable*. When S is \mathbb{R}^n for some n, we like to use the term *random vector*.



Let's consider a classroom of students as our Ω and give examples of a

• random element

• random variable

GPA

• random vector

To keep track of what values a random variable is "allowed" we use the following terminology3:

Definition 1.4. Given a set B, we say that a random variable Y is B—valued if

$$\mathbb{P}[Y \in B] = 1.$$

 $^{^{2}}$ In practice, people like to use the term $random\ variable$ even in more general context when there is no source of confusion. We will habitually do this.

³Read your lecture notes: https://web.ma.utexas.edu/users/gordanz/notes/discrete_probability_color.pdf