

Probabilità e incertezze di misura

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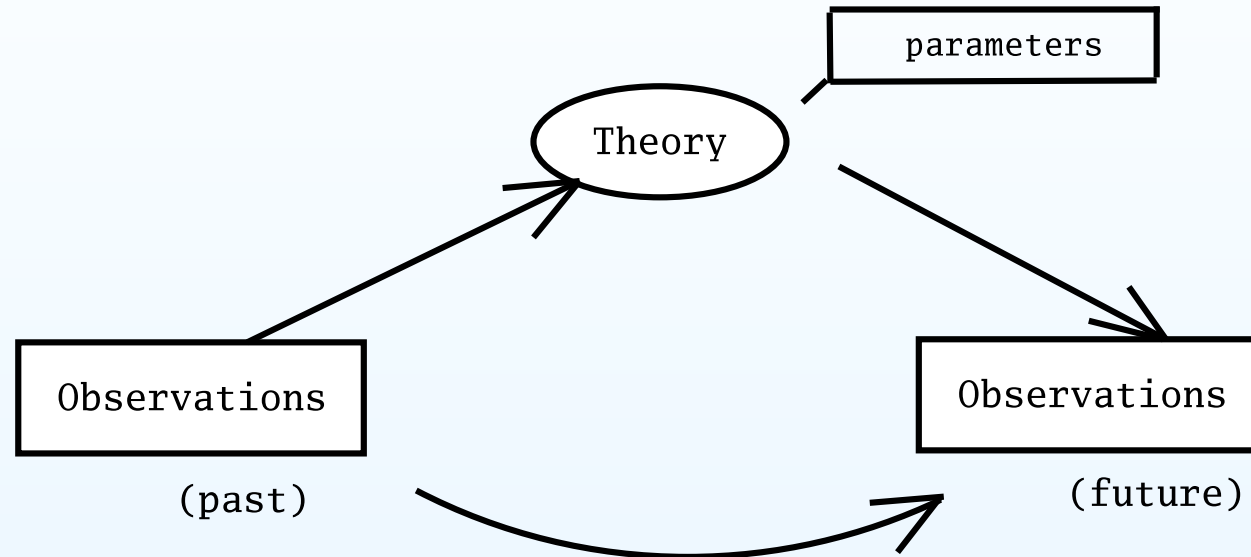
Piano dei due incontri

1. Rassegna critica e introduzione all'inferenza probabilistica
 - Quanto sono sensate e ben fondate le regolette per la valutazione dei cosiddetti "errori di misura"?
 - Per imparare dall'esperienza in modo quantitativo, facendo uso della logica dell'incerto, dobbiamo
 - rivedere il concetto di probabilità;
 - imparare ad ... imparare dall'esperienza.
2. Stima delle incertezze in misure dirette e indirette
 - Sorgenti delle incertezze di misura (*decalogo ISO*).
 - Applicazione dell'inferenza probabilistica alle misure sperimentali (semplice caso di errori gaussiani):
 - singola osservazione
 - campione di osservazioni
 - stima dei parametri di un andamento lineare
 - Propagazione delle incertezze

Scaletta del primo incontro

- Metodo scientifico: osservazioni e ipotesi
- Incertezza
- Cause \longleftrightarrow Effetti
“Il problema essenziale del metodo sperimentale” (Poincaré).
- L'esempio guida: il problema delle sei scatole.
“La probabilità è riferita a casi reali o non ha alcun senso” (de Finetti).
- *Fisichettume*: una rassegna critica.
- Falsificazionismo e *variazioni statistiche* ('test').
- Approccio probabilistico.
- *Cos'è la probabilità?* Regole di base della probabilità.
- Aggiornamento della probabilità alla luce delle osservazioni (formula di Bayes) \Rightarrow inferenza probabilistica (bayesiana)
- Conclusioni.

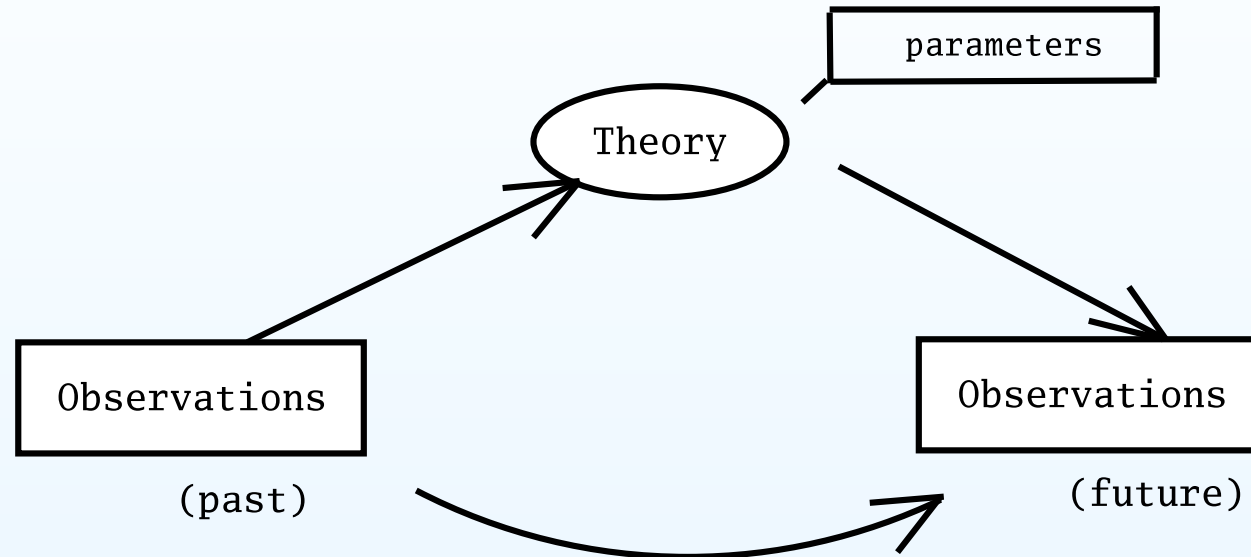
From past to future



Task of the 'physicist' (scientist, decision maker):

- Describe/understand the physical world
⇒ **inference** of laws and their parameters
- Predict observations
⇒ **forecasting**

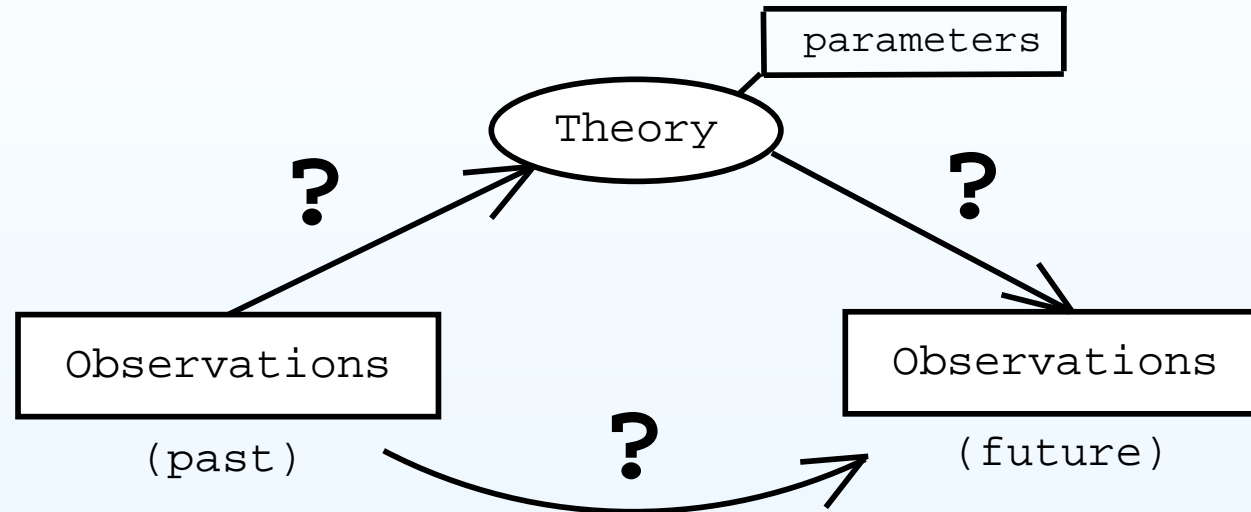
From past to future



Process

- neither automatic
- nor purely contemplative
 - ‘scientific method’
 - planned experiments (‘actions’) ⇒ **decision.**

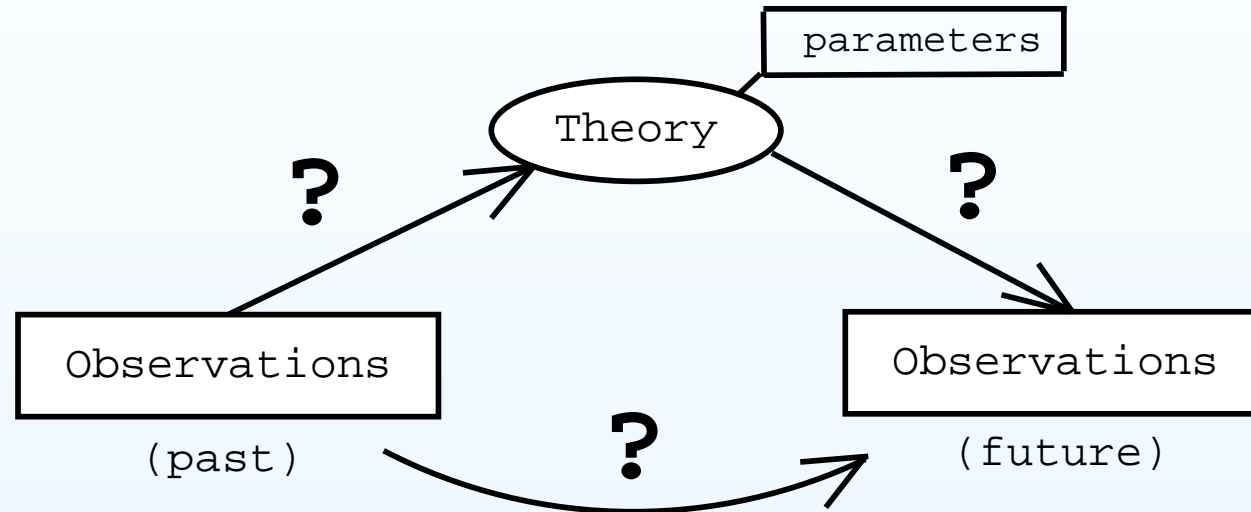
From past to future



⇒ **Uncertainty:**

1. Given the past observations, in general we are not sure about the theory parameter (and/or the theory itself)
2. Even if we were sure about theory and parameters, there could be internal (e.g. Q.M.) or external effects (initial/boundary conditions, 'errors', etc) that make the forecasting uncertain.

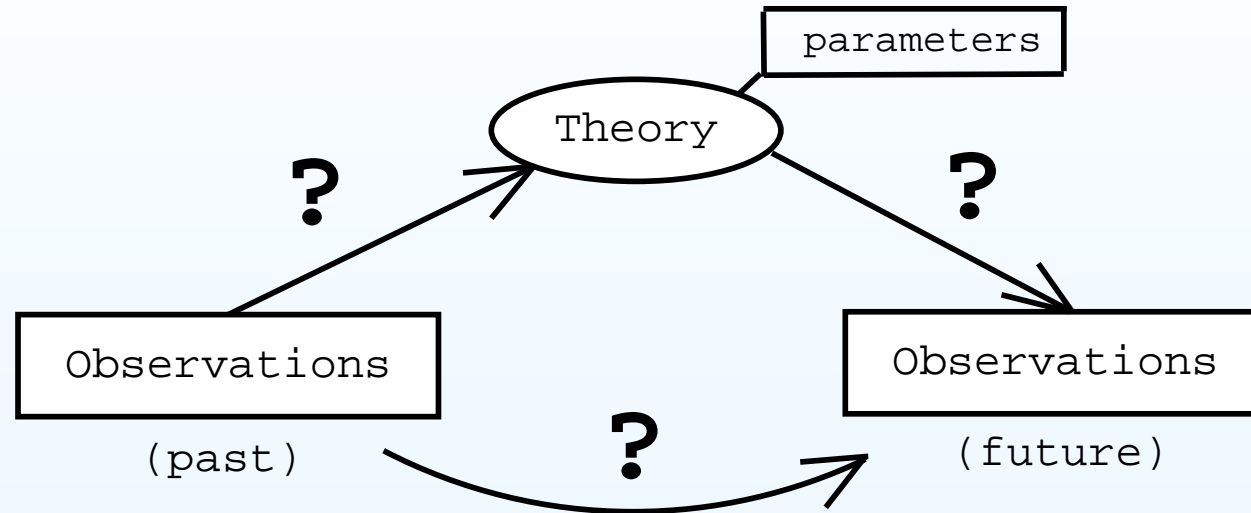
From past to future



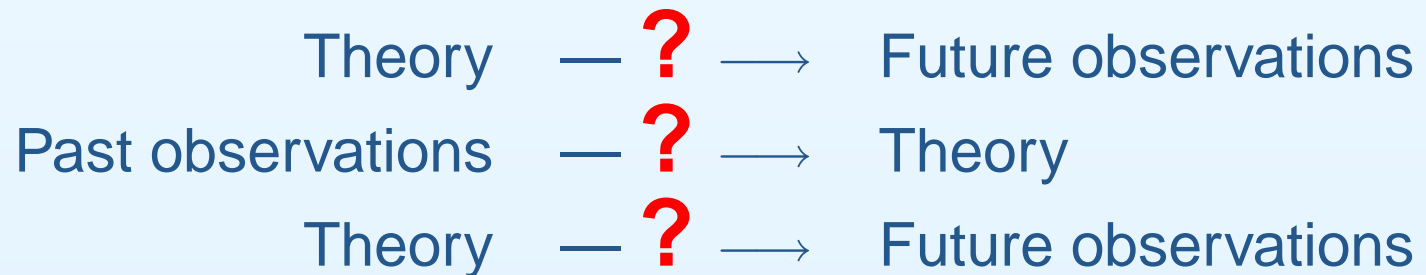
⇒ Decision

- What is the best action ('experiment') to take in order 'to be confident' that what we would like will occur? (Decision issues always assume uncertainty about future outcomes.)
- Before tackling problems of decision we need to learn to reason about uncertainty, possibly in a quantitative way.

From past to future



Deep reason of uncertainty



A simple example

- Three boxes each contains two balls: White-White, White-Black, Black-Black. We take randomly one of the box and extract one ball, e.g. White. We can extract the second ball from any of the three boxes.

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- **Decision problem:** From which box should we extract the second ball in order to have a second White?
- **Uncertainty:**
 - Which box have we taken?
 - What is the chance to get White from the same box, or from one of the remaining two, selected at random?

About predictions

Remember:

*“Prediction is very difficult,
especially if it’s about the future” (Bohr)*

About predictions

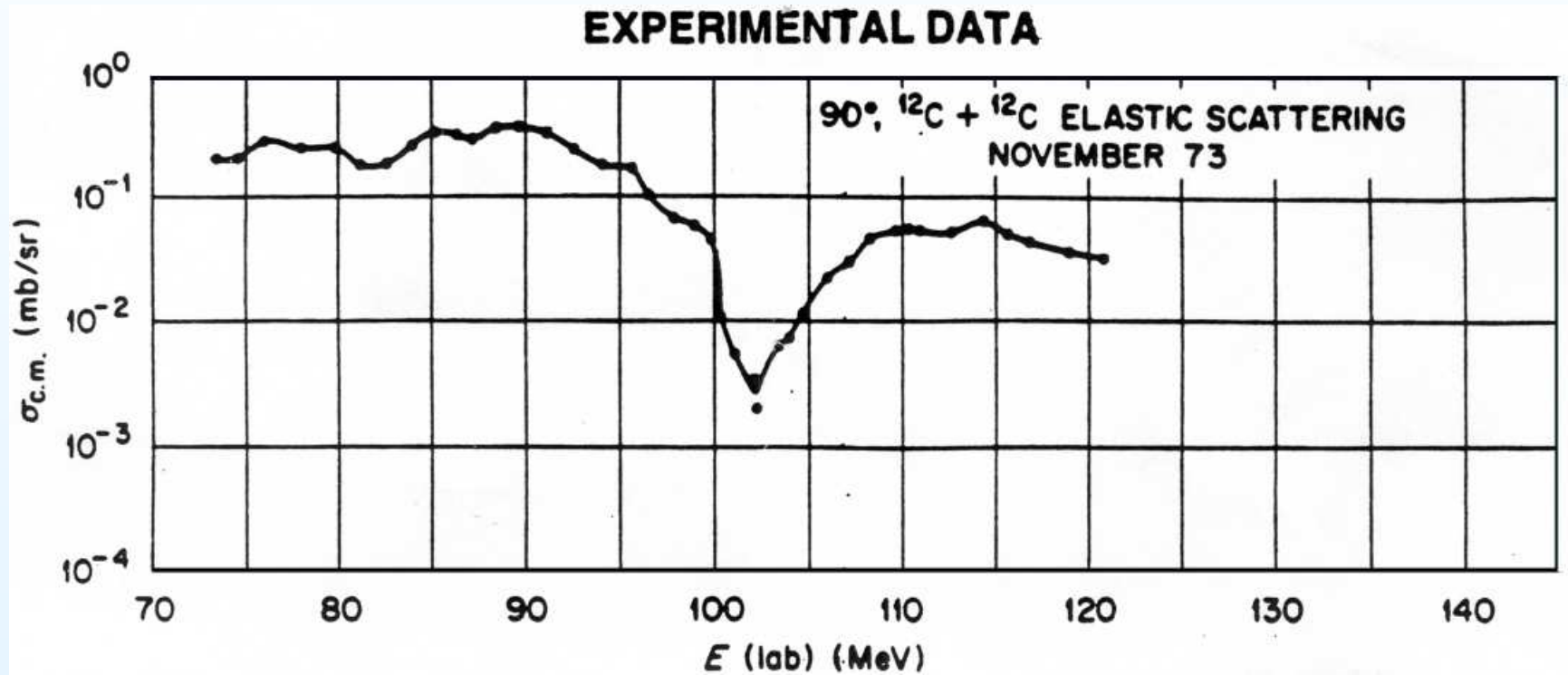
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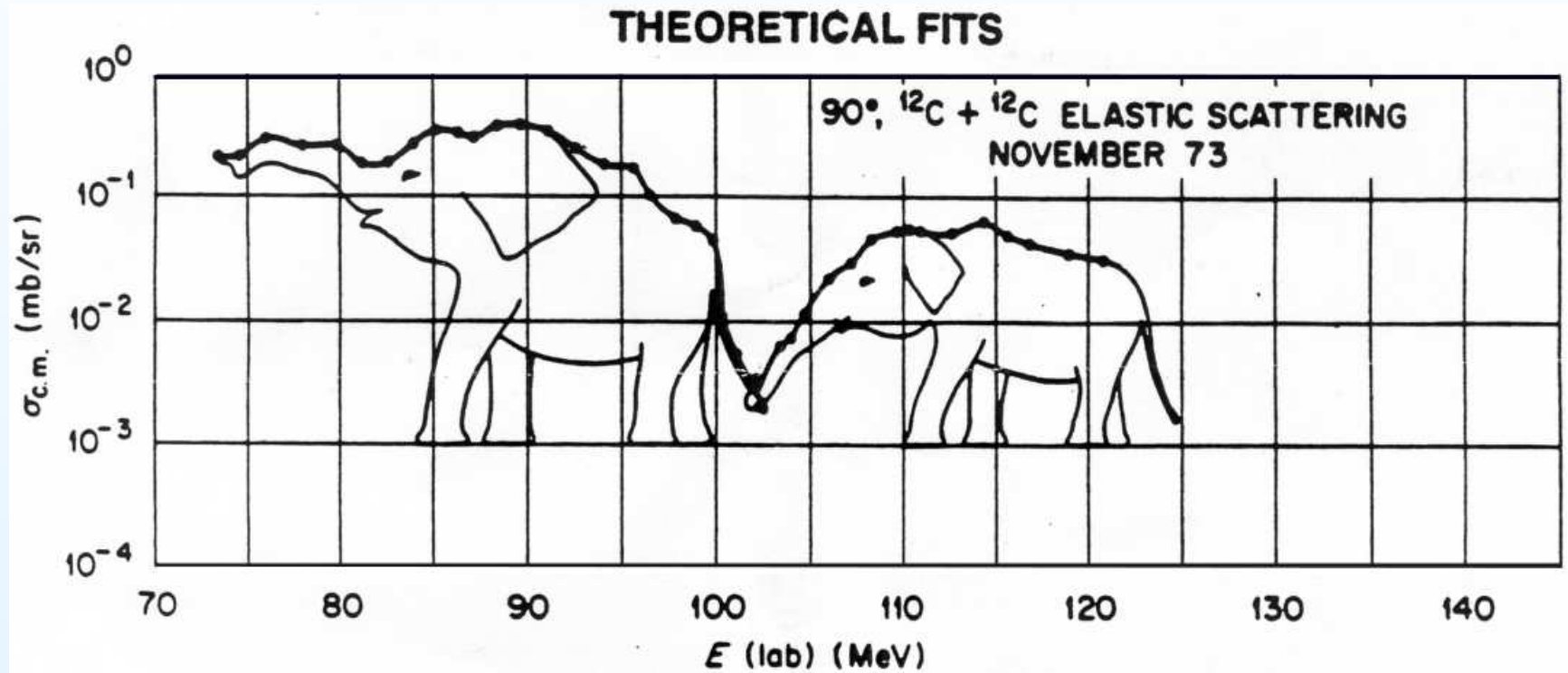
But, anyway:

*“It is far better to foresee even without
certainty than not to foresee at all”*
(Poincaré)

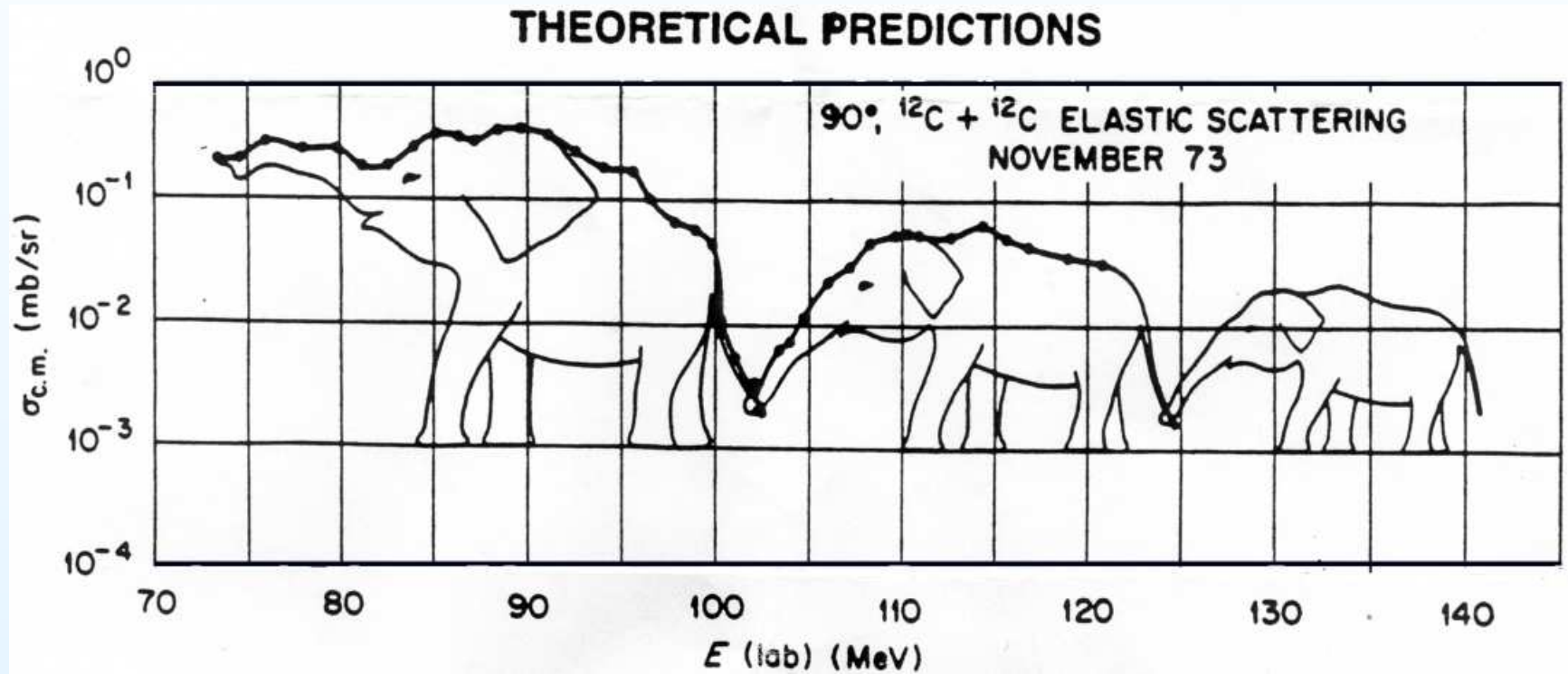
Inferential process



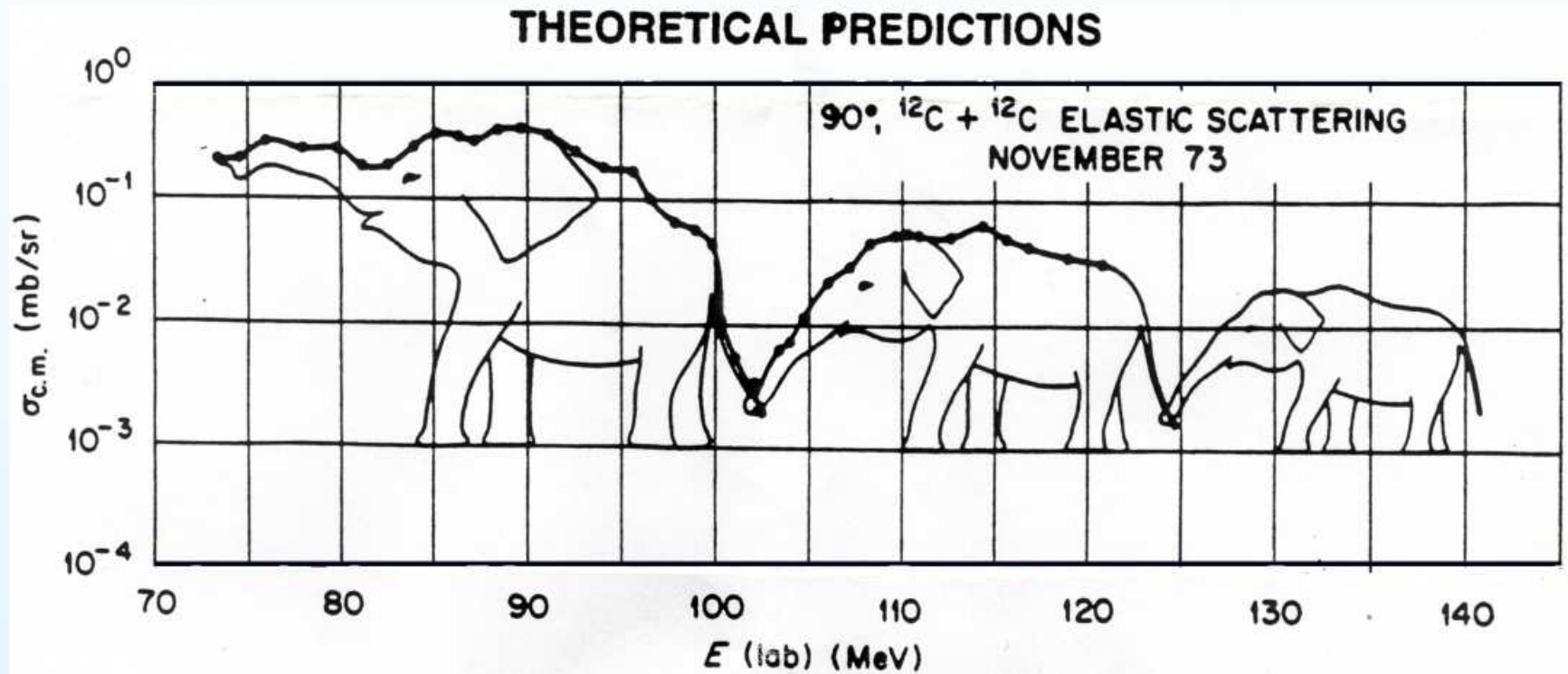
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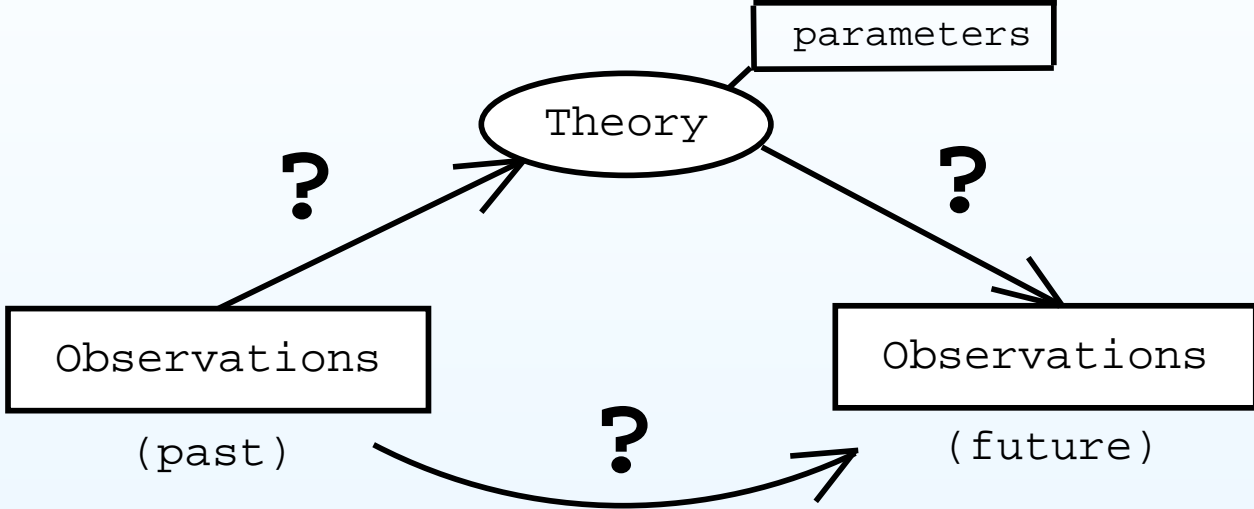


Inferential process



(S. Raman, *Science with a smile*)

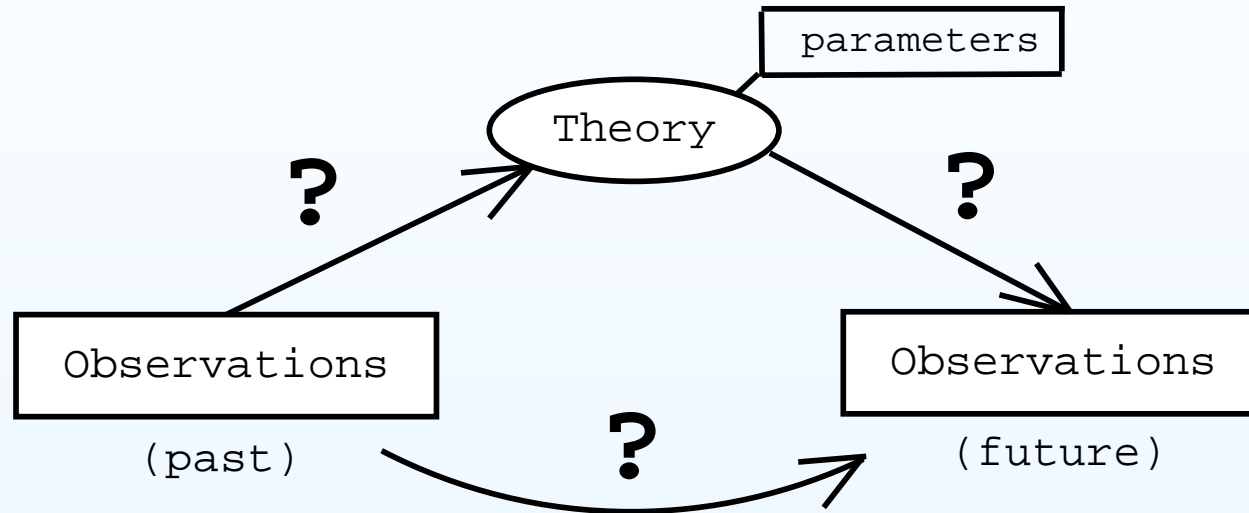
Deep source of uncertainty



Uncertainty:

Theory — ? —>
Past observations — ? —>
Theory — ? —> Future observations

Deep source of uncertainty



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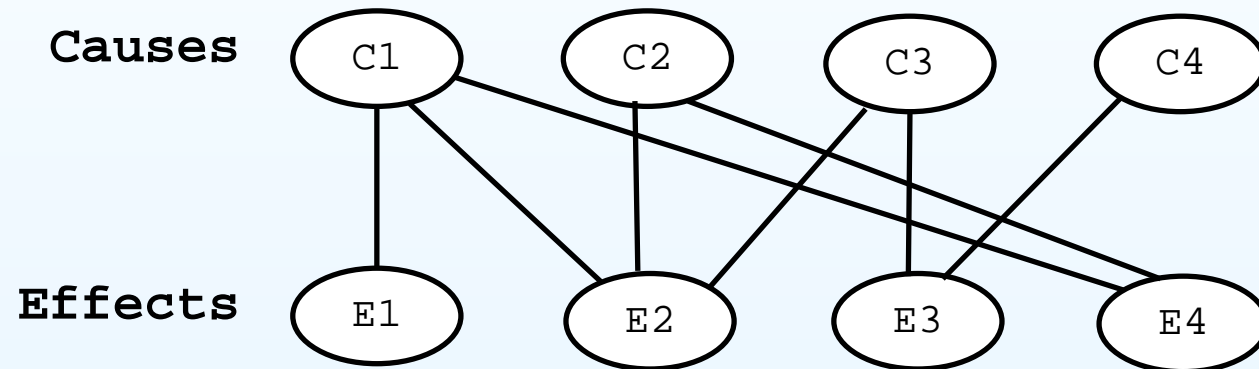
Theory — ? —> Future observations
Past observations — ? —> Theory
Theory — ? —> Future observations

⇒ **Uncertainty about causal connections**

CAUSE ⇔ **EFFECT**

Causes → effects

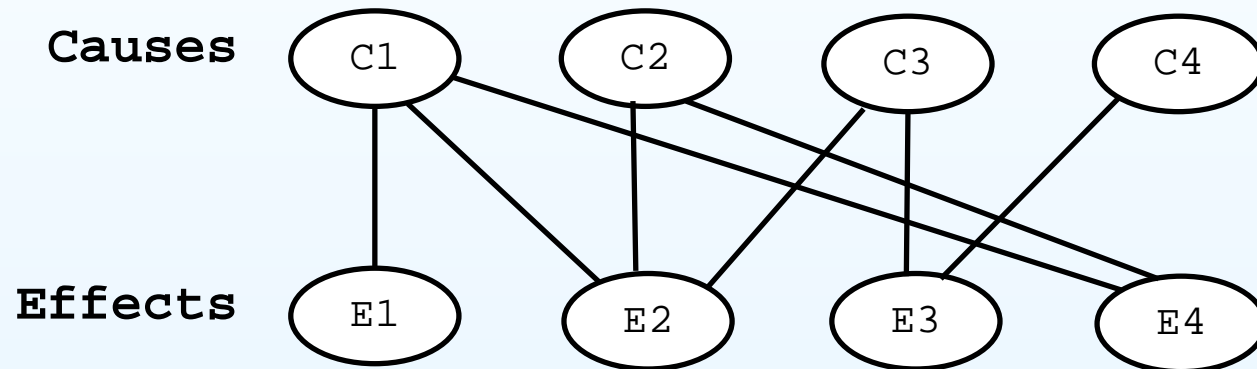
The same *apparent* cause might produce several, different effects



Given an observed effect, we are not sure about the exact cause that has produced it.

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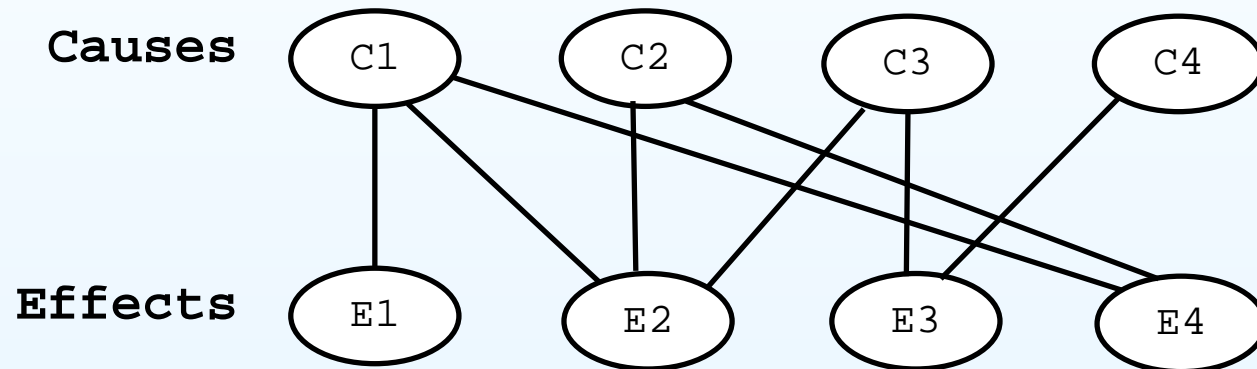
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Causes → effects

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$$E_2 \Rightarrow \{C_1, C_2, C_3\}?$$

The essential problem of the experimental method

“Now, these problems are classified as *probability of causes*, and are most interesting of all their scientific applications. I play at *écarté* with a gentleman whom I know to be perfectly honest. What is the chance that he turns up the king? It is $1/8$. This is a problem of the *probability of effects*.

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I play with a gentleman whom I do not know. He has dealt ten times, and he has turned the king up six times. What is the chance that he is a sharper? This is a problem in the probability of causes. It may be said that **it is the essential problem of the experimental method.**”

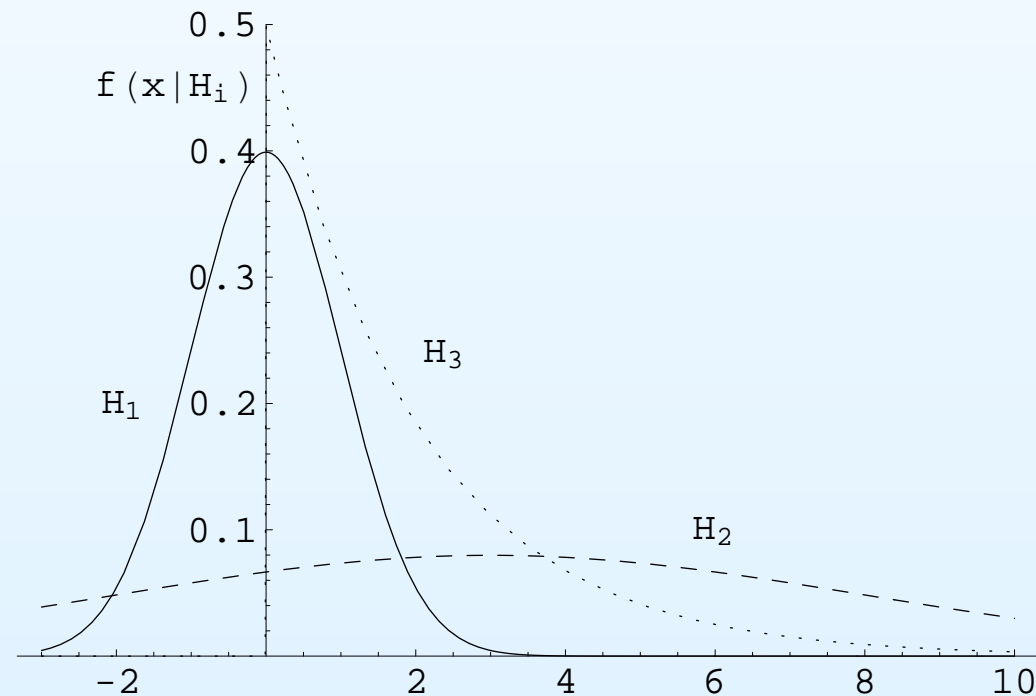
(H. Poincaré – *Science and Hypothesis*)

A numerical example

- Effect: number $x = 3$ extracted 'at random'
- Hypotheses: one of the following random generators:
 - H_1 Gaussian, with $\mu = 0$ and $\sigma = 1$
 - H_2 Gaussian, with $\mu = 3$ and $\sigma = 5$
 - H_3 Exponential, with $\tau = 2$

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- ⇒ Which one to prefer?

Note: ⇒ none of the hypotheses of this example can be excluded and, therefore, there is no way to reach a boolean conclusion. We can only state, somehow, our *rational preference*, based on the experimental result and our best knowledge of the behavior of each *model*.

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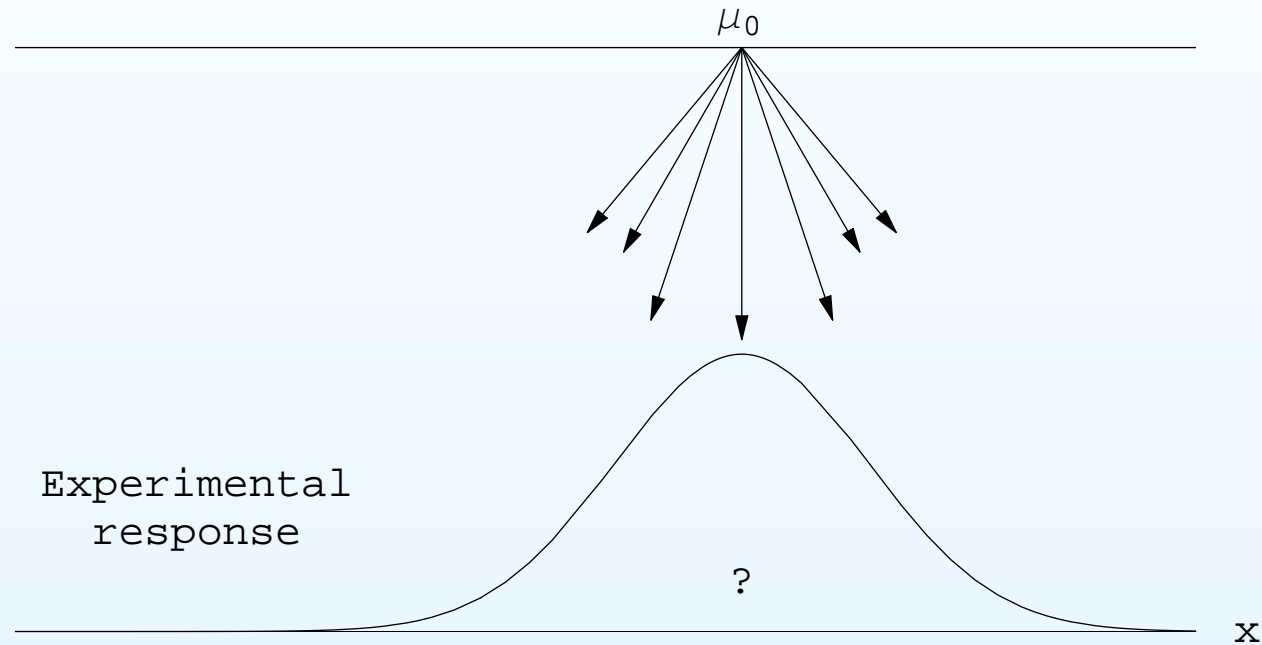
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We shall come back to this example

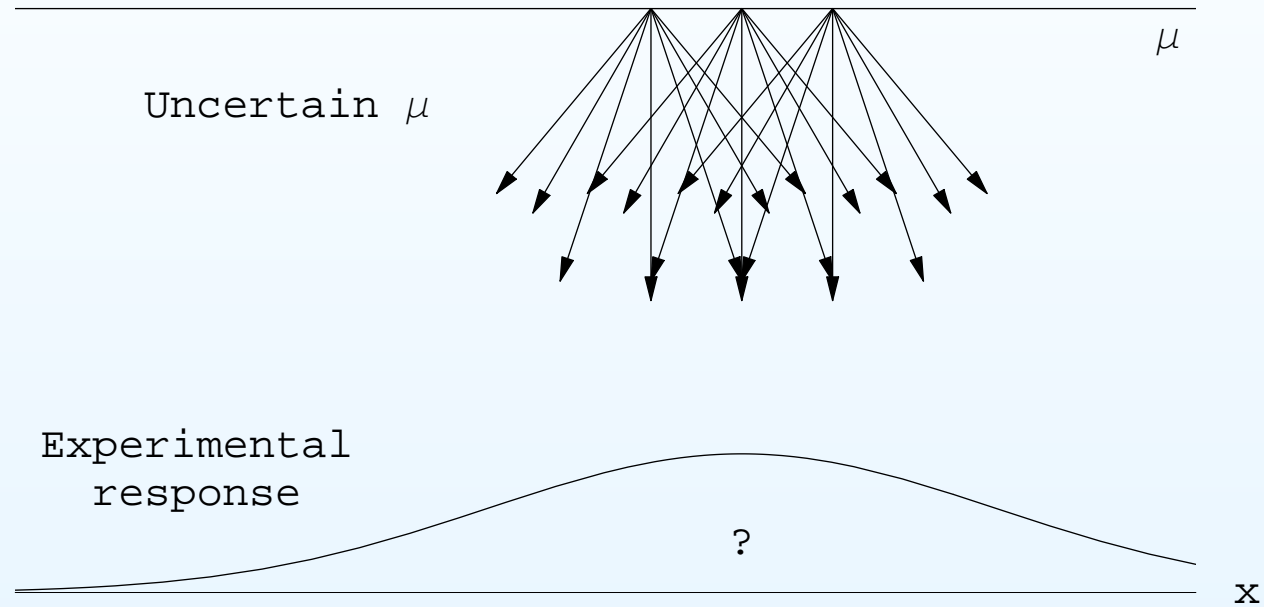
→ Let's now move to 'measuring true values'

From 'true value' to observations



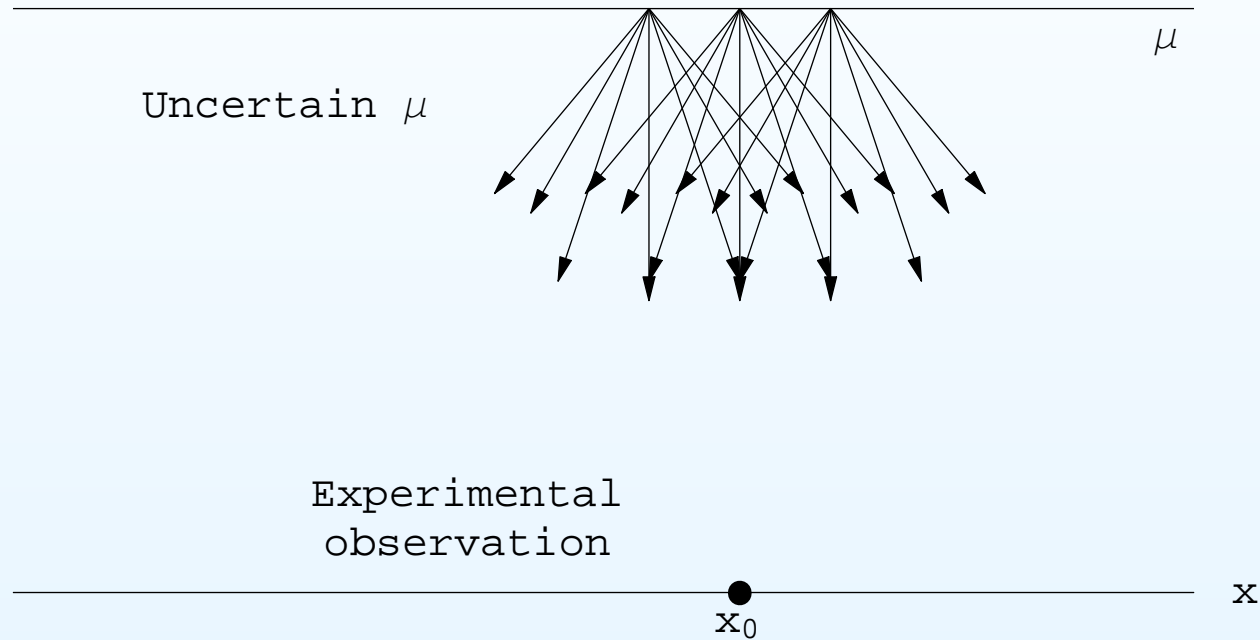
Given μ (exactly known) we are uncertain about x

From 'true value' to observations



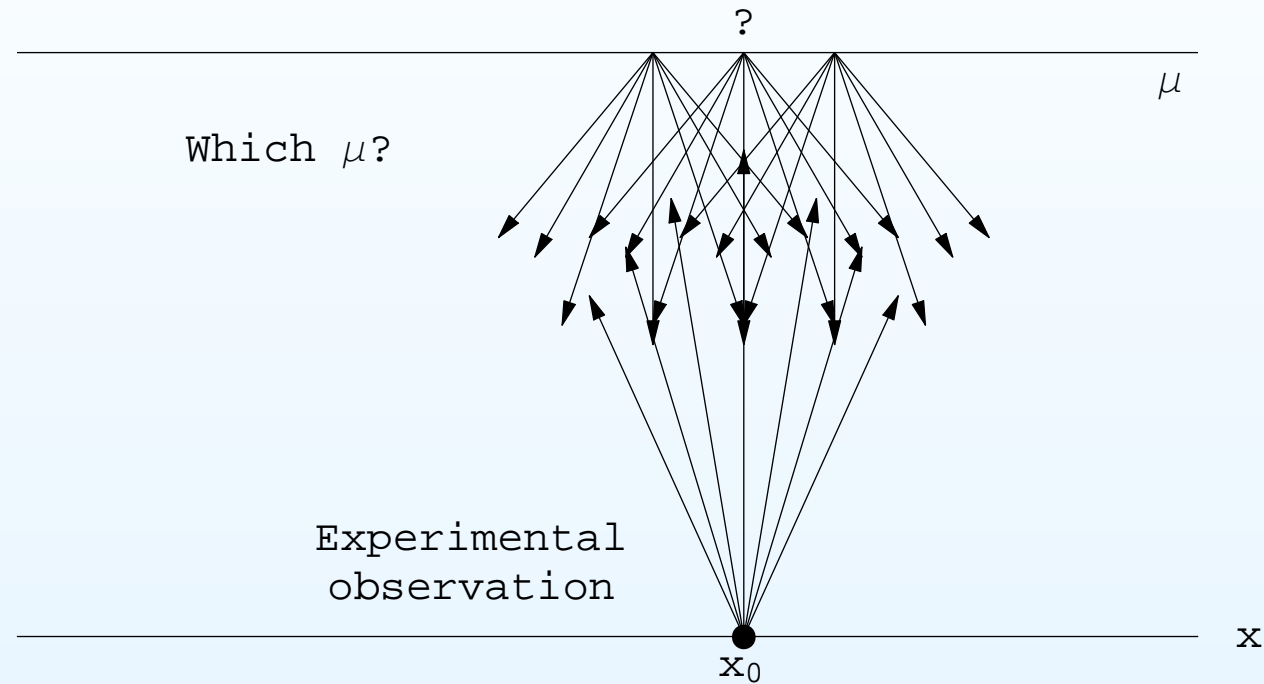
Uncertainty about μ makes us more uncertain about x

Inferring a true value



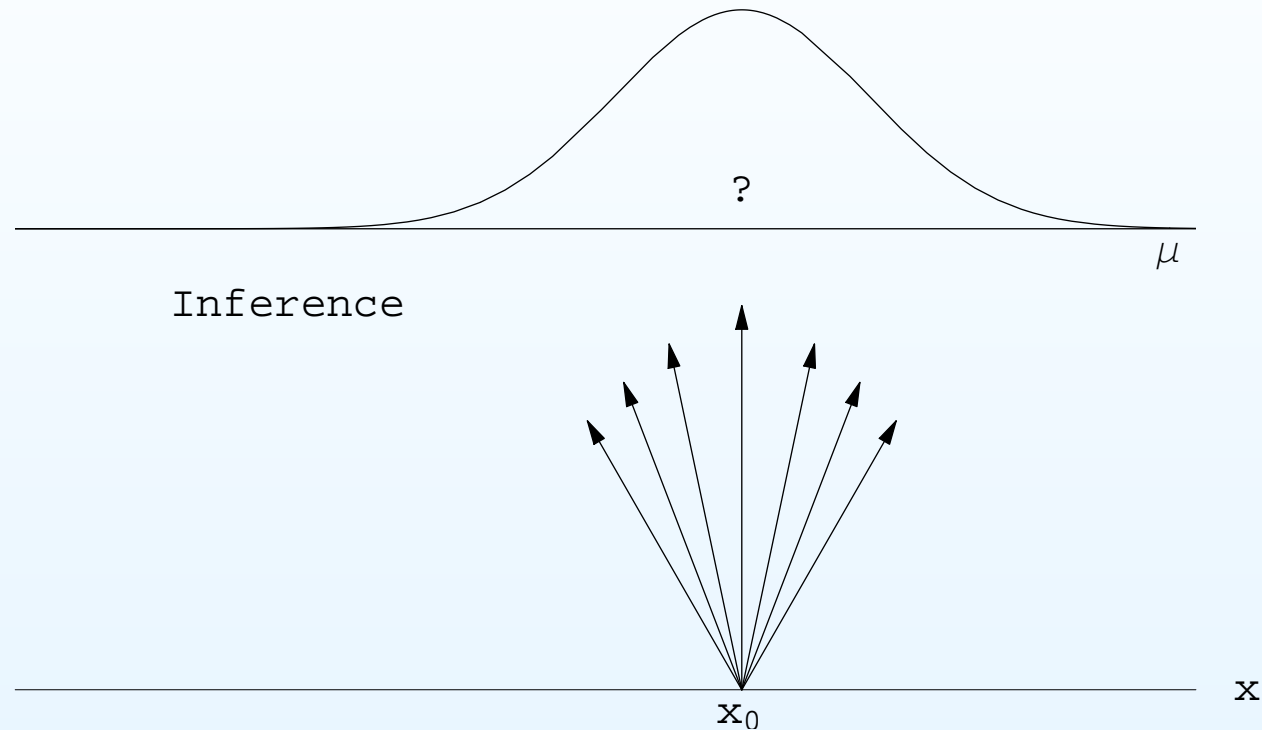
The observed data is certain: \rightarrow 'true value' uncertain.

Inferring a true value



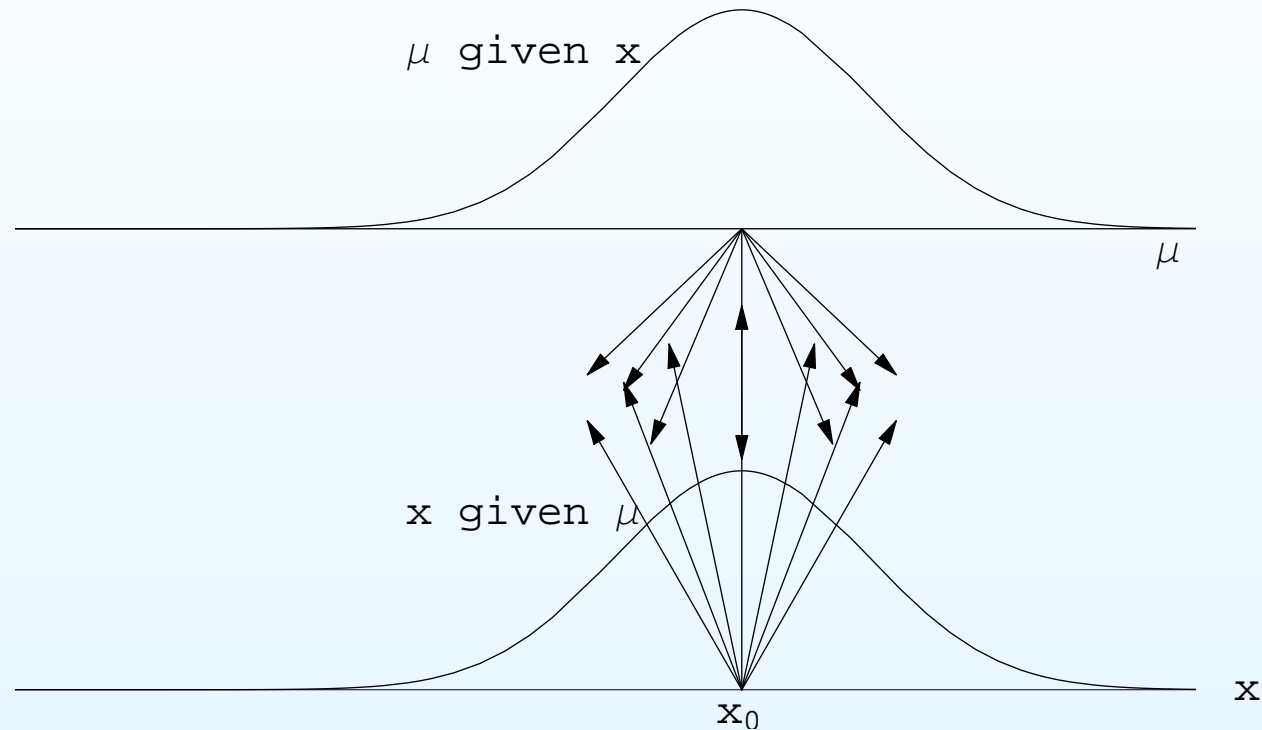
Where does the observed value of x comes from?

Inferring a true value



We are now uncertain about μ , given x .

Inferring a true value



Note the symmetry in reasoning.

Uncertainty

The human mind is used to live — and survive — in conditions of uncertainty and has developed mental categories to handle it.

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As a matter of fact, although we are in a constant state of uncertainty about many events which might or might not occur,

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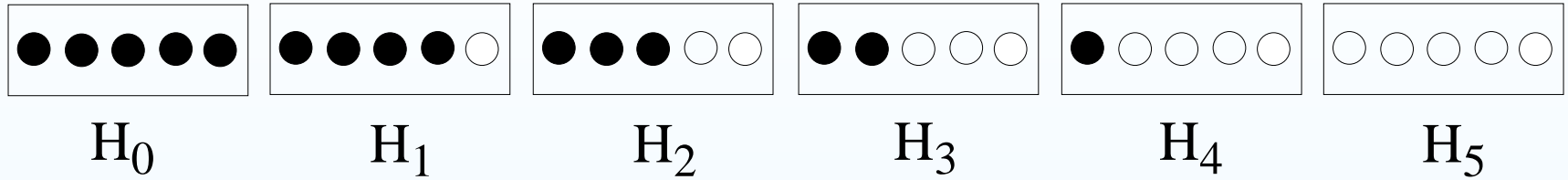
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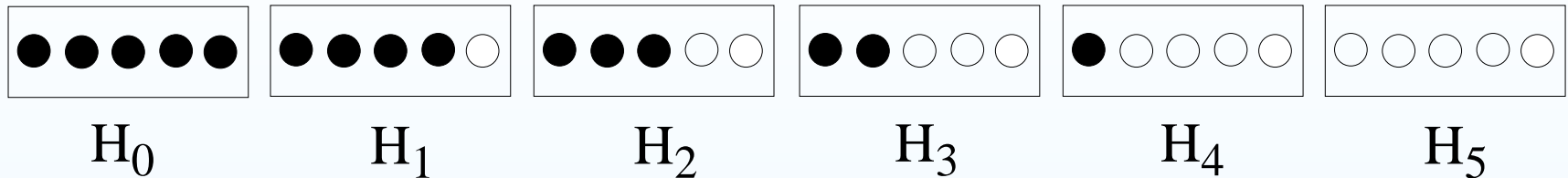
We can use similar expressions, all referring to the intuitive idea of **probability**.

The six box problem



Let us take randomly one of the boxes.

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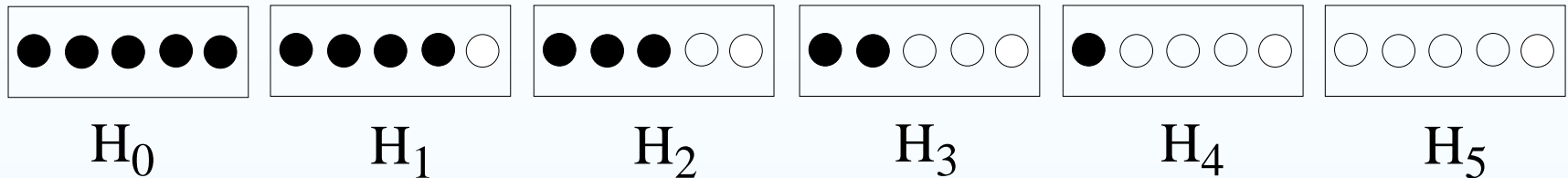
We are in a state of uncertainty concerning several *events*, the most important of which correspond to the following questions:

- (a) Which box have we chosen, H_0, H_1, \dots, H_5 ?
- (b) If we extract randomly a ball from the chosen box, will we observe a white ($E_W \equiv E_1$) or black ($E_B \equiv E_2$) ball?

Our certainty:

$$\begin{aligned} \bigcup_{j=0}^5 H_j &= \Omega \\ \bigcup_{i=1}^2 E_i &= \Omega. \end{aligned}$$

The six box problem

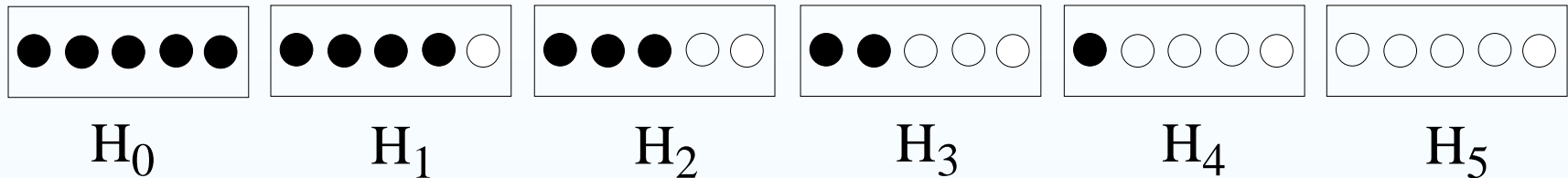


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 - And after a sequence of extractions?

The toy inferential experiment

The aim of the experiment will be to **guess** the content of the box without looking inside it, only extracting a ball, record its color and reintroducing in the box

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This toy experiment is conceptually very close to what we do in Physics

- try to guess what we cannot see (the electron mass, a branching ratio, etc)
... from what we can see (somehow) with our senses.

The rule of the game is that we are not allowed to watch inside the box! (As we cannot open and electron and read its properties, like we read the MAC address of a PC interface)

Doing Science in conditions of uncertainty

The constant status of uncertainty does not prevent us from doing Science (in the sense of Natural Science and not just Mathematics)

Doing Science in conditions of uncertainty

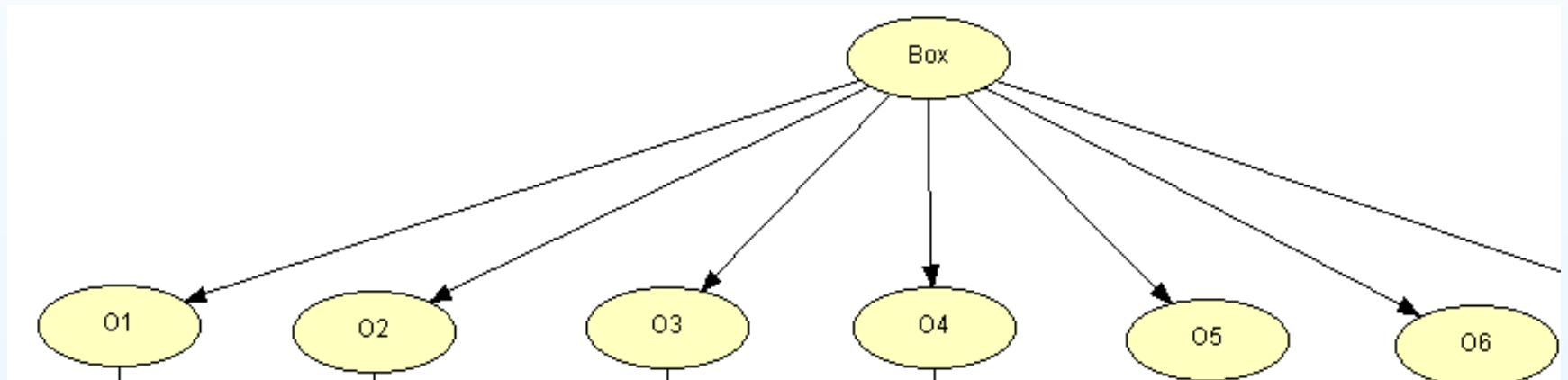
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Indeed

“It is scientific only to say what is more likely and what is less likely” (Feynman)

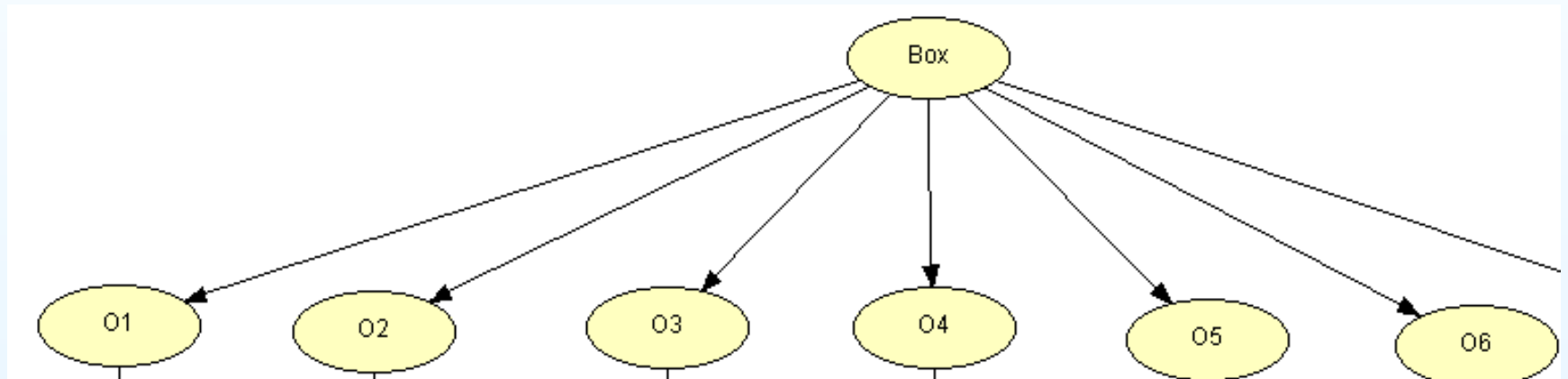
Cause-effect representation

box content \rightarrow observed color



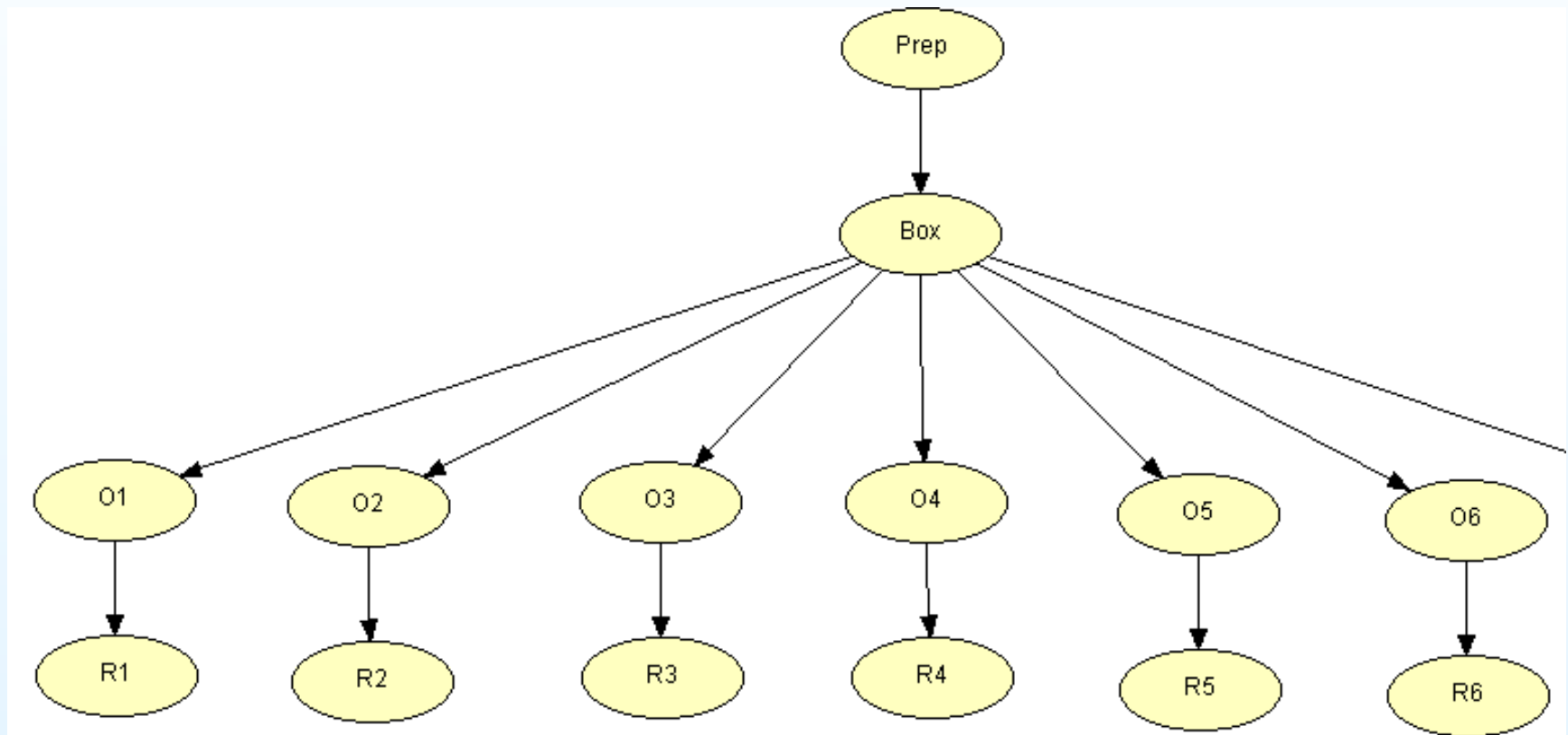
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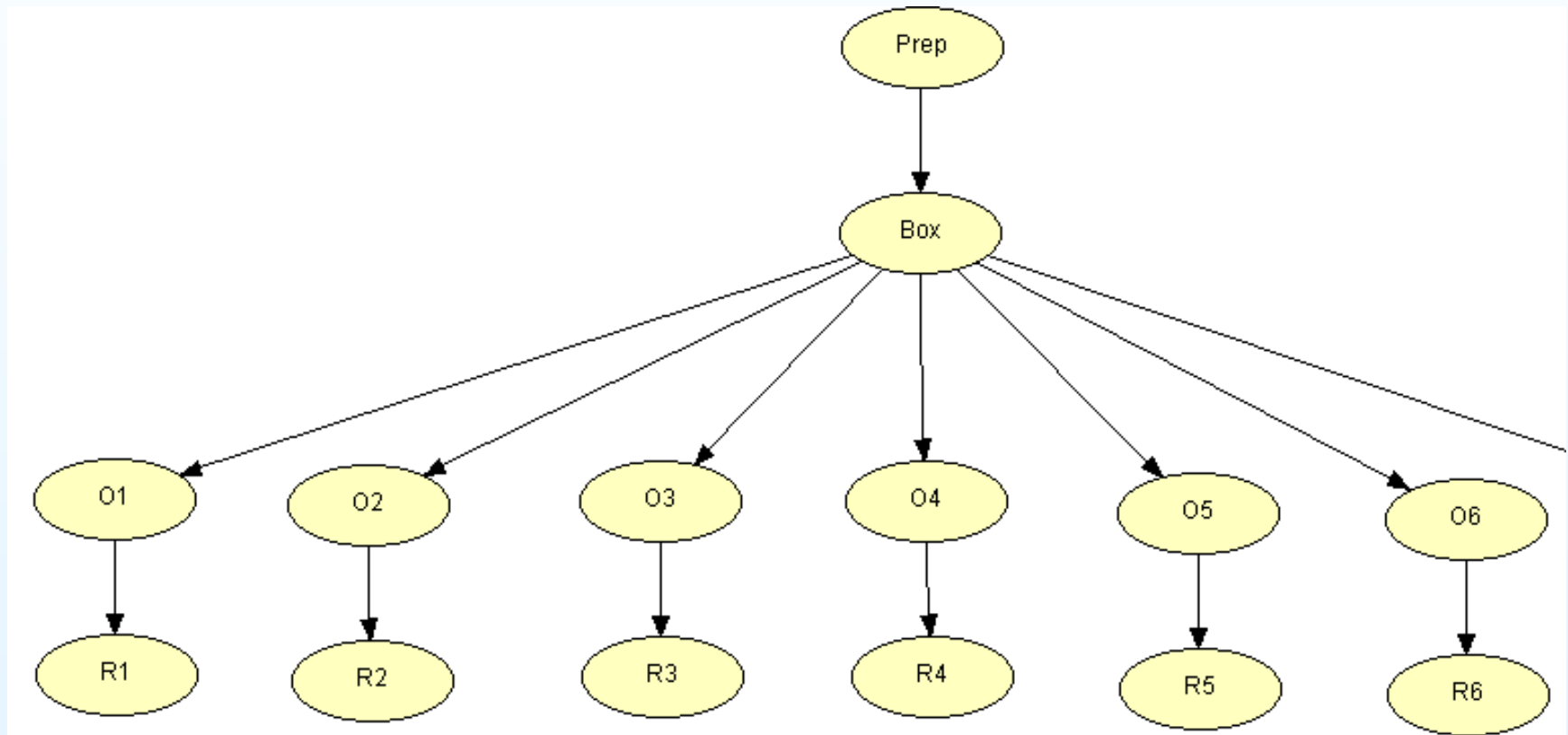


An effect might be the cause of another effect \longrightarrow

A network of causes and effects



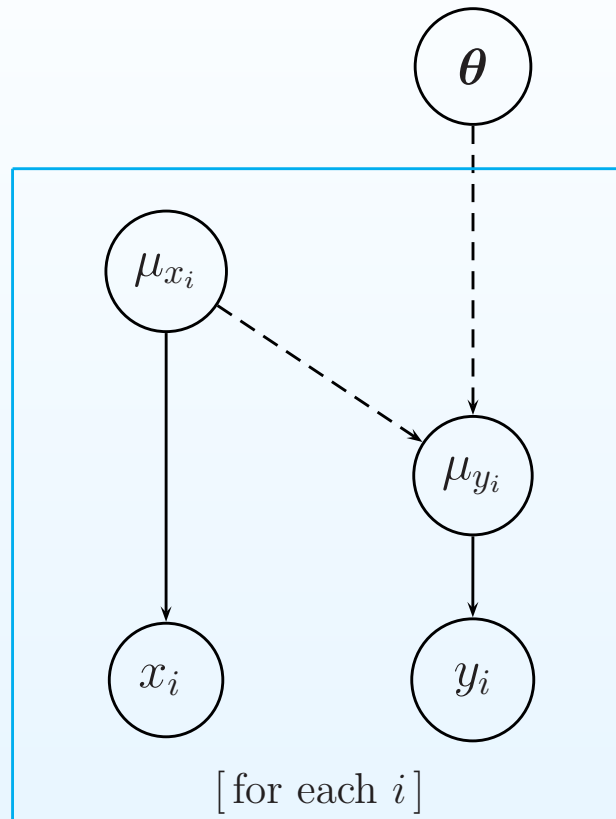
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and so on...

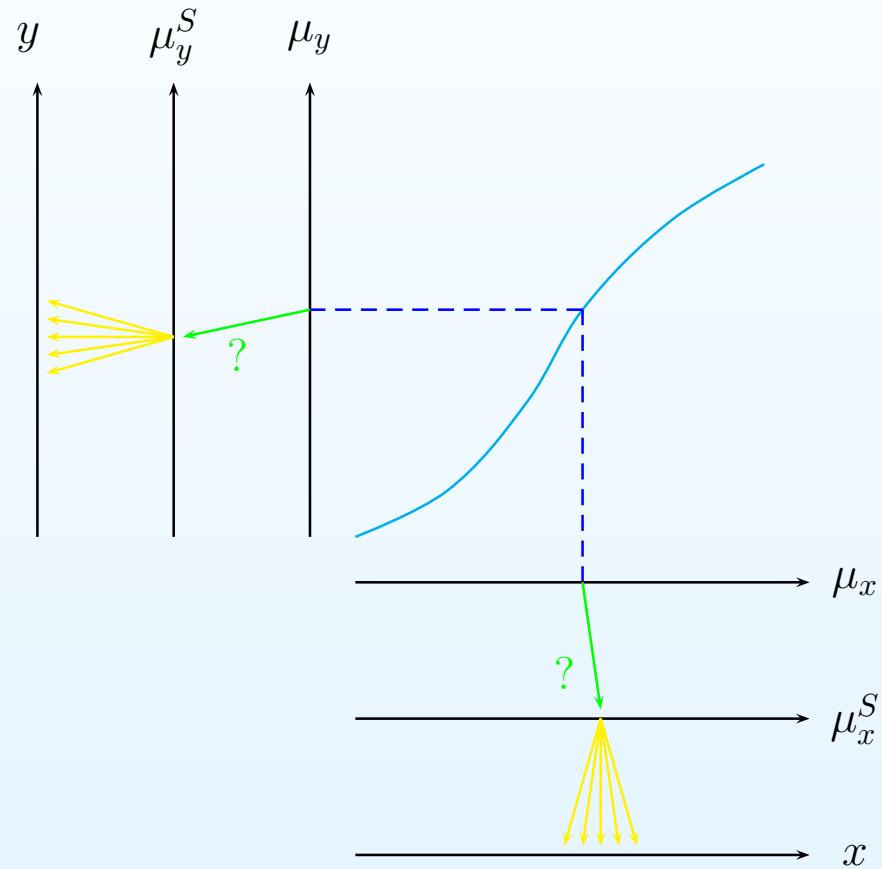
⇒ **Physics applications**

A different way to view fit issues



Deterministic link μ_x 's to μ_y 's
Probabilistic links $\mu_x \rightarrow x, \mu_y \rightarrow y$
(errors on both axes!)
 \Rightarrow aim of fit: $\{x, y\} \rightarrow \theta$

A different way to view fit issues



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- **... Fisichettume**

[Le varie formulette di “calcolo e propagazione degli errori”]

⇒ **Segue su lucidi:** vedi pp. 13-26 Ref. [2]

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It seems OK, but it is naive for several aspects.

Let start realizing that the method is analogous with method of the proof by contradiction of classical, deductive logic.

- Assume that a hypothesis is true
- Derive 'all' logical consequence
- If (at least) one of the consequences is known to be false, then the hypothesis is declared false.

Falsificationism? OK, but...

- What to do of all hypotheses that are not falsified? (Limbus? Get stuck?)

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- What to do of all hypotheses that are not falsified? (Limbus? Get stuck?)
- What to do is nothing of what can be observed is incompatible with the hypothesis (or with many hypotheses)?

E.g. H_i being a Gaussian $f(x | \mu_i, \sigma_i)$

⇒ Given any pair of parameters $\{\mu_i, \sigma_i\}$, all values of x between $-\infty$ and $+\infty$ are possible.

⇒ Having observed any value of x , none of H_i can be, strictly speaking, falsified.

Falsificationism and statistics

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in which the impossible is replaced by the improbable!

But from the impossible to the improbable there is **not just a question of quantity, but a question of quality.**

This mechanism, logically flawed, is particularly perverse, because deeply rooted in most people, due to education, but is not supported by logic.

⇒ Basically responsible of all fake claims of discoveries in the past decades.

[I am particularly worried about claims concerning our health, or the status of the planet, of which I have no control of the experimental data.]

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OK

~~B) if $C_i \xrightarrow{\text{small probability}} E$, and we observe E~~

~~NO~~

~~$\Rightarrow C_i$ has small probability to be true
"most likely false"~~

Example 1

Playing lotto

H : “I play honestly at lotto, betting on a rare combination”

E : “I win”

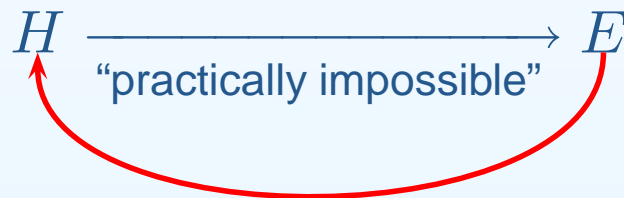
$H \xrightarrow{\text{“practically impossible”}} E$

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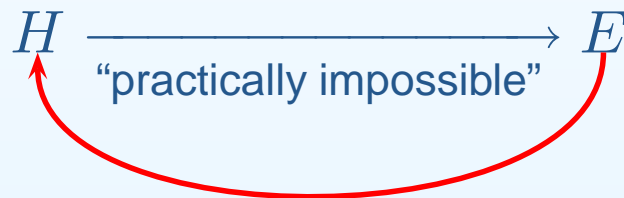
“practically to exclude”

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“practically to exclude”

⇒ almost certainly I have cheated...
(or it is false that I won...)

Example 2

An Italian citizen is selected at random to undergo an AIDS test. Performance of clinical trial is not perfect, as customary.

Toy model:

$$P(\text{Pos} \mid \text{HIV}) = 100\%$$

$$P(\text{Pos} \mid \overline{\text{HIV}}) = 0.2\%$$

$$P(\text{Neg} \mid \overline{\text{HIV}}) = 99.8\%$$

$H_1 = \text{'HIV'}$ (Infected)

$E_1 = \text{Positive}$

$H_2 = \overline{\text{'HIV'}}$ (Healthy)

$E_2 = \text{Negative}$

Example 2

An Italian citizen is selected at random to undergo an AIDS test. Performance of clinical trial is not perfect, as customary.

Toy model:

$$P(\text{Pos} \mid \text{HIV}) = 100\%$$

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Infected or healthy?

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Instead, $P(\text{HIV} | \text{Pos, random Italian}) \approx 45\%$

(We will see in the sequel how to evaluate it correctly)

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... which might result into **very bad decisions!**

'Standard' statistical tests, p-values, etc

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 - as far as logic is concerned, the situation is worsened (. . . although p-values ‘often, by chance work’).
- Mistrust statistical tests, unless you know the details of what it has been done.
→ You might take bad decisions!

Conflict: natural thinking \Leftrightarrow cultural superstructure

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- \Rightarrow **BUT** people think naturally in terms of probability of causes, and use p-values as if they were probabilities of null hypotheses. \Rightarrow **Terrible mistakes!**

Probabilistic reasoning

What to do?

⇒ Back to the past

Probabilistic reasoning

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⇒ **Back to the past**

But benefitting of

- Theoretical progresses in probability theory
- Advance in computation (both symbolic and numeric)
 - many frequentistic ideas had their *raison d'être* in the **computational barrier** (and many simplified – often simplistic – methods were ingeniously worked out)
 - **no longer an excuse!**

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⇒ Use consistently probability theory

- “It’s easy if you try”
- But first you have to recover the intuitive idea of probability.

Probability

What is probability?


Standard textbook definitions

$$p = \frac{\# \text{ favorable cases}}{\# \text{ possible equiprobable cases}}$$

$$p = \frac{\# \text{ times the event has occurred}}{\# \text{ independent trials under same conditions}}$$

Standard textbook definitions

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Standard textbook definitions

It is easy to check that ‘scientific’ definitions suffer of circularity

$$p = \frac{\# \text{ favorable cases}}{\# \text{ possible equally possible cases}}$$

$$p = \frac{\# \text{ times the event has occurred}}{\# \text{ independent trials under same conditions}}$$

Laplace: *“lorsque rien ne porte à croire que l’un de ces cas doit arriver plutôt que les autres”*

Pretending that replacing ‘equi-probable’ by ‘equi-possible’ is just cheating students (as I did in my first lecture on the subject...).

Standard textbook definitions

It is easy to check that ‘scientific’ definitions suffer of circularity, plus other problems

$$p = \frac{\# \text{ favorable cases}}{\# \text{ possible equiprobable cases}}$$

$$p = \lim_{n \rightarrow \infty} \frac{\# \text{ times the event has occurred}}{\# \text{ independent trials under same condition}}$$

Future \Leftrightarrow Past (believed so)

- $n \rightarrow \infty$:
- “*usque tandem?*”
 - “*in the long run we are all dead*”
 - It limits the range of applications

Definitions → evaluation rules

Very useful evaluation rules

$$A) \quad p = \frac{\# \text{ favorable cases}}{\# \text{ possible equiprobable cases}}$$

$$B) \quad p = \frac{\# \text{ times the event has occurred}}{\# \text{ independent trials under same condition}}$$

If the implicit beliefs are well suited for each case of application.

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If the implicit beliefs are well suited for each case of application.

BUT they cannot define the concept of probability!

Definitions → evaluation rules

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If the implicit beliefs are well suited for each case of application.

In the probabilistic approach we are going to see

- Rule *A* will be recovered immediately (under the assumption of equiprobability, when it applies).
- Rule *B* will result from a theorem (under well defined assumptions).

Probability

What is probability?

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It is what everybody knows what it is before going at school

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→ how much we are confident that something is true

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- how much we are confident that something is true
- how much we believe something

What is probability?

It is what everybody knows what it is before going at school

- how much we are confident that something is true
- how much we believe something
- “A measure of the degree of belief that an event *will* occur”

[Remark: ‘will’ does not imply future, but only uncertainty.]

Or perhaps you prefer this way...

“Given the state of our knowledge about everything that could possible have any bearing on the coming true¹ . . . ,

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“Given the state of our knowledge about everything that could possible have any bearing on the coming true¹ . . . , the numerical probability p of this event is to be a real number by the indication of which we try in some cases to setup a quantitative measure of the strength of our conjecture or anticipation, founded on the said knowledge, that the event comes true”

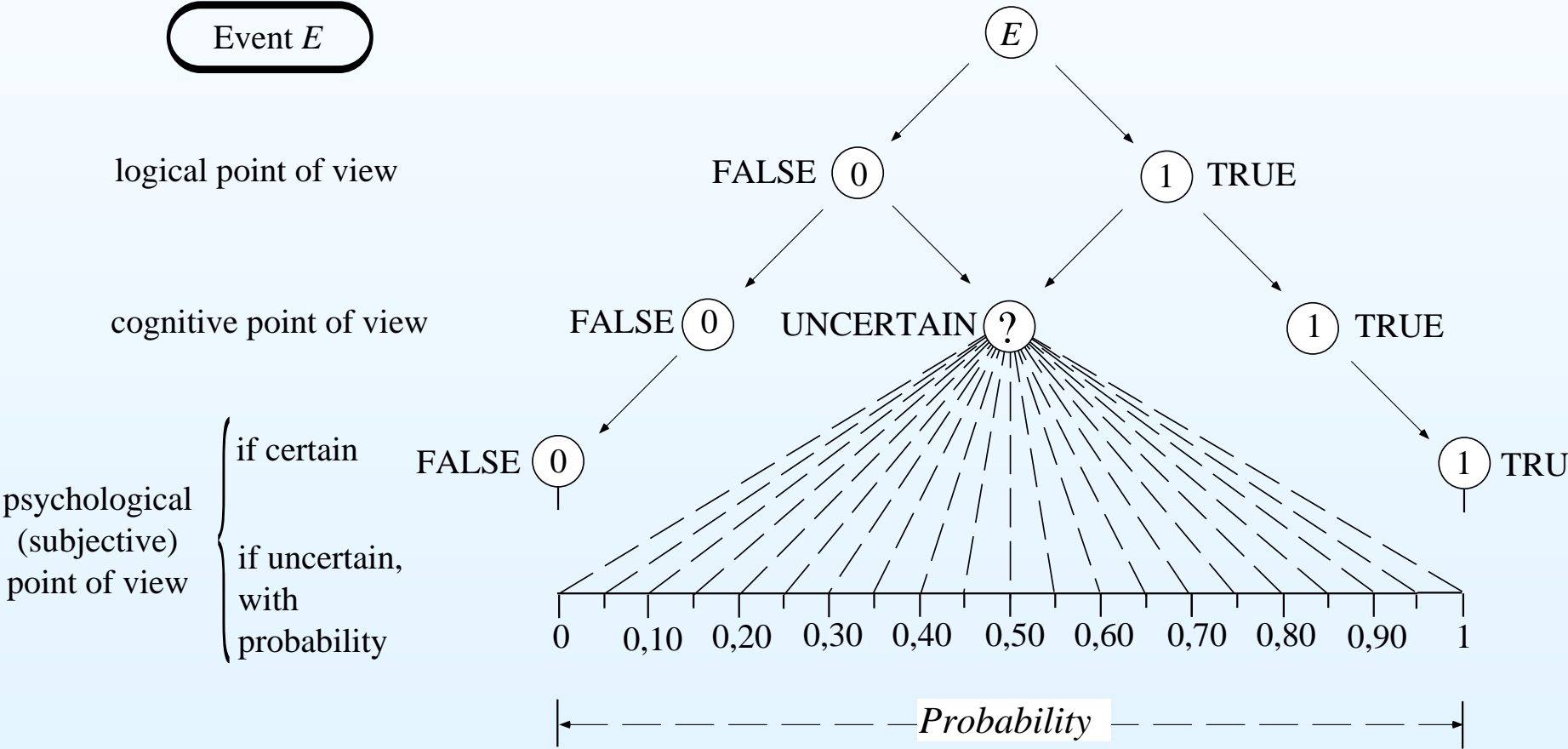
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*(E. Schrödinger, *The foundation of the theory of probability - I*, Proc. R. Irish Acad. 51A (1947) 51)*

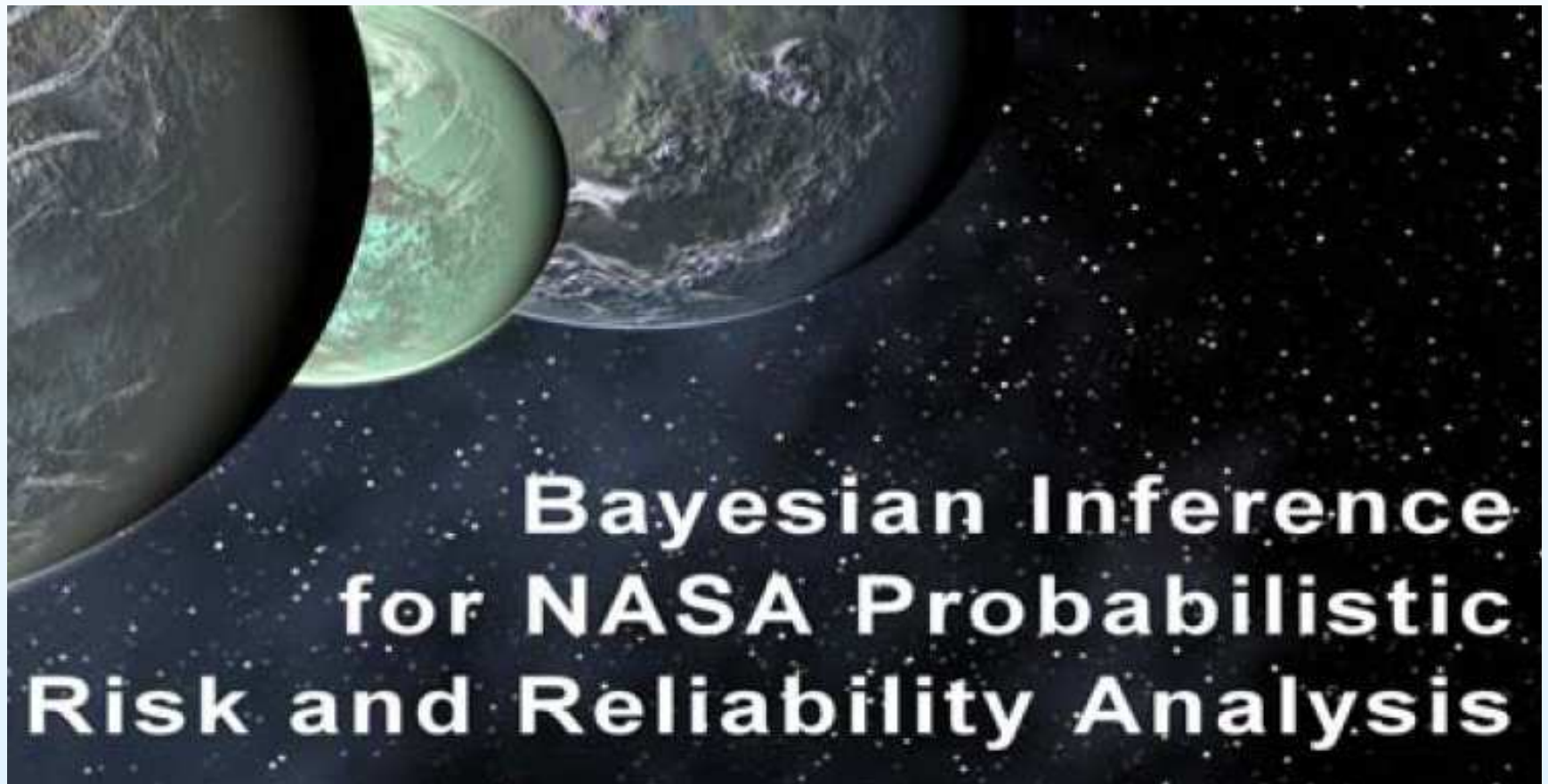
¹ *While in ordinary speech “to come true” usually refers to an event that is envisaged before it has happened, we use it here in the general sense, that the verbal description turns out to agree with actual facts.*

False, True and probable

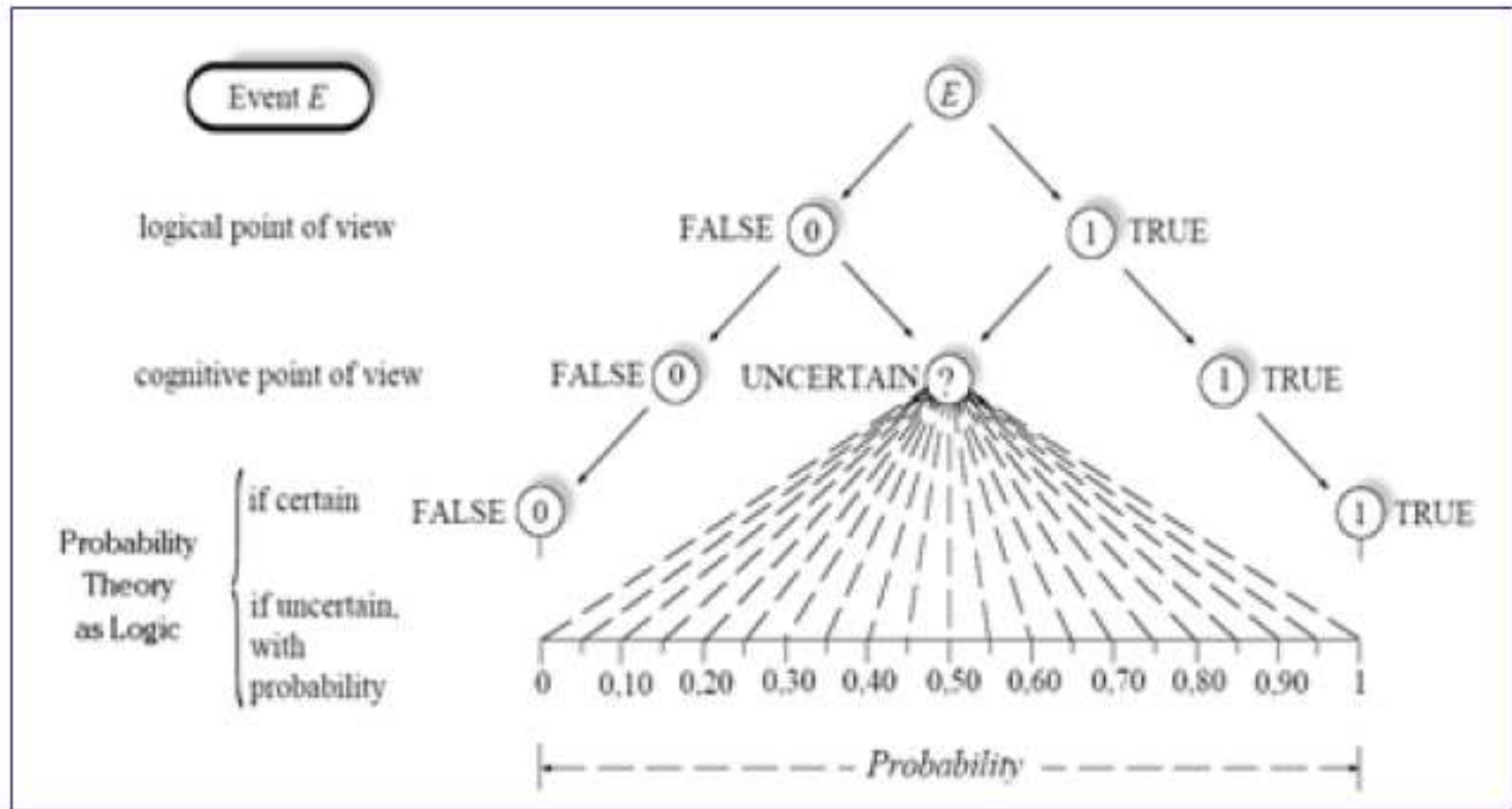


An helpful diagram

The previous diagram seems to help the understanding of the concept of probability



An helpful diagram



- Figure 2-1. Graphical abstraction of probability as a measure of information (adapted from "Probability and Measurement Uncertainty in Physics" by D'Agostini, [1995]).

(... but NASA guys are afraid of 'subjective', or 'psychological')