





# STOCHASTIC EVENTS

JAN C. WILLEMS

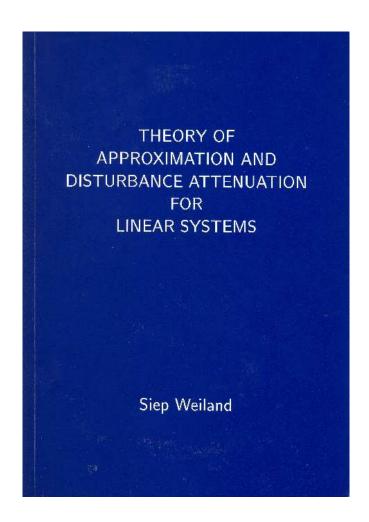
K.U. Leuven, Flanders, Belgium

On the occasion of the inaugural lecture of Siep Weiland



In honor of Siep Weiland on the occasion of his inauguration

# **Dissertation**



Dept. of Mathematics, University of Groningen, Jan. 4, 1991.

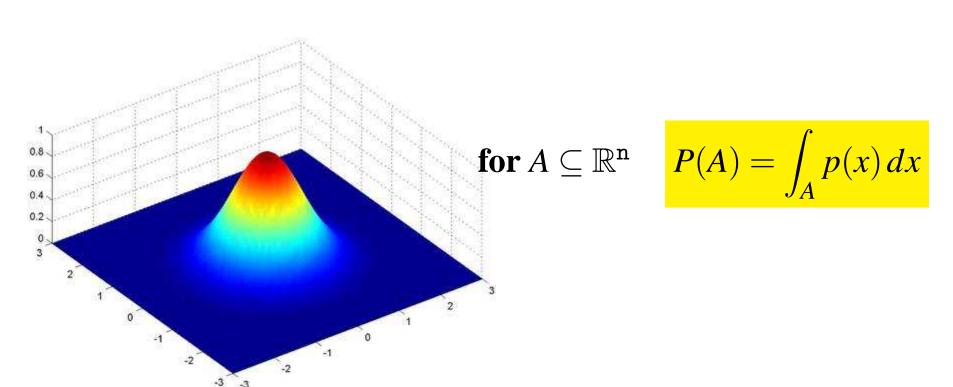
# Stochastic events

### Classical probability (as it is commonly taught)

Model a phenomenon stochastically; outcomes in  $\mathbb{R}^n$ .

#### **Usual framework:**

- probability distributions, probability density functions;
- ightharpoonup 'Every' subset of  $\mathbb{R}^n$  is assigned a probability.



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#### **Usual framework:**

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# **Thesis**

This is unduly restrictive, even for elementary applications.

# What this lecture does/does not do

#### It tries to

explain some basic probability ideas that should be taught.

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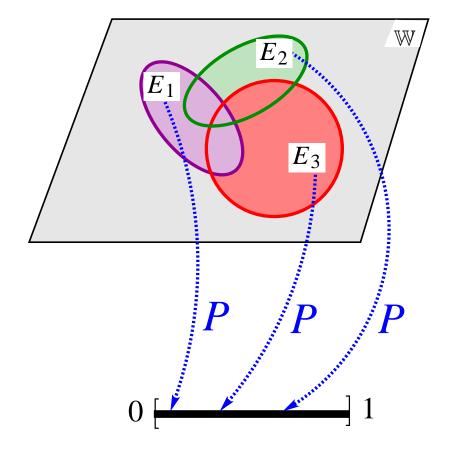
#### It tries to

explain some basic probability ideas that should be taught.

#### It does not address

- mathematical foundations of probability,
- interpretation of probability.

#### **Events**





A.N. Kolmogorov 1903 – 1987

A probability  $P(E) \in [0,1]$ is assigned to certain subsets E ('events') of the outcome space  $\mathbb{W}$ .

 $\mathscr{E}$  = the sets that are assigned a probability,

:= the class of 'measurable' subsets of  $\mathbb{W}$ .

### Main (not all) axioms

The events  $\mathscr{E}$  form a " $\sigma$ -algebra" of subsets of  $\mathbb{W}:\Rightarrow$ 

- $[E \in \mathscr{E}] \Rightarrow [E^{\text{complement}} \in \mathscr{E}]$

 $P:\mathscr{E}\to [0,1]$  is a probability measure : $\Rightarrow$ 

- $ightharpoonup P(\mathbb{W}) = 1,$

#### **Borel**

In applications the events often consist of the *Borel*  $\sigma$ -algebra.





Émile Borel 1871 – 1956

#### **Borel**

In applications the events often consist of the *Borel*  $\sigma$ -algebra.



Émile Borel 1871 – 1956

 $\mathscr E$  contains 'basically every' subset of  $\mathbb R^n$ .

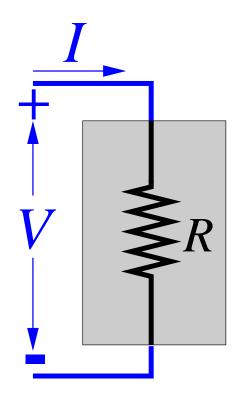
Allows to take probability distributions as the primitive concept, and avoids introducing  $\mathcal{E}$ .

# **Thesis**

Borel is unduly restrictive for system theoretic applications.

# Motivating examples

### **Ohmic resistor**



$$V = RI$$

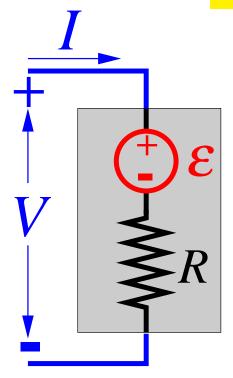
V: voltage across

I current through

R: resistance ( $\geq 0$ )

'Ohmic resistor'

# Noisy (or 'hot') resistor

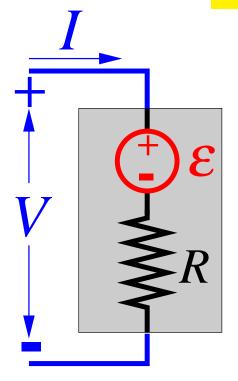


$$V = RI + \varepsilon$$

arepsilon gaussian zero mean variance  $\sim \sqrt{RT}$ 

'Johnson-Nyquist resistor'

# Noisy (or 'hot') resistor



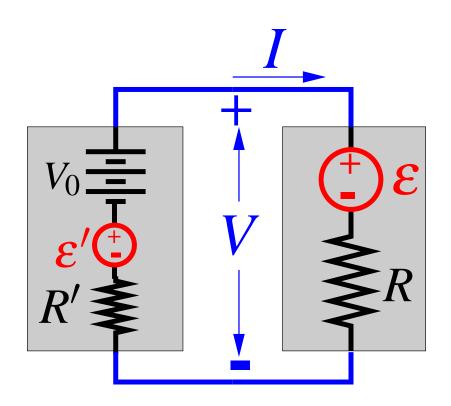
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'Johnson-Nyquist resistor'

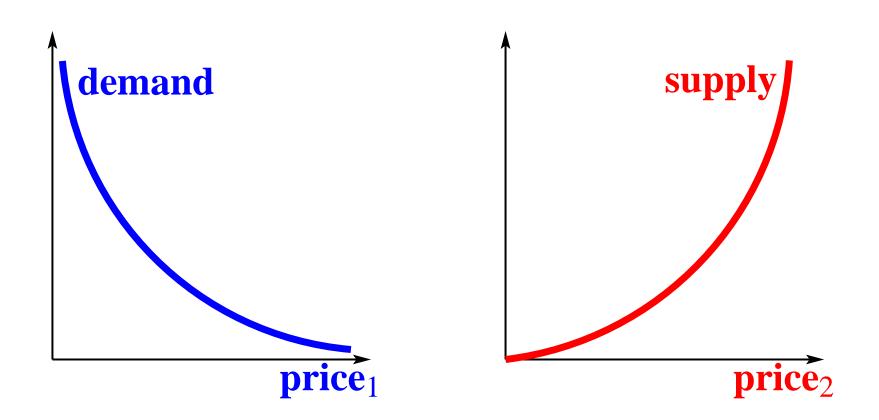
What is  $\begin{bmatrix} V \\ I \end{bmatrix}$  as a mathematical entity?

# Noisy resistor terminated by a voltage source

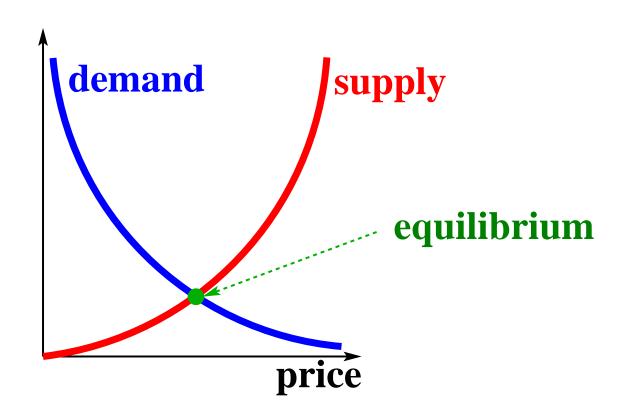


How do we deal with interconnection?

# **Deterministic price/demand/supply**



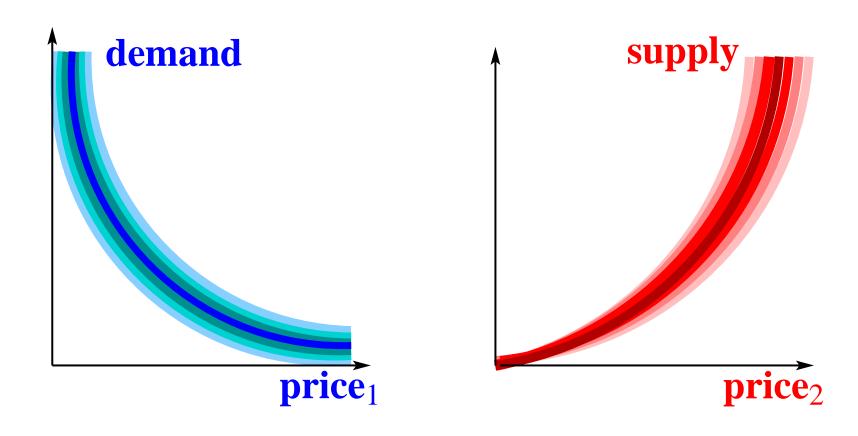
# **Deterministic price/demand/supply**



# 'Interconnection'

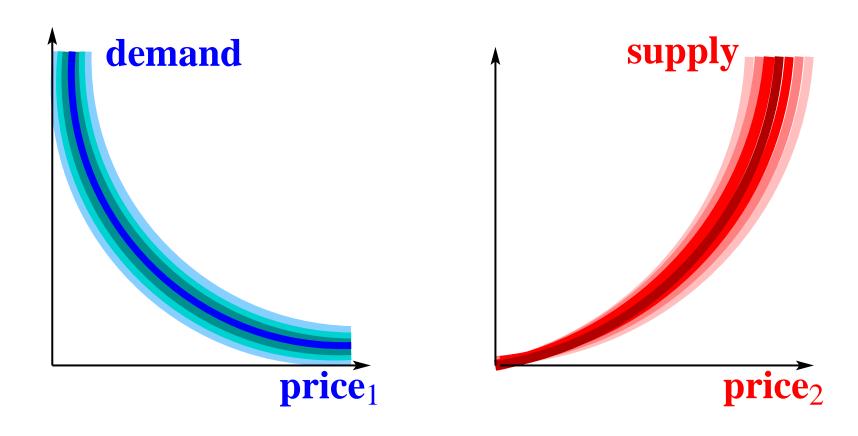
$$price_1 = price_2$$
,  $demand = supply$ .

# **Stochastic price/demand/supply**



(Only) certain regions of the  $\begin{bmatrix} price_1 \\ demand \end{bmatrix}$  and  $\begin{bmatrix} price_2 \\ supply \end{bmatrix}$  planes are assigned a probability.

# **Stochastic price/demand/supply**



(Only) certain regions of the  $\begin{bmatrix} price_1 \\ demand \end{bmatrix}$  and  $\begin{bmatrix} price_2 \\ supply \end{bmatrix}$  planes are assigned a probability.

How do we deal with equilibrium: supply = demand?

# Formal definitions

#### **Definition**

A *stochastic system* is a probability triple  $(\mathbb{W}, \mathcal{E}, P)$ 

- **▶** W a non-empty set, the *outcome space*,
- $\triangleright$  ε a σ-algebra of subsets of  $\mathbb{W}$ : the *events*,
- $ightharpoonup P: \mathscr{E} \to [0,1]$  a probability measure.
- $\mathcal{E}$ : the subsets that are assigned a probability.

Probability that outcomes  $\in E, E \in \mathscr{E}$ , is P(E).

Model  $\cong$   $\mathscr{E}$  and P;

 $\mathscr{E}$  is an essential part!

 $\mathcal{E}$  should not be taken for granted.

#### **Definition**

# A stochastic system is a probability triple $(\mathbb{W}, \mathcal{E}, P)$

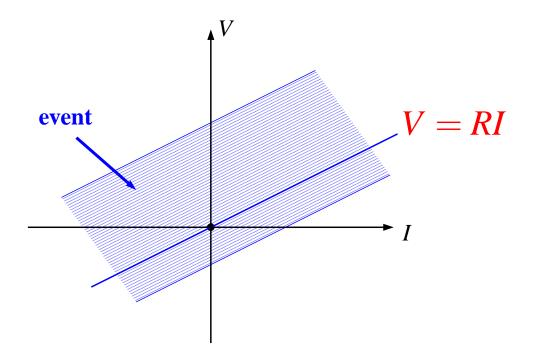
- **▶** W a non-empty set, the *outcome space*,
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# **'Classical'** stochastic system:

 $\mathbb{W} = \mathbb{R}^n$  and  $\mathscr{E} =$  'all' subsets of  $\mathbb{R}^n$ .

P specified by a probability distribution or a pdf.

# **Noisy resistor**

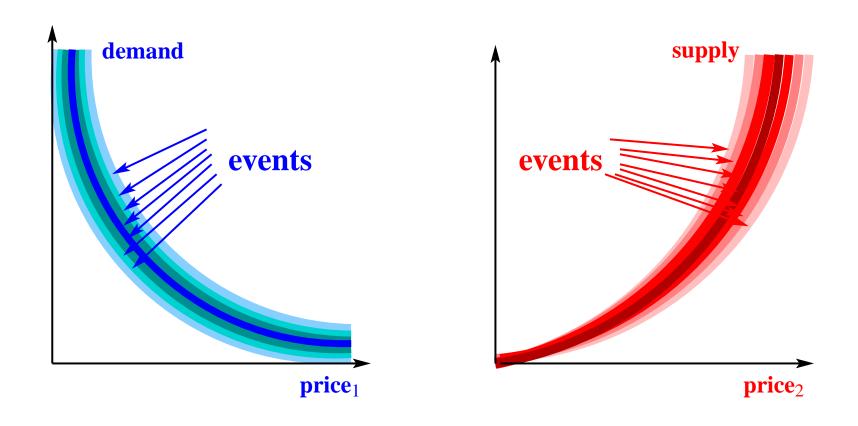


**Events:**  $\left\{ \begin{bmatrix} V \\ I \end{bmatrix} \in \mathbb{R}^2 \mid V - RI \in A \text{ with } A \text{ a Borel subset of } \mathbb{R} \right\}$ .

P(event) = gaussian measure of A.

Neither  $\begin{bmatrix} V \\ I \end{bmatrix}$  nor I nor V possess a pdf.

# **Stochastic price/demand/supply**



 $\mathscr{E}, \mathscr{E}'$  = the regions that are assigned a probability.

p, d, nor s are not classical real random variables.

# Linearity

# Linear stochastic system

# linear stochastic system

 $:\Leftrightarrow$  Borel probability on  $\mathbb{R}^n/\mathbb{L}$ ,

 $\mathbb{L} \subseteq \mathbb{R}^n$  a linear subspace, called the 'fiber'.

Note:  $\mathbb{R}^n/\mathbb{L}$  is a real vector space of dimension

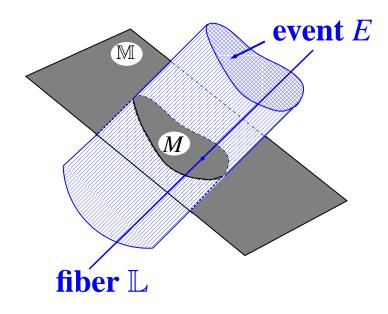
 $n-dimension(\mathbb{L})$ .

**Events:** cylinders with sides parallel to  $\mathbb{L}$ .

Subsets of  $\mathbb{R}^n$  as  $A + \mathbb{L}$ ,  $\mathbb{L}$  linear subspace, A Borel.

# Linearity

# linear stochastic system



Borel probability on  $\mathbb{M}$ .

**Example:** the noisy resistor.

Classical  $\Rightarrow$  linear!

**gaussian** : $\Leftrightarrow$  linear, probability on  $\mathbb M$  gaussian.

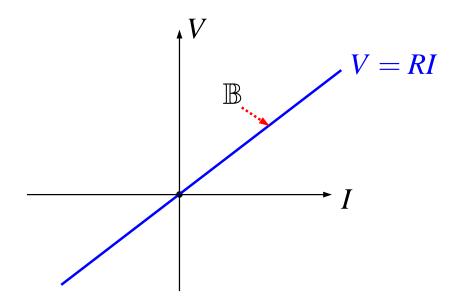
# **Deterministic system**

 $(\mathbb{W}, \mathcal{E}, P)$  is said to be *deterministic* if

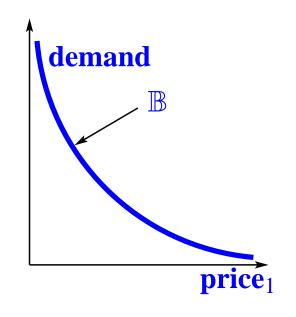
$$\mathscr{E} = \{\emptyset, \mathbb{B}, \mathbb{B}^{complement}, \mathbb{W}\} \text{ and } P(\mathbb{B}) = 1.$$

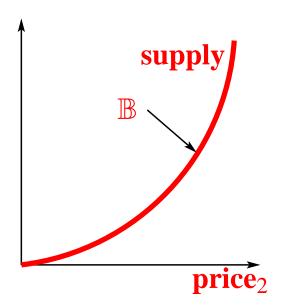
# **Deterministic examples**

# **Ohmic resistor:**

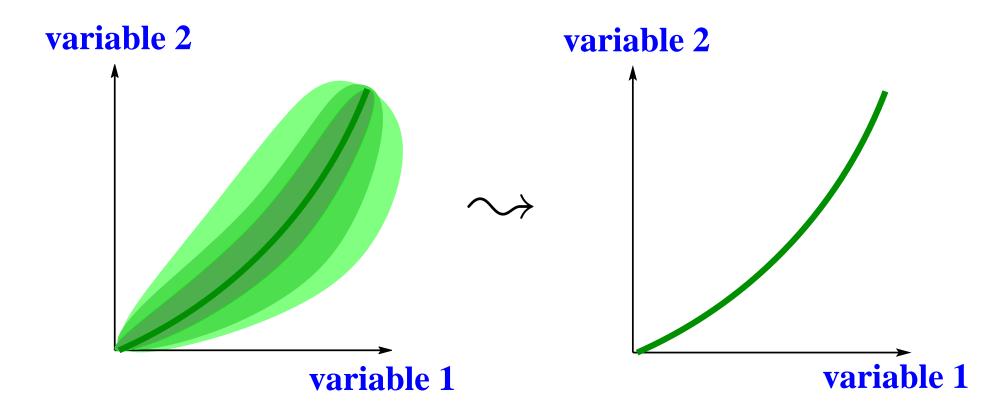


# **Economic example:**





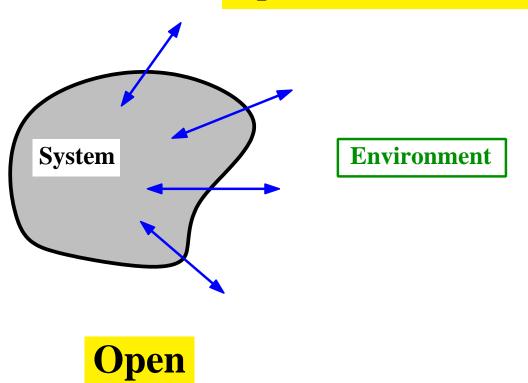
# The need for 'coarse' $\sigma$ -algebras



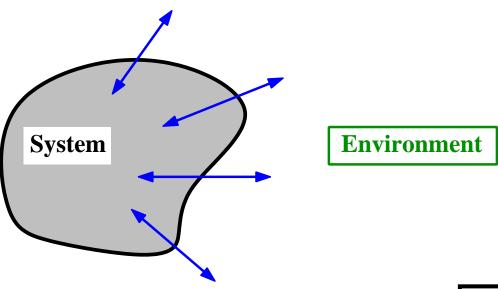
For a classical random vector, the deterministic limit  $\simeq$  a (singular) probability distribution. Awkward from the modeling point of view.

# Interconnection

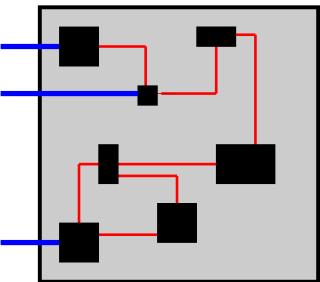
# **Open and connected**



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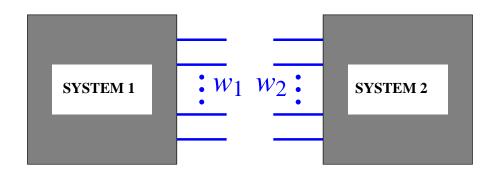


**Open** 

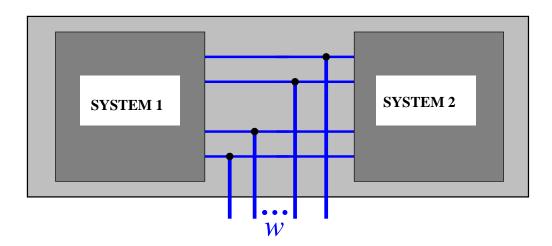


Connectable

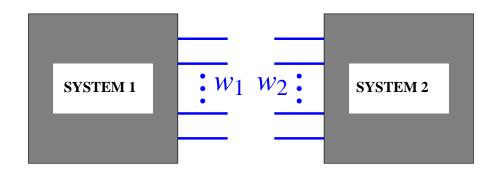
# Interconnection



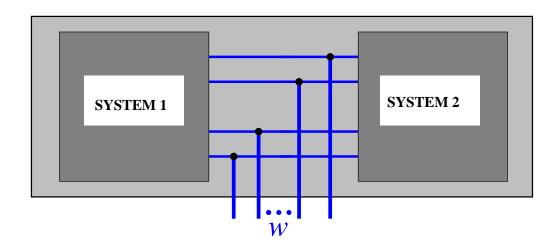




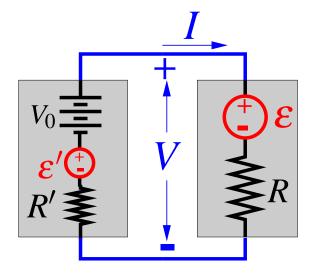
#### Interconnection



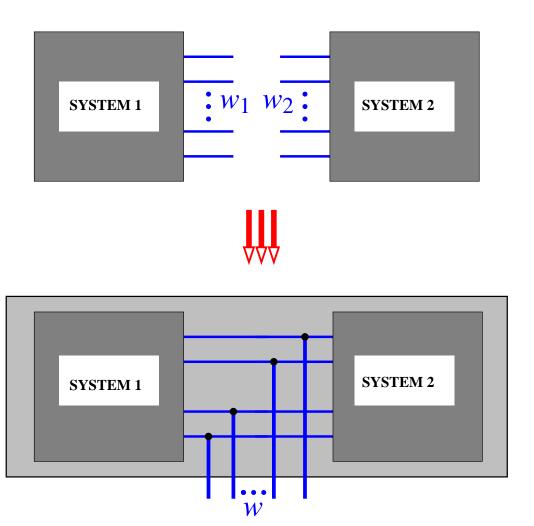




# **Example:**



#### Interconnection



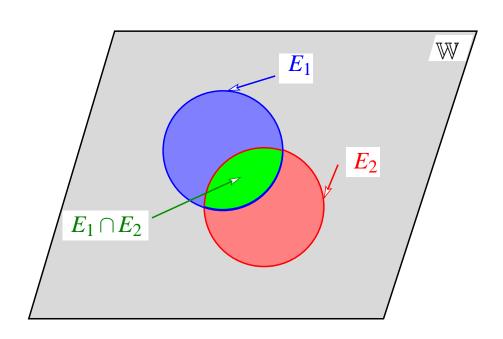
Can two distinct probabilistic laws

be imposed on the same set of variables?

#### Complementarity of $\sigma$ -algebras

 $\mathscr{E}_1$  and  $\mathscr{E}_2$  are complementary  $\sigma$ -algebras : $\Leftrightarrow$  for all nonempty sets  $E_1, E_1' \in \mathscr{E}_1, E_2, E_2' \in \mathscr{E}_2$ 

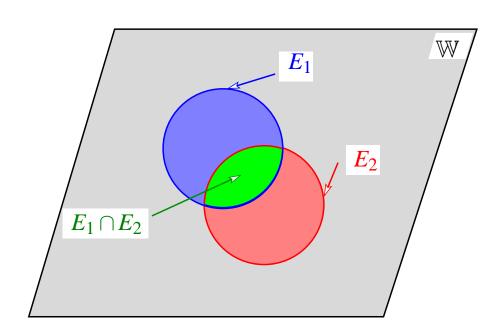
$$[\![E_1 \cap E_2 = E_1' \cap E_2']\!] \Rightarrow [\![E_1 = E_1' \text{ and } E_2 = E_2']\!].$$



#### Complementarity of $\sigma$ -algebras

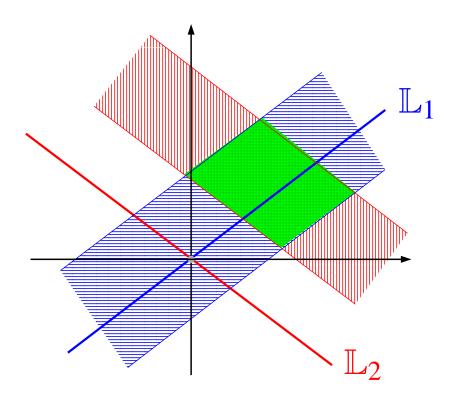
 $\mathscr{E}_1$  and  $\mathscr{E}_2$  are *complementary*  $\sigma$ -algebras : $\Leftrightarrow$  for all nonempty sets  $E_1, E_1' \in \mathscr{E}_1, E_2, E_2' \in \mathscr{E}_2$ 

$$[\![E_1 \cap E_2 = E_1' \cap E_2']\!] \Rightarrow [\![E_1 = E_1' \text{ and } E_2 = E_2']\!].$$



The intersection determines the intersectants.

# Linear example



$$\mathbb{L}_1 + \mathbb{L}_2 = \mathbb{R}^n$$

#### **Interconnection of complementary systems**

Let  $(\mathbb{W}, \mathcal{E}_1, P_1)$  and  $(\mathbb{W}, \mathcal{E}_2, P_2)$  be stochastic systems (independent). Assume complementarity.

Their *interconnection* is defined as

$$(\mathbb{W},\mathscr{E},P)$$

with  $\mathscr{E} :=$  the  $\sigma$ -algebra generated by 'rectangles'

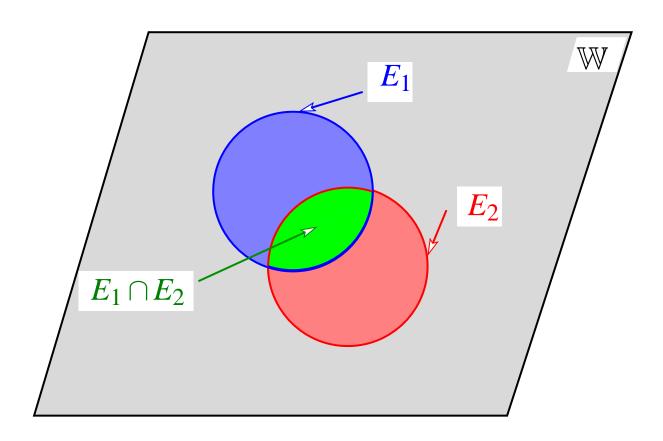
$$\{E_1 \cap E_2 \mid E_1 \in \mathscr{E}_1, E_2 \in \mathscr{E}_2\},\$$

and P defined through the rectangles by

$$P(E_1 \cap E_2) := P_1(E_1)P_2(E_2).$$

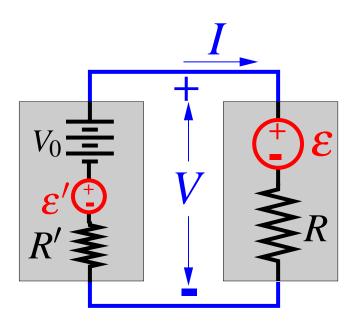
for  $E_1 \in \mathscr{E}_1, E_2 \in \mathscr{E}_2$ .

#### **Interconnection of complementary systems**

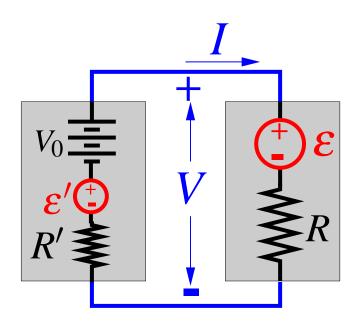


$$P(E_1 \cap E_2) := P_1(E_1)P_2(E_2).$$

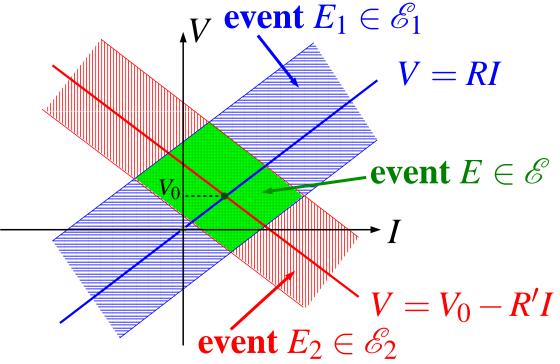
# Noisy resistor terminated by a voltage source



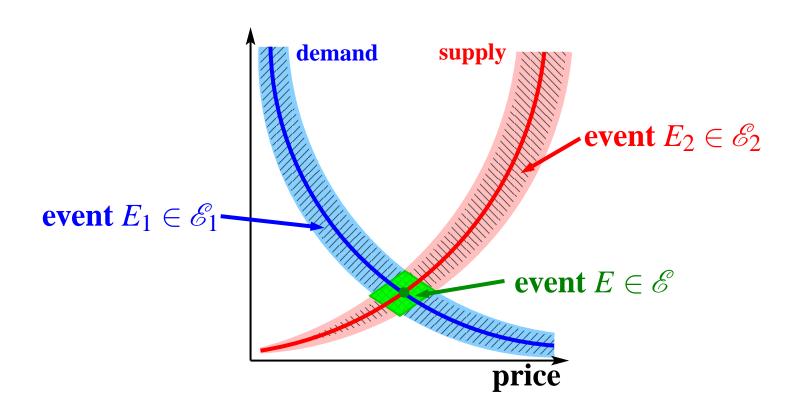
#### Noisy resistor terminated by a voltage source



$$P(E) = P_1(E_1)P_2(E_2)$$



#### **Equilibrium price/demand/supply**



$$P(E) = P_1(E_1)P_2(E_2).$$

# Open stochastic systems

#### **Open versus closed**

Consider  $\Sigma_1 = (\mathbb{R}^n, \mathscr{E}_1, P_1)$ .

If  $\mathscr{E}_1$  = the Borel  $\sigma$ -algebra, and  $\operatorname{support}(P_1) = \mathbb{R}^n$ , then  $\Sigma_1$  is interconnectable only with the free system  $(\mathbb{R}^n, \mathscr{E}_2, P_2)$ ,  $\mathscr{E}_2 = \{\emptyset, \mathbb{R}^n\}$ .  $\Rightarrow$  classical  $\Sigma_1 =$  'closed' system.

#### **Open versus closed**

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Coarse  $\mathcal{E}_1$ 

 $\Rightarrow \Sigma_1$  is interconnectable.

 $\Rightarrow$  'open' system.

#### **Open versus closed**

In the Kolmogorov philosophy, random variables, random vectors, and random processes are (measurable) functions defined on the probability space  $(\Omega, \mathcal{A}, P)$ .

We view the randomness as 'internal' to the system.

So, once the Gods choose  $\omega \in \Omega$ , all the random variables are determined.

The environment has no influence on the outcomes.

 $\Rightarrow$  'closed' systems.

# Conditional and constrained probability



#### **Conditional probability**

Given  $\Sigma = (\mathbb{W}, \mathscr{E}, P)$ .

Look at outcomes  $w \in \mathbb{S}$  with  $\mathbb{S} \subseteq \mathbb{W}$ .

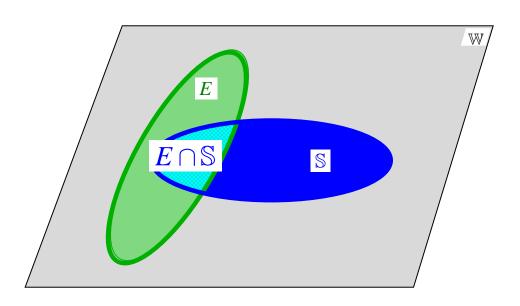
For  $\mathbb{S}$  an event,  $\mathbb{S} \in \mathscr{E}$ ,  $\sim$  conditional probability.

Assume P(S) > 0. Then

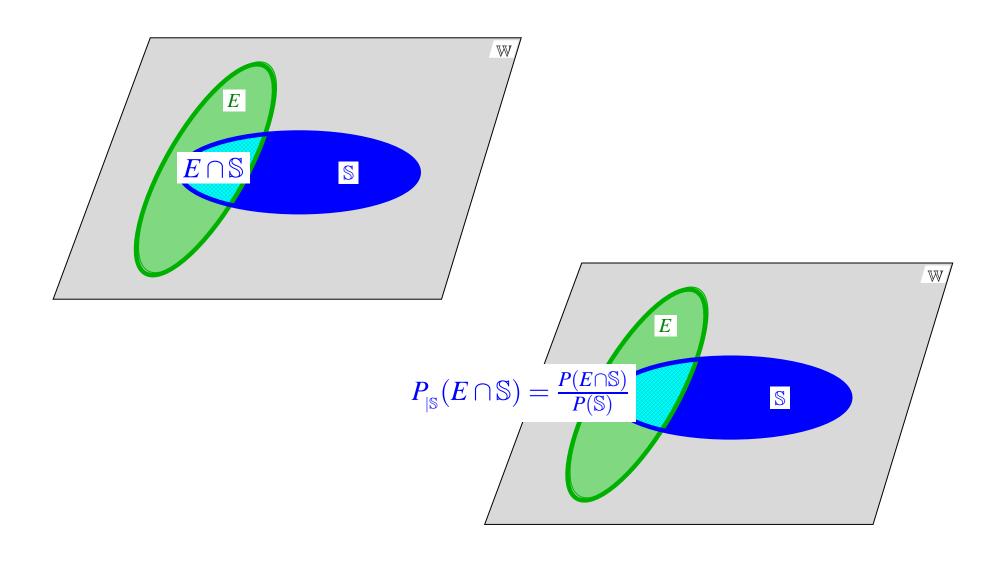
$$\Sigma_{|\mathbb{S}} := \left(\mathbb{S},\mathscr{E}\cap\mathbb{S},P_{|\mathbb{S}}
ight), ext{ with } P_{|\mathbb{S}}(E\cap\mathbb{S}) := rac{P(E\cap\mathbb{S})}{P(\mathbb{S})}.$$

The construction of  $P_{|\mathbb{S}}$  is more complicated when  $P(\mathbb{S})=0,$  but well-known.

# **Conditional probability**



### **Conditional probability**



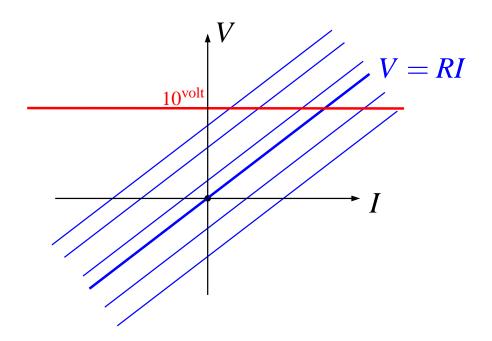
Let 
$$\Sigma = (\mathbb{W}, \mathscr{E}, P)$$
.

Impose the constraint 
$$w \in \mathbb{S}$$
 with  $\mathbb{S} \subset \mathbb{W}$ ,  $\mathbb{S} \notin \mathcal{E}$ .

What is the stochastic nature of the outcomes in  $\mathbb{S}$ ?

Is this a meaningful question?

#### **Noisy resistor**

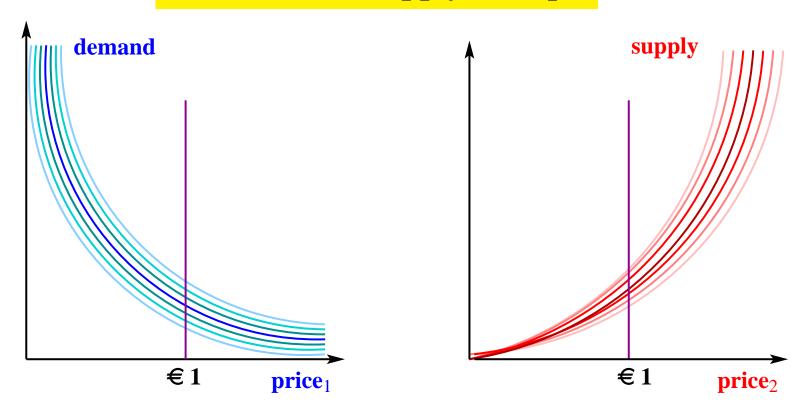


Impose  $V = 10^{\text{volt}}$ . What is the distribution of I?

$$V = RI + \varepsilon, V = 10^{\text{volt}} \Rightarrow I = \frac{V_0}{10} - \frac{\varepsilon}{10}.$$

I is a well-defined random variable!

#### Price/demand/supply example



Impose price  $= \le 1$ . Probability of demand, supply?

Let 
$$\Sigma = (\mathbb{W}, \mathscr{E}, P)$$
.

Impose the constraint  $w \in \mathbb{S}$  with  $\mathbb{S} \subset \mathbb{W}$ ,  $\mathbb{S} \notin \mathscr{E}$ .

What is the stochastic nature of the outcomes in  $\mathbb{S}$ ?

Is this a meaningful question? Yes, it is!

Constraining  $\simeq$  interconnection of  $\Sigma = (\mathbb{W}, \mathscr{E}, P)$  and the deterministic system with behavior  $\mathbb{S}$ .

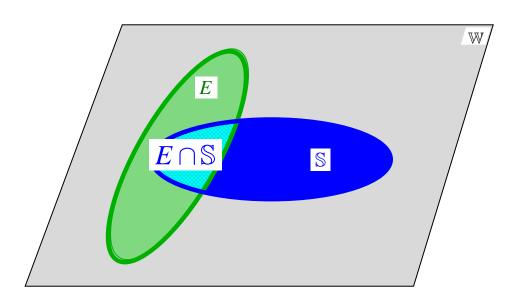
### **Assume complementarity:**

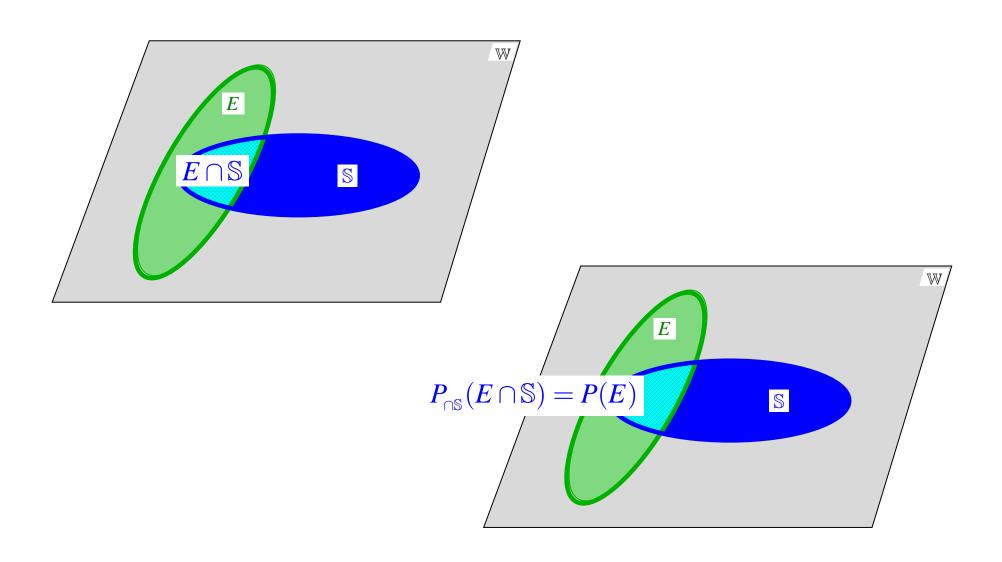
$$\llbracket E_1, E_2 \in \mathscr{E} \text{ and } E_1 \cap \mathbb{S} = E_2 \cap \mathbb{S} \rrbracket \Rightarrow \llbracket E_1 = E_2 \rrbracket$$

#### Interconnection $\sim$

$$\Sigma_{\cap \mathbb{S}} = (\mathbb{S}, \mathscr{E} \cap \mathbb{S}, P_{\cap \mathbb{S}})$$
 with  $P_{\cap \mathbb{S}}(E \cap \mathbb{S}) := P(E)$ .

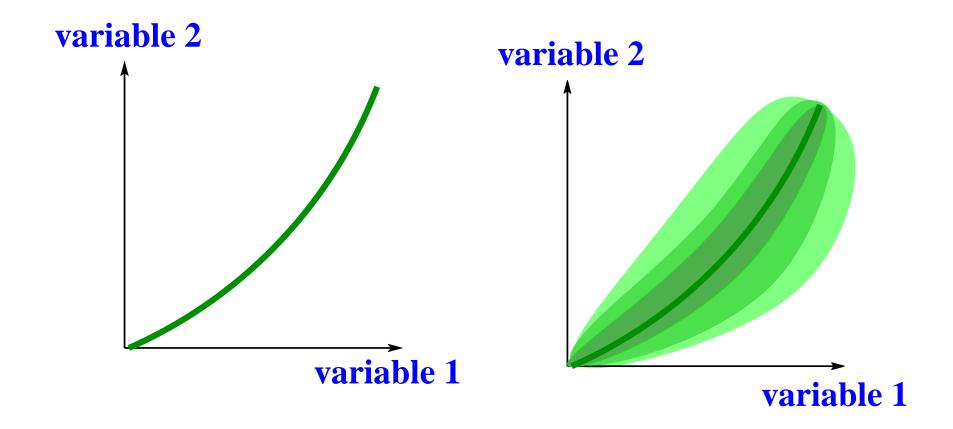
 $P_{0S} =$  "probability of w constrained by  $w \in S$ ".



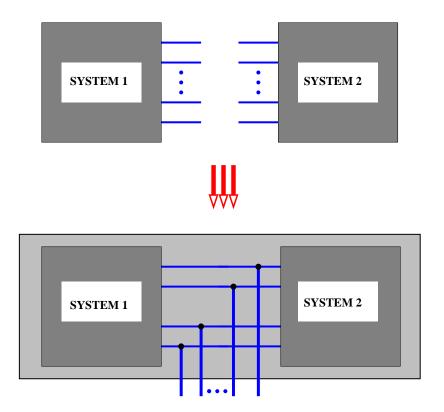


# Conclusions

The Borel  $\sigma$ -algebra is inadequate even for elementary applications.



Complementary stochastic systems can be interconnected: two distinct laws imposed on one set of variables.



 Open stochastic systems require a coarse σ-algebra.

Classical random vectors imply closed systems.

 Open stochastic systems require a coarse σ-algebra.

Classical random vectors imply closed systems.

ightharpoonup notion of 'constrained probability'.

**Future work** 

### **Urgent:**

Generalization to stochastic processes.

#### Where to find more

Reference: Open stochastic systems, IEEE Tr. AC, submitted.

# Copies of the lecture frames available from/at

http://www.esat.kuleuven.be/~jwillems



Professor Siep, het ga je goed!

# Thank you

Thank you