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Axiom 2: For a set $A \subseteq \Omega$ we have $0 \leq \mathbb{P}(A) \leq 1$.

Axiom 3: For a disjoint collection of events A_1, A_2, \dots (where $A_i \subseteq \Omega$)

$$\mathbb{P} \left(\bigcup_{i=1}^{\infty} A_i \right) = \sum_{i=1}^{\infty} \mathbb{P}(A_i)$$

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- ▶ $\mathcal{P}(\Omega)$? Recall $\mathcal{P}(\Omega) = \{A : A \subseteq \Omega\}$. Seems like a great choice!

Towards a formal definition of \mathbb{P}

Probability measure \mathbb{P} can be defined as a set-function $\mathbb{P} : \mathcal{P}(\Omega) \rightarrow [0, 1]$ that satisfies the following 3 axioms.

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- ▶ Is there a perceivable problem with this definition?
- ▶ The following counter-example will construct a set-function \mathbb{P} for which you cannot assign valid probabilities to every subsets in Ω without violating these axioms.

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- ▶ $\mathcal{P}(\mathbb{R})$ is unimaginably complex!

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- ▶ $\mathcal{F} = \{\emptyset, \Omega, \mathbb{R}_-, \mathbb{R}_+\}$. This is all that this \mathbb{P} can measure!

Towards sigma-algebra

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- ▶ For example, if $B \in \mathcal{F}$, then $B^c \in \mathcal{F}$. Also, \emptyset and Ω in \mathcal{F} .
- ▶ A domain with such nice properties is called as a *sigma-algebra*.

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When Ω is countable and finite, we will consider power-set $\mathcal{P}(\Omega)$ as the domain.

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- ▶ For a collection of sets $\mathcal{C} \subset 2^\Omega$, the σ -algebra generated by \mathcal{C} , denoted by $\sigma(\mathcal{C})$ is the smallest σ -algebra containing \mathcal{C} . Here $\sigma(\mathcal{C}) = \{\cap \mathcal{F} : \mathcal{C} \subseteq \mathcal{F}\}$.

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- The trio $(\Omega, \mathcal{F}, \mathbb{P})$ is called as a probability space.

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- ▶ Recall that when $|\Omega| < \infty$, we consider $\mathcal{F} = 2^{\Omega}$.

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Probability space for $U[0, 1]$

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- ▶ Lets call it $\mathcal{F}^{++} =$
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- ▶ Continuing on these lines, the resulting sigma algebra is called a borel-sigma algebra $\mathcal{B}[0, 1]$.

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- ▶ How would you define $\mathcal{B}(\mathbb{R}^2)$?

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HW
- ▶ What is $P(A \cup B \cup C)$?
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- ▶ $\bigcup_{\omega \in \Omega} \{\omega\}$ is an uncountable disjoint union!