



OPEN STOCHASTIC SYSTEMS

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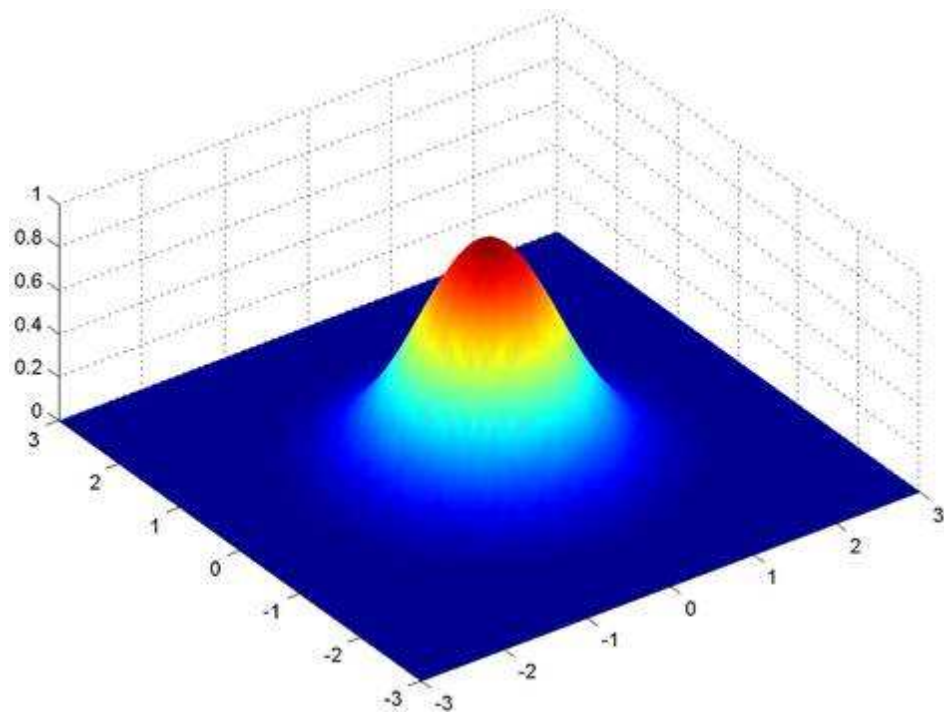
The idea

Theme

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

Usual framework:

- ▶ probability distributions, probability density functions;
- ▶ means that the event σ -algebra consists of the Borel sets.
 \leadsto ‘Every’ subset of \mathbb{R}^n is assigned a probability.



for $A \subseteq \mathbb{R}^n$ Borel,

$$P(A) = \int_A p(x) dx$$

Theme

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

Usual framework:

- ▶ **probability distributions, probability density functions;**
- ▶ **means that the event σ -algebra consists of the Borel sets.**
 \rightsquigarrow **‘Every’ subset of \mathbb{R}^n is assigned a probability.**

Thesis

*This is unduly restrictive,
even for elementary applications.*

What this lecture does/**does not**

It tries to

- ▶ **explain some probability ideas that should be taught,**
- ▶ **in the setting of orthodox mathematical probability theory.**

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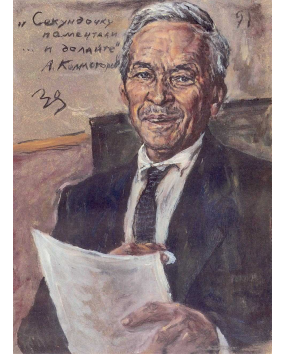
- ▶ **explain some probability ideas that should be taught,**
- ▶ **in the setting of orthodox mathematical probability theory.**

It does not address

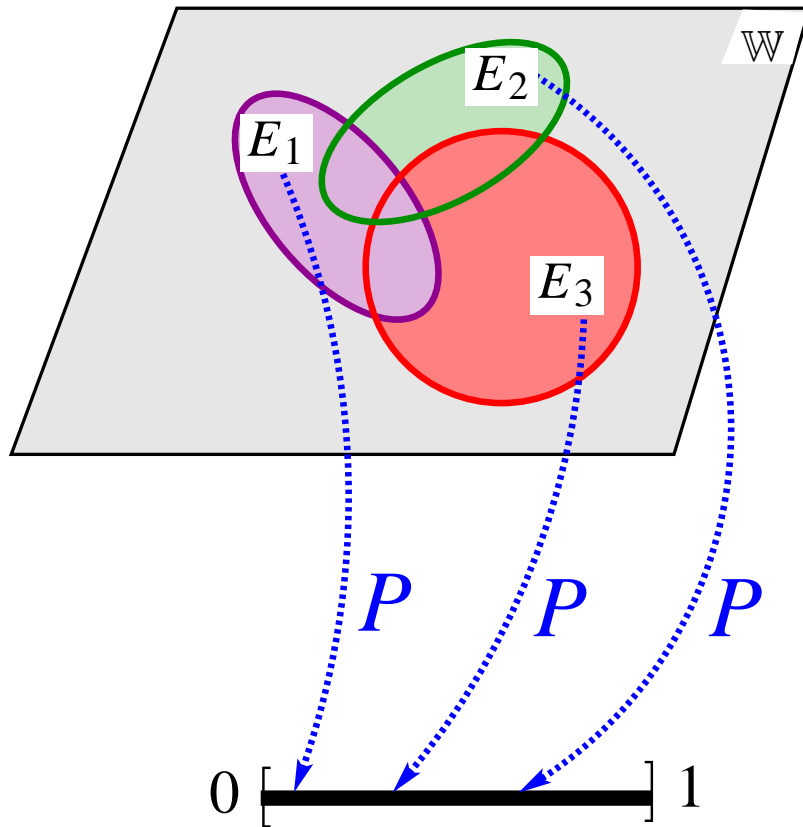
- ▶ **mathematical foundations of probability,**
- ▶ **interpretation of probability.**

Basic 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} .

\mathcal{E} = the sets that are assigned a probability,
:= the class of 'measurable' subsets of \mathbb{W} .

Main (not all) axioms

The events \mathcal{E} form a σ -algebra of subsets of $\mathbb{W} \Rightarrow$

- ▶ $[[E \in \mathcal{E}] \Rightarrow [E^{\text{complement}} \in \mathcal{E}]$
- ▶ $[[E_1, E_2 \in \mathcal{E}] \Rightarrow [[E_1 \cap E_2 \in \mathcal{E}, E_1 \cup E_2 \in \mathcal{E}]$

$P : \mathcal{E} \rightarrow [0, 1]$ is a **probability measure** \Rightarrow

- ▶ $P(\mathbb{W}) = 1,$
- ▶ $[[E_1, E_2 \in \mathcal{E} \text{ and } E_1 \cap E_2 = \emptyset]$
 $\Rightarrow [[P(E_1 \cup E_2) = P(E_1) + P(E_2)]]$ (P is **additive**).

Borel

In applications the measurable sets often consist of the *Borel σ -algebra*.



Émile Borel
1871 – 1956

$\mathcal{B}(\mathbb{R}^n)$:= the Borel σ -algebra on \mathbb{R}^n ;

random variable: $\mathbb{W} = \mathbb{R}$ (or \mathbb{C}), and $\mathcal{E} = \mathcal{B}(\mathbb{R})$

random vector: $\mathbb{W} = \mathbb{R}^n$, and $\mathcal{E} = \mathcal{B}(\mathbb{R}^n)$

random process: a family of random variables, etc.

$\mathcal{B}(\mathbb{R}^n)$ contains ‘basically every’ subset of \mathbb{R}^n .

Borel



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In applications the measurable sets often consist of the *Borel σ -algebra*.

$\mathcal{B}(\mathbb{R}^n)$ = the Borel σ -algebra on \mathbb{R}^n ;

$\mathcal{B}(\mathbb{R}^n)$ contains ‘basically every’ subset of \mathbb{R}^n .

Allows to take probability distributions as the primitive concept, avoids introducing \mathcal{E} *ab initio*.

Thesis

*Borel is unduly restrictive
for system theoretic applications.*

Borel

In applications the measurable sets often consist of the *Borel σ -algebra*.



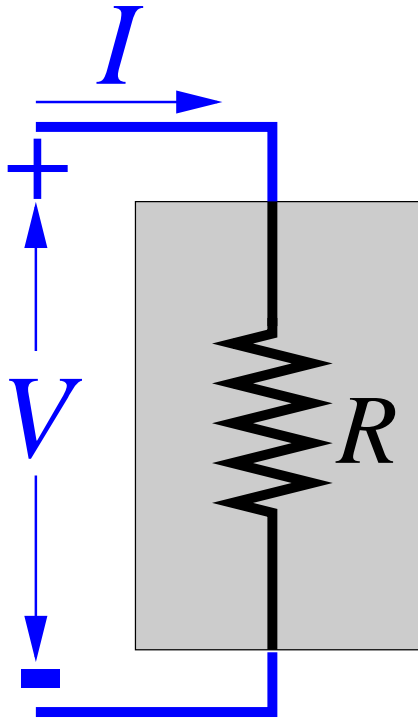
Émile Borel
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Borel is usually assumed for many basic concepts, as

- ▶ random variable, random vector,
- ▶ independence of random variables,
- ▶ marginal measure, conditioning,
- ▶ random process,
- ▶ Brownian motion, Markov process, etc.

Motivating examples

Ohmic resistor



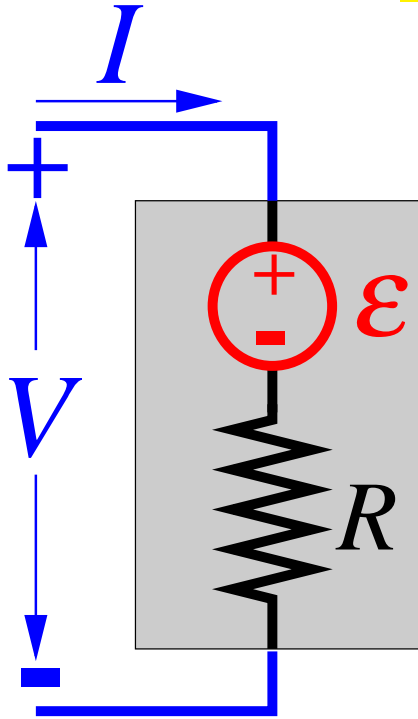
$$V = RI$$

V : voltage across
 I current through

R : resistance (≥ 0)

‘Ohmic resistor’

Noisy (or 'hot') resistor



$$V = RI + \varepsilon$$

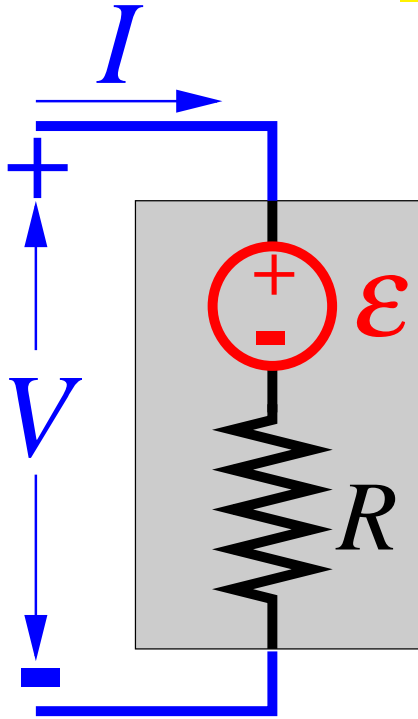
ε gaussian

zero mean

variance $\sim \sqrt{RT}$

'Johnson-Nyquist resistor'

Noisy (or 'hot') resistor



$$V = RI + \varepsilon$$

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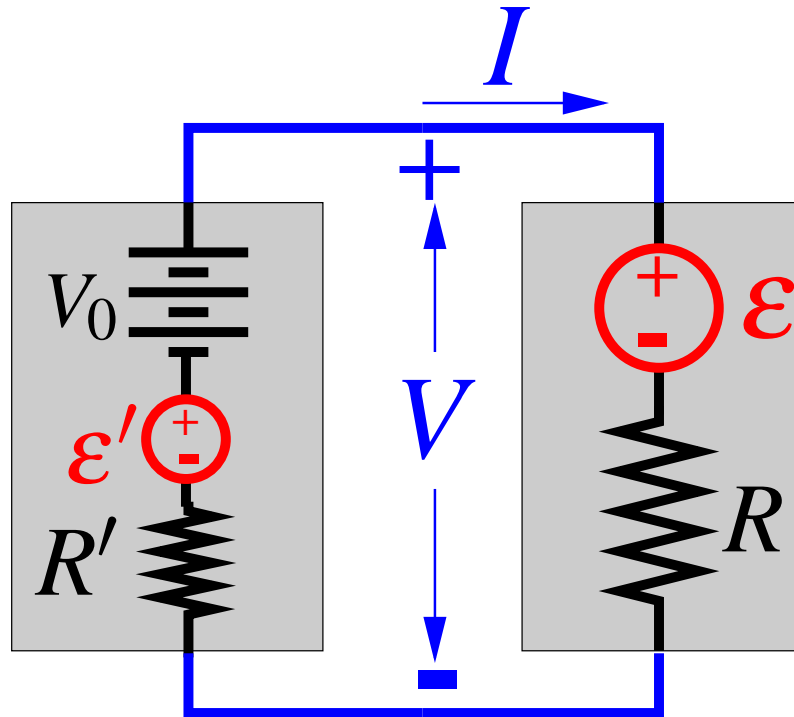
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‘Johnson-Nyquist resistor’

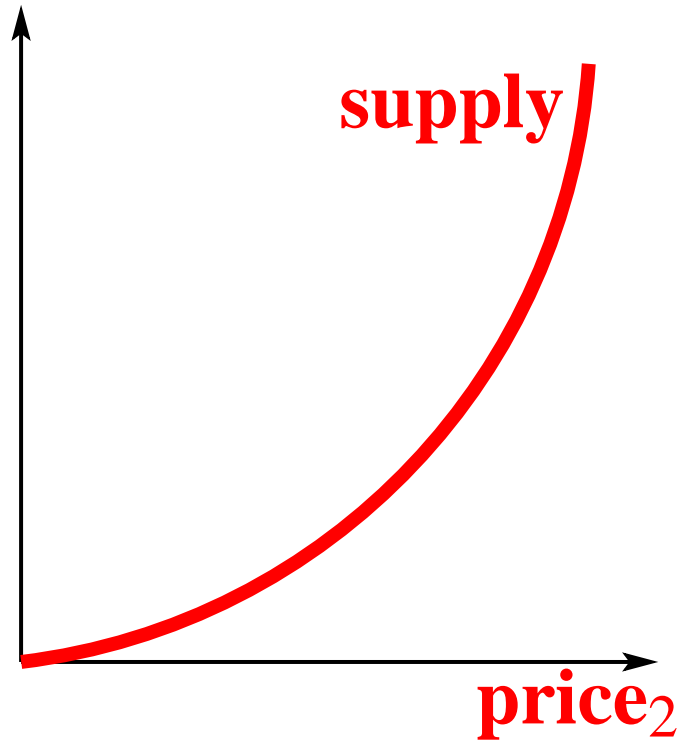
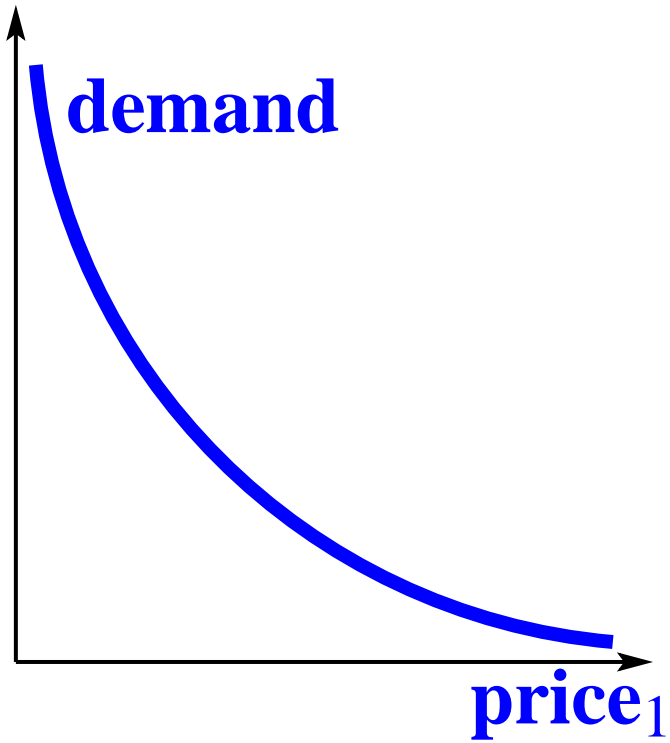
What is $\begin{bmatrix} V \\ I \end{bmatrix}$ as a mathematical entity?

Noisy resistor terminated by a voltage source

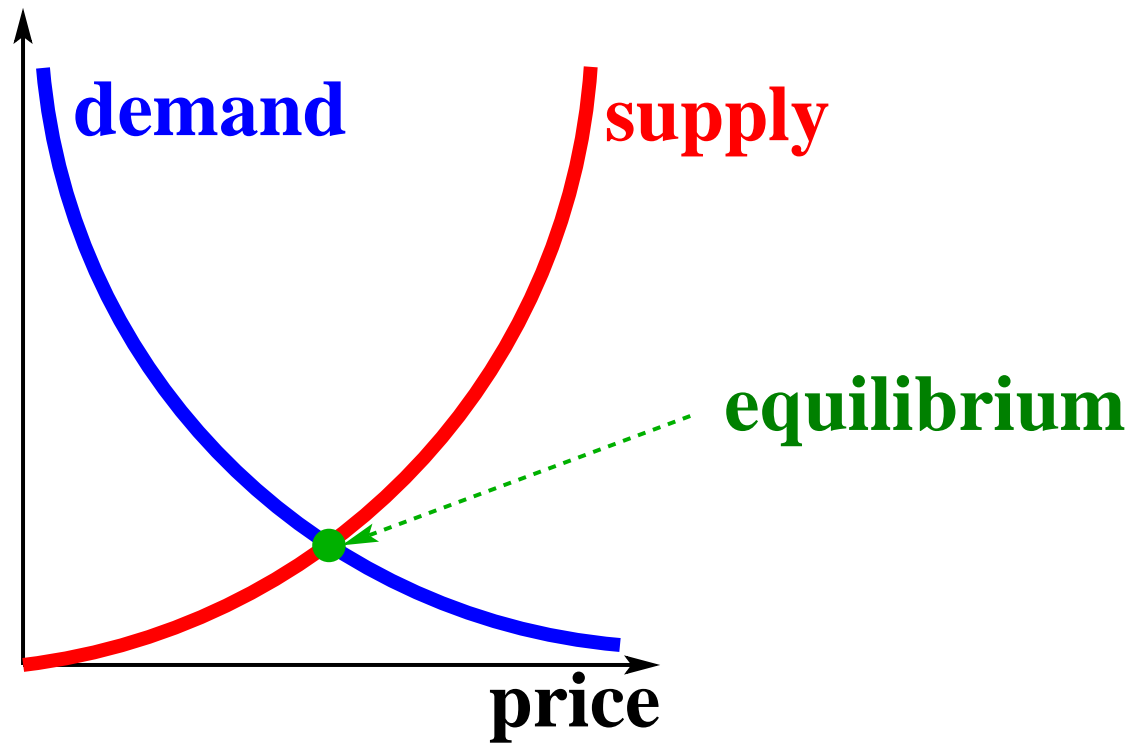


How do we deal with interconnection?

Deterministic price/demand/supply



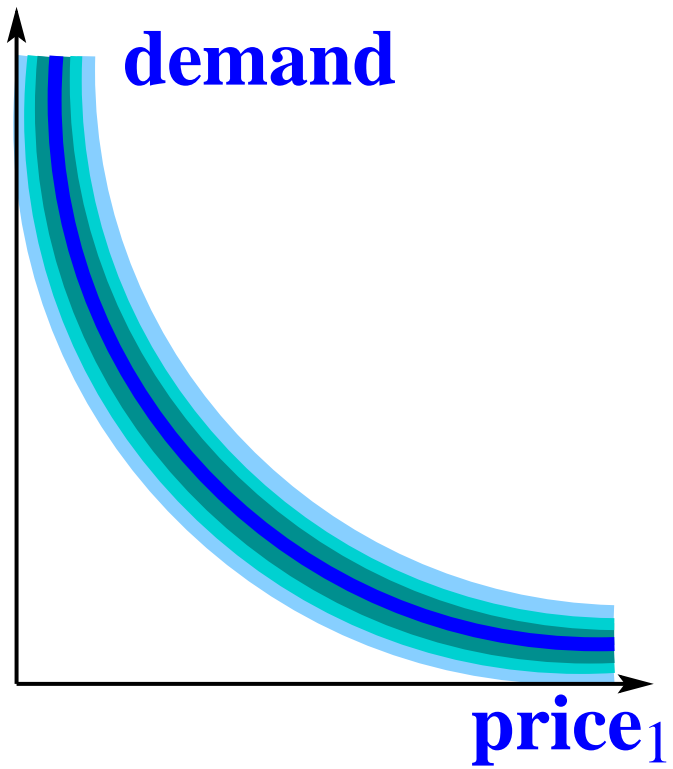
Deterministic price/demand/supply



‘Interconnection’

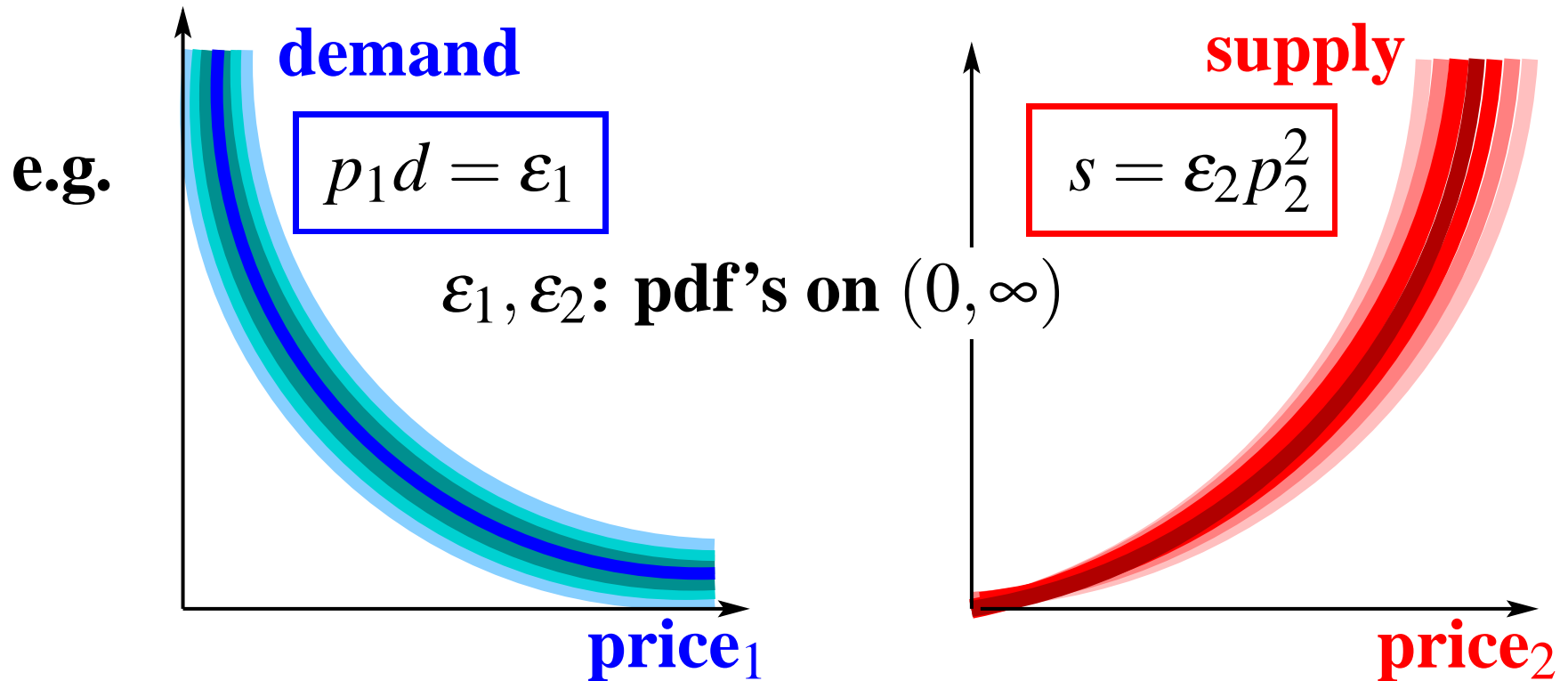
$$\text{price}_1 = \text{price}_2, \quad \text{demand} = \text{supply}.$$

Stochastic price/demand/supply



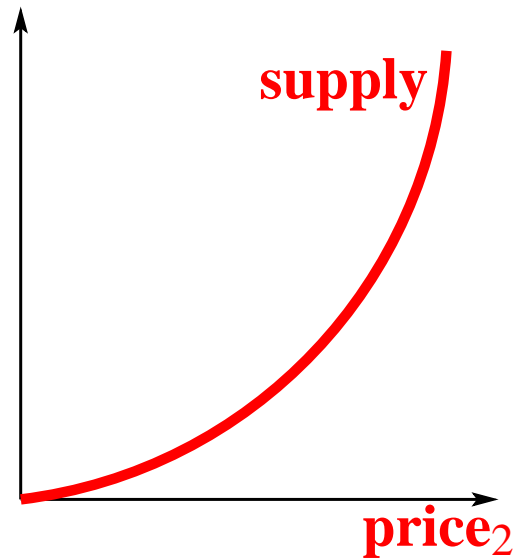
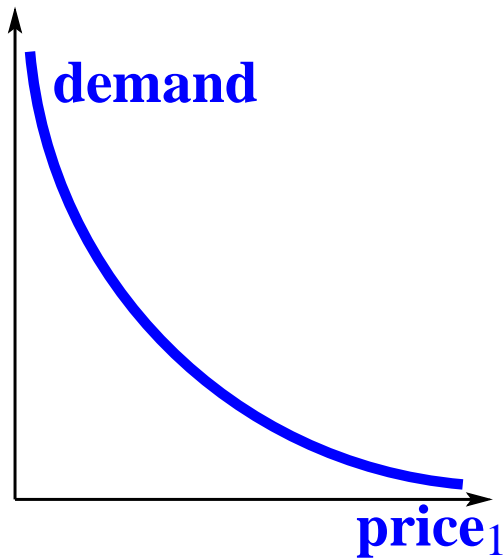
(Only) certain regions of the $\begin{bmatrix} \text{price}_1 \\ \text{demand} \end{bmatrix}$ and $\begin{bmatrix} \text{price}_2 \\ \text{supply} \end{bmatrix}$ planes are assigned a probability.

Stochastic price/demand/supply



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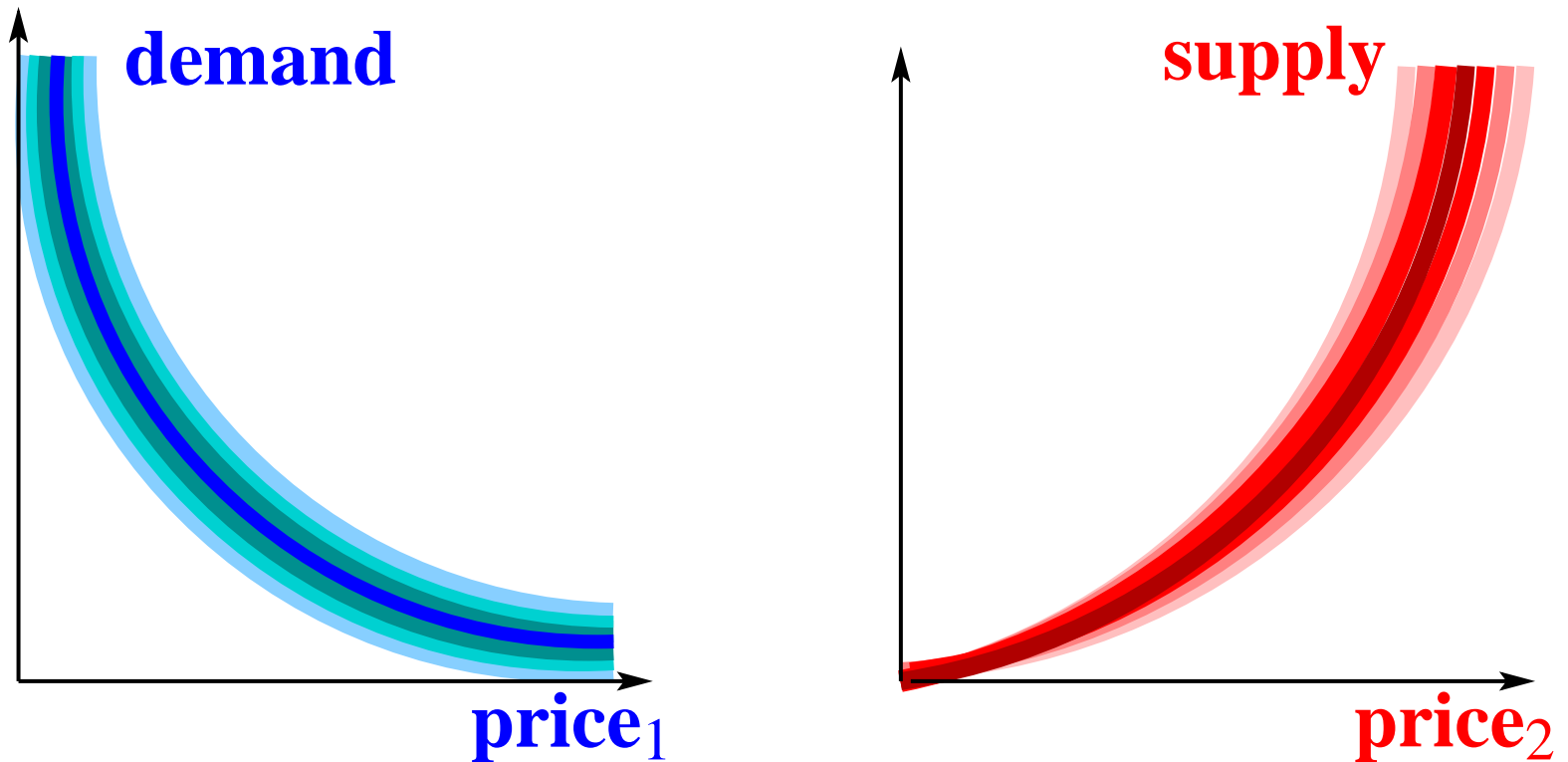
Deterministic price/demand/supply



$(p_1, d) \in$ **characteristic w.p. 1.**

$(p_2, s) \in$ **characteristic w.p. 1.**

Stochastic price/demand/supply



(Only) certain regions of the $\begin{bmatrix} \text{price}_1 \\ \text{demand} \end{bmatrix}$ and $\begin{bmatrix} \text{price}_2 \\ \text{supply} \end{bmatrix}$ planes are assigned a probability.

How do we deal with equilibrium: supply = demand?

Formal definitions

Definition

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

- ▶ \mathbb{W} a non-empty set, the *outcome space*,
- ▶ \mathcal{E} a σ -algebra of subsets of \mathbb{W} : the *events*,
- ▶ $P : \mathcal{E} \rightarrow [0, 1]$ a *probability measure*.

\mathcal{E} : the subsets that are assigned a probability.

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

Model \cong \mathcal{E} and P ;

\mathcal{E} is an essential part!

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

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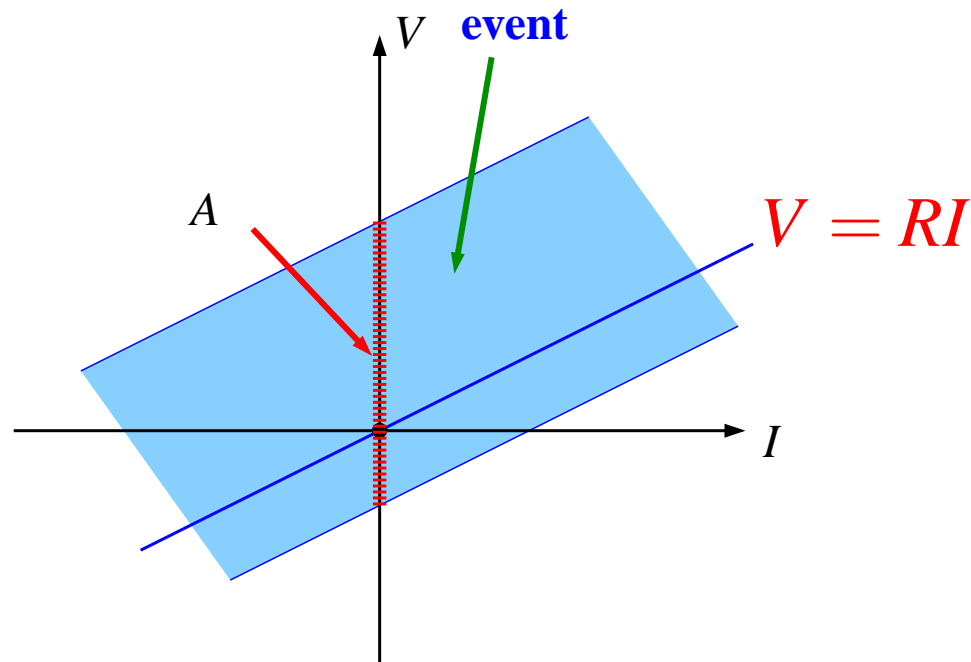
‘Classical’ stochastic system:

$\mathbb{W} = \mathbb{R}^n$ and \mathcal{E} = the Borel subsets of \mathbb{R}^n .

P specified by a probability distribution or a pdf.

\mathcal{E} is inherited from the topology of the outcome space, it does not involve the randomness.

Noisy resistor



$V = RI + \varepsilon$: **stoch. system, outcomes** $\begin{bmatrix} V \\ I \end{bmatrix}$, $\mathbb{W} = \mathbb{R}^2$.

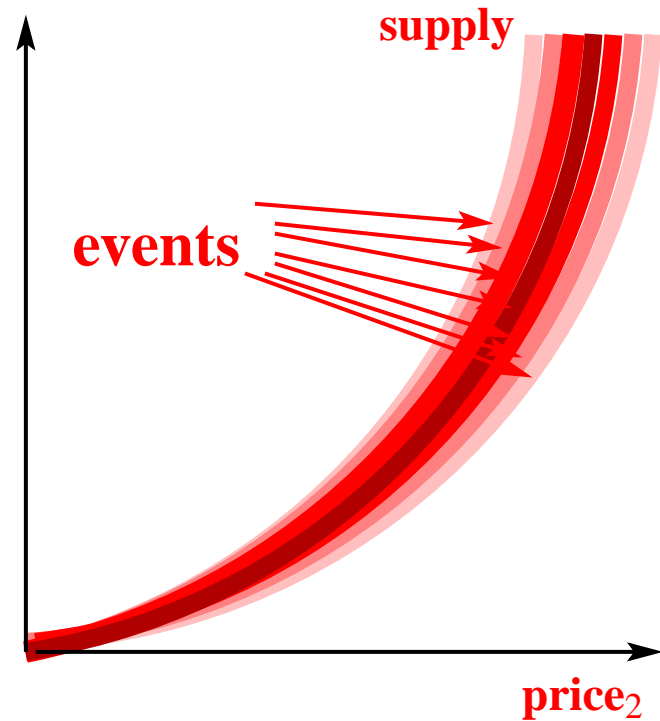
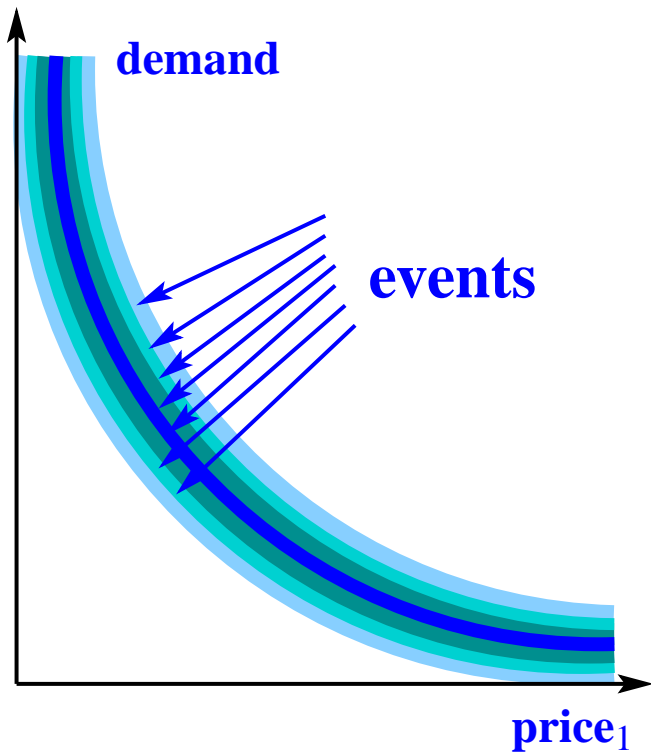
Events: $\left\{ \begin{bmatrix} V \\ I \end{bmatrix} \in \mathbb{R}^2 \mid V - RI \in A \text{ with } A \text{ a Borel subset of } \mathbb{R} \right\}$.

$P(\text{event}) =$ **gaussian measure of A.**

V nor I are **not** classical real random variables.

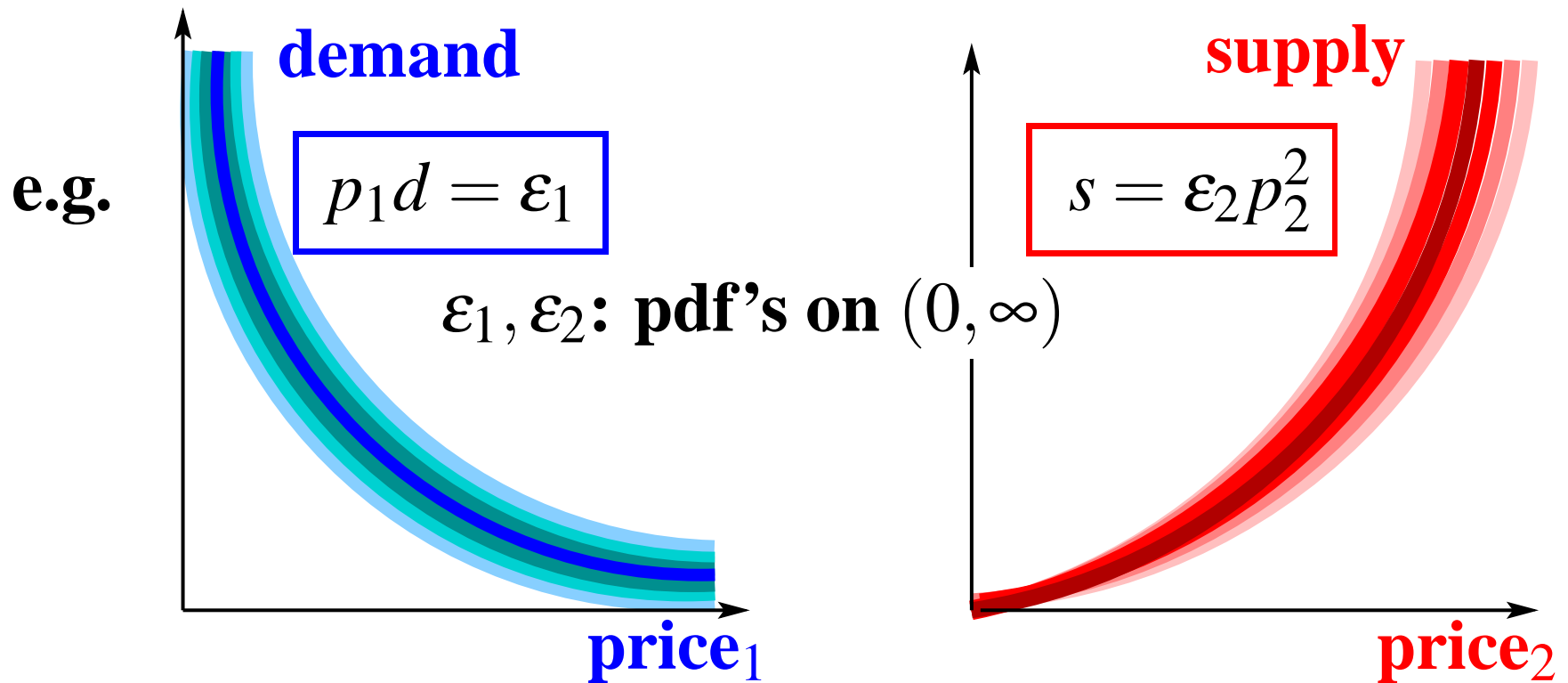
Neither $\begin{bmatrix} V \\ I \end{bmatrix}$ nor I nor V possess a pdf.

Stochastic price/demand/supply



\mathcal{E} , \mathcal{E}' = the regions that are assigned a probability.
 p , d , nor s are not classical real random variables.

Stochastic price/demand/supply



Often the events can be parameterized (as here by $\varepsilon_1, \varepsilon_2$).

σ -algebra definition is more elementary/desirable/general.

Linearity

Atoms

Let $(\mathbb{W}, \mathcal{E})$ be a measurable space.

$E \in \mathcal{E}$ is said to be **atomic** : \Leftrightarrow

$$[[E' \in \mathcal{E}, E' \subseteq E]] \Rightarrow [[E' = E \text{ or } E' = \emptyset]].$$

Examples:

► For $(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$ the atoms are the singletons.

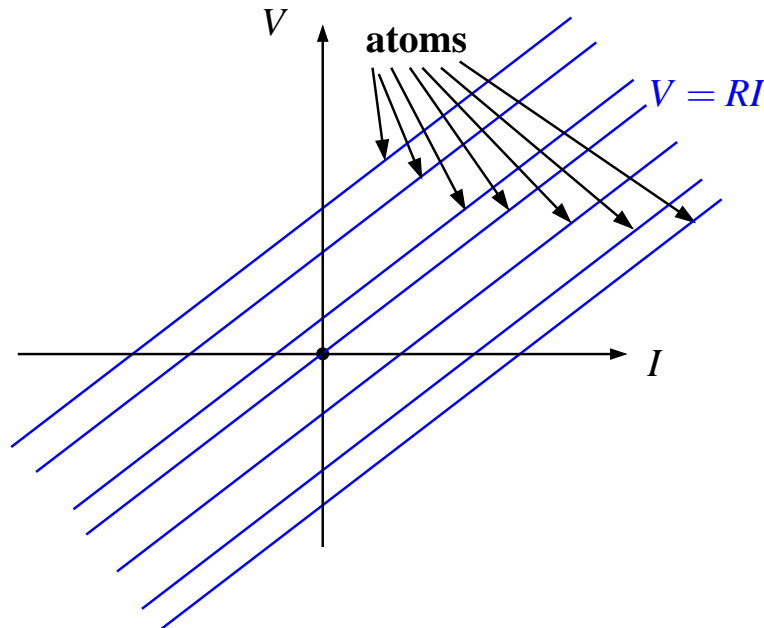
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- For the noisy resistor the atoms are the lines parallel to $V = RI$.



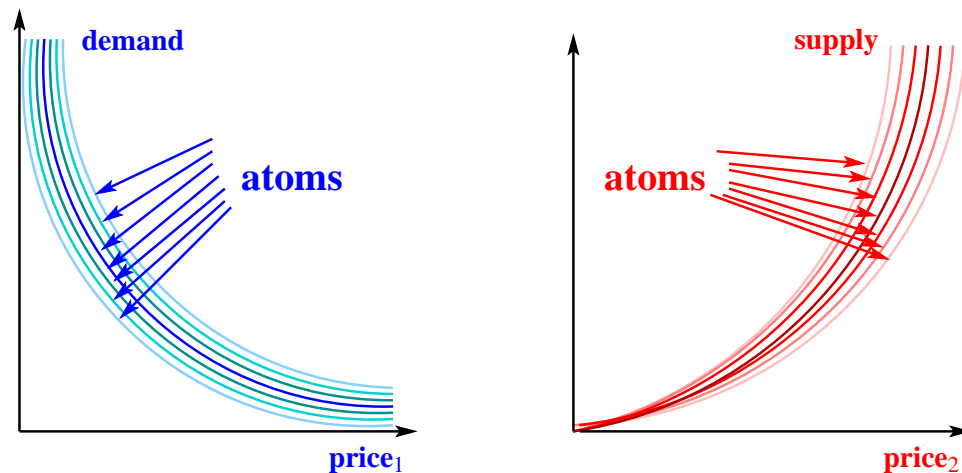
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- ▶ **Economic example: atoms are (p_1, d) hyperbolas; (p_2, s) parabolas.**



Linear stochastic system

$(\mathbb{R}^n, \mathcal{E}, P)$ is said to be a *linear stochastic system*

$:\Leftrightarrow \exists \mathbb{L}$, a linear subspace of \mathbb{R}^n , such that

\mathbb{L} is an atomic event, and

$$\llbracket E \in \mathcal{E} \text{ is atomic} \rrbracket \Leftrightarrow \llbracket E = E + \mathbb{L} \rrbracket.$$

I.e., all atoms are cylinders with sides parallel to \mathbb{L} .

However, in the remainder of this lecture we will use the following more restrictive definition.

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

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

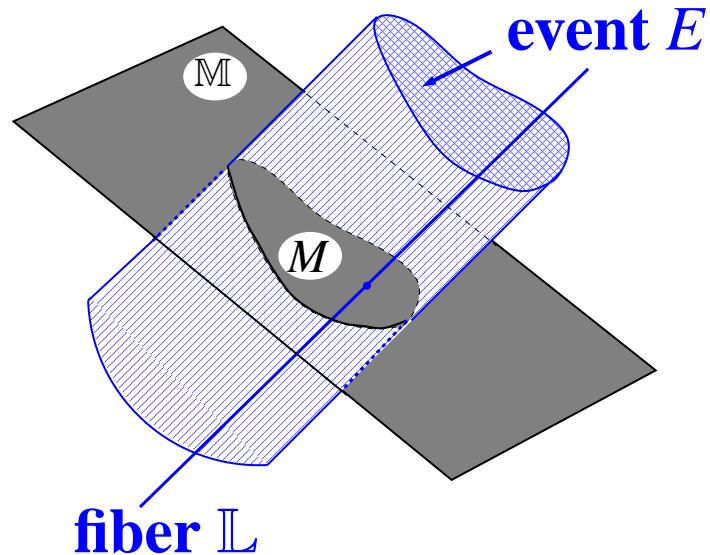
Subsets of \mathbb{R}^n as $A + \mathbb{L}$, \mathbb{L} linear subspace, A Borel.

Linearity

linear stochastic system

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



Borel probability on $M \cong \mathbb{R}^n/\mathbb{L}$, $(M \oplus \mathbb{L} = \mathbb{R}^n)$.

Classical \Rightarrow linear!

gaussian $:\Leftrightarrow$ **linear, Borel probability gaussian.**

Deterministic system

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

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

Deterministic system

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Atoms of deterministic system: \mathbb{B} and $\mathbb{B}^{\text{complement}}$.

Linear deterministic $\Leftrightarrow \mathbb{B}$ linear subspace of \mathbb{R}^n .

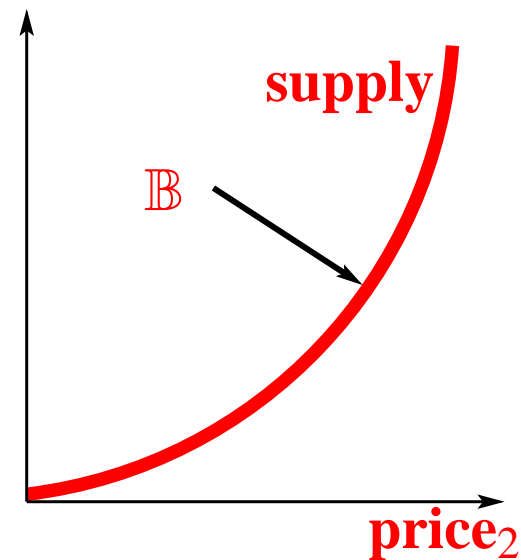
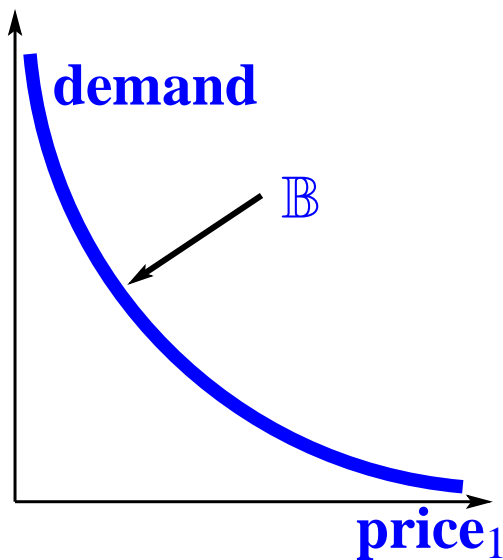
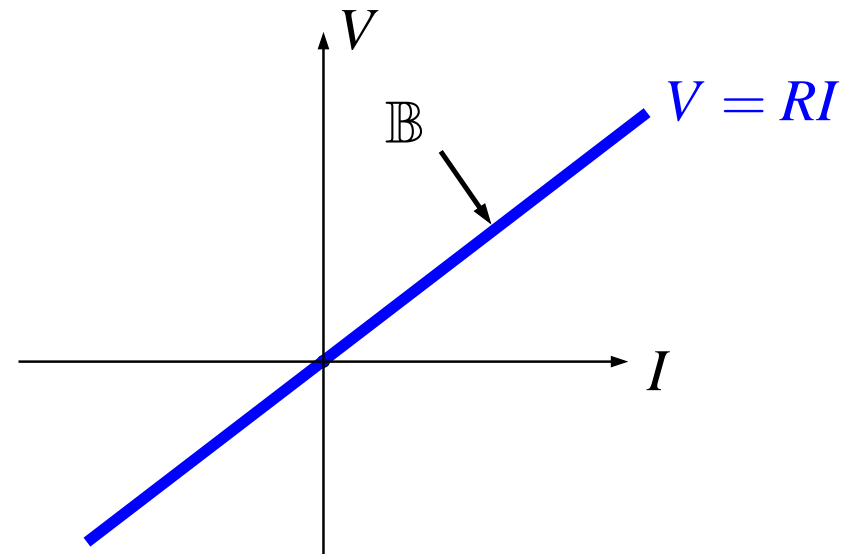
\mathbb{B} is called the **behavior**.

Deterministic examples

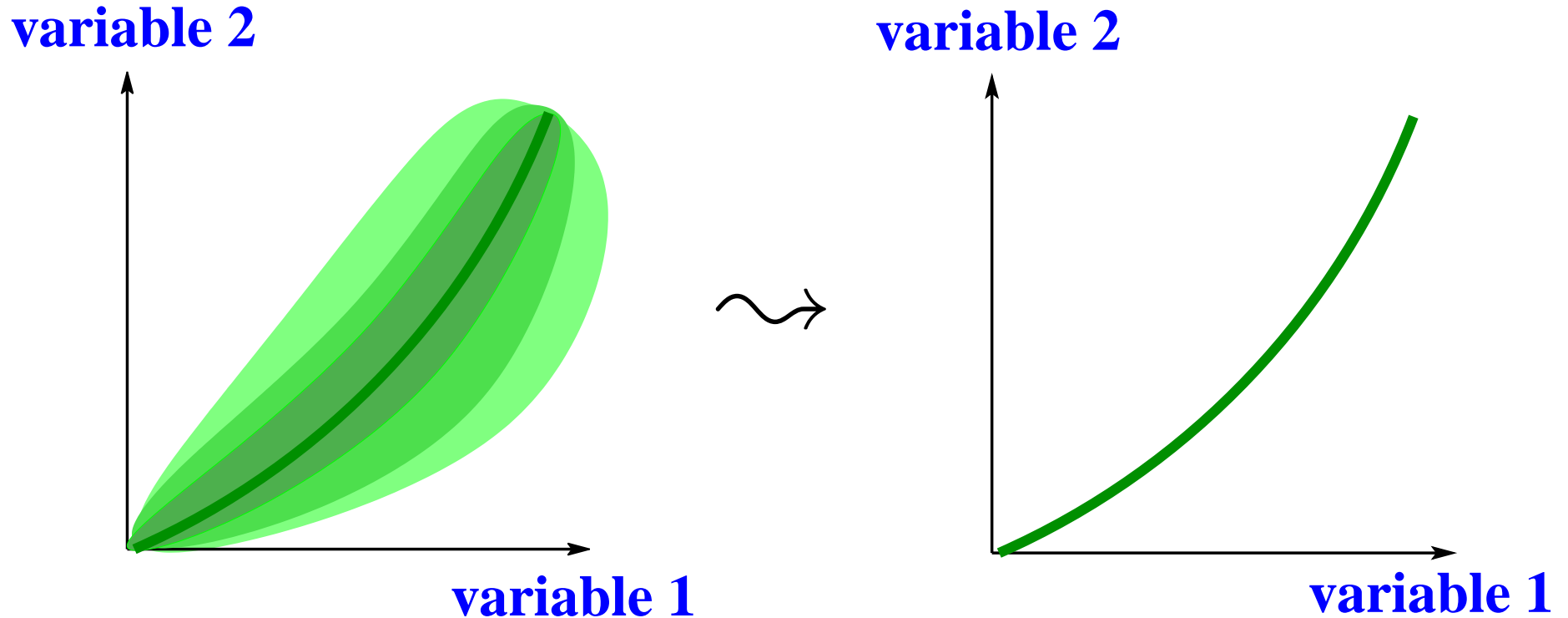
An Ohmic resistor,

$$\mathbb{W} = \mathbb{R}^2,$$

$$\mathbb{B} = \left\{ \begin{bmatrix} V \\ I \end{bmatrix} \in \mathbb{R}^2 \mid V = RI \right\}.$$



The need for 'coarse' σ -algebras

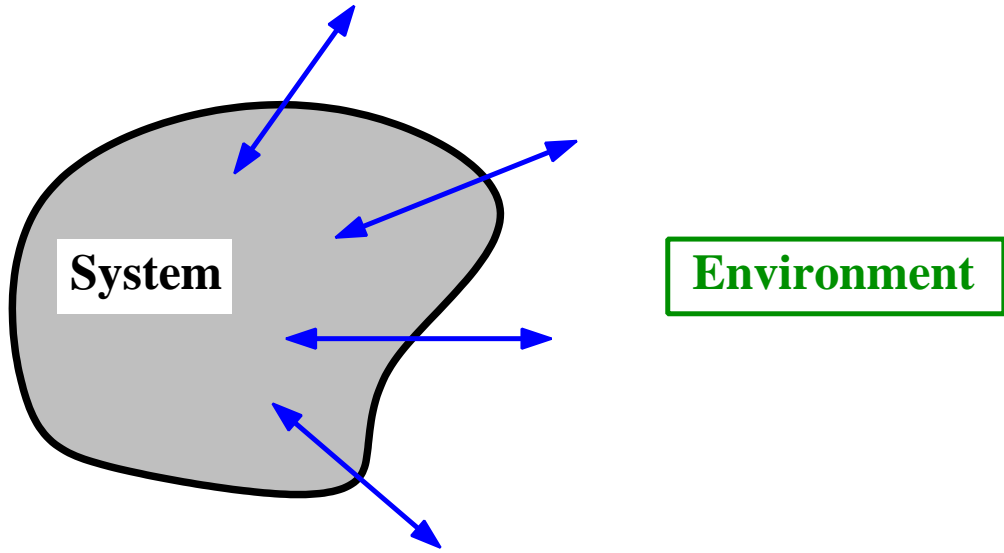


**For a classical random vector, the deterministic limit
 \simeq a (singular) probability distribution.**

Awkward from the modeling point of view.

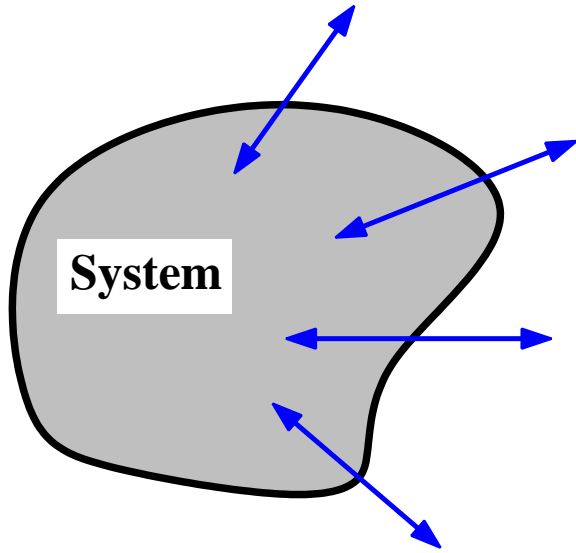
Interconnection

System theoretic musts



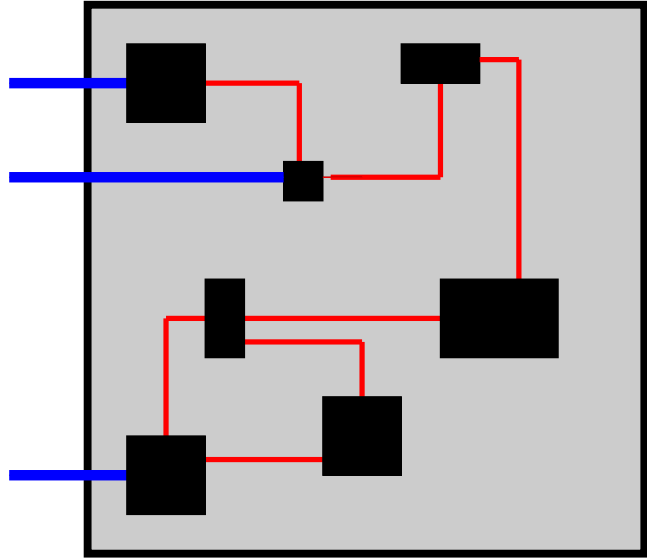
Open

System theoretic musts



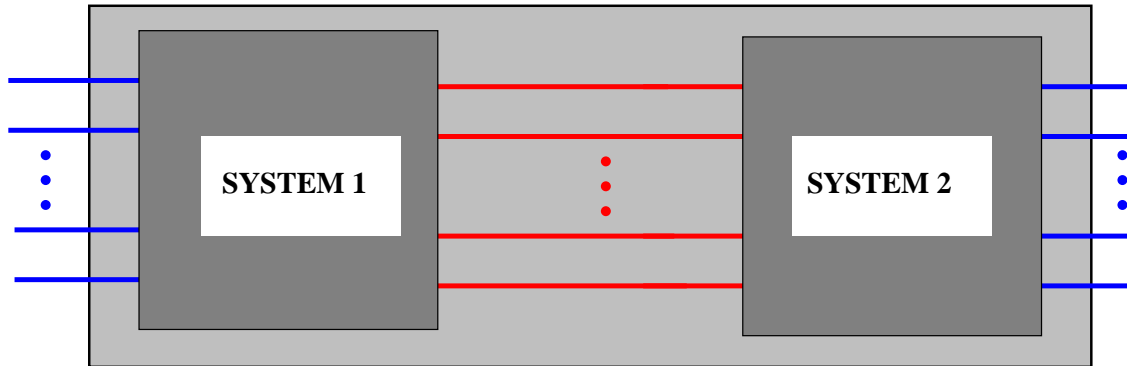
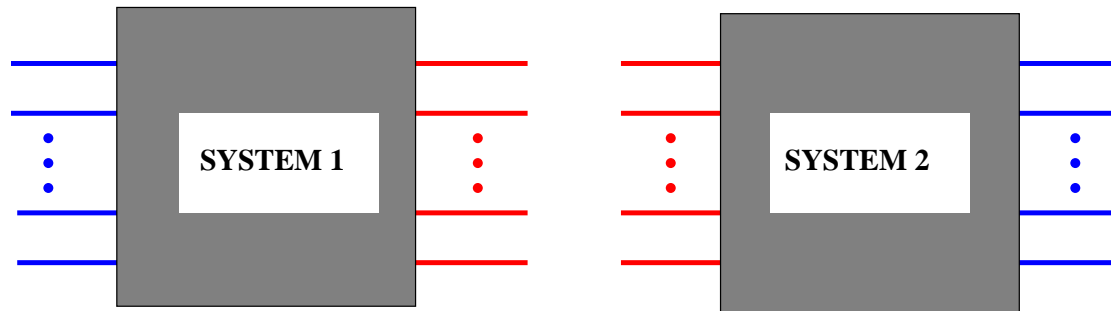
Environment

Open

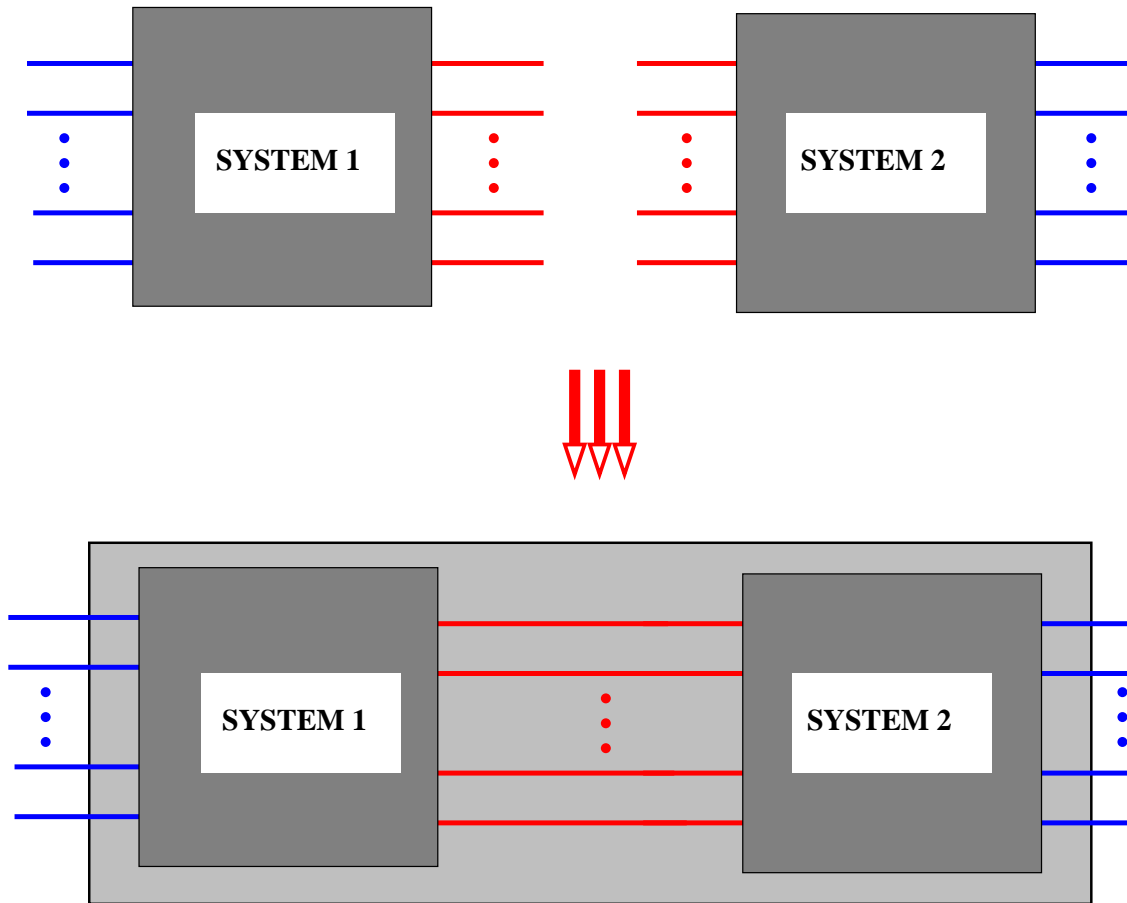


Connectable

Interconnection

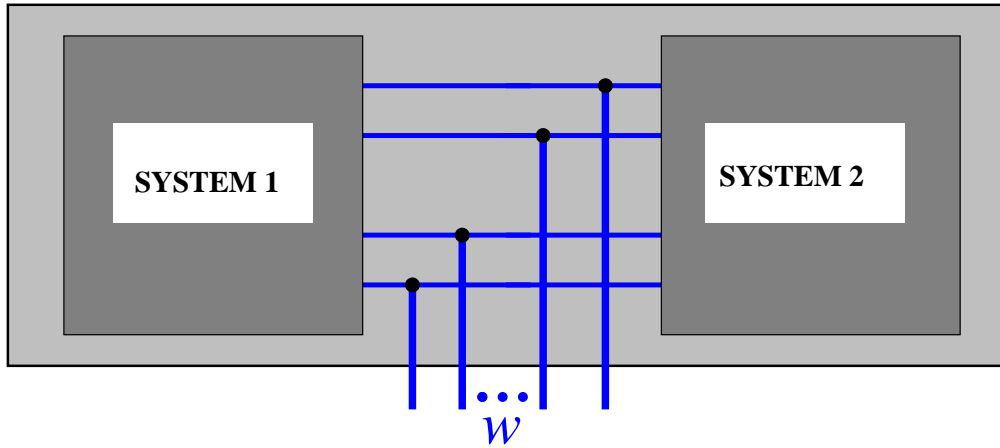
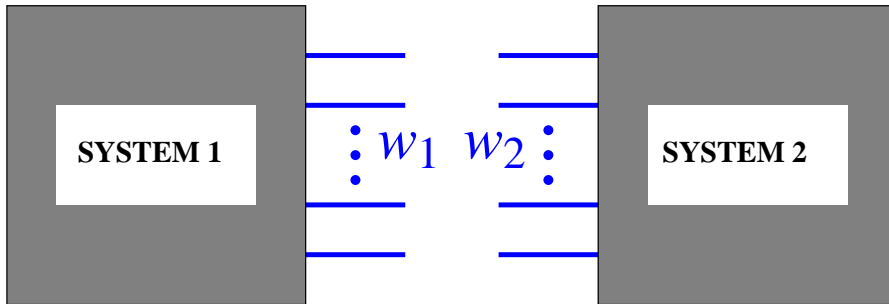


Interconnection

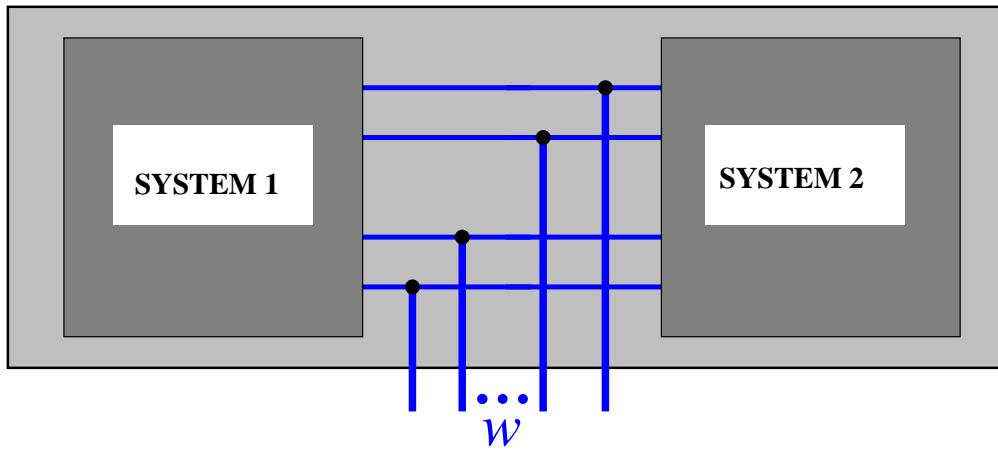
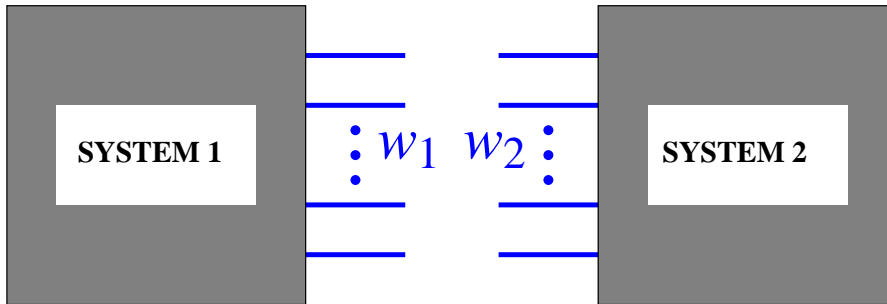


Consider the following (seemingly) special case.

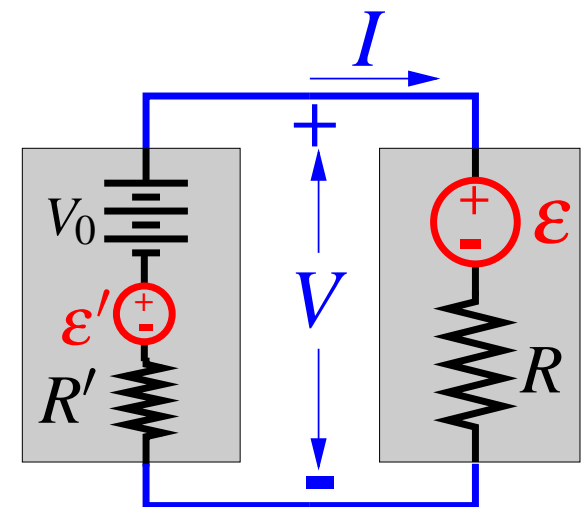
Interconnection



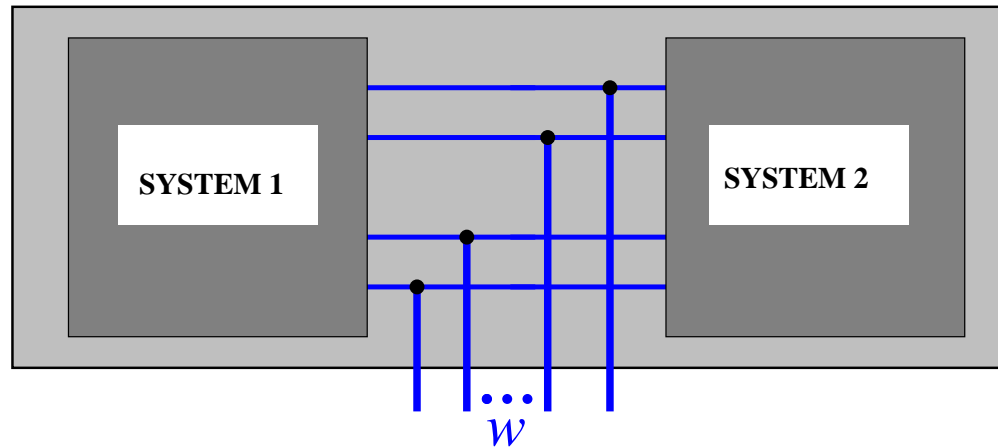
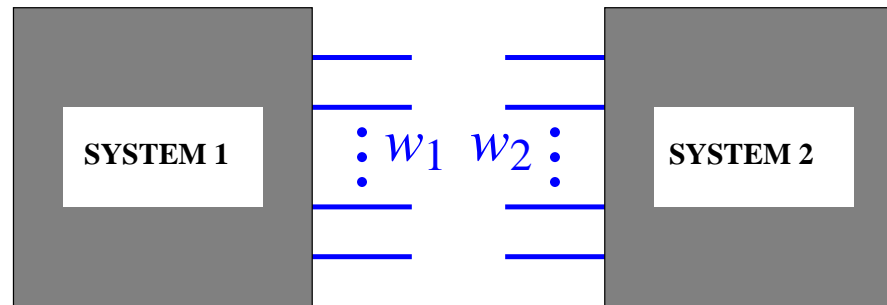
Interconnection



Example:



Interconnection

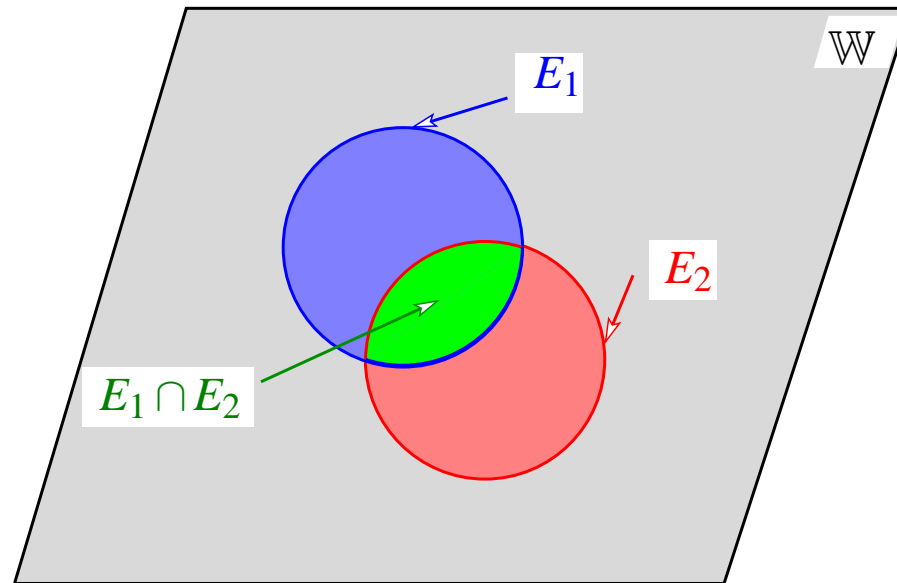


**Can we impose two distinct probabilistic laws
on the same set of variables?**

Complementarity of σ -algebras

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

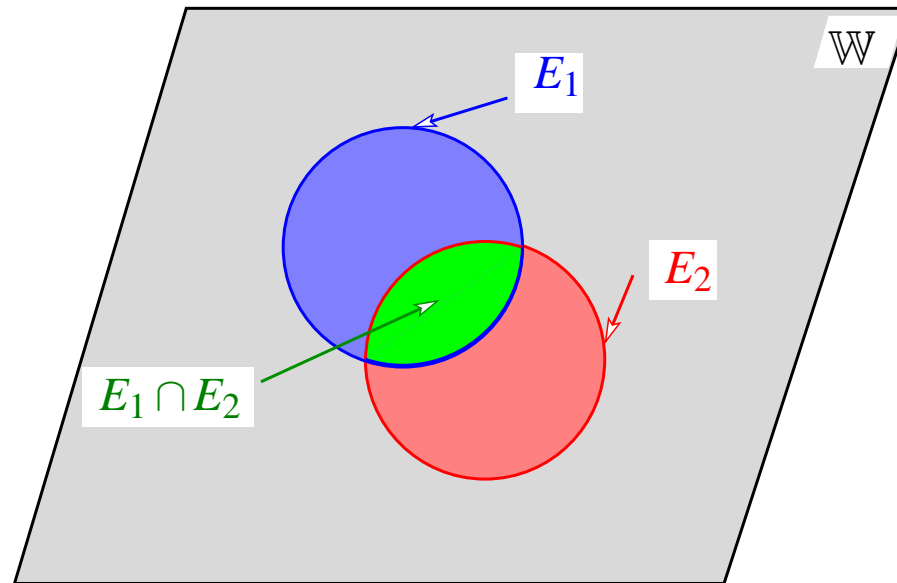
$$[[E_1 \cap E_2 = E'_1 \cap E'_2]] \Rightarrow [[E_1 = E'_1 \text{ and } E_2 = E'_2]].$$



Complementarity of σ -algebras

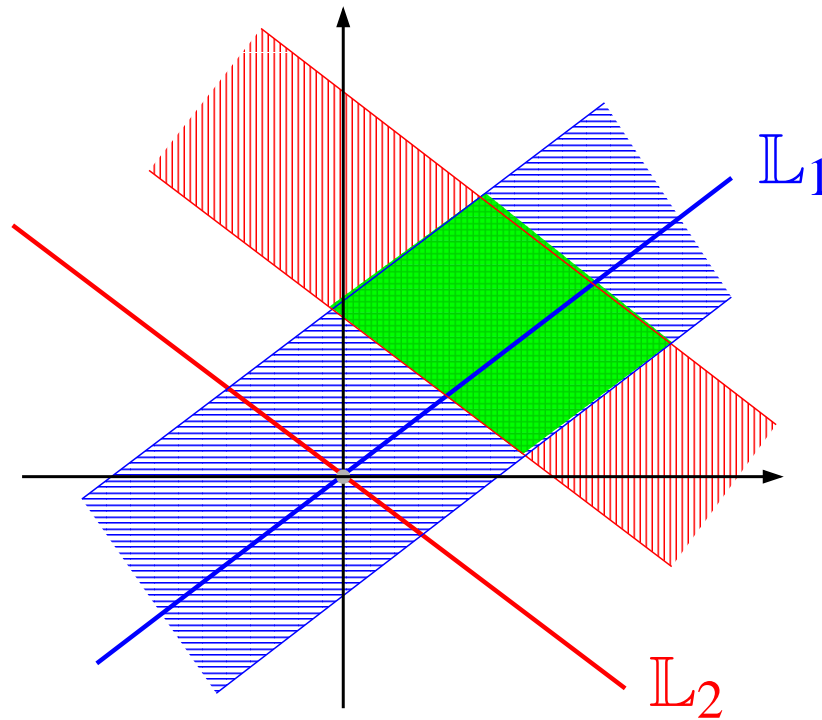
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$$[[E_1 \cap E_2 = E'_1 \cap E'_2]] \Rightarrow [[E_1 = E'_1 \text{ and } E_2 = E'_2]].$$



The intersection determines the intersectants.

Linear example

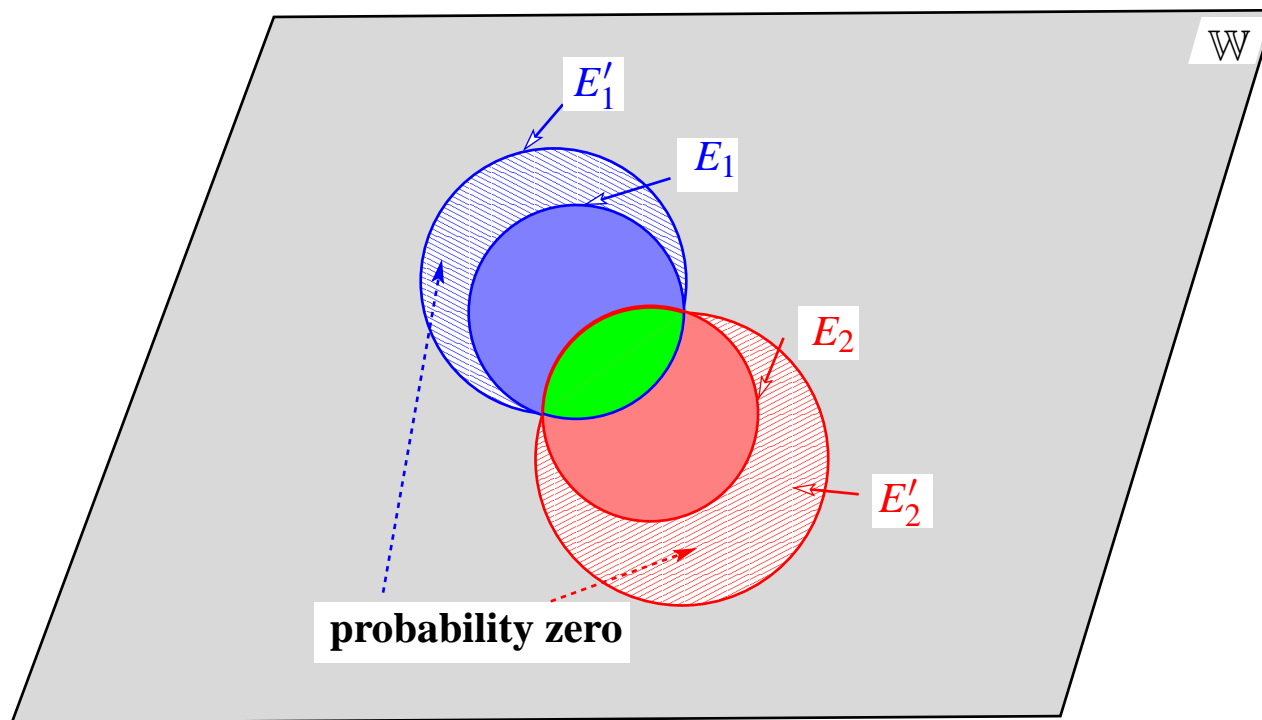


$$L_1 + L_2 = \mathbb{R}^n$$

Complementarity of systems

$(\mathbb{W}, \mathcal{E}_1, P_1)$ and $(\mathbb{W}, \mathcal{E}_2, P_2)$ are said to be **complementary** $:\Leftrightarrow$ for $E_1, E'_1 \in \mathcal{E}_1$ and $E_2, E'_2 \in \mathcal{E}_2$:

$$\llbracket E_1 \cap E_2 = E'_1 \cap E'_2 \rrbracket \Rightarrow \llbracket P_1(E_1)P_2(E_2) = P_1(E'_1)P_2(E'_2) \rrbracket.$$



Intersection \Rightarrow product of probabilities of intersectants.

Interconnection of complementary systems

Let $(\mathbb{W}, \mathcal{E}_1, P_1)$ and $(\mathbb{W}, \mathcal{E}_2, P_2)$ be stochastic systems (stochastically independent). Assume complementarity. Their *interconnection* is defined as

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

with $\mathcal{E} :=$ the σ -algebra generated by ‘rectangles’

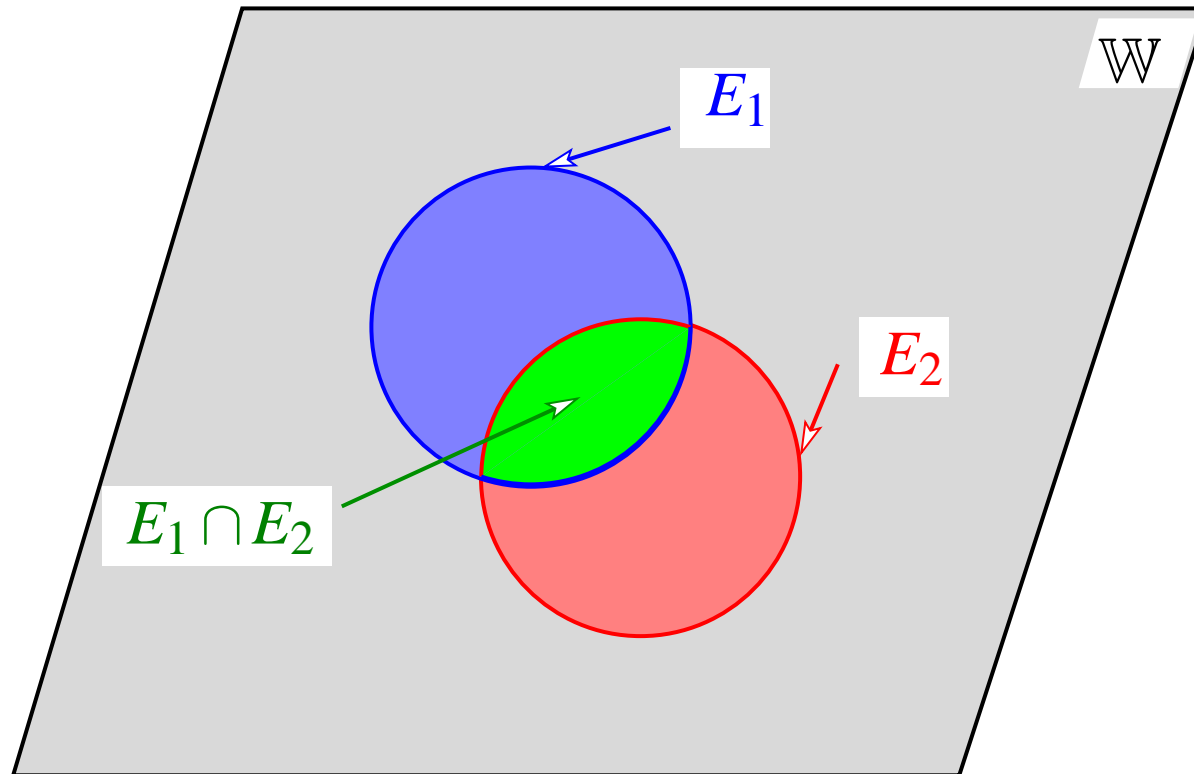
$$\{E_1 \cap E_2 \mid E_1 \in \mathcal{E}_1, E_2 \in \mathcal{E}_2\},$$

and P defined through the rectangles by

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

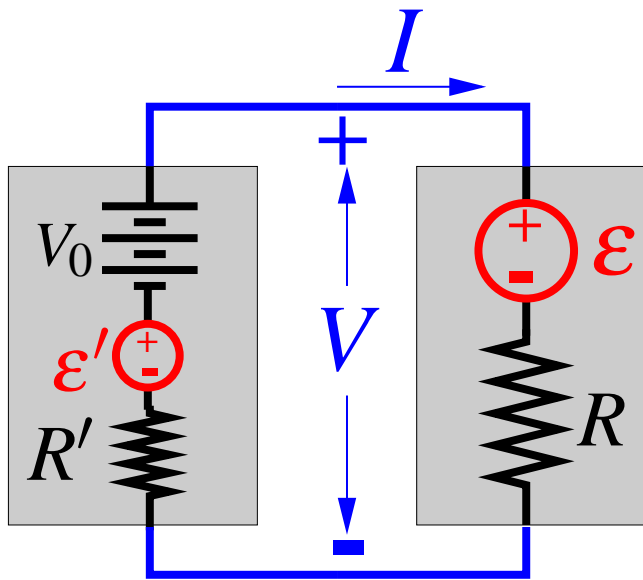
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Interconnection of complementary systems

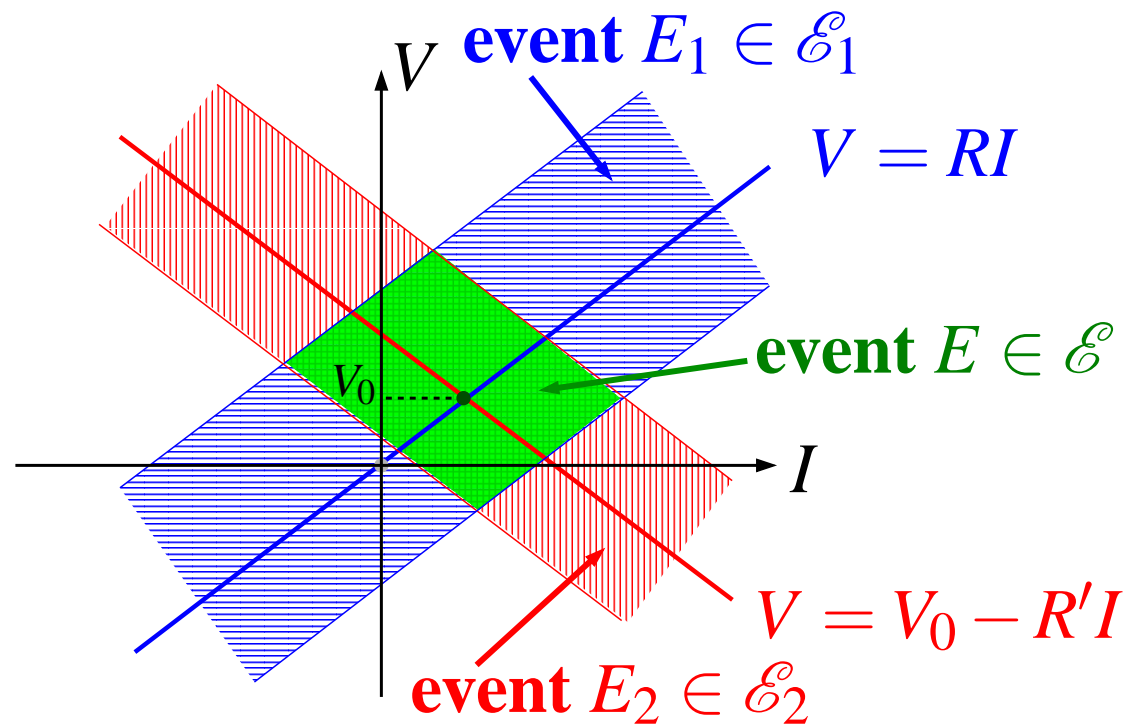
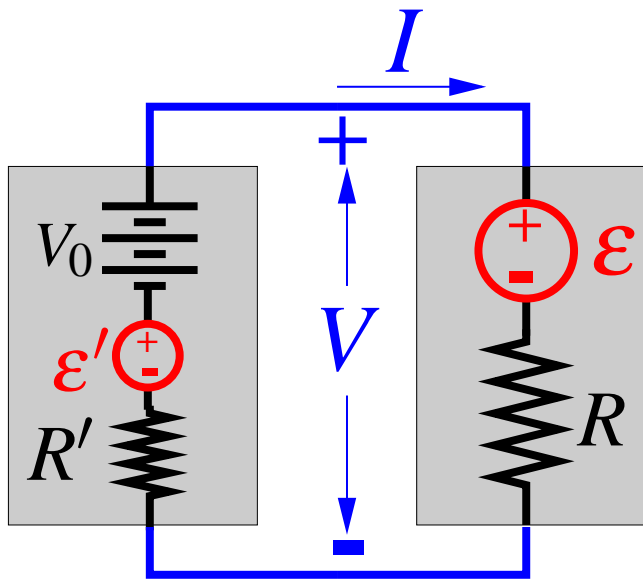


$$P(E_1 \cap E_2) := P_1(E_1)P_2(E_2).$$

Noisy resistor terminated by voltage source

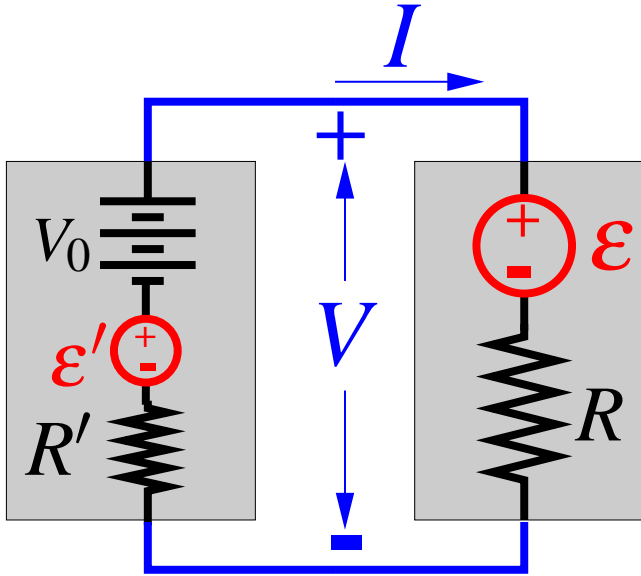


Noisy resistor terminated by voltage source



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

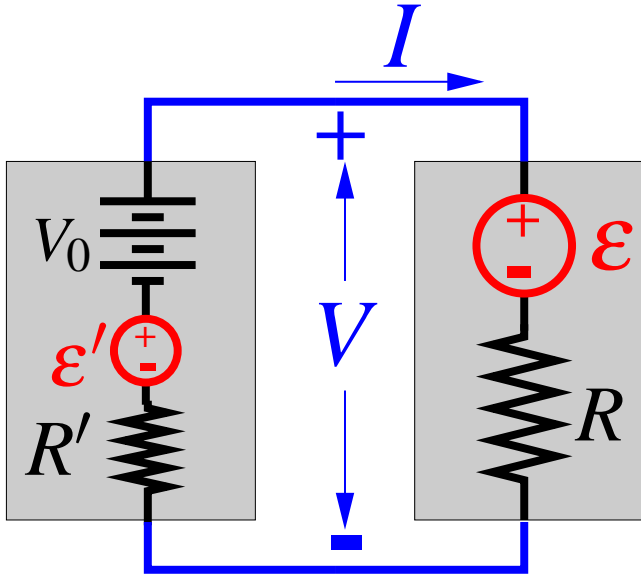
In terms of equations



$$V = RI + \varepsilon$$

$$V = V_0 - R'I + \varepsilon'$$

In terms of equations



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$$V = V_0 - R'I + \varepsilon'$$

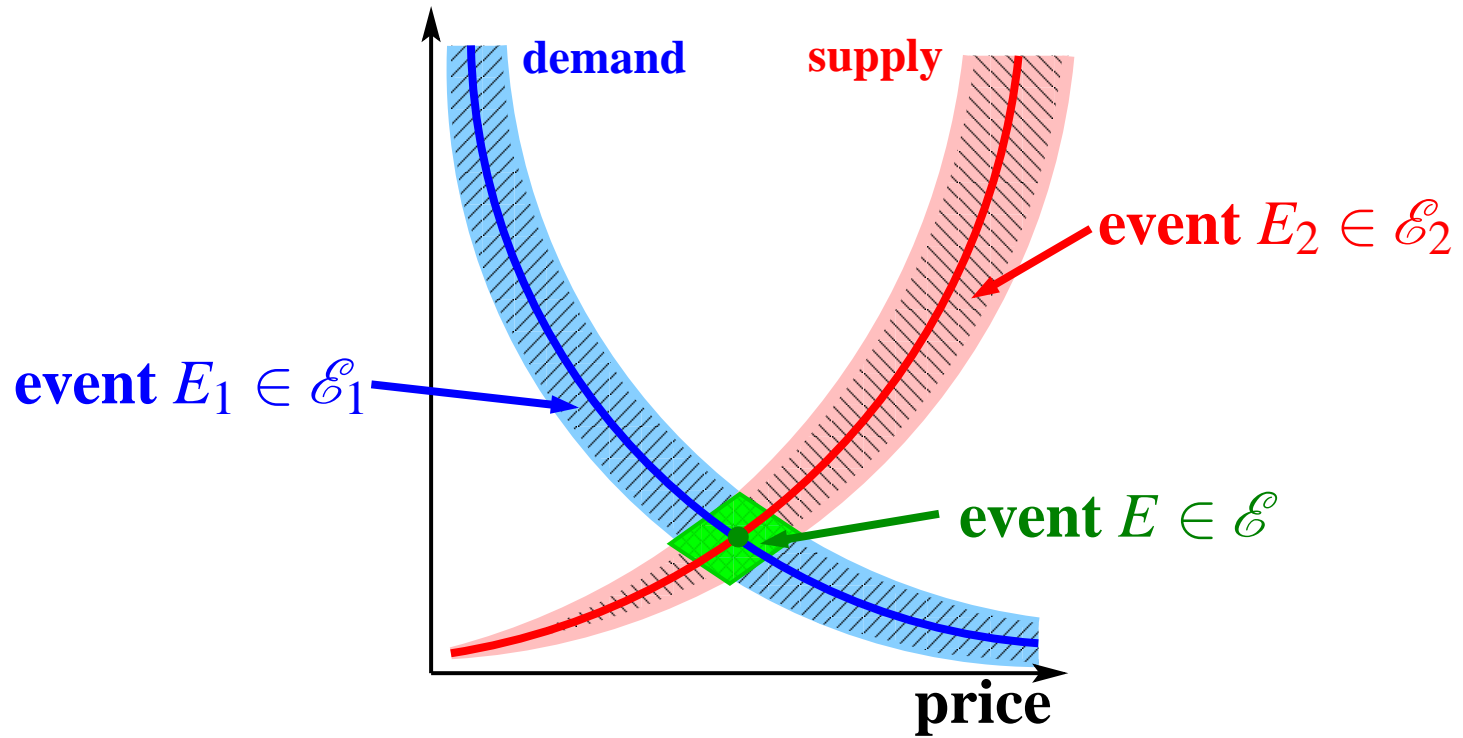
$$V = \frac{1}{R + R'} (R'\varepsilon + R(V_0 + \varepsilon'))$$

$$I = \frac{1}{R + R'} (-\varepsilon + V_0 + \varepsilon').$$

Shows that $\begin{bmatrix} V \\ I \end{bmatrix}$ is indeed a classical random vector.

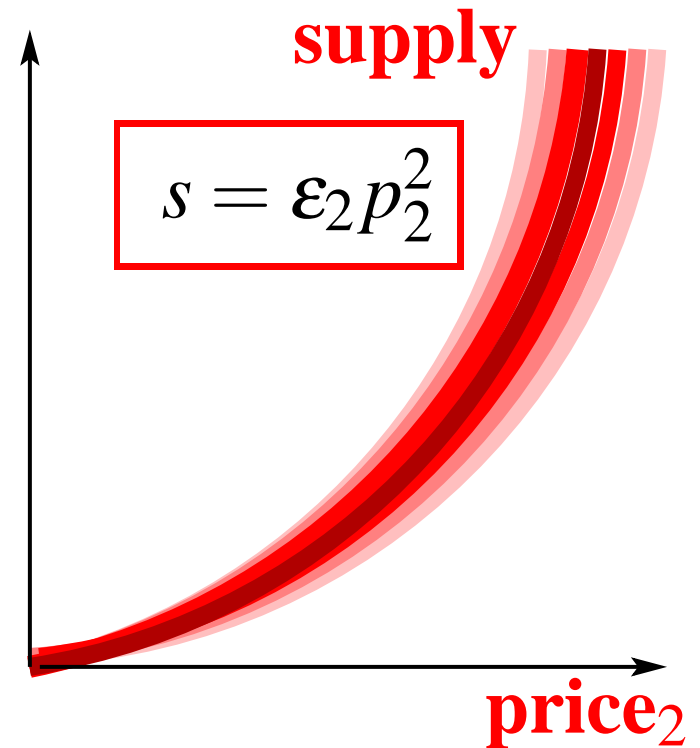
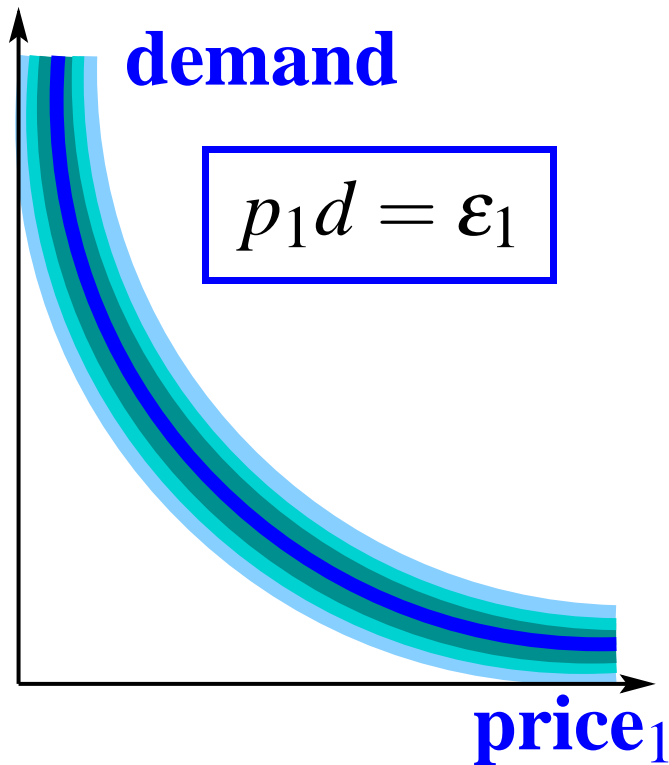
Complementarity \Rightarrow this construction geometrically.

Equilibrium price/demand/supply



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

In terms of equations



$$p_1 = p_2 = \sqrt[3]{\frac{\varepsilon_1}{\varepsilon_2}}, \quad d = s = \sqrt[3]{\varepsilon_1^2 \varepsilon_2}.$$

Complementarity \Rightarrow this construction geometrically.

Open stochastic systems

Open versus closed

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

**If \mathcal{E}_1 = the Borel σ -algebra, and $\text{support}(P_1) = \mathbb{R}^n$,
then Σ_1 is interconnectable only with the free system
 $(\mathbb{R}^n, \mathcal{E}_2, P_2)$, $\mathcal{E}_2 = \{\emptyset, \mathbb{R}^n\}$.**

\Rightarrow classical $\Sigma_1 =$ ‘closed’ system.

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\Rightarrow classical $\Sigma_1 =$ ‘closed’ system.

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 (Ω, \mathcal{A}, P) .

We view the randomness as ‘internal’ to the system.

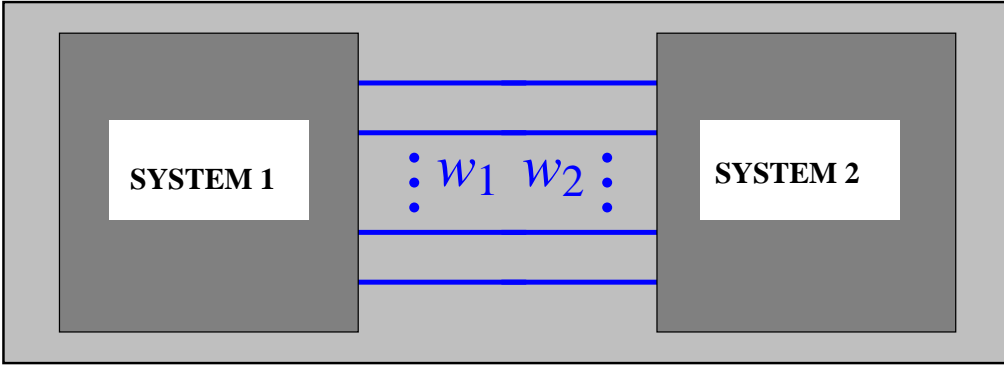
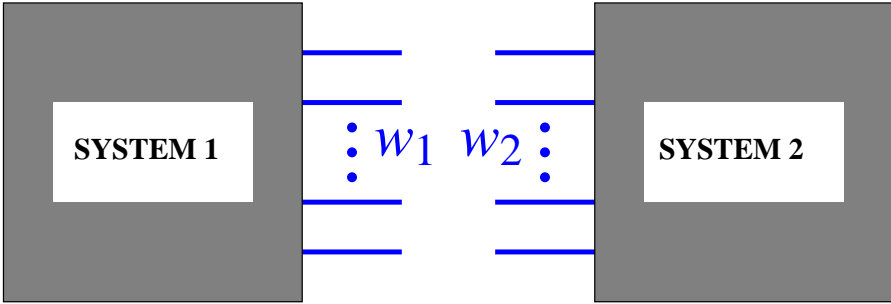
So, once the Gods choose $\omega \in \Omega$, all the random variables are determined.

The environment has no influence on the outcomes.

\Rightarrow ‘closed’ systems.

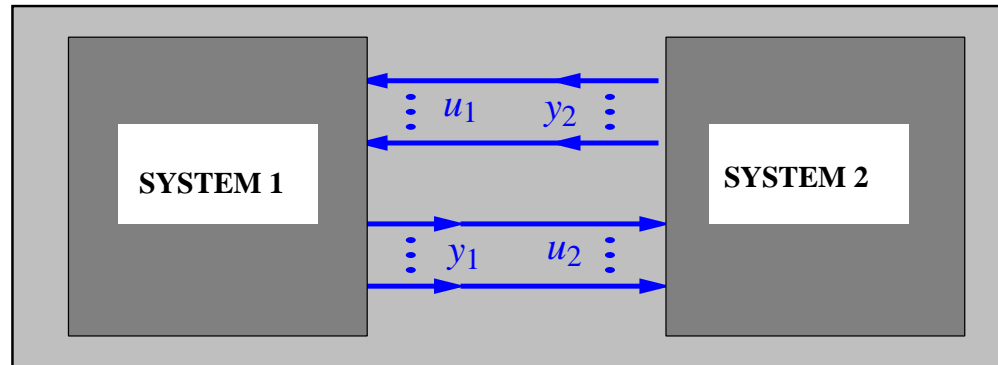
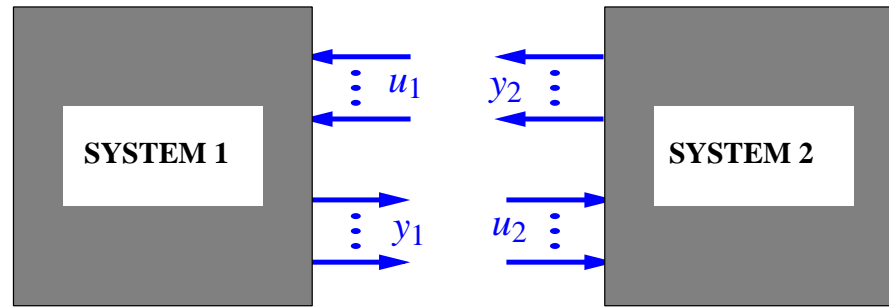
Interconnection \Leftrightarrow variable sharing

Variable sharing



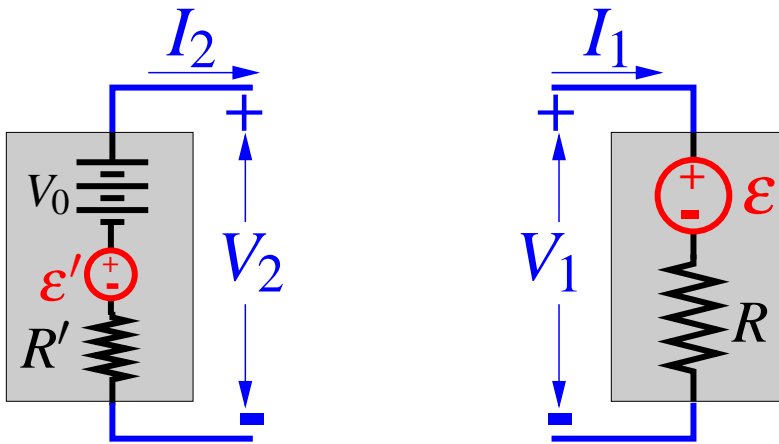
$$w_1 = w_2$$

Output-to-input assignment



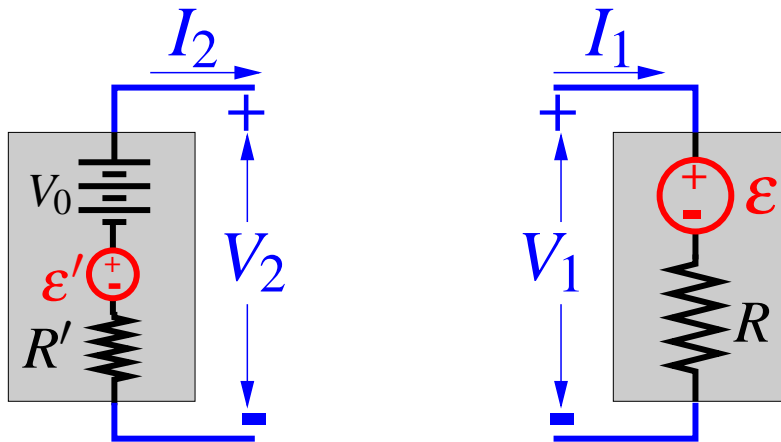
$$u_1 = y_2, \quad u_2 = y_1$$

Resistor interconnection

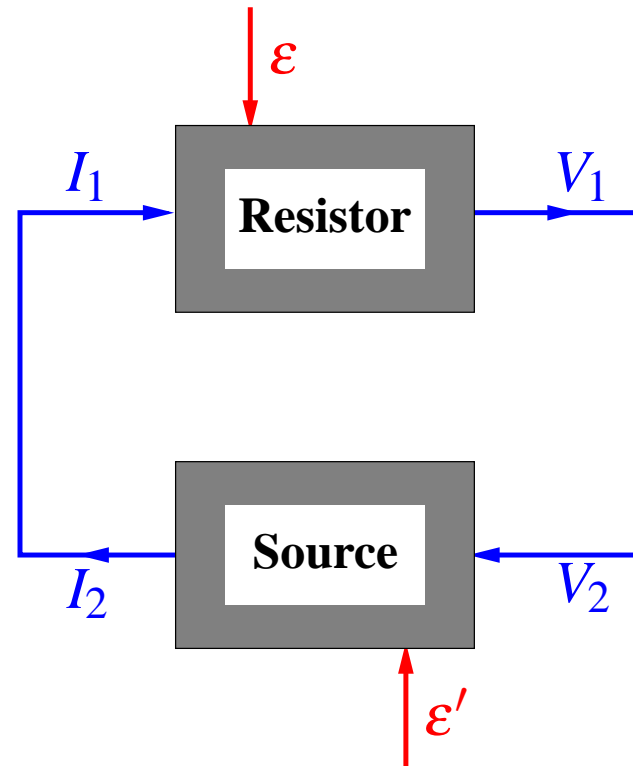


$$V_1 = V_2, \quad I_1 = I_2$$

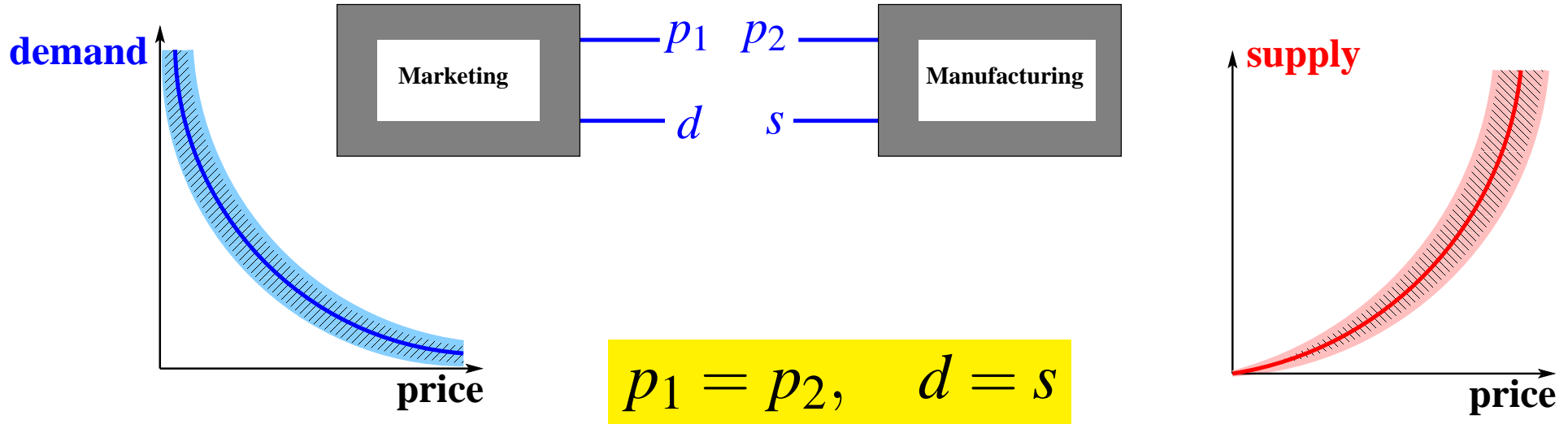
Resistor interconnection



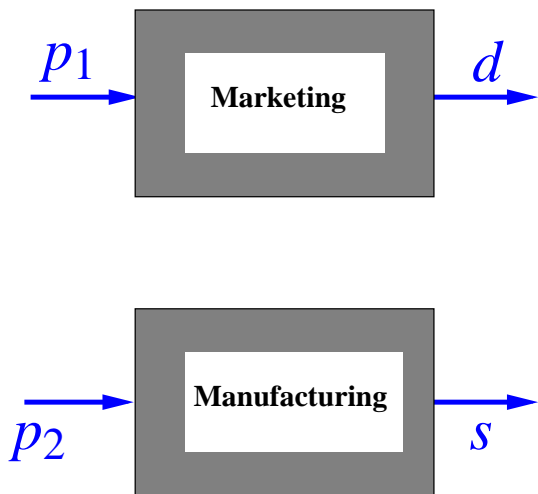
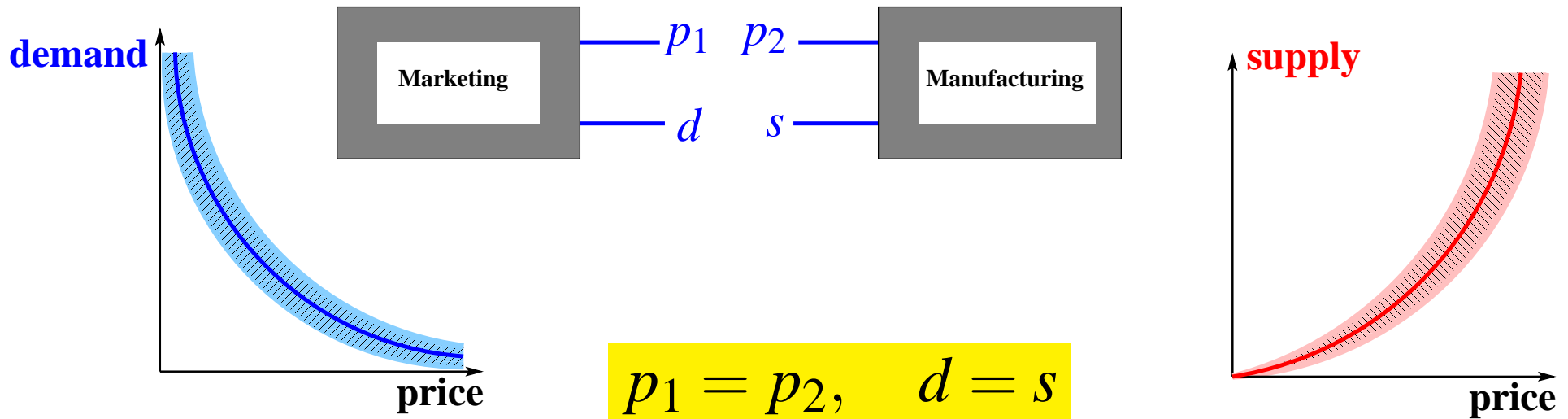
$$V_1 = V_2, \quad I_1 = I_2$$



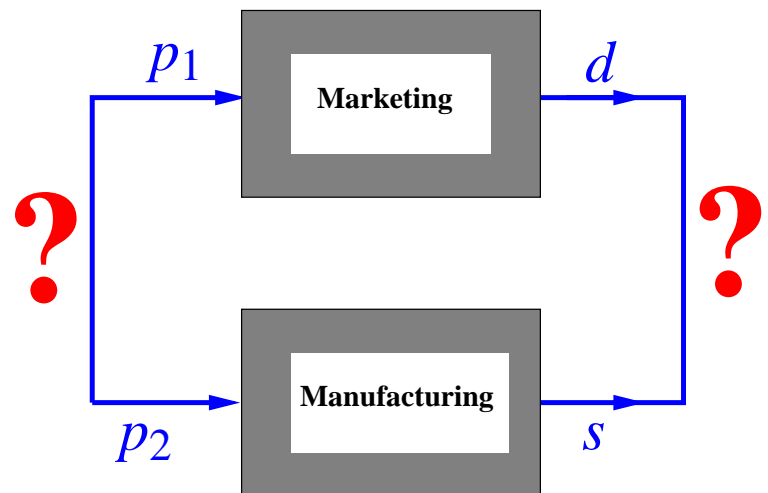
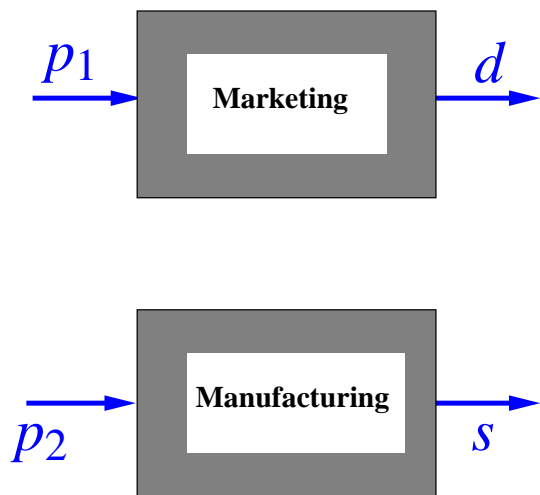
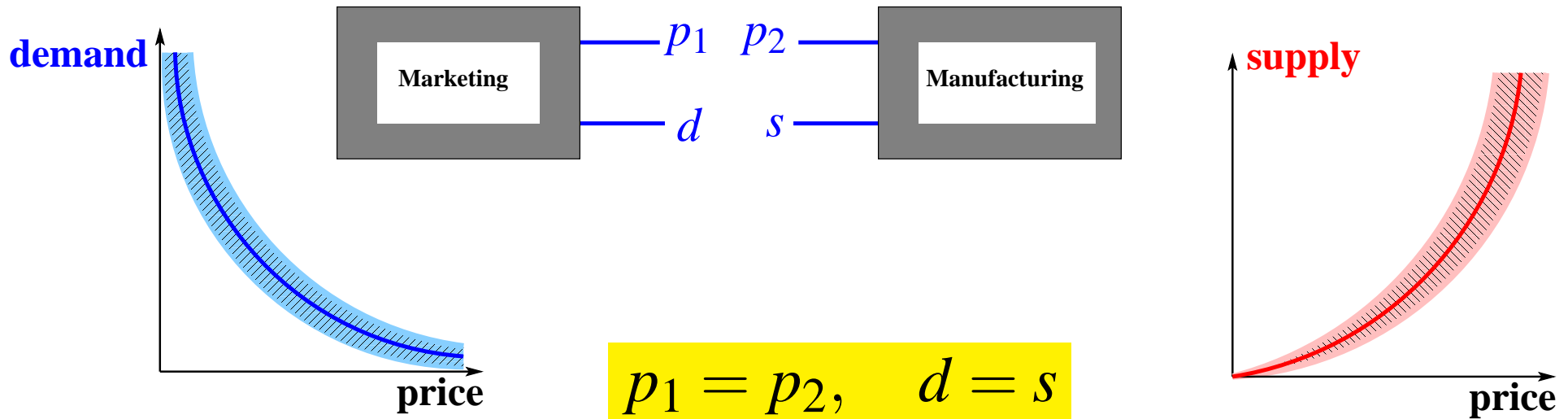
Price/demand/supply interconnection



Price/demand/supply interconnection



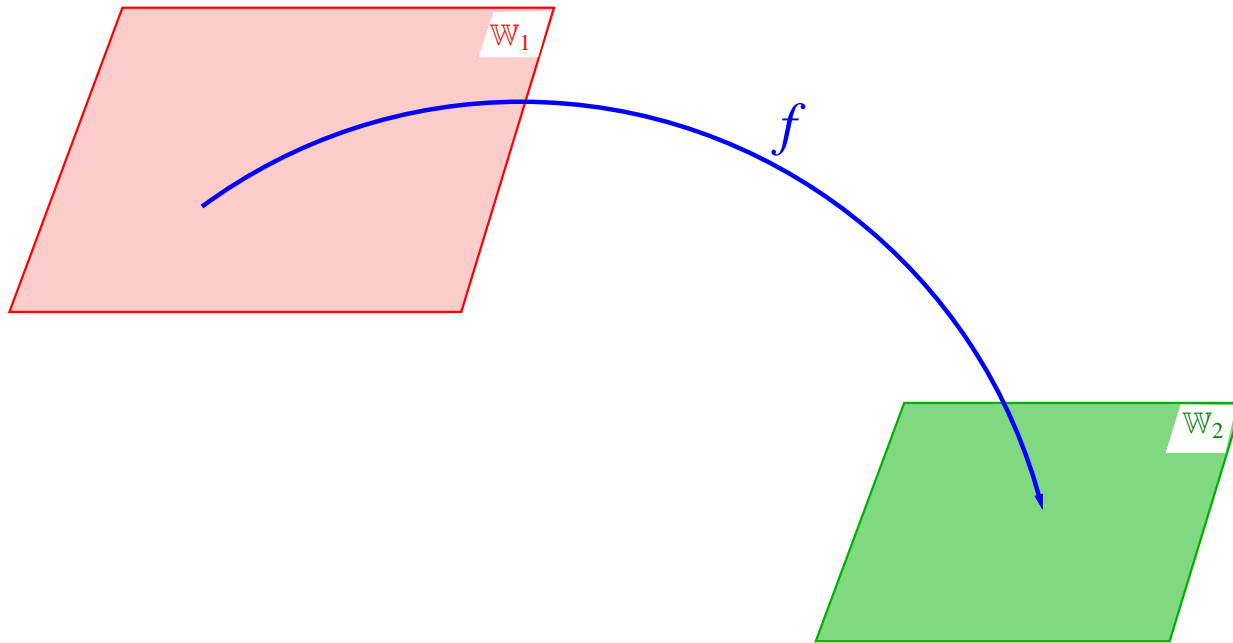
Price/demand/supply interconnection



Functions

Problem

Consider the stochastic system $(\mathbb{W}_1, \mathcal{E}_1, P_1)$,
and the map $\mathbb{W}_1 \xrightarrow{f} \mathbb{W}_2$.

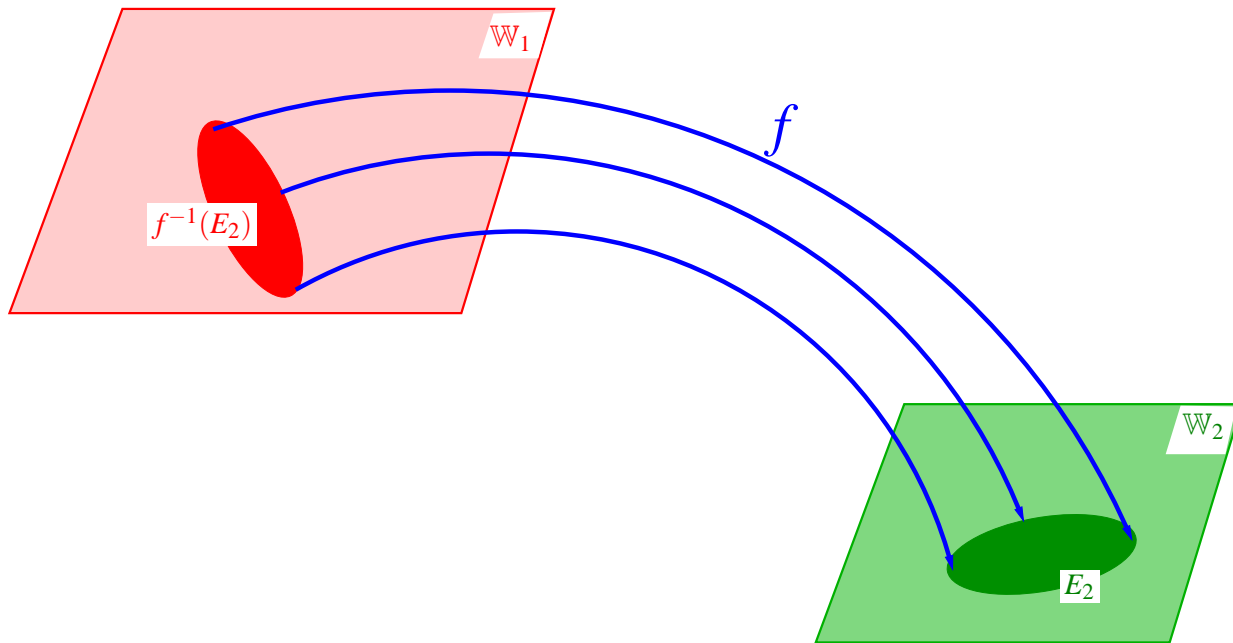


Which stochastic system on \mathbb{W}_2 does f generate?

Pullback construction

If $(\mathbb{W}_1, \mathcal{E}_1) \xrightarrow{f} (\mathbb{W}_2, \mathcal{E}_2)$ is a measurable map: \Leftrightarrow

$$[[E_2 \in \mathcal{E}_2]] \Rightarrow [[f^{-1}(E_2) \in \mathcal{E}_1]],$$



then $P_2(E_2) := P_1(f^{-1}(E_2)) \rightsquigarrow (\mathbb{W}_2, \mathcal{E}_2, P_2)$.

Pullback construction

If $(\mathbb{W}_1, \mathcal{E}_1) \xrightarrow{f} (\mathbb{W}_2, \mathcal{E}_2)$ is a measurable map: \Leftrightarrow

$$\llbracket E_2 \in \mathcal{E}_2 \rrbracket \Rightarrow \llbracket f^{-1}(E_2) \in \mathcal{E}_1 \rrbracket,$$

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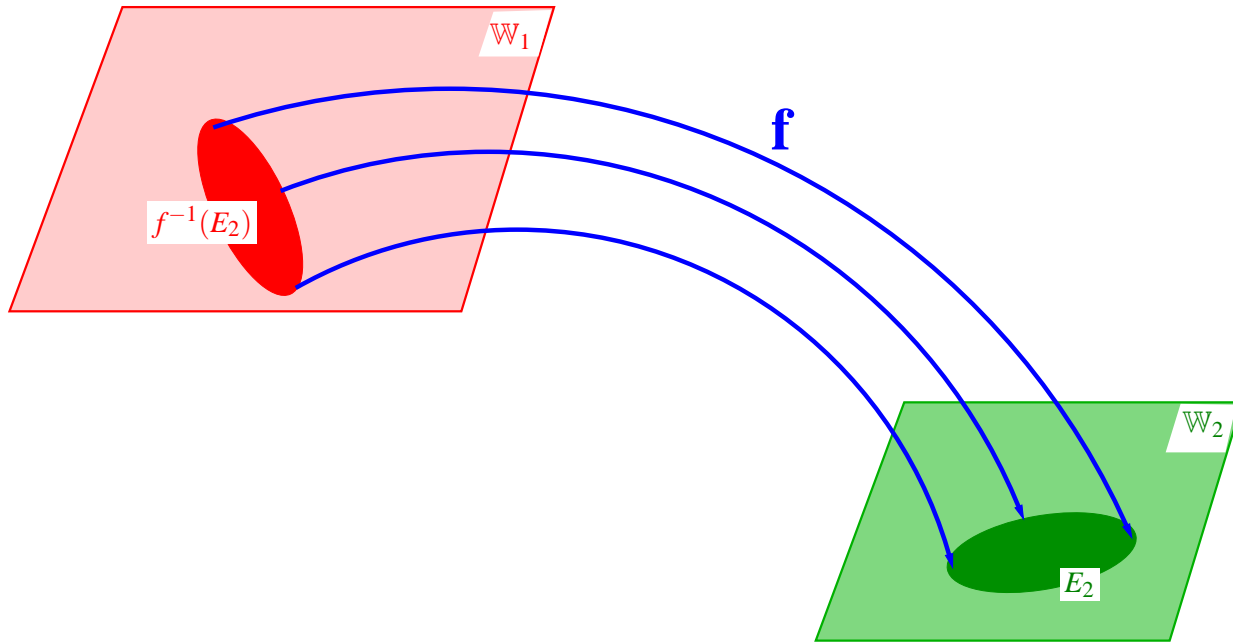
Usually, it is assumed that \mathcal{E}_2 is given, say as $\mathcal{B}(\mathbb{R}^n)$, and that f is measurable.

However, since the events are part of the stochastic phenomenon, \mathcal{E}_2 ought to be constructed.

Construction of \mathcal{E}_2

$$\mathcal{E}_2 := \{E_2 \subseteq \mathbb{W}_2 \mid f^{-1}(E_2) \in \mathcal{E}_1\}.$$

\mathcal{E}_2 is a σ -algebra, \rightsquigarrow stochastic system $(\mathbb{W}_2, \mathcal{E}_2, P_2)$
with $P_2(E_2) = P_1(f^{-1}(E_2))$.



$\mathcal{E}_2 =$ those subsets to which a probability can be assigned.

\mathcal{E}_2 is modeled, not obtained from the topology on \mathbb{W}_2 .

Example

Noisy resistor, $V = RI + \varepsilon$, $R \neq 0$.

• $f : \begin{bmatrix} V \\ I \end{bmatrix} \mapsto V.$

\rightsquigarrow **stochastic system** $(\mathbb{R}, \{\emptyset, \mathbb{R}\}, P_2).$

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• $f : \begin{bmatrix} V \\ I \end{bmatrix} \mapsto I.$

\rightsquigarrow **stochastic system** $(\mathbb{R}, \{\emptyset, \mathbb{R}\}, P_2).$

• $f : \begin{bmatrix} V \\ I \end{bmatrix} \mapsto V - RI.$

\rightsquigarrow **stochastic system** $(\mathbb{R}, \mathcal{B}(\mathbb{R}), P_2)$

with $P_2 = \mathcal{N}(0, \sim \sqrt{RT}).$

Independence

Independence of stochastic variables

Independence of events is a measure theoretic concept. Does not need adjustment.

Let $(\mathbb{W}, \mathcal{E}, P)$ be a stochastic system.

Consider $\mathbb{W} \xrightarrow{f_1} \mathbb{W}_1, \quad \mathbb{W} \xrightarrow{f_2} \mathbb{W}_2$

\rightsquigarrow stochastic systems $(\mathbb{W}_1, \mathcal{E}_1, P_1)$, and $(\mathbb{W}_2, \mathcal{E}_2, P_2)$.

Are the outcomes w_1 and w_2 stochastically independent?

Independence of stochastic variables

$$\mathbb{W} \xrightarrow{f_1} \mathbb{W}_1, \quad \mathbb{W} \xrightarrow{f_2} \mathbb{W}_2.$$

Consider also $\mathbb{W} \xrightarrow{(f_1, f_2)} \mathbb{W}_1 \times \mathbb{W}_2$.

$$\rightsquigarrow \Sigma_{12} = (\mathbb{W}_1 \times \mathbb{W}_2, \mathcal{E}_{12}, P_{12}).$$

Independence $:\Leftrightarrow (\mathbb{W}_1 \times \mathbb{W}_2, \mathcal{E}_{12}, P_{12})$

is the product of $(\mathbb{W}_1, \mathcal{E}_1, P_1)$ **and** $(\mathbb{W}_2, \mathcal{E}_2, P_2)$.

Noisy resistor $R \neq 0$: V and I are not independent.

$R = 0$: V and I are independent.

Conditioning & Constraining

Conditional probability

Let $\Sigma = (\mathbb{W}, \mathcal{E}, P)$. Look at the outcomes $\boxed{w \in \mathbb{S}}$.

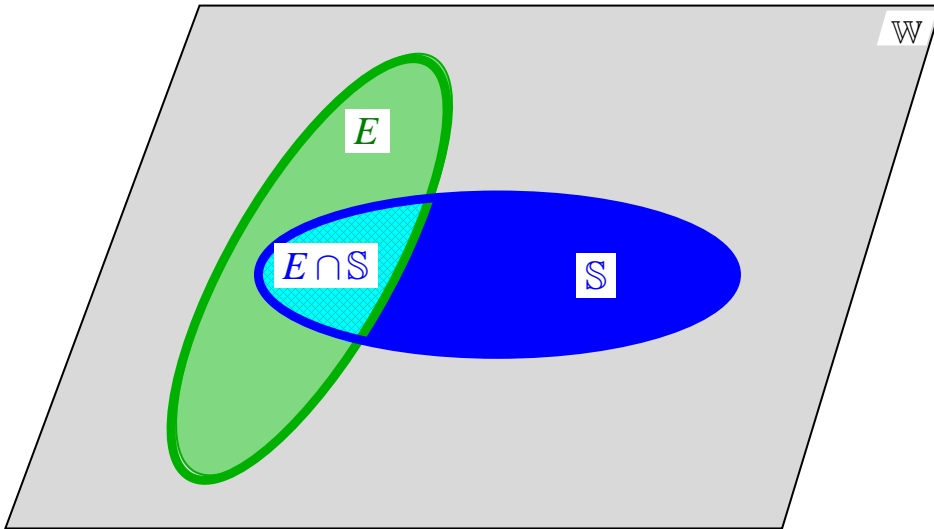
When \mathbb{S} is an event, that is $\mathbb{S} \in \mathcal{E}$,

\rightsquigarrow conditional probability. Assume $P(\mathbb{S}) > 0$. Then

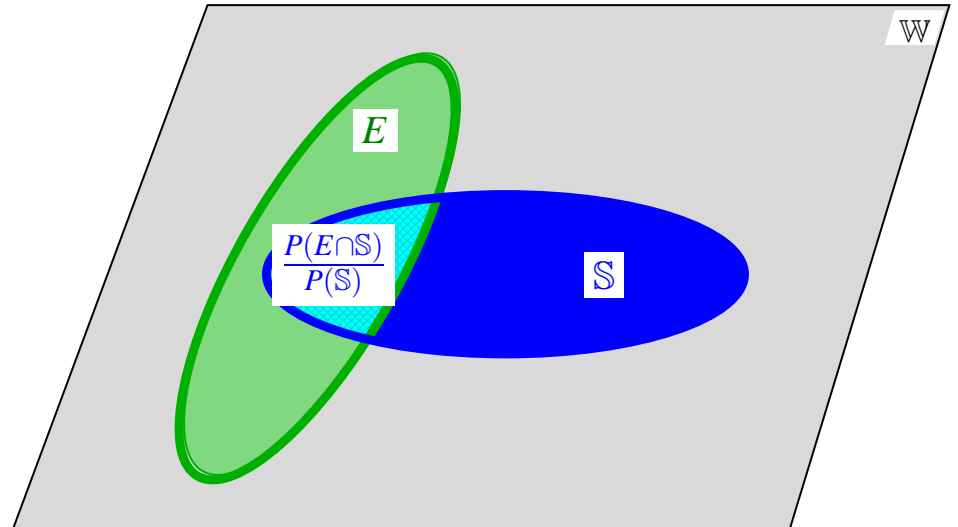
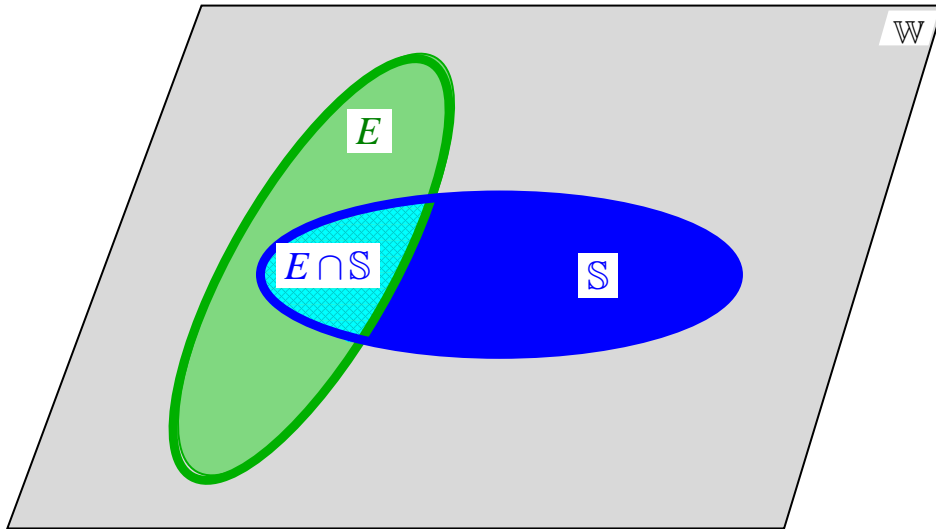
$$\Sigma|_{\mathbb{S}} := (\mathbb{S}, \mathcal{E} \cap \mathbb{S}, P(\cdot|\mathbb{S})), \text{ with } P(E \cap \mathbb{S}|\mathbb{S}) := \frac{P(E \cap \mathbb{S})}{P(\mathbb{S})}.$$

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

Conditional probability



Conditional probability



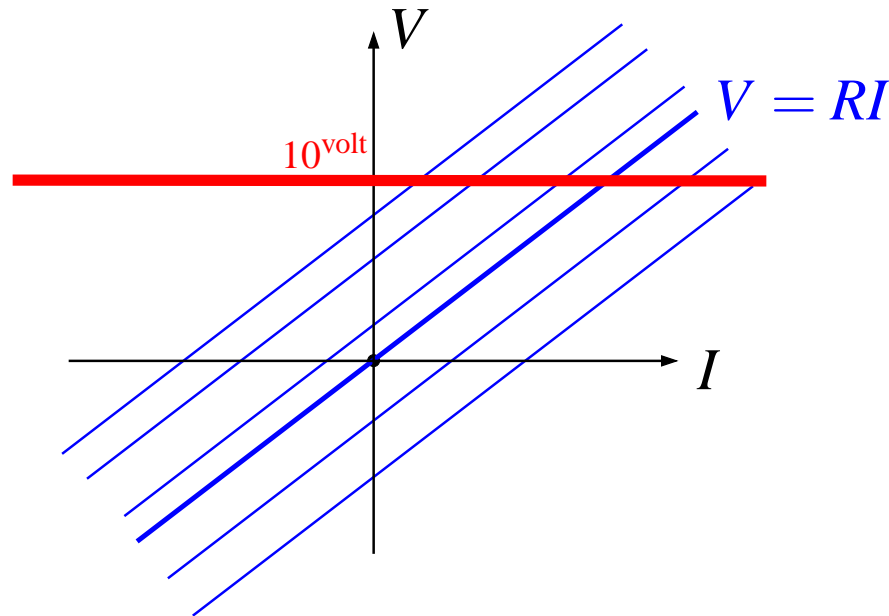
Constrained probability

Let $\Sigma = (\mathbb{W}, \mathcal{E}, P)$. Impose the constraint $\boxed{w \in \mathbb{S}}$
with $\mathbb{S} \subset \mathbb{W}$, $\boxed{\mathbb{S} \notin \mathcal{E}}$.

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

Is this a meaningful question?

Noisy resistor

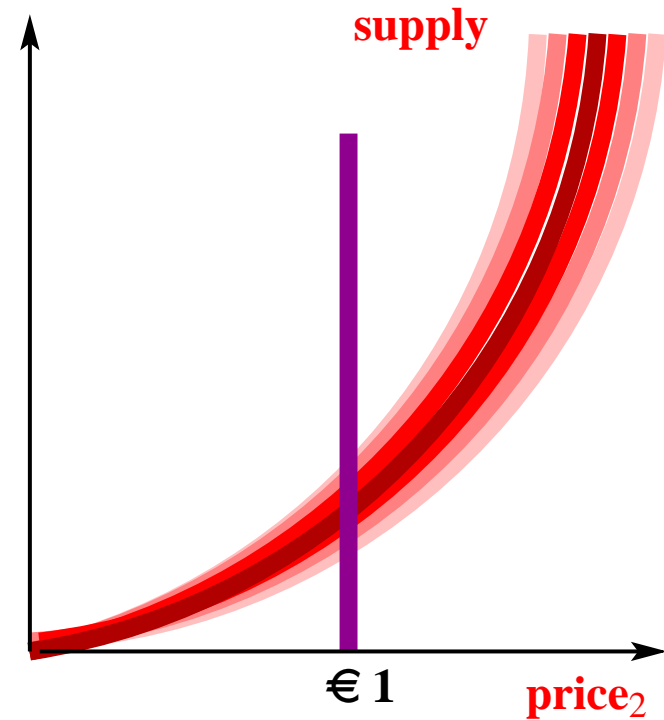
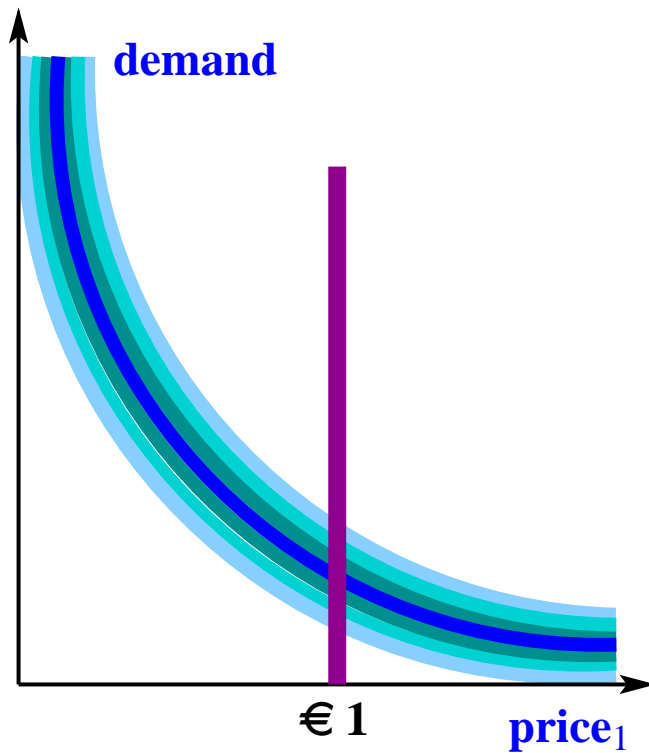


Impose $V = 10^{\text{voltage}}$. What is the distribution of I ?

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

I is a well-defined random variable!

Price/demand/supply example



Impose price = € 1. Distribution of demand, supply?

$$p_1 d = \varepsilon_1, p_1 = e 1 \Rightarrow d = \varepsilon_1; s = \varepsilon_2 p_2^2, p_2 = e 1 \Rightarrow 2 = \varepsilon_2.$$

d, s become well-defined random variables.

Constrained probability

Let $\Sigma = (\mathbb{W}, \mathcal{E}, P)$. Impose the constraint $\boxed{“w \in S”}$
with $S \subset \mathbb{W}$, $\boxed{S \notin \mathcal{E}}$.

What is the stochastic nature of the outcomes in S ?

Is this a meaningful question? **Yes!**

Constrained probability

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

Assume complementarity:

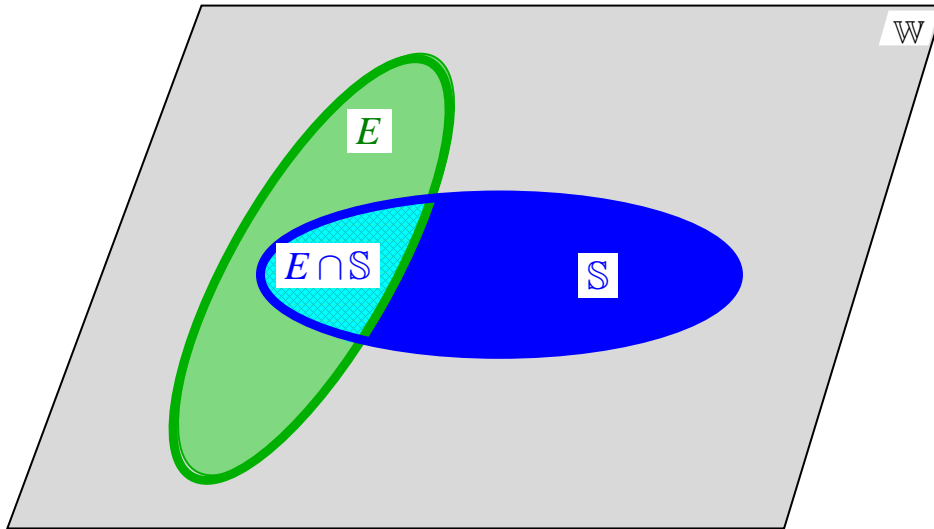
$$\llbracket E_1, E_2 \in \mathcal{E} \text{ and } E_1 \cap \mathbb{S} = E_2 \cap \mathbb{S} \rrbracket \Rightarrow \llbracket P(E_1) = P(E_2) \rrbracket$$

Interconnection \rightsquigarrow

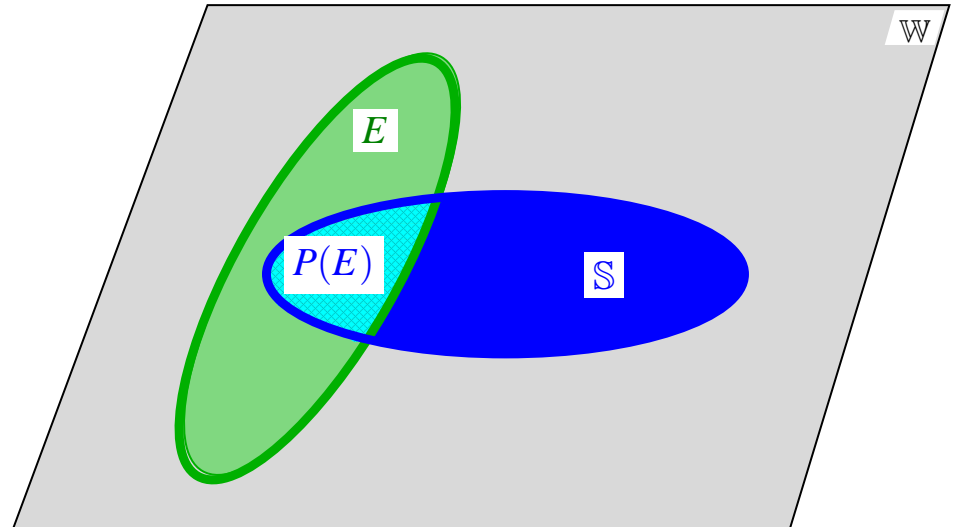
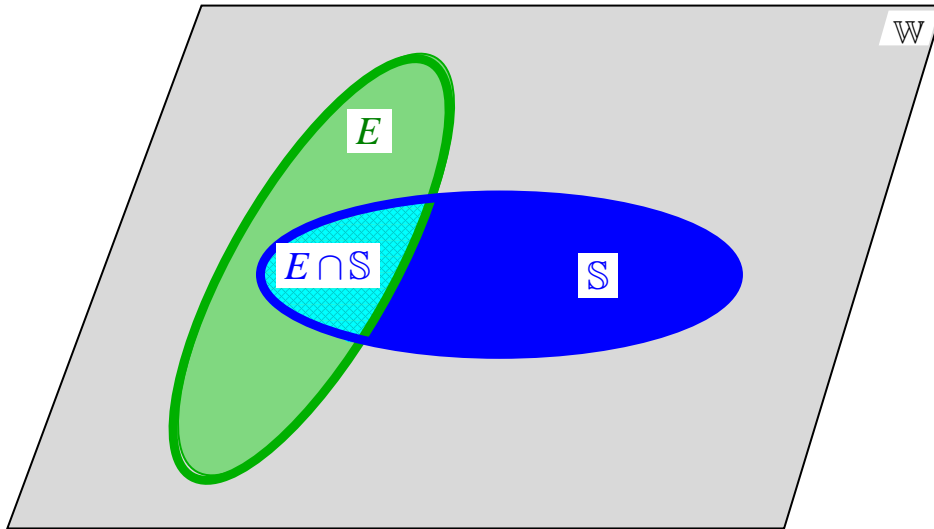
$$\Sigma|_{\mathbb{S}} = (\mathbb{S}, \mathcal{E} \cap \mathbb{S}, P(\cdot|\mathbb{S})) \quad \text{with} \quad P(E \cap \mathbb{S}|\mathbb{S}) := P(E).$$

$P(\cdot|\mathbb{S}) =$ “probability of w constrained by $w \in \mathbb{S}$ ”.

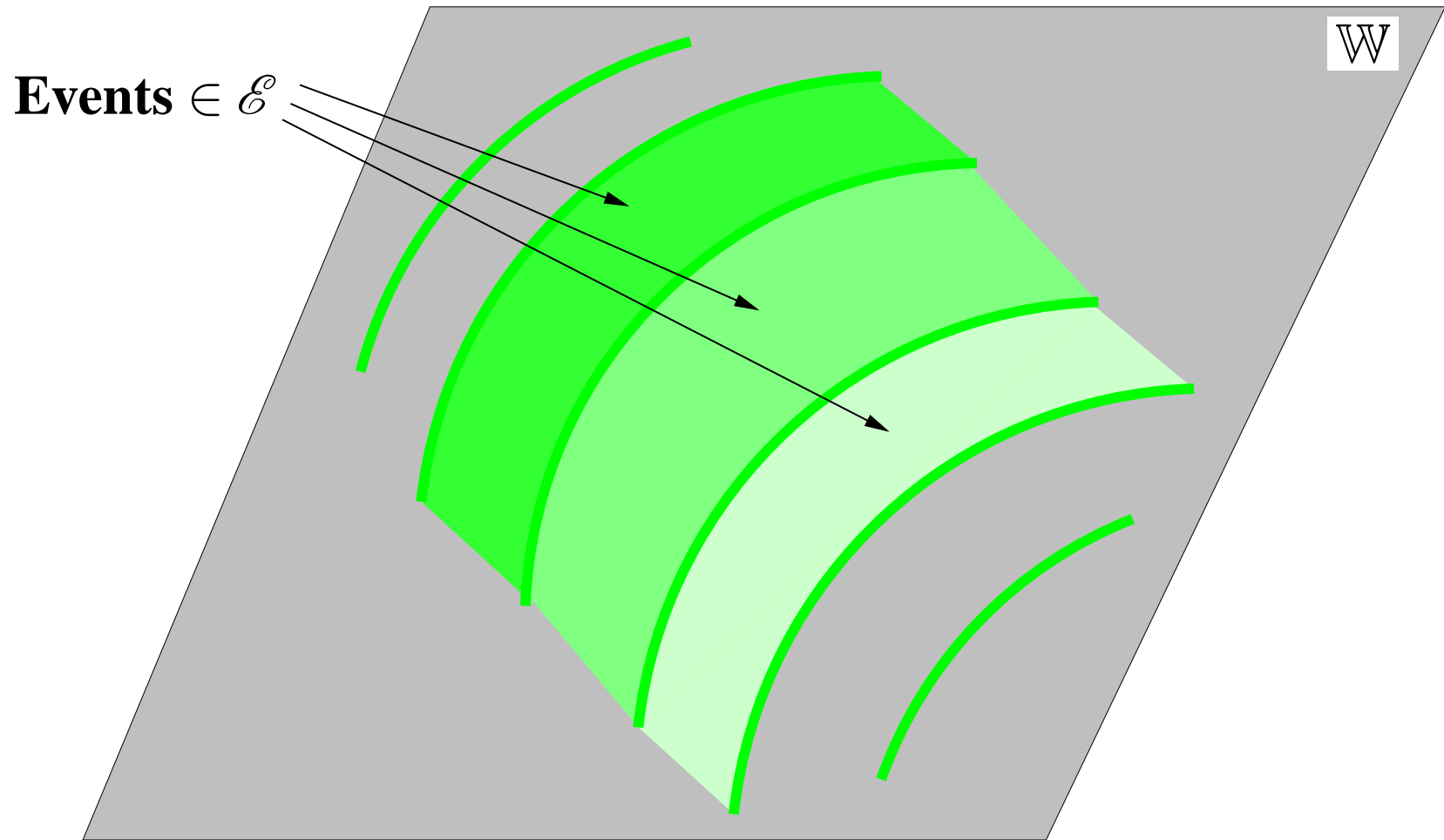
Constrained probability



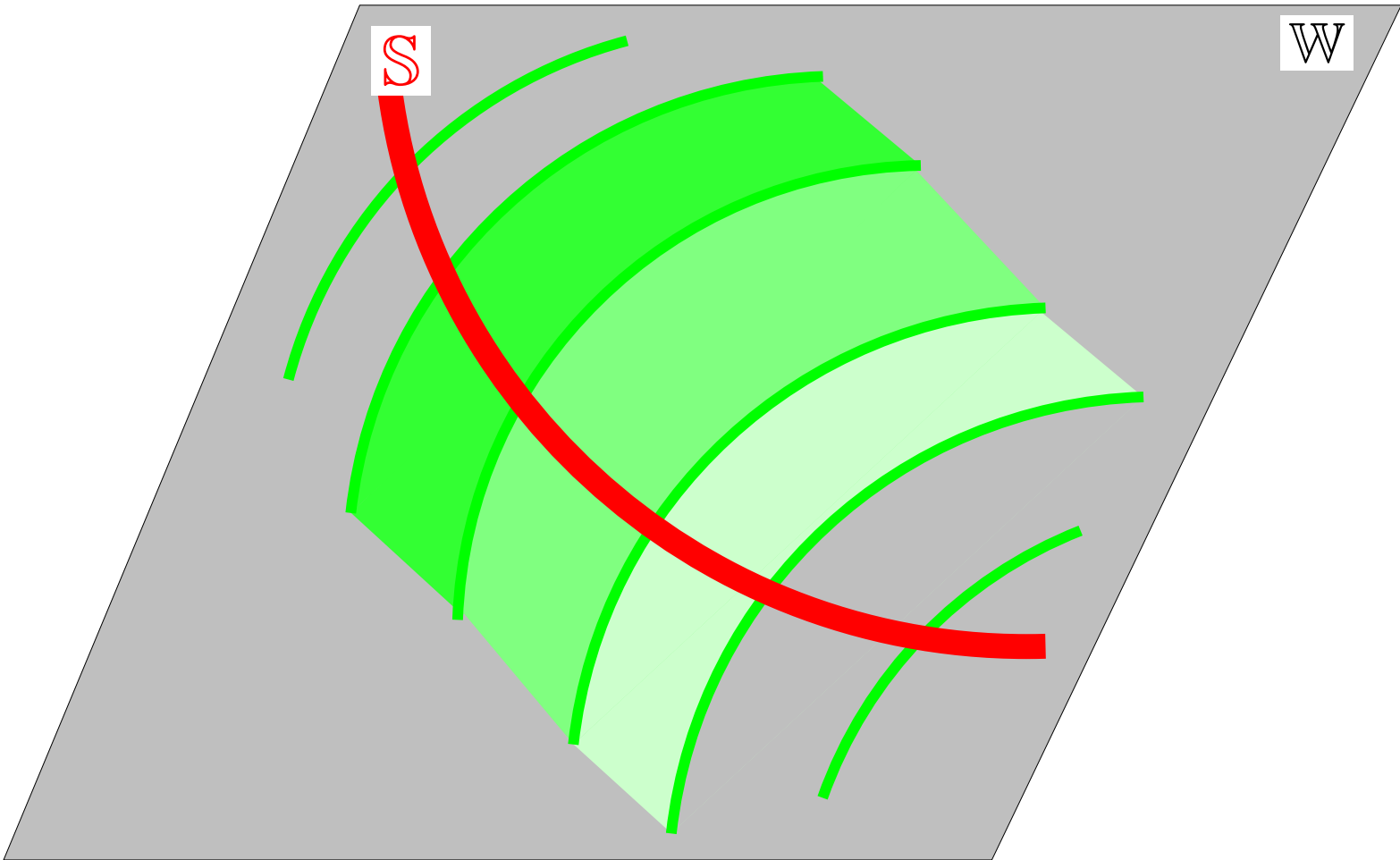
Constrained probability



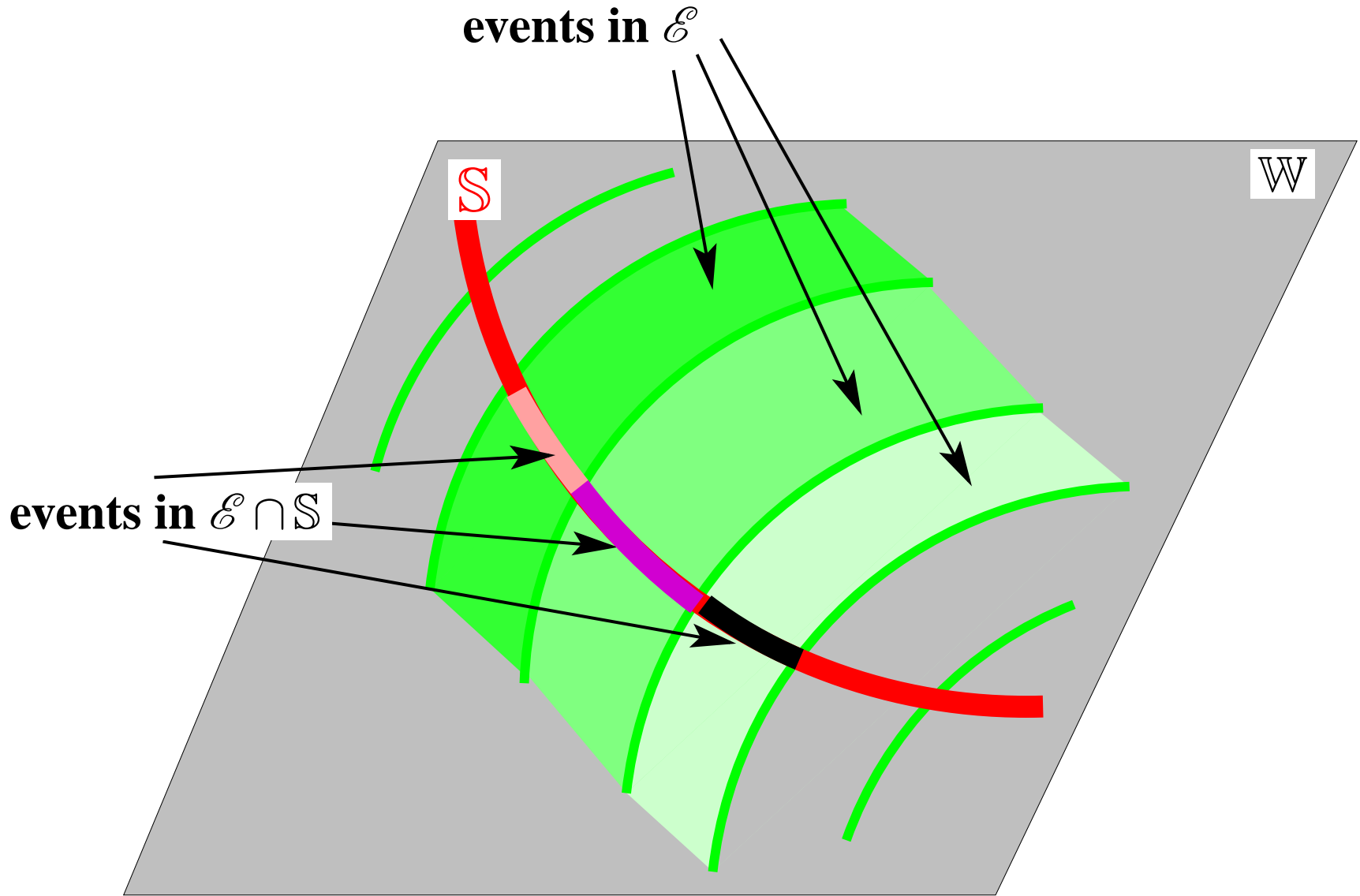
In pictures



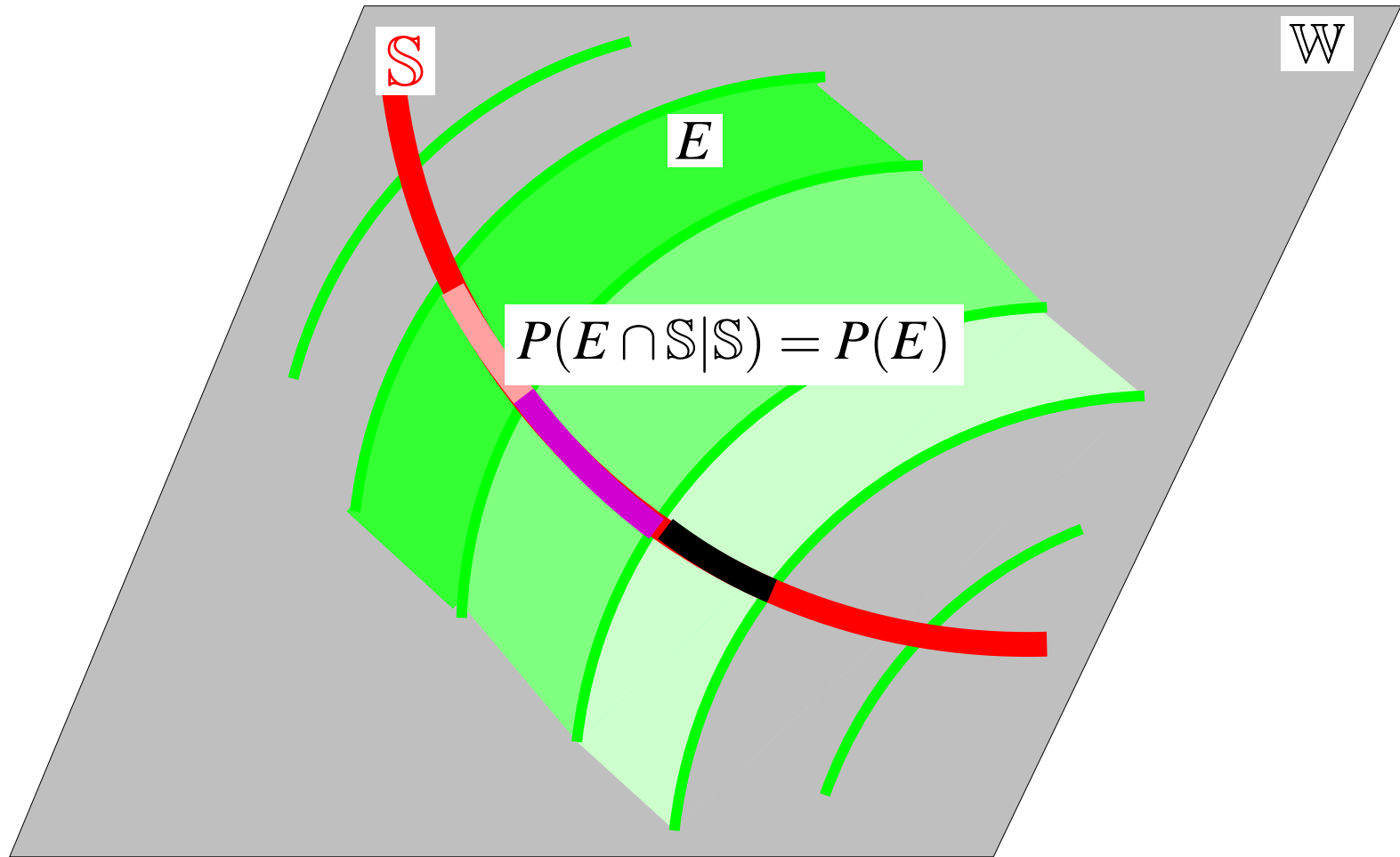
In pictures



In pictures

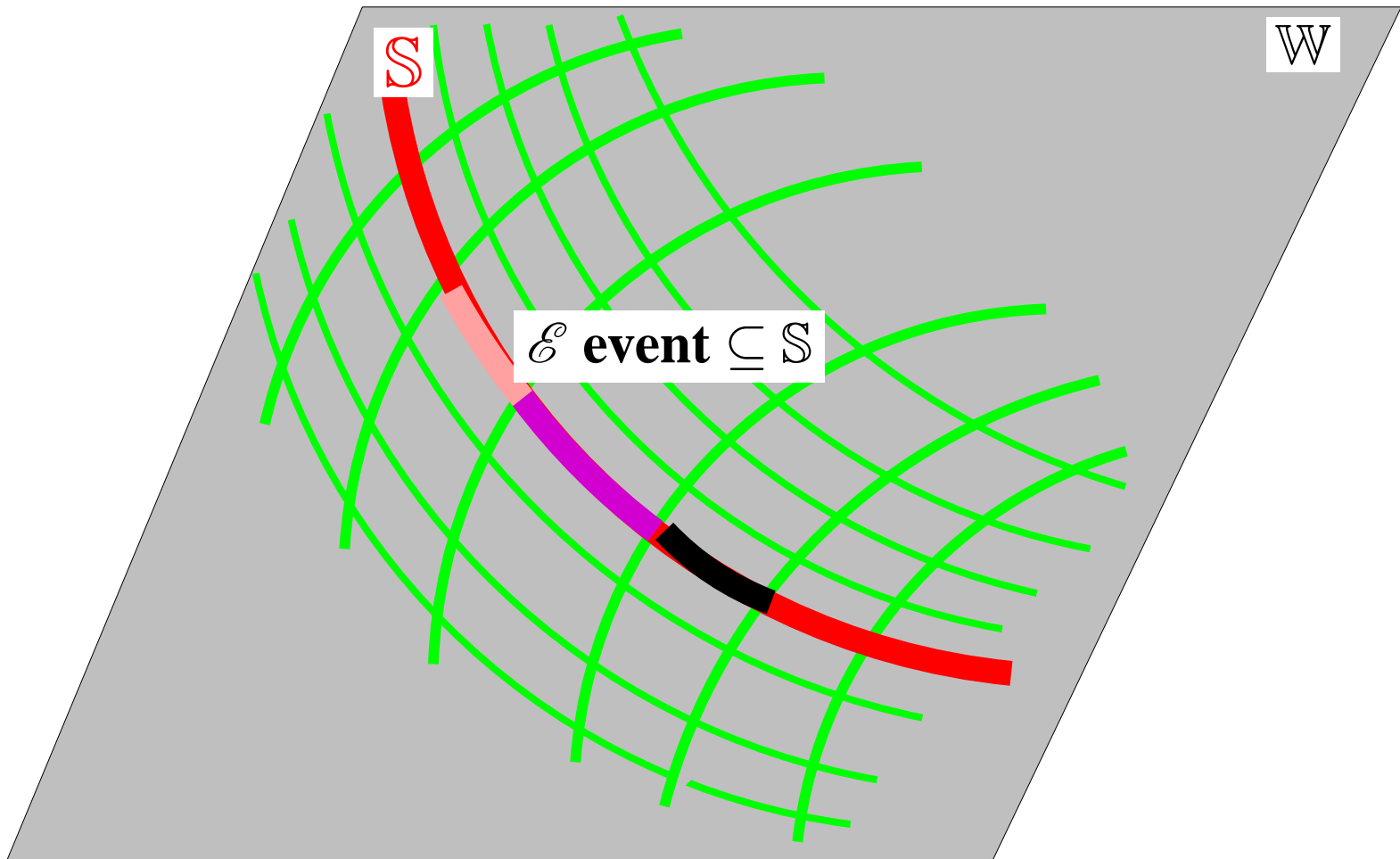


In pictures

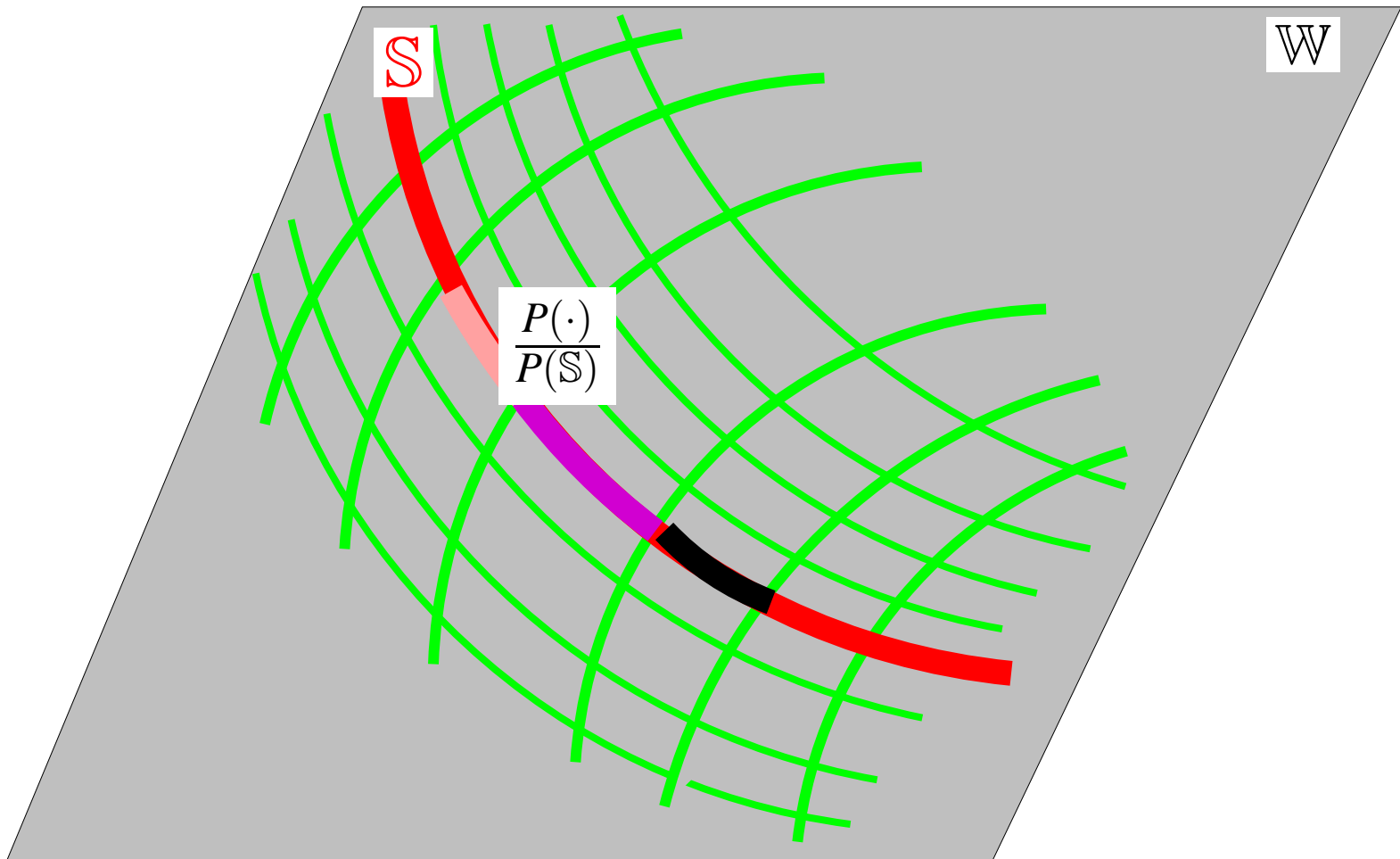


Probability of E drawn globally into S .

Contrast with conditional probability



Contrast with conditional probability



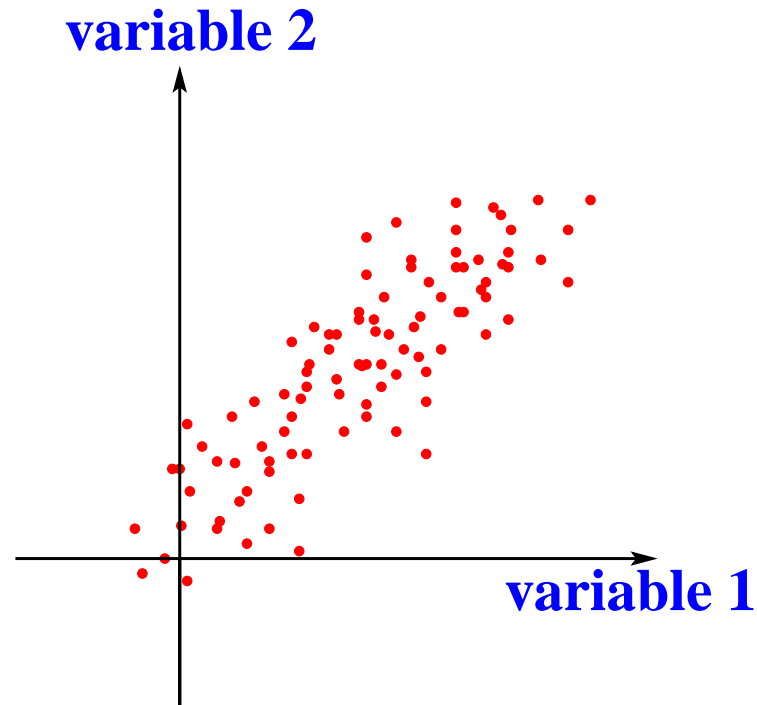
Probability locally computed, with renormalization.

Constraining and conditioning

Is there a point of view from which yields both concepts are special cases of one unifying idea?

Identification

Sampling

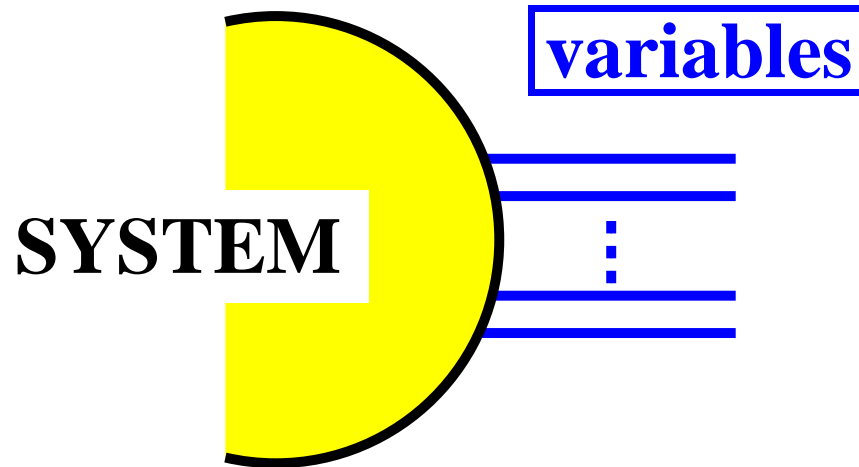


System identification: deduce the stochastic model

\mathcal{E} and P

from the samples.

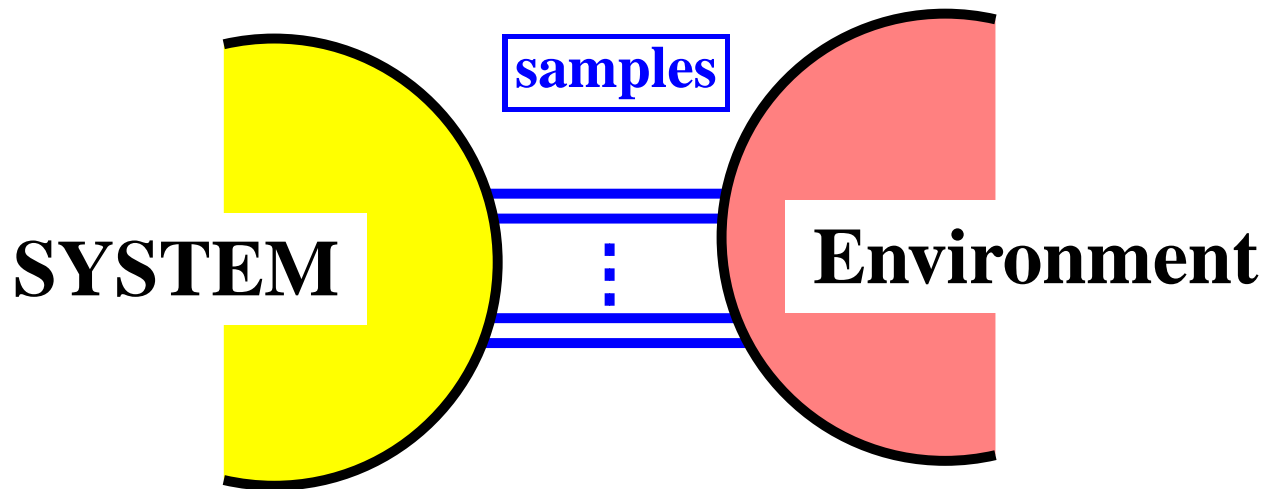
Measurements



**Data collection implies observing a stochastic system
*in interaction with an environment.***

Measurements

**Data collection implies observing a stochastic system
*in interaction with an environment.***



Measurements

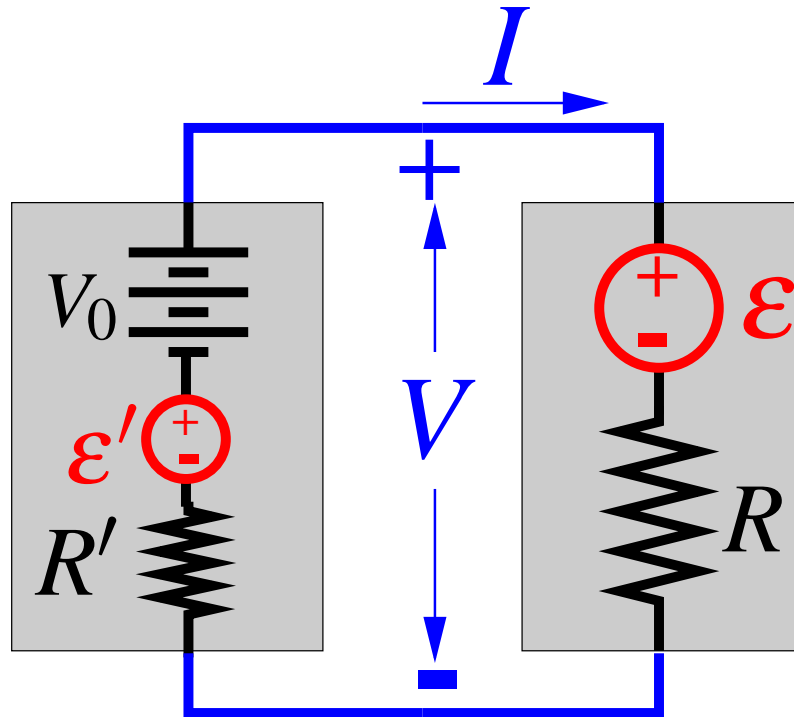
Data collection implies observing a stochastic system *in interaction with an environment.*

Is it possible to disentangle the laws of a system from the laws of the environment?

In engineering, it may be possible to set the experimental conditions.

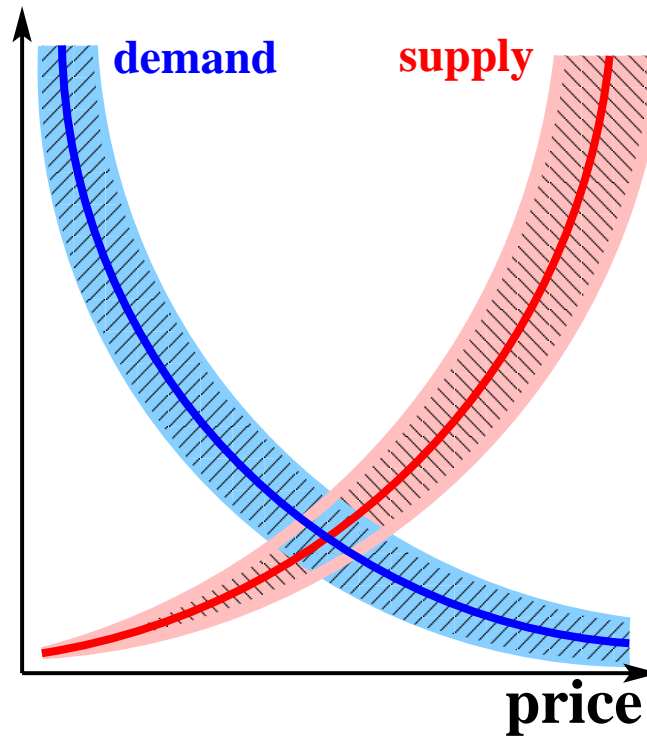
In economics and the social sciences (and biology?), data often gathered passively *‘in vivo’*.

Disentangling



Can R and σ be deduced by sampling (V, I) ?

Disentangling



**Can the price/demand characteristic be deduced
by sampling (p, d) in equilibrium?**

SYSID for gaussian systems

Let Σ_1 and Σ_2 be complementary gaussian systems and assume that the interconnection $\Sigma_1 \wedge \Sigma_2$ is a classical random system.

Sampling \rightsquigarrow the mean and covariance of $\Sigma_1 \wedge \Sigma_2$.

Assume that the covariance is nonsingular.

SYSID for gaussian systems

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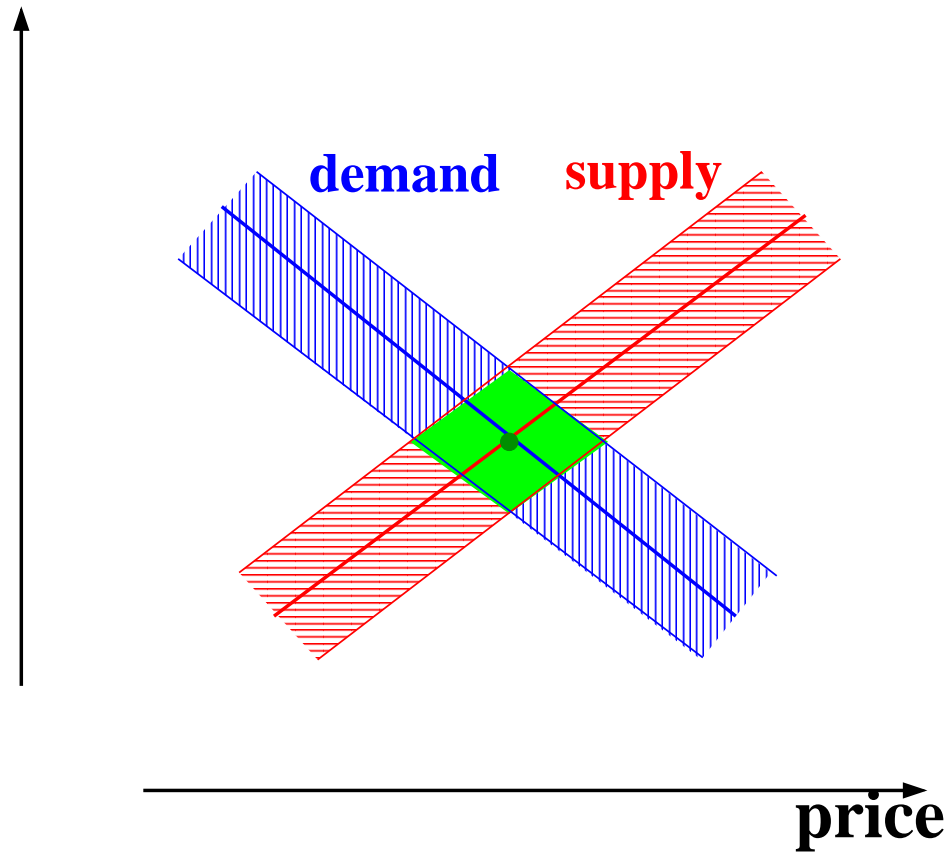
Sampling \rightsquigarrow the mean and covariance of $\Sigma_1 \wedge \Sigma_2$.

Assume that the covariance is nonsingular.

Given the fiber of **either Σ_1 **or** Σ_2 , then all the other parameters of Σ_1 and Σ_2 can be deduced from $\Sigma_1 \wedge \Sigma_2$.**

The fiber of Σ_1 can be chosen freely.

Linearized gaussian price/demand/supply



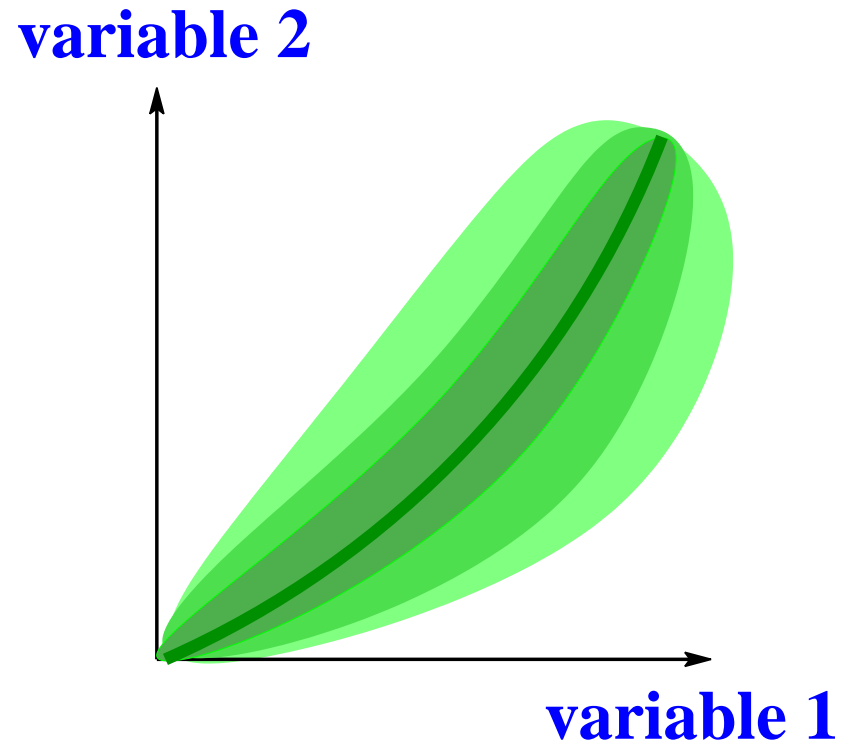
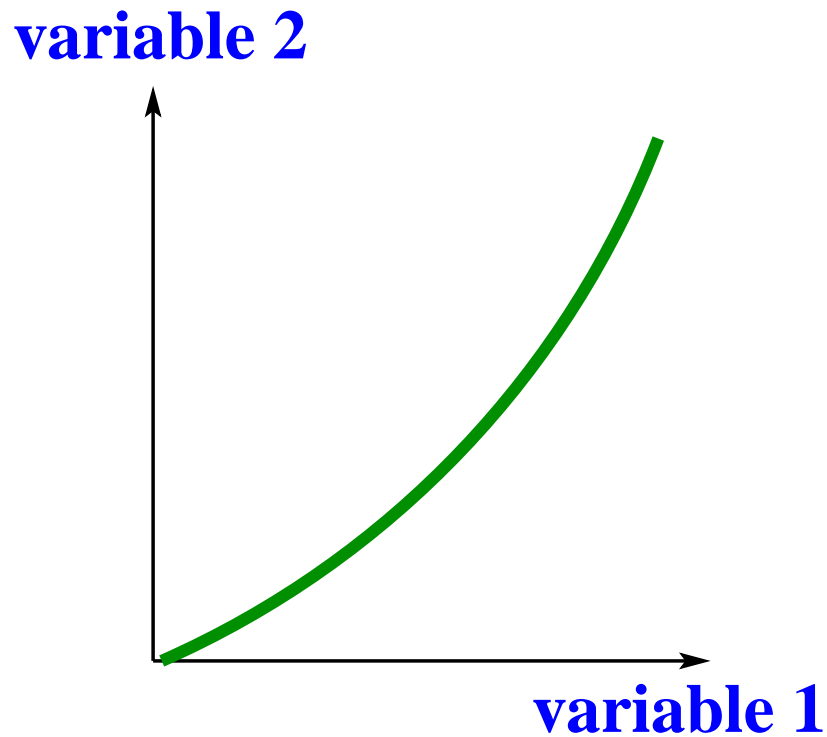
Identifiability provided one of the fibers is known.

Sampling alone does not give these elasticities.

Conclusions

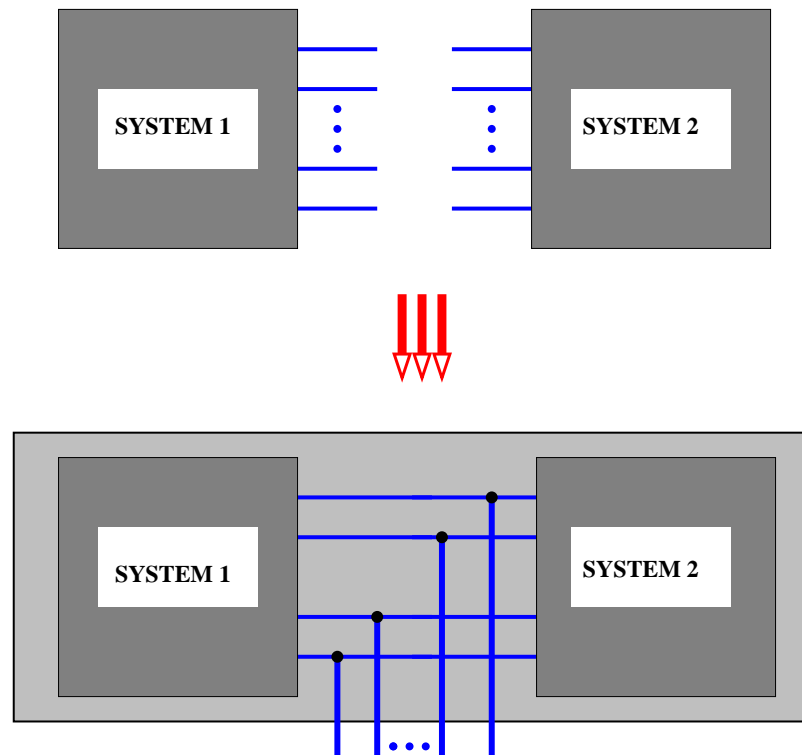
Stochastic systems

- ▶ **The Borel σ -algebra is inadequate even for elementary applications.**



Stochastic systems

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



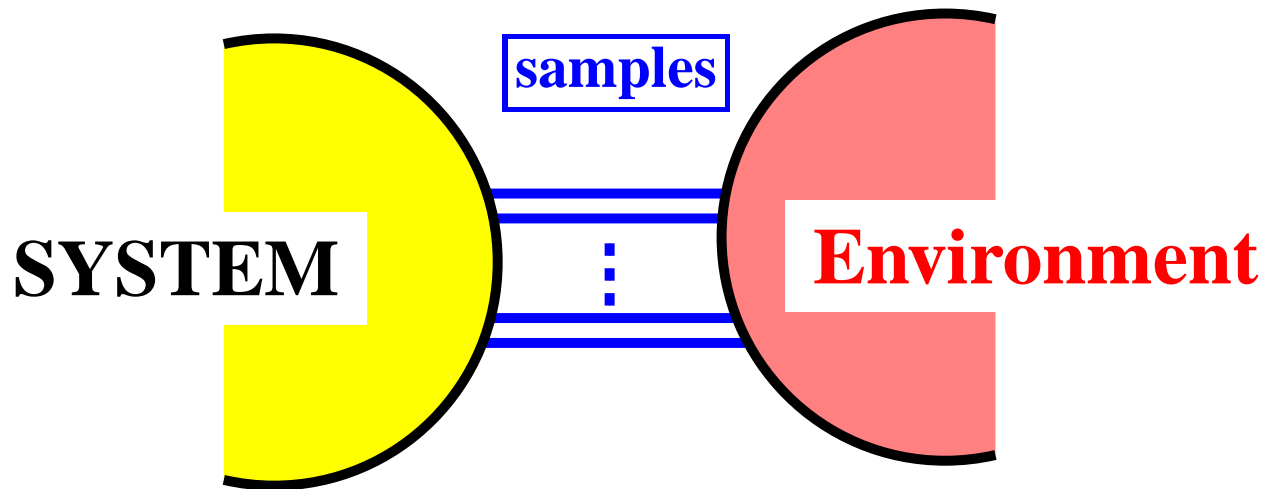
Stochastic systems

- ▶ **Open stochastic systems require a coarse σ -algebra.**

Classical random vectors imply closed systems.

SYSID

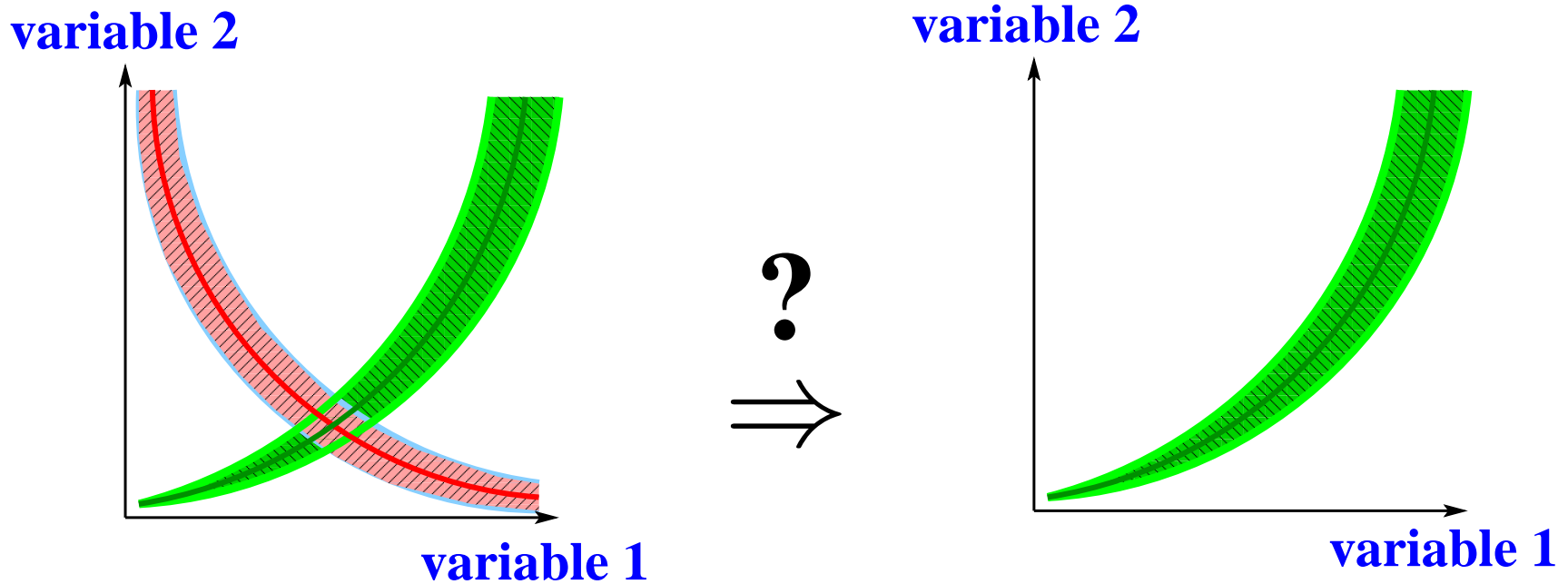
- ▶ **Measurements are the result of interaction with an environment.**



Modeling from data requires disentanglement.

SYSID

- ▶ **Modeling from data requires disentanglement.**
The data alone are insufficient for identifiability.



Future work

Urgent:

Generalization to stochastic processes.

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

Copies of the lecture frames available from/at

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

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Thank you

Thank you

Thank you

Thank you

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