# **18.175:** Lecture 1 Probability spaces and $\sigma$ -algebras

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#### Probability spaces and $\sigma$ -algebras

Distributions on  $\ensuremath{\mathbb{R}}$ 

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- ► continuity from above: measures of sets A<sub>i</sub> in decreasing sequence converge to measure of intersection ∩<sub>i</sub>A<sub>i</sub>

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- Thus  $[0,1) = \bigcup \tau_r(A)$  as r ranges over rationals in [0,1).
- ▶ If P(A) = 0, then  $P(S) = \sum_{r} P(\tau_r(A)) = 0$ . If P(A) > 0 then  $P(S) = \sum_{r} P(\tau_r(A)) = \infty$ . Contradicts P(S) = 1 axiom.

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- Most mainstream probability and analysis takes the third approach. But good to be aware of alternatives (e.g., axiom of determinacy which implies that all sets are Lebesgue measurable).

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- Why does this notion make sense? If F<sub>i</sub> are σ-fields (for i in possibly uncountable index set I) does this imply that ∩<sub>i∈I</sub>F<sub>i</sub> is a σ-field?

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# Can we classify set of all probability measures on $\mathbb{R}$ ?

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- ► Theorem: for each right continuous, non-decreasing function F, tending to 0 at -∞ and to 1 at ∞, there is a unique measure defined on the Borel sets of ℝ with P((a, b]) = F(b) F(a).