

Randomness – Beyond Lebesgue Measure

Jan Reimann

Department of Mathematics
University of California, Berkeley

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Measures on Cantor Space

Outer measures from premeasures

Approximate sets from outside by open sets and weigh with a general measure function.

- ▶ A **premeasure** is a function $\rho : 2^{<\omega} \rightarrow \mathbb{R}_0^+ \cup \{\infty\}$.
- ▶ One can obtain an **outer measure** μ_ρ from ρ by letting

$$\mu_\rho(X) = \inf_{C \subseteq 2^{<\omega}} \left\{ \sum_{\sigma \in C} \rho(\sigma) : \bigcup_{\sigma \in C} U_\sigma \supseteq X \right\},$$

where U_σ is the **basic open set** induced by σ .
(Set $\mu_\rho(\emptyset) = 0$.)

The resulting $\mu = \mu_\rho$ is a countably subadditive, monotone set function, an **outer measure**.

Measures on Cantor Space

Types of measures

Probability measures: based on a premeasure ρ which satisfies

- ▶ $\rho(\emptyset) = 1$ and
- ▶ $\rho(\sigma) = \rho(\sigma \frown 0) + \rho(\sigma \frown 1)$.

Hausdorff measures: based on a premeasure ρ which satisfies

- ▶ If $|\sigma| = |\tau|$, then $\rho(\sigma) = \rho(\tau)$.
- ▶ $\rho(n)$ is nonincreasing.
- ▶ $\rho(n) \rightarrow 0$ as $n \rightarrow \infty$.
- ▶ For example: $\rho(\sigma) = 2^{-|\sigma|^s}$, $s \geq 0$.

Measures on Cantor Space

Nullsets

The way we constructed outer measures, $\mu(A) = 0$ is equivalent to the existence of a sequence $(W_n)_{n \in \omega}$, $W_n \subseteq 2^{<\omega}$, such that for all n ,

$$A \subseteq \bigcup_{\sigma \in W_n} U_\sigma \quad \text{and} \quad \sum_{\sigma \in W_n} \rho(\sigma) \leq 2^{-n}.$$

Thus,

every nullset is contained in a G_δ nullset.

Randomness

Effective G_δ sets

By requiring that the covering nullset is **effectively G_δ** , we obtain a notion of **effective nullsets**.

Definition

- ▶ A **test relative to $z \in 2^\omega$** is a set $W \subseteq \mathbb{N} \times 2^{<\omega}$ which is c.e. in z .
- ▶ Given a natural number $n \geq 1$, an **n -test** is a test which is c.e. in $\emptyset^{(n-1)}$.
- ▶ A real x **passes** a test W if $x \notin \bigcap_n U(W_n)$, where $W_n = \{\sigma : (n, \sigma) \in W\}$.

Hence a real passes a test W if it is not in the G_δ -set represented by W .

Randomness

Martin-Löf tests

To test for randomness, we want to ensure that W actually describes a nullset.

Definition

Suppose μ is a measure on 2^ω . A test W is **correct for μ** if for all n ,

$$\sum_{\sigma \in W_n} \mu(U_\sigma) \leq 2^{-n}.$$

Any test which is correct for μ will be called a **test for μ** .

Accordant definitions for **n -tests**, **arithmetical tests** are straightforward.

Randomness

Representation of measures

An effective test for randomness should have access to the measure it is testing for.

- ▶ Therefore, represent it by an infinite binary sequence.
- ▶ As a measure on 2^ω is completely determined by its values on the cylinder sets (i.e. by the underlying premeasure ρ), it seems reasonable to represent these values via approximation by rational intervals.

Definition

Given a premeasure ρ , define its rational representation r_ρ by letting, for all $\sigma \in 2^{<\omega}$, $q_1, q_2 \in \mathbb{Q}$,

$$\langle \sigma, q_1, q_2 \rangle \in r_\rho \Leftrightarrow q_1 < \rho(\sigma) < q_2.$$

Randomness

Representation of measures

The condition $q_1 < \rho(\sigma) < q_2$ induces a **subbasis for the weak topology** on the space of probability measures.

- ▶ More general, if a space X is Polish, so is the space $M(X)$ of all probability measures on X (under the weak topology). Also, if X is compact metrizable, so is $M(X)$ (**Prokhorov metric**).

This yields various ways to represent a measure: Cauchy sequences, list of basic open balls it is contained in, etc.

Randomness

Tests for Arbitrary Measures

Definition

Suppose ρ is a premeasure on 2^ω and $z \in 2^\omega$. A real is μ_ρ - z -random if it passes all $r_\rho \oplus z$ -tests which are correct for μ_ρ .

Hence, a real x is random with respect to an arbitrary measure μ_ρ if and only if it passes all tests which are enumerable in the representation r_ρ of the underlying premeasure ρ .

The Initial Question

Question

What is the logical/computability theoretic structure of random reals?

Making Reals Random

Image Measures

Let μ be a probability measure and $f : 2^\omega \rightarrow 2^\omega$ be a continuous (Borel) function.

Define a new measure μ_f by setting

$$\mu_f(\sigma) = \mu(f^{-1}U_\sigma)$$

Observation: If $\mu\{x\} = 0$ for all $x \in 2^\omega$, then, for

$$F(x) = \mu\{y : y <_{\mathbb{R}} x\},$$

it holds that $\mu_f = \lambda$.

Making Reals Random

Randomness Conservation

Idea

If the transformation f is computable in z , then it should preserve randomness, i.e. it should map a μ - z -random real to a μ_f - z -random one.

Making Reals Random

Computable measures

Note: If μ is a computable measure, then an atom of μ is μ -random iff it is computable.

Theorem (Levin, Kautz)

If a real is noncomputable and random with respect to a computable probability measure, then it is Turing equivalent to a λ -random real.

Non-Trivial Randomness

The atomic case

Note that every real is trivially random with respect to some μ if it is an atom of μ .

- ▶ We are interested in the case when a real is non-trivially random.

Theorem (Reimann and Slaman)

For any real x , the following are equivalent.

- (i) *There exists (a representation of) a measure μ such that $\mu(\{x\}) = 0$ and x is 1-random for μ .*
- (ii) *x is not computable.*

Non-Trivial Randomness

The atomic case

Features of the proof:

- ▶ Conservation of randomness:
- ▶ Randomness of cones:
 - ▶ Kucera's coding argument shows that every degree above \emptyset' contains a λ -random.
 - ▶ Relativize this using the Posner-Robinson Theorem.
 - ▶ Conclude that every non-recursive real x is Turing equivalent to some λ -z-random real for some real z .
- ▶ A basis theorem for relative randomness.

Randomness for Continuous Measures

In the proof there is no control over the measure that makes x random.

- ▶ Atoms cannot be avoided.
- ▶ Uses a special (though natural) representation of $M(2^\omega)$ as a particular Π_1^0 class.

Question

What if one admits only **continuous probability measures**?

Randomness for Continuous Measures

Characterizing randomness for continuous measures

Theorem (Reimann and Slaman)

Let x be a real. For any $z \in 2^\omega$, the following are equivalent.

- (i) x is truth-table equivalent to a λ - z -random real.*
- (ii) x is random for a continuous (dyadic) measure recursive in z .*
- (iii) There exists a functional Φ recursive in z which is an order-preserving homeomorphism of 2^ω such that $\Phi(x)$ is λ - z -random.*

This is an effective version of the [classical isomorphism theorem](#) for continuous probability measures.

The Class NCR

Question

Which level of logical complexity guarantees continuous randomness?

Let NCR_n be the set of all reals which are not n -random relative to any continuous measure.

- ▶ **Kjos-Hanssen and Montalban:** Every member of a countable Π_1^0 class is contained in NCR_1 . (It follows that elements of NCR_1 can be found at arbitrary high levels of the hyperarithmetical hierarchy.)
- ▶ **Reimann and Slaman:** $\text{NCR}_1 \subseteq \Delta_1^1$ (by arguments tailored for $n = 1$).

Example of higher order: Kleene's \mathcal{O} cannot be 3-random with respect to a continuous measure.

Upper Bounds for Continuous Randomness

In general, can we give a distinct bound on NCR_n like in the case $n = 1$?

- ▶ There is some evidence that NCR_n grows very quickly with n .
- ▶ Can we give an upper bound?

Theorem (Reimann and Slaman)

For all n , NCR_n is countable.

NCR_n is Countable

Main Features of the Proof

- ▶ Show that the complement of NCR_n contains an upper Turing cone.
 - ▶ Show that the complement of NCR_n contains a Turing invariant and cofinal Borel set. We can use the set of all y that are Turing equivalent to some $z \oplus R$, where R is $(n+1)$ -random relative to z .
 - ▶ Use [Martin's result on Borel Turing sets](#) to infer that the complement of NCR_n contains a cone.
- ▶ Go on to show that the elements of NCR_n are definable at a rather low level of the constructible universe.
 - ▶ $\text{NCR}_n \subseteq L_{\beta_n}$, where β_n is the least ordinal such that $L_{\beta_n} \models \text{ZFC}^- + \text{there exist } n \text{ many iterates of the power set of } \omega$, where ZFC^- is Zermelo-Fraenkel set theory without the Power Set Axiom.

NCR_n is Countable

Is the metamathematics necessary?

Question

Do we need to use metamathematical methods to prove the countability of NCR_n ?

We make **fundamental use of Borel determinacy**; this suggests to analyze the metamathematics in this context.

Borel Determinacy and Iterates of the Power Set

Friedman's result

The necessity of iterates of the power set is known from a result by Friedman.

- ▶ Martin's proof of Borel determinacy starts with a description of a Borel game and produces a winning strategy for one of the players.
- ▶ The more complicated the game is in the Borel hierarchy, the more iterates of the power set of the continuum are used in producing the strategy.

Theorem (Friedman)

$ZFC^- \not\vdash$ All Σ_5^0 -games on countable trees are determined.

Martin later improved this to Σ_4^0 .

Borel Determinacy and Iterates of the Power Set

Friedman's result

Inductively one can infer from Friedman's result that in order to prove full Borel determinacy, a result about sets of reals, one needs infinitely many iterates of the power set of the continuum.

- ▶ The proof works by showing that there is a model of ZFC^- for which Σ_4^0 -determinacy does not hold.
- ▶ This model is L_{β_0} .

NCR and Iterates of the Power Set

We can work along similar lines to obtain a result concerning the countability of NCR_n .

Theorem

For every k , the statement

For every n , NCR_n is countable.

cannot be proven in

$\text{ZFC}^- + \text{there exists } k \text{ many iterates of the power set of } \omega.$

- ▶ The proof (for $k = 0$) shows that there is an n such that NCR_n is cofinal in the Turing degrees of L_{β_0} . Hence, NCR_n is not countable in L_{β_0} .
- ▶ The witnesses for NCR_n are **master codes of models** L_α for certain $\alpha < \beta_0$.

Hausdorff Measures

Question

What is the computability theoretic relation between randomness for Hausdorff measures and randomness for (continuous) probability measures?

- ▶ We denote Hausdorff premeasures by h , and the corresponding measure by \mathcal{H}^h .

Hausdorff Measures

Hausdorff measures and algorithmic entropy

Hausdorff randomness can be interpreted as a **degree of incompressibility**.

Theorem

Let h be a computable Hausdorff premeasure. A real x is \mathcal{H}^h -random if and only if there exists a constant c such that for all n ,

$$K(x \upharpoonright_n) \geq -\log h(n) - c.$$

- ▶ K denotes the **prefix-free Kolmogorov complexity**.

Hausdorff Measures

Hausdorff measures and computable measures

Theorem

For every computable dimension function h there is a real x such that x is \mathcal{H}^h -random but not random with respect to any computable measure.

- ▶ **Proof:** Join a 1-generic and a λ -random real with appropriate density.

Hausdorff Measures and Randomness

Non-extractibility results

Theorem (Kjos-Hanssen, Merkle, and Stephan; R. and Slaman)

There exists computable, unbounded, nondecreasing function h and a real x such that for all n ,

$$K(x \upharpoonright_n) \geq h(n) \quad (*)$$

and x does not compute a Martin-Löf random real.

How close to $h(n) = n$ can h be? (The Dimension Problem)

Hausdorff Measures and Randomness

Non-extractibility results – strong reducibilities

Theorem (Reimann and Nies)

For each rational r , $0 \leq r \leq 1$, there is a real $x \leq_{\text{wtt}} \emptyset'$ such that

$$\liminf_n \frac{K(x \upharpoonright_n)}{n} = r \quad \text{and} \quad (\forall z \leq_{\text{wtt}} x) \liminf_n \frac{K(z \upharpoonright_n)}{n} \leq r.$$

Probability Measures and Hausdorff Measures

Theorem

For every rational s and every real x such that x is \mathcal{H}^s -random, there exists a probability measure μ such that x is μ -random and for some $c > 0$

$$\mu(U_\sigma) \leq c 2^{-|\sigma|^s} \quad \text{for all } \sigma \in 2^\omega. \quad (*)$$

- ▶ Here \mathcal{H}^s denotes the Hausdorff measure based on the premeasure $2^{-|\sigma|^s}$.

Probability Measures and Hausdorff Measures

Pulling back measure effectively

Assume that Φ and Ψ are Turing functionals such that

$$\Psi(x) = R \quad \text{and} \quad \Phi(R) = x,$$

where R is λ -random.

Let $\text{Pre}(\sigma) := \{\tau \in 2^\omega : \Phi(\tau) \supseteq \sigma \ \& \ \Psi(\sigma) \sqsubseteq \tau\}$.

To define a measure μ with respect to which x is random, we satisfy two requirements:

- (1) $\text{Pre}(\sigma) := \{\tau \in 2^\omega : \Phi(\tau) \supseteq \sigma \ \& \ \Psi(\sigma) \sqsubseteq \tau\}$.
- (2) $\lambda(U_{\text{Pre}(\sigma)}) \leq \mu(U_\sigma) \leq \lambda(U_{\Psi(\sigma)})$

The configurations given by (2) induce a Π_1^0 class P of probability measures (with respect to a certain Cauchy representation).

Probability Measures and Hausdorff Measures

A basis theorem for relative randomness

We want to show that for some $\mu \in P$, x is μ -random.

Note that if (V_n) were a μ -test covering x , then $\Phi^{-1}(V_n)$ would be a λ - r_μ -test covering R .

So, what we need to show is that R is λ - r_μ -random for some $\mu \in M$.

Theorem (indep. by Downey, Hirschfeldt, Miller, and Nies)

If $B \subseteq 2^\omega$ is nonempty and Π_1^0 , then, for every R which is λ -random there is $z \in B$ such that R is λ - z -random.

The proof is essentially a **compactness argument**.

Probability Measures and Hausdorff Measures

Constructing measures

Construct an effectively closed set of measures.

- ▶ Work along a computable tree $T \subseteq 2^{<\omega}$ whose infinite paths are \mathcal{H}^s -random.
- ▶ Define sequence of uniformly computable measures $\{\mu^n\}$. Each μ^n can be seen as an approximation to a final μ , knowing only the paths of T up to length n .
- ▶ For $n \in \mathbb{N}$, define μ_n^n such that for all $\sigma \in 2^n$

$$\mu_n^n \upharpoonright U_\sigma = \begin{cases} 2^{(1-s)n} \lambda \upharpoonright U_\sigma, & \text{if } \sigma \in T, \\ 0, & \text{if } \sigma \notin T. \end{cases}$$

Probability Measures and Hausdorff Measures

Constructing measures

- ▶ Modify μ_n^n downward to ensure (*) at all levels $\leq n$. Define μ_{n-1}^n by requiring that

$$\mu_{n-k-1}^n \upharpoonright_{U_\sigma} = \gamma(\sigma) \mu_{n-k}^n \upharpoonright_{U_\sigma}$$

where $\gamma(\sigma) = \min\{1, 2^{-(n-k-1)s} (\mu_{n-k}^n U_\sigma)^{-1}\}$.

- ▶ We stop as soon as $[T] \subseteq U_\sigma$ for some $\sigma \in 2^{k_0}$ and define $\mu^n = \mu_{k_0}^n$. k_0 can be determined effectively and μ^n is a computable measure, as the μ_m^n are.
- ▶ Finally, note that every real is computable in a λ -random real. (Gacs, Kucera)
- ▶ Combine the Gacs-reduction with the basis theorem to obtain the desired continuous probability measure.