

# Randomness in a dynamical system

Kenshi Miyabe

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## Randomness on a Cantor space

- ❖ AIT
- ❖ Random sequence
- ❖ Questions?
- ❖ Computability
- ❖ Martin-Löf randomness
- ❖ Complexity
- ❖ Martingale
- ❖ van Lambalgen's Theorem
- ❖ An extension

Random closed set

Generalization

Randomness and Ergodic Theory

Probability

# Randomness on a Cantor space

## Algorithmic randomness is a subfield of Algorithmic Information Theory.

AIT is the result of putting Shannon's information theory and Turing's computability theory into a cocktail shaker and shaking vigorously. (G. J. Chaitin)

- From “for almost every” to “for all random points”
- Random deficiency
- Computable points with statistical properties
- Three approaches to a concept.

# Random sequence

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$2^\omega$ : Cantor space of infinite binary sequences

$\mu$ : uniform Bernoulli measure on  $2^\omega$

$T$ : left shift,  $T(x_0x_1x_2\dots) = x_1x_2x_3\dots$

$T$  preserves  $\mu$ .

For each  $A \subseteq 2^\omega$ , frequency of  $T^n(x) \in A$  is  $\mu(A)$  a.e.

**Theorem 1** (Kučera 1985, Bienvenu et al.). *A point  $x \in 2^\omega$  is Martin-Löf random iff  $(\exists n)T^n(x) \notin A$  for all c.e. open set  $A$  with  $\mu(A) < 1$ .*

Moreover there exists universal c.e. open set  $A$ .

# Questions?

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- New interpretation via three approaches
- Another relation between a property in ergodic theory and a random concept
- Random deficiency
- Other operator  $T$
- General space

# Computability

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A set  $A \subseteq \mathbb{N}$  is *computable* iff  $1_A : \mathbb{N} \rightarrow \{0, 1\}$  is computable.

A set  $A$  is *c.e.* iff  $A = \text{dom}(f)$  for a partial computable function.

A real  $r \in [0, 1]$  is *computable* iff its binary expansion is computable.

A real  $r$  is *c.e.* iff  $\{q \in \mathbb{Q} : q < r\}$  is c.e.

There exist a c.e. set and a c.e. real that are not computable.

base: a set of  $[\sigma] = \{A : \sigma \preceq A\}$

A open set  $U$  is *c.e.* iff  $U = \bigcup_{\sigma \in S} [\sigma]$  for a c.e. set  $S$ .

$\mu$ : generated from  $\mu([\sigma]) = 2^{-|\sigma|}$

We often identify  $2^\omega$  with  $[0, 1]$ .

# Martin-Löf randomness

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Typicalness: No effective null set contains the point.

**Definition 2** (Martin-Löf 1966).  $\{U_n\}$ : a seq. of open sets  
 $\{U_n\}$  is a test iff it is uniformly c.e. and  $\mu(U_n) \leq 2^{-n}$ .

A real  $A$  passes the test iff  $A \notin \bigcap_n U_n$ .

$A$  is ML-random iff  $A$  passes all tests.

There exists a universal test.

The set of ML-random reals has measure 1.

# Complexity

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Incompressibility: The prefix is hard to compress or describe.

**Definition 3.** A set  $A \subseteq 2^*$  is prefix-free iff no string is a prefix of another.

A function  $f : 2^* \rightarrow 2^*$  is prefix-free iff its domain is prefix-free.

A machine is a partial computable function.

The Kolmogorov complexity of  $\sigma$  is

$$K(\sigma) = \min\{|\tau| : V(\tau) = \sigma\}.$$

where  $V$  is a universal prefix-free machine.

**Proposition 4 (Levin, Schnorr).**  $A$  is ML-random iff  $K(A \upharpoonright n) > n - O(1)$ .



# Martingale

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Unpredictability: It is hard to predict the next bit.

**Definition 5.** A function  $d : 2^* \rightarrow \mathbb{R}^+ \cup \{0\}$  is a martingale iff

$$d(\sigma) = \frac{d(\sigma 0) + d(\sigma 1)}{2}.$$

**Proposition 6 (Schnorr 1971).**  $A$  is ML-random iff  $\sup_n d(A \upharpoonright n) < \infty$  for all c.e. martingales.

There exists a universal c.e. martingale.

# van Lambalgen's Theorem

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Two parts of random sequence is relatively random each other.

Let  $A \oplus B = A(0)B(0)A(1)B(1) \dots$ .

**Theorem 7** (van Lambalgen's Theorem 1987).  $A \oplus B$  is *ML-random* iff  $A$  is *ML-random* and  $B$  is *A-ML-random*.

This theorem holds for many randomnesses and is a criterion for a proper randomness.

# An extension

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**Theorem 8** (M. 2010).  $A_i$ : ML-random relative to  $\bigoplus_{j<i} A_j$   
 $\exists B_i$  s.t.  $B_i =^* A_i$  and  $\bigoplus B_i$  is ML-random.

**Theorem 9** (Bienvenu et al.). In the above theorem we can replace  $B_i =^* A_i$  with deleting finite prefixes or adding finite strings.

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**Random closed set**

- ❖ s-gale
- ❖ Effective Hausdorff dimension
- ❖ Weakly s-ML-randomness
- ❖ Other dimensions
- ❖ Random closed set

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# Random closed set

# *s-gale*

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$d$ : a martingale

$h$ : an order (unbounded non-decreasing)

**Definition 10** ( $h$ -success set of  $d$ ; Schnorr 1971).

$$S_h[d] = \left\{ A : \limsup_n \frac{d(A \upharpoonright n)}{h(n)} = \infty \right\}.$$

**Definition 11** ( $s$ -gale: Lutz 2000).

$$f(\sigma) = 2^{-s}(f(\sigma 0) + f(\sigma 1)).$$

**Theorem 12** (Lutz 2000).  $X \subseteq 2^\omega$

$$\begin{aligned} \dim_H(X) &= \inf \{ s : (\exists f) X \subseteq S[f] \} \\ &= \inf \{ s : (\exists d) X \subseteq S_{2^{(1-s)n}}[d] \} \end{aligned}$$

# Effective Hausdorff dimension

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**Definition 13** (Effective Hausdorff dimension; Lutz 2000).  
*d*: a c.e. martingale

$$\dim(X) = \inf\{s : (\exists d) X \subseteq S_{2^{(1-s)n}}[d]\}$$

We also have a direct characterization with a cover.

**Theorem 14** (Mayordomo 2002). For  $A \in 2^\omega$ ,

$$\dim(X) = \liminf_n \frac{K(A \upharpoonright n)}{n}.$$

# Weakly $s$ -ML-randomness

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**Definition 15** (Tadaki 2002).  $\{U_n\}$  is a test for weak  $s$ -ML-randomness iff they are uniformly c.e. sets of strings with  $\sum_{\sigma \in V_n} 2^{-s|\sigma|} \leq 2^{-n}$ .  $A$  is weakly  $s$ -ML-random iff  $A \notin \bigcap_n [V_n]$ .

$A$  is weakly  $s$ -random if  $K(A \upharpoonright n) \geq sn - O(1)$ .

**Theorem 16** (Tadaki 2002).  $A$  is weakly  $s$ -ML-random iff  $A$  is weakly  $s$ -random.

**Proposition 17.**

$$\dim(A) = \sup\{s : A \text{ is weakly } s\text{-random}\}.$$

A characterization by martingales is unknown.

# Other dimensions

Some researchers also study the relation with the following dimension.

- box counting dimension
- packing dimension

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# Random closed set

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- Barmpalias et al. (2007) introduced it.
- Kjos-Hanssen (2009) uses Galton-Watson trees to obtain a similar notion.
- Axon uses a notion of “random closed set” in probability theory.

Diamondstone et al. (2009) proved the equivalence of the first two.

Axon proved the equivalence on  $2^\omega$ .

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**Generalization**

- ❖ Type-2 Theory of Effectivity
- ❖ Representation
- ❖ CTS
- ❖ Representations
- ❖ ML-randomness over a CTS
- ❖ Randomness over a CMS
- ❖ Computability of measures
- ❖ By martingales
- ❖ Complexity randomness

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# Generalization

# Type-2 Theory of Effectivity

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Let  $Y_0, \dots, Y_n \in \{\Sigma^*, \Sigma^\omega\}$  and  $Y = Y_1 \times \dots \times Y_n$ .

A function  $f : \subseteq Y \rightarrow Y_0$  is *computable* if it is computed by Type-2 machine.

Informally, a *Type-2 machine* is a Turing machine, which reads from input tapes with finite or infinite inscription, operates on work tapes and write one-way to an output tape.

On  $\Sigma^*$  we consider the discrete topology.

On  $\Sigma^\omega$  we consider the topology generated by the base  $\{w\Sigma^\omega : w \in \Sigma^*\}$  of open sets.

Every computable function is continuous.

# Representation

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A *representation* of a set  $M$  is a surjective function  $\gamma : \subseteq Y \rightarrow M$  where  $Y \in \{\Sigma^*, \Sigma^\omega\}$ .

*Example 18.* (i)  $\nu_{\mathbb{Q}} : \subseteq \Sigma^* \rightarrow \mathbb{Q}$

(ii)  $\rho : \subseteq \Sigma^\omega \rightarrow \mathbb{R}$

A point is  $\gamma$ -*computable* if it has a computable representation by  $\gamma$ .

A function  $f : \subseteq M_1 \rightarrow M_2$  is  $(\gamma_1, \gamma_2)$ -*computable* if it has a computable realization.

$$\begin{array}{ccc} M_1 & \xrightarrow{f} & M_2 \\ \uparrow \gamma_1 & & \gamma_2 \uparrow \\ Y_1 & \xrightarrow{\text{comp}} & Y_2 \end{array}$$

**Definition 19** (Hertling and Weihrauch 2009). *A computable topological space is a 4-tuple  $\mathbf{X} = (X, \tau, \beta, \nu)$  such that*

- $(X, \tau)$  is a topological  $T_0$ -space,
- $\nu : \subseteq \Sigma^* \rightarrow \beta$  is a notation of a base  $\beta$  of  $\tau$ ,
- $\text{dom}(\nu)$  is recursive and
- $\nu(u) \cap \nu(v) = \bigcup \{ \nu(w) : (u, v, w) \in S \}$  for all  $u, v \in \text{dom}(\nu)$  for some r.e. set  $S \subseteq (\text{dom}(\nu))^3$ .

# Representations

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**Definition 20.** Let  $\mathbf{X} = (X, \tau, \beta, \nu)$  be a computable topological space.

Define a representation  $\delta : \subseteq \Sigma^\omega \rightarrow X$  of the points as

$$x = \delta(p) \iff (\forall w \in \Sigma^*)(w \ll p \iff x \in \nu(w))$$

and a representation  $\theta : \subseteq \Sigma^\omega \rightarrow \tau$  of the set of open sets as

$$W = \theta(p) \iff \begin{cases} w \ll p \Rightarrow w \in \text{dom}(\nu) \\ W = \bigcup \{ \nu(w) : w \ll p \} \end{cases} .$$

# ML-randomness over a CTS

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**Definition 21** (Hertling and Weihrauch 2003).  $\{U_n\}$ : open sets

$\{U_n\}$  is a test iff uniformly  $\theta$ -comp. with  $\mu(U_n) \leq 2^{-n}$ .  
 $x$  is measure  $\mu$ -random iff  $x \notin \bigcap_n U_n$  for each test.

Universality needs an unnatural condition.

# Randomness over a CMS

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Randomness over a computable metric space gets much attention.

- Gács (2005) first gave a restricted characterization by complexity.
- Hoyrup and Rojas (2009) gave a complete characterization by complexity.

The most important point was computability of measures.



# Computability of measures

$M(X)$ : the space of bounded non-negative Borel measures

**Proposition 22.**  $G$ : the finite union of base sets  
A countable subbase of the  $A$ -topology  $\tau_A$ :

$$\{\mu : \mu(G) > q\}, \{\mu : \mu(X) < q\}$$

**Theorem 23.**  $(M(X), \tau_A, \beta_A, \nu_A)$  is a CTS.

**Definition 24.** A measure is computable if it is a computable point in the CTS.

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# By martingales

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Let  $(X, \mathcal{A}, \mu)$  be a measure space.

A *filtration* is a sequence of sub- $\sigma$ -algebra  $(\mathcal{A}_n)$  such that

$\mathcal{A}_n \subseteq \mathcal{A}_{n+1}$  for each  $n$ .

A sequence of  $\mathcal{A}$ -measurable functions  $(f_n, \mathcal{A}_n)$  is called a *supermartingale* if  $\int f_n d\mu < \infty$  and  $\int_A f_n d\mu \geq \int_A f_{n+1} d\mu$  for all  $A \in \mathcal{A}_n$ .

**Theorem 25.** A point  $x$  is measure  $\mu$ -random iff  $\sup_n f_n(x) < \infty$  for each  $([\nu_{\mathbb{N}}, \delta], \bar{\rho}_{<})$ -computable supermartingale  $(f_n, \mathcal{A}_n)$

*Proof idea.* Let  $U_{k,m} = \{y : \sup_{n \leq m} f_n(y) > 2^k\}$  and use Doob's maximal inequality. □

# Complexity randomness

Suppose that the base is complete.

**Definition 26.** *A point  $x$  is complexity  $\mu$ -random iff*

$$x \in \xi(u) \Rightarrow K(u) \geq -\log \mu\xi(u) - O(1)$$

*where  $\xi(u) = \nu(u)^c$ .*

**Theorem 27.** *Suppose that the space has almost decidability and the measure is effectively regular. Then two randomnesses coincide and van Lambalgen's Theorem holds for it.*

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# Randomness and Ergodic Theory

# Effective Birkhoff's ergodic theorem

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**Definition 28.**  $T$  is  $\ln^2$ -ergodic if  $\forall f, g$

$$\gamma_n(f, g) = \left| \frac{1}{n} \sum_{i < n} \int f \circ T^i \cdot g d\mu - \int f d\mu \int g d\mu \right| \leq \frac{c_{f,g}}{(\ln(n))^2}$$

**Theorem 29** (Birkhoff's ergodic theorem for random points; V'yugin '97, Nandakumar '08, Hoyrup and Rojas '09).

$(X, \mu)$ : a CPS

$T : X \rightarrow X$  an effectively measurable measure-preserving map

$f \in L^1(X, \mu)$  an effectively measurable function

$T$ :  $\ln^2$ -ergodic

Then  $\exists n(c, \epsilon)$  comp. function s.t.

if  $x \in K_c$  then for all  $n > n(c, \epsilon)$ ,

$$\left| \frac{1}{n} \sum_{i < n} \int f \circ T^i \cdot g d\mu - \int f d\mu \int g d\mu \right| < \epsilon$$

The effectiveness of the convergence speed can be also seen in Davie('01).

# Other results

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- Under a appropriate condition computable typical points are dense.
- Not all computable dynamical systems have a computable invariant measure.
- The complexity of the orbits of random points equals the Kolmogorov-Sinaï entropy of the system,

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**Probability**

- ❖ What is probability?
- ❖ Von Mises's philosophy
- ❖ Collective
- ❖ Sigma-algebra
- ❖ Probability axioms
- ❖ Random sequence
- ❖ Practically impossible
- ❖ End

# Probability

# What is probability?

- Frequency probability
- Subjective probability
- Propensity probability

Modern mathematical formalization is by Kolmogorov (1933).

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# Von Mises's philosophy

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Von Mises took a very different approach.

- His first paper on this subject is in 1919.
- His standpoint is empiricism.
- He claims that probability can be defined only for repetitive events.
- First the collective, then the probability (as its frequency).

# Collective

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A collective is a set of sequences s.t.

- their limits of relative frequency exist and are the same
- relative frequency are not affected by appropriate place selections

Ville (1940) constructed a sequence for which LIL does not hold.

# ***Sigma-algebra***

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$\Sigma_0$  is an *algebra* on  $S$  iff

(i)  $S \in \Sigma_0,$

(ii)  $F \in \Sigma_0 \Rightarrow F^c = S \setminus F \in \Sigma_0,$

(iii)  $F, G \in \Sigma_0 \Rightarrow F \cup G \in \Sigma_0.$

$\Sigma$  is a  $\sigma$ -*algebra* iff it is an algebra and

$$F_n \in \Sigma \Rightarrow \bigcup_n F_n \in \Sigma$$

# Probability axioms

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$P$  is a *finite probability function* on  $\Sigma_0$  iff

(i)  $P(A) \geq 0$  for  $A \in \Sigma_0$ ,

(ii)  $P(S) = 1$ ,

(iii) (finite additivity)

$$A \cap B = \phi \Rightarrow P(A \cup B) = P(A) + P(B).$$

$P$  is a *probability function* on  $\Sigma$  iff it satisfies above and (countable additivity)  $P(\bigcup F_n) = \sum_n P(F_n)$  for pairwise disjoint  $F_n \in \Sigma$ .

Note that  $P$  need not to be a measure.

# Random sequence

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Generalization

Randomness and Ergodic Theory

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❖ What is probability?

❖ Von Mises's philosophy

❖ Collective

❖ Sigma-algebra

❖ Probability axioms

❖ Random sequence

❖ Practically impossible

❖ End

Now we can consider a CTS  $X^{\mathbb{N}}$ .

We define frequency as

$$F(A) = \lim_n \frac{\#\{i \leq n : x_i \in A\}}{n}.$$

**Theorem 30.** *Suppose that the base is uniformly almost decidable.*

*$\mathcal{A}$ : the minimal algebra containing the base*

$$F|_{\mathcal{A}} = \mu|_{\mathcal{A}}$$

We can not extend to Borel sets.

# Practically impossible

$E \in \mathcal{A}$ : an observable set

$$\mu(E) = 0 \Rightarrow E \cap \text{cRND}(\mu) = \emptyset$$

$$\mu(E) = 1 \Rightarrow \text{cRND}(\mu) \subseteq E$$

We need restricted countable additivity.

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Thank you!