

From Sample Probabilities to Random Processes

Life as a (somewhat frustrated)
measure non-theorist

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Random Processes

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But how are they characterized?

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ALSO:

want to compute global probabilities for the entire process.

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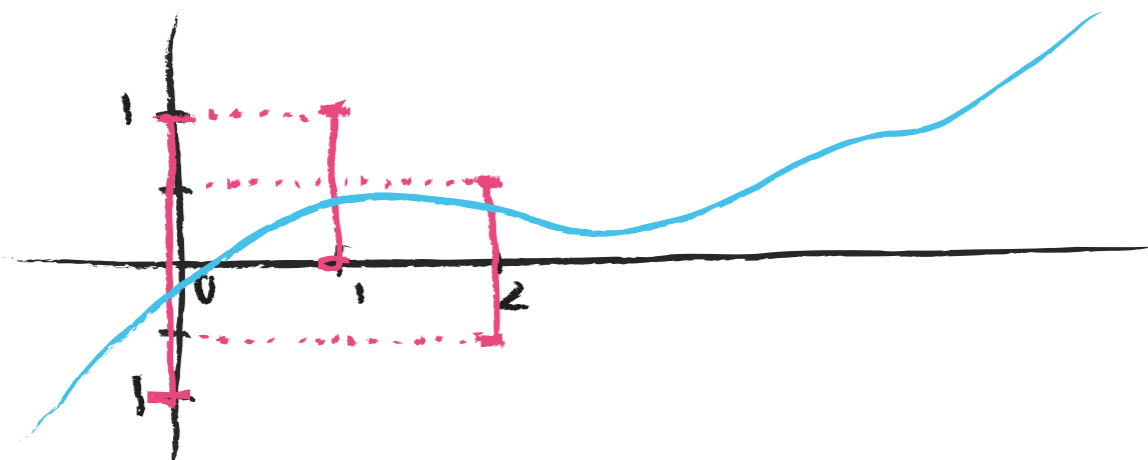
Cylinder sets = $\pi_{T_1}^{-1} A$

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Example:

$$T_1 = \{0, 1, 2\}$$

$$A = [-1, 1] \times [0, 1] \times [-1/2, 1/2]$$



$\pi_{T_1}^{-1} A =$ set of all functions

that satisfy

$$f(0) \in [-1, 1],$$

$$f(1) \in [0, 1],$$

$$f(2) \in [-1/2, 1/2].$$

Precise formulation of the problem:

We have a finitely additive set function

on the algebra $\mathcal{A} = \bigcup_{\substack{T_1 \subset T \\ \text{finite}}} \pi_{T_1}^{-1} \mathcal{G}_{T_1}$
defined as

Borel σ -algebra

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→ σ -algebra of cylinder sets

Kolmogorov's Extension Theorem (Kolmogorov, 1933):

for $G = \sigma(\mathcal{A})$:

Yes (if the finite measures are consistent).

SMALL PRINT

μ_T 's must possess an approximating compact class, which all Borel measures on \mathbb{R}^T , T finite, do possess.



Consistency = Given $T_2 \subset T_1 \subset T$ with $\#T_1 < \infty$,

$$M_{T_2}(A) = M_{T_1}(\pi_{T_1 \rightarrow T_2}^{-1} A) \quad \forall A \in \mathcal{G}_{T_2}$$

where $\pi_{T_1 \rightarrow T_2} : \mathbb{R}^{T_1} \rightarrow \mathbb{R}^{T_2}$ is the

natural projection $\mathbb{R}^{T_1} \rightarrow \mathbb{R}^{T_2}$.

why is this unsatisfactory?

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sets in $\sigma(\mathcal{A})$ are countably determined
(σ -combinations of finitely determined
cylinder sets)...

Many interesting sets are left out
(cannot compute their probability):

e.g. $\{f : |f| < a\}$ $C = \{f \text{ continuous on } \mathbb{R}\}$

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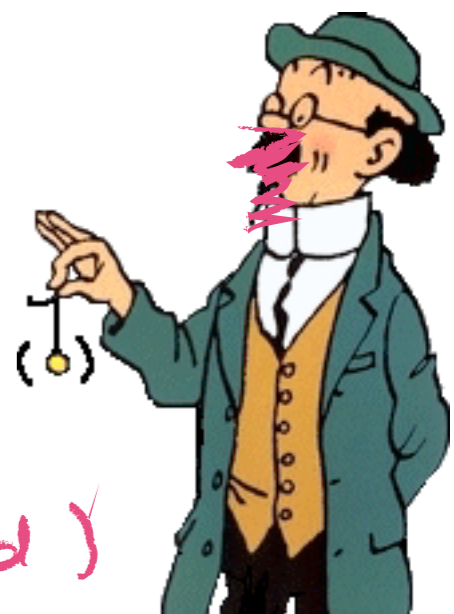
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(remove the beard)



Measures on linear spaces

Random processes on linear spaces

E = linear space;

f = random element of E^* ,

the algebraic dual of E

(= random linear functional on E);

μ = measure on σ -algebra of Borel cylinder sets in E^* associated with f .

notation: $f(e_i) = \langle e_i, f \rangle \rightsquigarrow R.V.$

How to characterize μ ?

RECALL:

Given

any lin. indep. vector $E_1 = (e_1, \dots, e_n) \in E^n$



Define

Projection $\pi_1: E^* \rightarrow \mathbb{R}^n: f \mapsto (f(e_1), \dots, f(e_n))$

For $A \in \text{Borel}(\mathbb{R}^n)$,

$\pi_1^{-1}A$ is a Borel cylinder set.

Consistency for linear processes: Given

$$E_1 = (e_1, \dots, e_m) \in E^n \quad \text{and}$$

$$E_2 = (\tilde{e}_1, \dots, \tilde{e}_m) = M E_1, \quad \begin{matrix} \text{matrix } \in \mathbb{R}^{n \times m} \end{matrix}$$

$$M_{E_2}(A) = M_{E_1}(M^{-1}A) \quad \forall A \in \text{Bowl}(\mathbb{R}^m)$$

Equivalently:


$$f\left(\sum \alpha_i e_i\right) = \sum \alpha_i \underbrace{f(e_i)}_{\text{R.V.}}$$

in distribution

Zorn's lemma
(\equiv Axiom of Choice)



E has a
(Hamel) basis T



f is uniquely determined
by $f(t), t \in T$



Problem is reduced to Kolmogorov's thm.

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enter Generalized Random Processes
of Gelfand & Co.

Generalized Random Process:

Family of R.V.s

$$\{f(e) \mid e \in E\}$$

where E is a topological vector space.

- consistency defined as before;

- New notion: Continuity:

$$\left. \begin{array}{l} e_{1,k} \rightarrow e_1 \\ \vdots \\ e_{n,k} \rightarrow e_n \\ \text{in } E \end{array} \right\} \Rightarrow (f(e_{1,k}), \dots, f(e_{n,k})) \rightarrow (f(e_1), \dots, f(e_n))$$

in distribution
 $M_{E_{1,k}} \rightarrow M_E$

Naturally, if μ is a measure on E'
(= continuous dual), consistency and continuity
are satisfied. ① ②

Q: When are ①, ② sufficient for extending
a cylinder measure to a σ -additive one?
on E'

(RECALL: consistency alone was sufficient to
have a measure on E^* .)

Minlos (1958) - proof of Gel'fand's conjecture:

For finite-dim. measures w/ ①, ② to uniquely extend to a measure on E' , it is sufficient that E be nuclear.



Examples of nuclear spaces:

\mathcal{S} = Schwartz space

→ random processes in \mathcal{S}'

\mathcal{D} = space of compactly supp.
test functions

→ random distributions

Characteristic functionals:

one way to define/construct
finite dim. measures fulfilling ①, ②.

NOT TODAY !!!

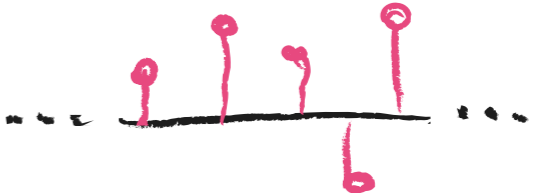


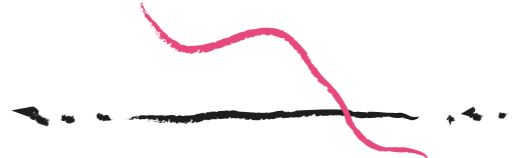
QUESTIONS ?

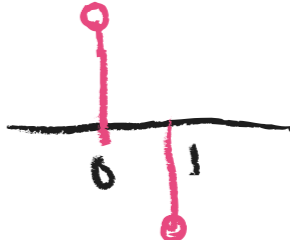
Note on notation:

$A^B = \text{set of all functions } B \rightarrow A$
sets

Examples:

$\mathbb{R}^{\mathbb{Z}}$ = set of all real sequences 

$\mathbb{R}^{\mathbb{R}}$ = set of all functions $\mathbb{R} \rightarrow \mathbb{R}$ 

$\mathbb{R}^{\{0,1\}}$ = set of all functions $\{0,1\} \rightarrow \mathbb{R}$ 

Proximity measure :

Probability measure:

\mathcal{G} = set system $\subset \mathcal{P}(X)$

closed under \cup \cap \setminus \rightarrow algebra
 $\sigma\cup$ $\sigma\cap$ \setminus \rightarrow σ -algebra

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$\mu: \mathcal{G} \rightarrow [0, 1]$

isotone, $\mu(\emptyset) = 0$

pairwise disjoint $A \rightarrow \mu\left(\bigcup_{A \in \mathcal{A}} A\right) = \sum_{A \in \mathcal{A}} \mu(A)$

\mathcal{A} finite \rightarrow additivity

\mathcal{A} countable \rightarrow σ -additivity