Classical mechanics ...

\ldots viewed as a classical probability theory with a dynamical law

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Random variables

Stochastic kernels and spectral representation Deterministic kernels and sharp measurements

Reminder of the Kolmogorov definition (1)

- Abstract measurable space (Ω, \mathcal{F}) , $\mathcal{F} \subseteq \mathcal{P}(\Omega)$.
 - $\Omega\in \mathcal{F}$,
 - $A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}$,
 - $(A_n)_{n\in\mathbb{N}}\in\mathcal{F}\Rightarrow\cup_{n\in\mathbb{N}}A_n\mathcal{F}.$
- Probability measure on $\Omega\in\mathcal{F},$ i.e. $\mathbb{P}:\mathcal{F}\rightarrow[0,1]$
 - $\mathbb{P}(\Omega) = 1$,
 - $\mathbb{P}(\sqcup_{n\in\mathbb{N}}A_n) = \sum_{n\in\mathbb{N}}\mathbb{P}(A_n).$
- Concrete measurable space $(\mathbb{X}, \mathcal{X})$.
- \mathbb{X} -valued random variable: any $(\mathcal{F}, \mathcal{X})$ -measurable map $X : \Omega \to \mathbb{X}$.



Random variables

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Reminder of the Kolmogorov definition (2)

Remark

Probability \mathbb{P} on (Ω, \mathcal{F}) does not intervene directly in the definition of X. It induces nevertheless a probability \mathbb{P}_X on $(\mathbb{X}, \mathcal{X})$, the law of X, by

$$\mathcal{X} \ni A \mapsto \mathbb{P}_X(A) := \mathbb{P}(X^{-1}(A)) = \mathbb{P}(\{\omega \in \Omega : X(\omega) \in A\}).$$

Remark

Important in definition of X the concrete space X. The abstract space Ω is irrelevant.

Blackboard 1: 3 ways to toss a coin ...



Probability theory

Postulates of classical mechanics Insufficiency of classical probability to describe Nature

Random variables

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Reducibility of classical randomness



Figure: From: Diaconis, Holmes, Montgomery, Dynamical bias in the coin toss, SIAM Review 2007.



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Reminder on stochastic kernels

Definition

Let (Ω, \mathcal{F}) and $(\mathbb{X}, \mathcal{X})$ measurable spaces. Map

 $K: \Omega \times \mathcal{X} \to [0, 1]$

is stochastic kernel from (Ω, \mathcal{F}) to $(\mathbb{X}, \mathcal{X})$ if

- $\forall \omega \in \Omega, \mathcal{K}(\omega, \cdot)$ probability on \mathcal{X} , and
- $\forall A \in \mathcal{X}, K(\cdot, A)$ measurable function.

Blackboard 2: example of 2 coins



Probability theory

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Action of K

• $K(\omega, \cdot)$ probability. Hence

$$egin{array}{rcl} b\mathcal{X}
i f &\mapsto & extsf{K}f \in b\mathcal{F} \ Kf(\omega) &:= & \int_{\mathbb{X}} K(\omega, dx) f(x). \end{array}$$

• $K(\cdot, A)$ (bounded) measurable function. Hence

$$\mathcal{M}_1(\mathcal{F}) \ni \mu \quad \mapsto \quad \mu K \in \mathcal{M}_1(\mathcal{X}) \ \mu K(A) \quad := \quad \int_{\Omega} \mu(d\omega) K(\omega, A).$$

Blackboard 3: contravariant and covariant functors.

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Probability theory

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Deterministic kernel K

Definition

Stochastic kernel K is deterministic if

$$\forall \omega \in \Omega, \exists ! x := x_{\mathcal{K}}(\omega) \in \mathbb{X} : \mathcal{K}(\omega, A) = \epsilon_{x}(A) = \mathbb{1}_{A}(x).$$

Blackboard 4: stochastic matrices and extremal stochastic matrices. Blackboard 5: equivalence $X \leftrightarrow K$ for discrete r.v. Blackboard 6: equivalence $X \leftrightarrow K$ for continuous r.v.



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Random variables and detrerministic kernels

For X r.v. on $(\Omega, \mathcal{F}, \mu)$ and values in $(\mathbb{X}, \mathcal{X})$, kernel $K := K_X$

$$K(\omega, A) = \mathbb{1}_{X^{-1}(A)}(\omega) = \epsilon_{X(\omega)}(A)$$

conveys exactly same information as X.

$$\begin{aligned} \forall \omega \in \Omega, X(\omega) &= \int_{\mathbb{X}} \epsilon_{X(\omega)}(dx)x = \int_{\mathbb{X}} K(\omega, dx)x = (K \mathrm{id}_{\mathbb{X}})(\omega), \\ \forall A \in \mathcal{X}, \mathbb{P}_{X}(A) &= \mu(X^{-1}(A)) = \int_{\Omega} \mu(d\omega) \mathbb{1}_{X^{-1}(A)}(\omega) \\ &= \int_{\Omega} \mu(d\omega) K(\omega, A) = (\mu K)(A). \end{aligned}$$



Archetypal example of a physical sharp measurement

- $X \leftrightarrow K_X$ with K_X deterministic kernel.
- For fixed X (hence K_X) define sharp elementary observable:

$$\mathcal{X} \ni A \mapsto M(A) := K(\cdot, A) = \mathbb{1}_{X^{-1}(A)} \in b\mathcal{F}.$$

- Random variable recovery: $X(\omega) = \int_{\mathbb{X}} M(dx)(\omega)x$.
- Precise preparation of the system $\mu \in \mathcal{M}_1(\mathcal{F})$.
- Measurement: $\mathbf{S} \times \mathbf{O} \ni (\mu, M) \mapsto \pi^{\mu}_{M} \in \mathcal{M}_{1}(\mathcal{X})$, where

$$\pi^{\mu}_{M}(A) = \int_{\Omega} \mu(d\omega) M(A)(\omega) = \int_{\Omega} \mu(d\omega) K(\omega, A) = (\mu K)(A).$$

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Gambling with classical dice I

- Dice shows up a face $\omega \in \Omega := \{1, 2, \dots, 6\}.$
- Gambler's net gain determined by the random variable X:

$$X(\omega) = [(\omega - 1 \mod 3) - 1] \in \mathbb{X} := \{-1, 0, 1\}.$$

Two ways to represent information conveyed by X: either as a 6-dimensional vector V or as a 6 × 3 stochastic deterministic matrix K:

$$V := \begin{pmatrix} -1 \\ 0 \\ 1 \\ -1 \\ 0 \\ 1 \end{pmatrix}; \mathcal{K} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$



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Gambling with classical dice II

Observable *M* is the family $M = (M_x)_{x \in \mathbb{X}}$ of elementary observables , where

$$M_{\mathbf{X}}(\omega) := K(\omega, \mathbf{X}) = \mathbb{1}_{\mathbf{X}}(X(\omega)) = \mathbb{1}_{\mathbf{X}^{-1}(\{\mathbf{X}\})}(\omega) = \delta_{\mathbf{X}(\omega)}(\{\mathbf{X}\}).$$

$$M_{-1} := \begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}; M_0 := \begin{pmatrix} 0 \\ 1 \\ 0 \\ 1 \\ 0 \end{pmatrix}; M_1 := \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 1 \end{pmatrix}.$$

Remark

- $\forall x \in \mathbb{X}, \ \omega \in \Omega, \ M_x(\omega) \ge 0 \text{ and } M_x^2(\omega) = M_x(\omega).$ (i.e. M_x projections).
- $\sum_{x \in \mathbb{X}} M_x = 1$. ((M_x) $_{x \in \mathbb{X}}$ resolution of identity).
- $X = \sum_{x \in \mathbb{X}} M_x x$. ("Spectral decomposition" of X).

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Gambling with classical dice III

For preparation of dice in state μ ∈ M₁(F), measurement determines a probability π^μ_M ∈ M₁(X) by

$$\pi^{\mu}_{M}(x) = \mu(\{\omega \in \Omega : X(\omega) = x\})$$
$$= \sum_{\omega \in X^{-1}(\{x\})} \mu(\omega)$$
$$= \langle \mu, M_{x} \rangle$$
$$= \sum_{\omega \in \Omega} \mu(\omega) M_{x}(\omega).$$

Definition

Observable decomposable into family of projections (M_x) called **sharp**.



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Gambling with classical dice IV

• Example of 2 different preparations of the system "dice":

$$\mu_1 = (\frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}); \ \mu_2 = (\frac{1}{32}, \frac{1}{32}, \frac{1}{16}, \frac{1}{8}, \frac{1}{4}, \frac{1}{2})$$

• Corresponding probability measures in $\mathcal{M}_1(\mathbb{X})$:

$$\pi_M^{\mu_1} = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3}); \ \pi_M^{\mu_2} = (\frac{5}{32}, \frac{9}{32}, \frac{18}{32}).$$

• Average value $\mathbb{E}_{\mu}(X) = \sum_{x \in \mathbb{X}} \pi^{\mu}_{M}(x) x$:

$$\mathbb{E}_{\mu_1}(X) = 0; \ \mathbb{E}_{\mu_2}(X) = -\frac{5}{32} + \frac{18}{32} = \frac{13}{32}.$$



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Randomised gambling with classical dice I

Again gambler's net gain (observable) $\leftrightarrow K$ but now K genuine stochastic matrix, e.g.

Exercise

What is the significance of the vector V?

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Randomised gambling with classical dice II

Remark

- $\forall x \in \mathbb{X}, \ \omega \in \Omega, \ M_x^2(\omega) \le M_x(\omega)$. (M_x not projections).
- $\sum_{x \in \mathbb{X}} M_x = 1$. ((M_x)_{$x \in \mathbb{X}$} resolution of identity).
- $\pi_M^{\mu}(x) = \langle \mu, M_x \rangle = \sum_{\omega \in \Omega} \mu(\omega) M_x(\omega)$. (But (M_x) do not provide spectral decomposition of X).
- But still average gain in state μ given by $\mathbb{E}_{\mu} X = \sum_{x \in \mathbb{X}} \pi^{\mu}_{M}(x) x.$

Definition

Resolution of identity $M = (M_x)$ with M_x positive but not necessarily projections called **unsharp or randomised observable**.



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Randomised gambling with classical dice III

With previous μ_1 and μ_2 :

$$\begin{aligned} \pi_{M}^{\mu_{1}} &= \mu_{1} \mathcal{K} = \left(\frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}\right) \begin{pmatrix} \frac{4}{5} & 0 & \frac{1}{5} \\ 0 & 1 & 0 \\ \frac{1}{5} & 0 & \frac{4}{5} \\ 0 & 1 & 0 \\ \frac{1}{5} & 0 & \frac{4}{5} \\ \end{pmatrix} &= \left(\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right); \\ \pi_{M}^{\mu_{2}} &= \mu_{2} \mathcal{K} = \left(\frac{1}{32}, \frac{1}{32}, \frac{1}{16}, \frac{1}{8}, \frac{1}{4}, \frac{1}{2}\right) \begin{pmatrix} \frac{4}{5} & 0 & \frac{1}{5} \\ 0 & 1 & 0 \\ \frac{1}{5} & 0 & \frac{4}{5} \\ 0 & 1 & 0 \\ \frac{1}{5} & 0 & \frac{4}{5} \\ 0 & 1 & 0 \\ \frac{1}{5} & 0 & \frac{4}{5} \\ 0 & 1 & 0 \\ \frac{1}{5} & 0 & \frac{4}{5} \\ \end{pmatrix} &= \left(\frac{38}{160}, \frac{45}{160}, \frac{77}{160}\right). \\ \mathbb{E}_{\mu_{1}}(\mathcal{X}) &= 0; \mathbb{E}_{\mu_{2}}(\mathcal{X}) = \frac{39}{160}. \end{aligned}$$

Phase space and states The dynamical law Observables Physical measurement Composite systems

Postulates of classical mechanics Phase space and states

Postulate (Phase space and states)

Phase space: a measurable space (Ω, \mathcal{F}) . **States:** possible preparations of the system $S = \mathcal{M}_1(\mathcal{F})$.

Set **S** is **convex**

$$\mu_1, \mu_2 \in \mathbf{S}, \lambda \in [0, 1] \Rightarrow \lambda \mu_1 + (1 - \lambda) \mu_2 \in \mathbf{S}.$$

Extremal points, i.e. states without non-trivial convex decomposition, are the **pure states** $\mathbf{S}_{p} = \{\epsilon_{\omega}, \omega \in \Omega\} \simeq \Omega$.



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RFN

Postulates of classical mechanics Dynamical law

Postulate (Dynamical law)

Time evolution of isolated^a system: a measurable invertible function $T : \Omega \rightarrow \Omega$.

^aNot exchanging mass or energy with environment.

T is a r.v. hence $\leftrightarrow K_T$ deterministic stochastic kernel.

Definition

State
$$\mu \in S$$
 invariant if $T_*\mu := \mu K_T = \mu$, i.e.

$$\forall A \in \mathcal{F}, \mu(T^{-1}A) = \mu(A).$$

Phase space and states The dynamical law **Observables** Physical measurement Composite systems

Postulates of classical mechanics Sharp general and elementary observables

Postulate (Observables)

- General sharp X-valued observables: random variables^a on phase space (Ω, F) taking values in (X, X).
- **Elementary sharp observables:** The {0,1}-valued spectral components M_x .
- General (unsharp) X-valued observables described by decompositions of identity (M_x) into positive but not necessarily projective components.

^aRecall r.v. $X \leftrightarrow$ deterministic stochastic kernel $K \leftrightarrow (M_x) : X = \sum_x M_x x$.

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Phase space and states The dynamical law Observables **Physical measurement** Composite systems

Postulates of classical mechanics Measurement

For any $X \in \mathbf{O}$ consider family of questions $(M(A))_{A \in \mathcal{X}}$.

 $M(A) = 1 \Leftrightarrow X \in A.$

Make system interact with **measuring apparatus** that determines whether question gets positive answer.

Postulate (Physical measurement)

For system in state μ ask question M(A) and determine probability of positive answer.

$$\mathbf{S} \times \mathbf{O} \ni (\mu, M(A)) \mapsto \pi^{\mu}_{M}(A) = \mu K(A).$$



Phase space and states The dynamical law Observables Physical measurement Composite systems

Postulates of classical mechanics Composite systems

System composed of N subsystems, each with its own phase space $(\Omega_i, \mathcal{F}_i)$ for i = 1, ..., N.

Postulate (Composite system)

The phase space (Ω, \mathcal{F}) of the composite system is

$$\begin{aligned} \Omega &= \times_{i=1}^{N} \Omega_i, \\ \mathcal{F} &= \otimes_{i=1}^{N} \mathcal{F}_i = \sigma(\times_{i=1}^{N} \mathcal{F}_i). \end{aligned}$$

States are not necessarily product measures! Blackboard 7: Two illustrative examples.



Hidden variables hypothesis

- In the next lecture, postulates of Quantum Mechanics.
- QM never been contradicted by experiment up to now.
- Nevertheless, "measurement postulate" so counter-intuitive that physicists searched ways of circumvention.
- One of the criticism on this postulate concerns the irreducibility¹ of quantum randomness it imposes.
- One attempt of circumvention was the "hidden variables" hypothesis².
- Personal view: Situation similar to aether hypothesis in EM.

¹Einstein's aphorism: "God does not play dice with the world". ²Bohm, A suggested interpretation of the quantum theory in terms of "NUMERSITE OF RENT" "hidden" variables. I+II, Phys. Rev. 85:166–179, 180–193 (1952).

Hidden variables, Bell's inequalities, the Orsay experiment

Experiments with polarisers



Experimental facts:

- When photon passes through first polariser in direction α emerges polarised in that direction.
- When second polariser encountered in direction β photon passes through with probability $\cos^2(\alpha \beta)$.
- If photon initially already polarised in direction α, nothing changes if the first polariser is removed.

Bell's inequalities

If hidden variables \Rightarrow Kolmogorov theory holds.

Proposition (Four-variable Bell's inequality)

 X_1, X_2, Y_1, Y_2 arbitrary quadruple of $\{0,1\}\text{-valued random variables. Then}$

$$\mathbb{P}(X_1 = Y_1) \leq \mathbb{P}(X_1 = Y_2) + \mathbb{P}(X_2 = Y_2) + \mathbb{P}(X_2 = Y_1).$$

Proof.

R.v. being $\{0, 1\}$ -valued, enough to check on all 16 possible realisations of quadruple $(X_1(\omega), X_2(\omega), Y_1(\omega), Y_2(\omega))$ that

$$\{X_1 = Y_1\} \subseteq \{[X_1 = Y_2] \lor [X_2 = Y_2] \lor [X_2 = Y_1]\}.$$



Hidden variables, Bell's inequalities, the Orsay experiment

The Orsay experiment

Aspect, Dalibard, Roger. Experimental test of Bell's inequalities using time-varying analyzers, Phys. Rev. Lett., 49: 1804–1807 (1982).





Experimental refutation of hidden variables hypothesis I

- $X_{\alpha} := 1 \Leftrightarrow \{ \text{left photon passes if polariser oriented in } \alpha \}.$
- $Y_{\beta} := 1 \Leftrightarrow \{ \text{right photon passes if polariser oriented in } \beta \}.$
- Experimental fact: $\mathbb{P}(X_{\alpha} = Y_{\beta}) = \sin^2(\alpha \beta)$.
- Bell's inequalities:

$$\mathbb{P}(X_{\alpha_1}=Y_{\beta_1}) \leq \mathbb{P}(X_{\alpha_1}=Y_{\beta_2}) + \mathbb{P}(X_{\alpha_2}=Y_{\beta_1}) + \mathbb{P}(X_{\alpha_2}=Y_{\beta_2}).$$

• With choice $\alpha_1 = 0$, $\alpha_2 = \pi/3$, $\beta_1 = \pi/2$, and $\beta_2 = \pi/6$:

$$\sin^2(\pi/2) \le \sin^2(-\pi/6) + \sin^2(-\pi/6) + \sin^2(\pi/6)$$

or else $\Rightarrow 1 \leq 1/4 + 1/4 + 1/4$.



Experimental refutation of hidden variables hypothesis II

- Orsay experiment can be seen as game you play against nature and you always loose!
- Exist other experiments³ refuting hidden variables, e.g. by use of Kochen-Specker theorem.



³Including by Chilean groups, e.g. Saavedra, Concepción.