The Set Structure of Precision

Rabanus Derr, Robert C. Williamson

The "Foundations of Machine Learning Systems" Group

http://fm.ls





Led by Robert C. Williamson

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Why does ISIPTA take place?

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Imprecise Probability

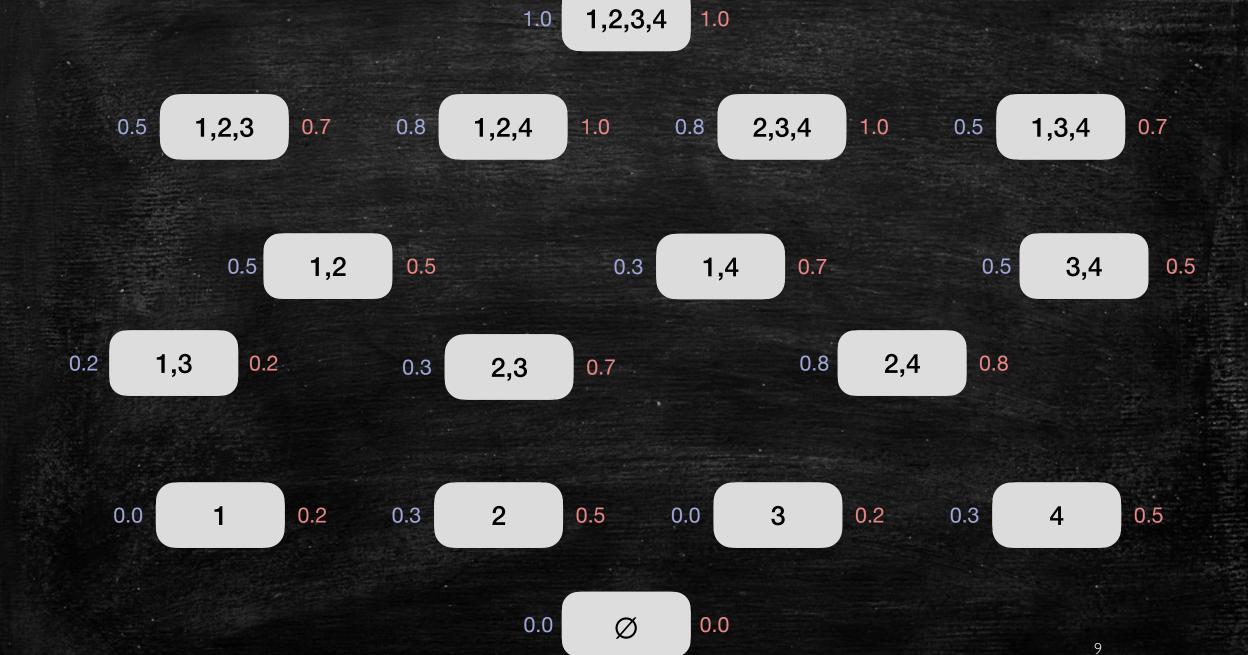
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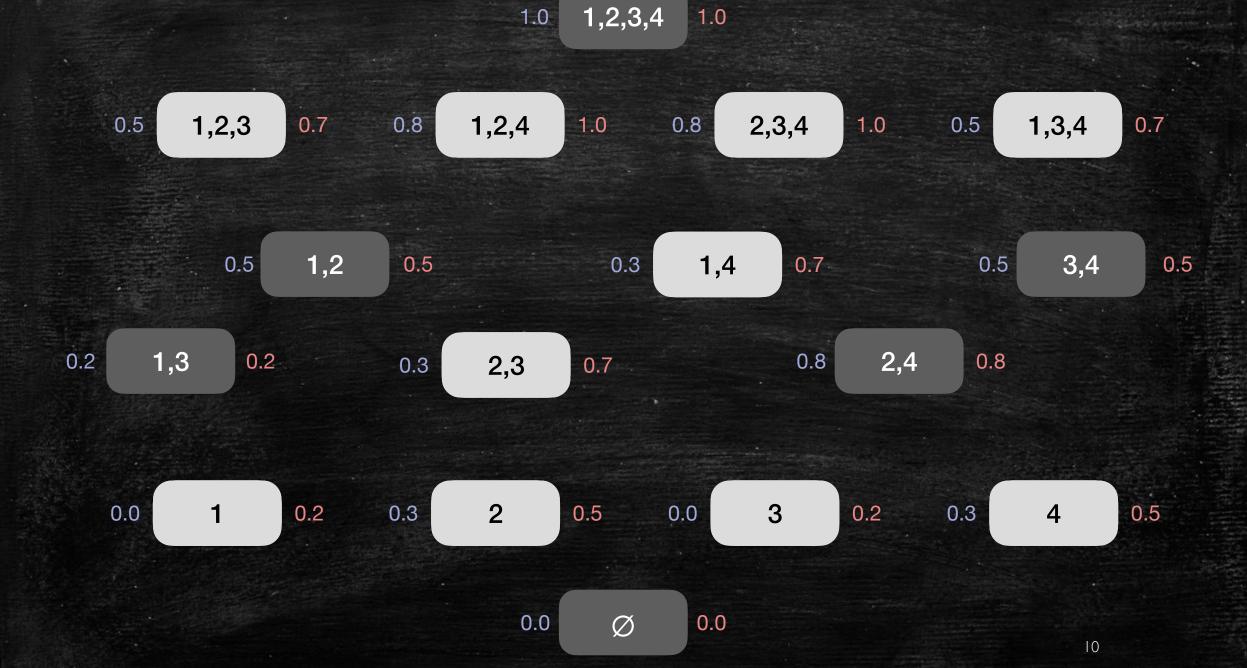
Imprecise Probability

On which events are imprecise probabilities precise?

1,2,3,4 2,3,4 1,2,3 1,2,4 1,3,4 1,2 3,4 1,4 2,4 1,3 2,3 2 3 4

Ø





On which events are imprecise probabilities precise?

A. Normalization Imprecise Probability B. Conjugacy C. Subadditivity D. Superadditivity

=>(Pre-)Dynkin-Systems

$$\mathcal{D} \subseteq 2^{\{1,2,3,4\}}$$

$$\mathcal{D} \subseteq 2^{\{1,2,3,4\}} \quad \emptyset$$

 $I. \emptyset \in \mathcal{D}$

$$\mathcal{D} \subseteq 2^{\{1,2,3,4\}}$$



- $I. \emptyset \in \mathcal{D}$
- $2. A \in \mathcal{D}$

$$\mathcal{D} \subseteq 2^{\{1,2,3,4\}}$$

1.
$$\emptyset \in \mathcal{D}$$

2. $A \in \mathcal{D} \implies A^c \in \mathcal{D}$

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$$I. \emptyset \in \mathcal{D}$$

$$2. A \in \mathcal{D} \implies A^c \in \mathcal{D}$$

3.
$$A, B \in \mathcal{D}, A \cap B = \emptyset \implies A \cup B \in \mathcal{D}$$

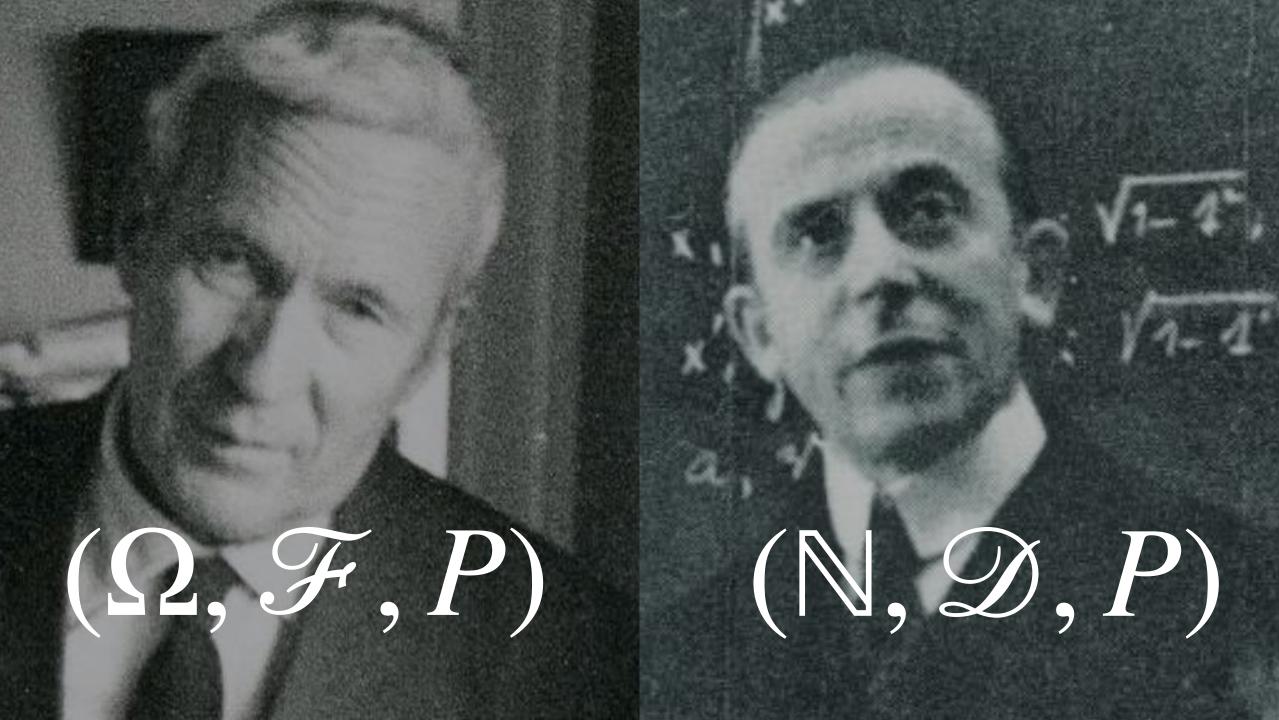
1.
$$\emptyset \in \mathcal{D}$$

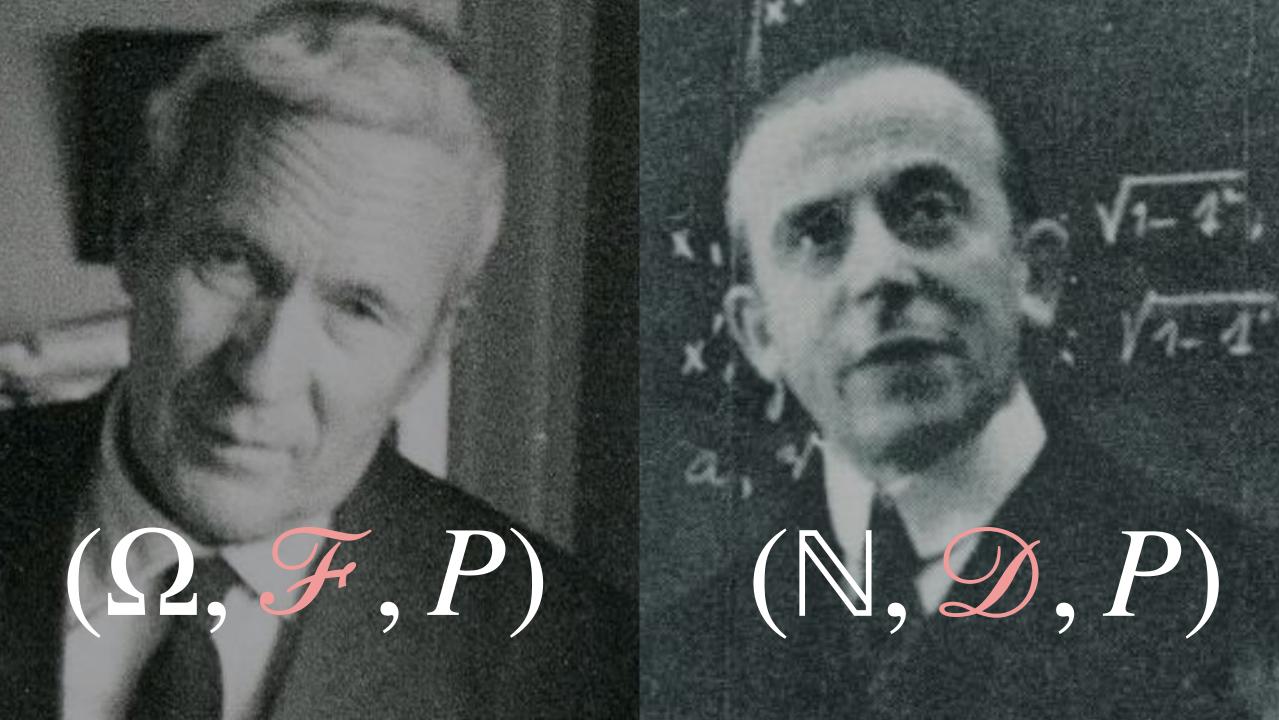
2. $A \in \mathcal{D} \implies A^c \in \mathcal{D}$
3. $A, B \in \mathcal{D}, A \cap B = \emptyset \implies A \cup B \in \mathcal{D}$

How we actually started...

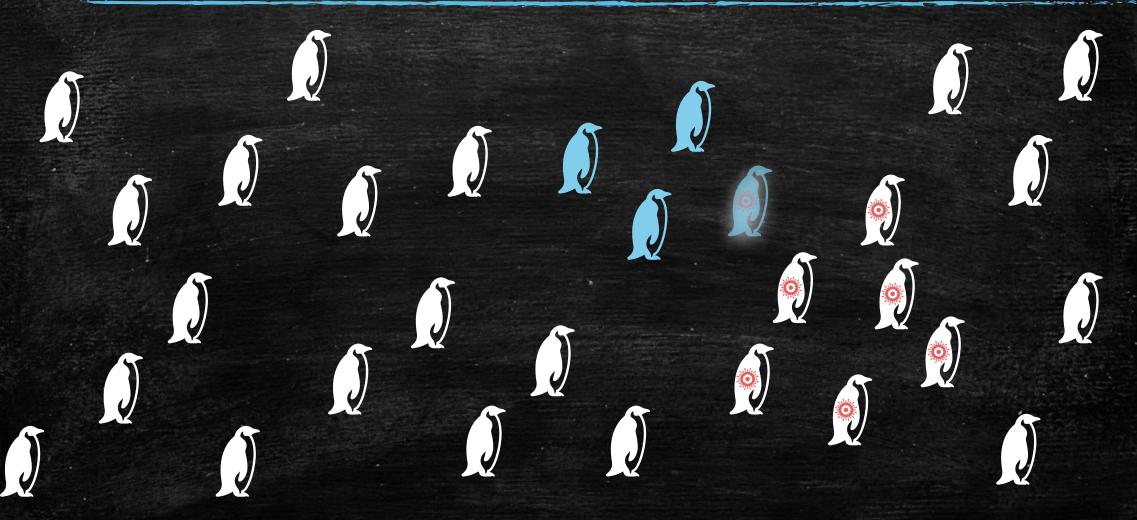
The same story told from a different start



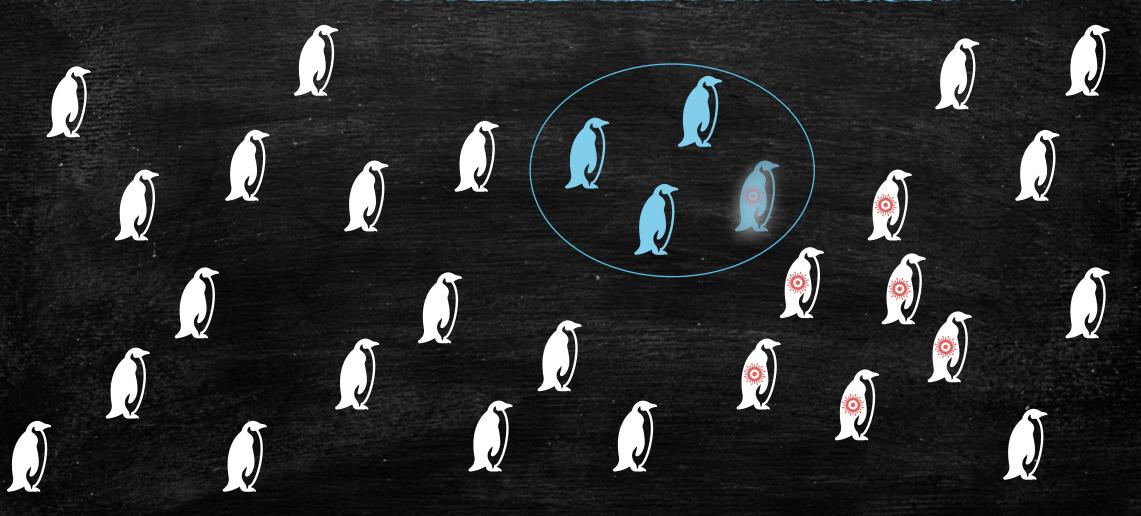




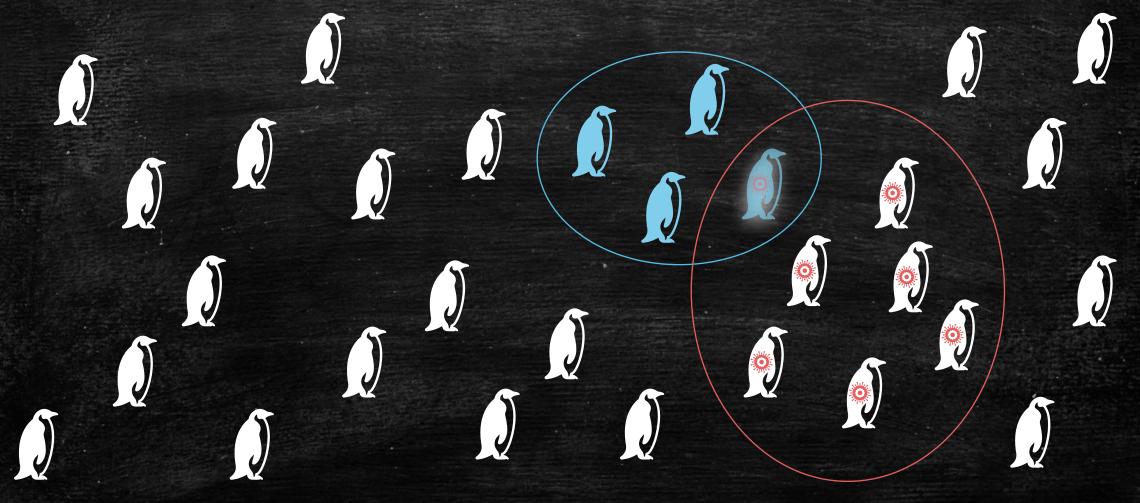
Unmeasurable Penguin Colonies



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Discrete Chebyshev Classifiers

Elad Eban* Elad Mezuman[†] Amir Globerson* ELADE@CS.HUJI.AC.IL ELAD.MEZUMAN@MAIL.HUJI.AC.IL GAMIR@CS.HUJI.AC.IL

†Edmond and Lily Safra Center for Brain Sciences. The Hebrew University of Jerusalem

*The Selim and Rachel Benin School of Computer Science and Engineering. The Hebrew University of Jerusalem

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Coherent risk measures induced by partially specified probabilities

Ehud Lehrer*

December 6, 2007

A preliminary draft

Discrete Chebyshev Classifiers

Elad Eban*
Elad Mezuman†
Amir Glo

ELADE@CS.HUJI.AC.IL ELAD.MEZUMAN@MAIL.HUJI.AC.IL

†Edmond *The Seli

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SUBJECTIVE PROBABILITIES ON SUBJECTIVELY UNAMBIGUOUS EVENTS

By Larry G. Epstein and Jiankang Zhang¹

This paper suggests a behavioral definition of (subjective) ambiguity in an abstract setting where objects of choice are Savage-style acts. Then axioms are described that deliver probabilistic sophistication of preference on the set of unambiguous acts. In particular, both the domain and the values of the decision-maker's probability measure are derived from preference. It is argued that the noted result also provides a decision-theoretic foundation for the Knightian distinction between risk and ambiguity.

KEYWORDS: Ambiguity, Knightian uncertainty, subjective probability, probabilistic sophistication. Coherent risk measures induced by partially specified probabilities

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QUANTUM PROBABILITY SPACES

STANLEY P. GUDDER

1. Introduction. In [5] P. Suppes introduced the notion of a quantum probability space. He noted that such spaces may be used to describe the position and momentum of a quantum mechanical particle but cannot be used for more general systems. This author has considered quantum probability spaces not only because they are an interesting example of a nonclassical logic but because quantum mechanical phenomena are seen to develop in a quite transparent fashion in this case.

bV

partially specified probabilities

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partially specified probabilities

Ehud Lehrer*

G. SCHURZ

H. LEITGEB

Finitistic and Frequentistic

Approximation of

Probability Measures with or

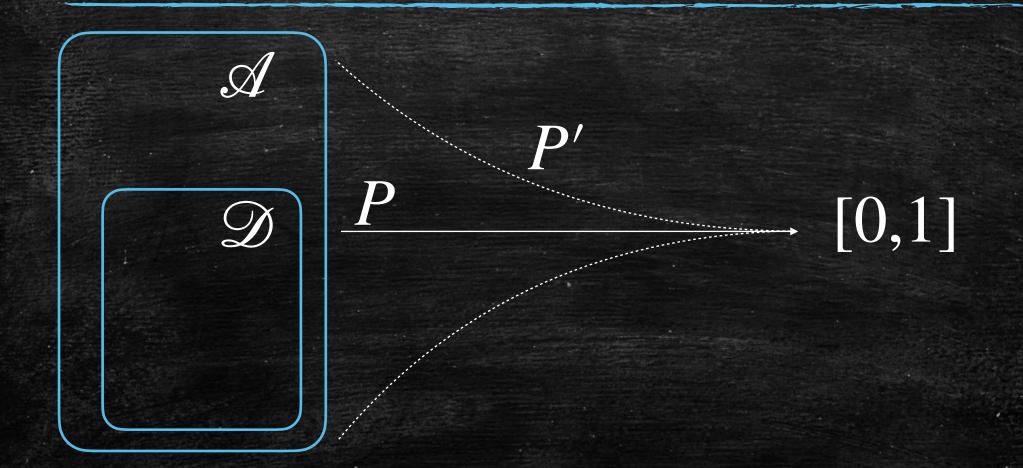
without σ -Additivity

Better be Extendable

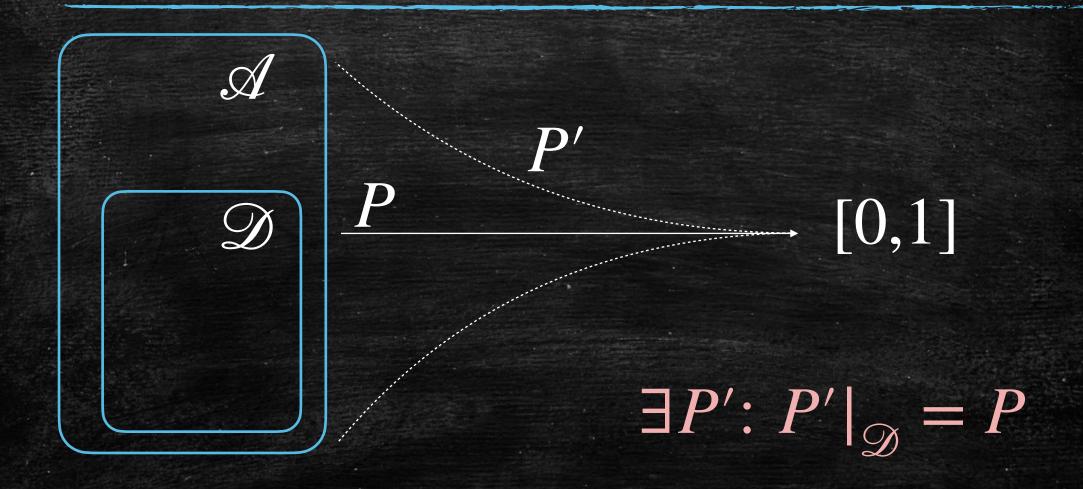


 $P \longrightarrow [0,1]$

Better be Extendable



Better be Extendable



Extendability = Coherence



Imprecise Probability on Set Algebra

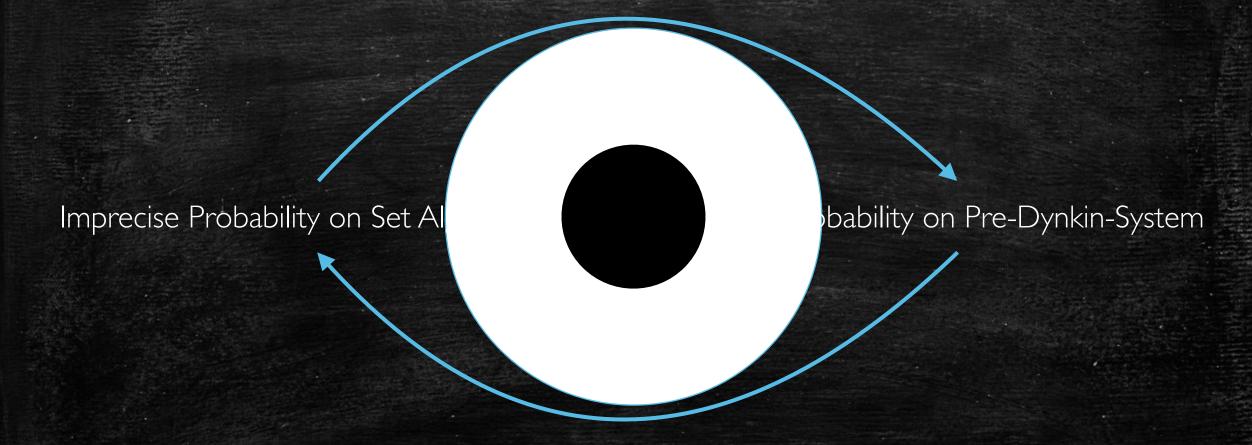
Probability on Pre-Dynkin-System

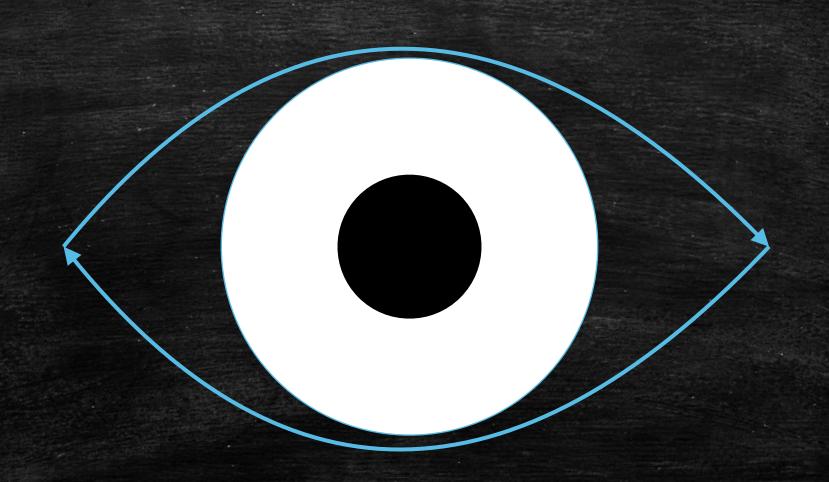
Imprecise Probability on Set Algebra

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