

# The Metamathematics of Randomness

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January 26, 2007

# (Original) Motivation

Effective extraction of randomness

In my PhD-thesis I studied the computational power of reals effectively random for Hausdorff measures.

## The Dimension Problem

Can we efficiently extract uniform randomness from such reals?

Examples:

- ▶ Von Neumann's trick
- ▶ randomness extractors in computational complexity
- ▶ factors of dynamical systems (Sinai, Ornstein)

# Motivation

## The basic paradigm

This lead eventually to another question:

Which reals are random with respect to  
a (continuous) probability measure?

The answers to this question took an unexpected turn.

# 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**  $\mu_\rho$  from  $\rho$  by letting

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

where  $N_\sigma$  is the **basic open set** induced by  $\sigma$ .  
(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  $\rho$  which satisfies

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

For probability measures it holds that  $\mu_\rho(N_\sigma) = \rho(\sigma)$ .

**Hausdorff measures:** based on a premeasure  $\rho$  which satisfies

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

The actual definition of Hausdorff measures is more complicated, but we are **only interested in nullsets**.

# 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  $n$ ,

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

Thus,

every nullset is contained in a  $G_\delta$  nullset.

# Randomness for Outer Measures

## Effective $G_\delta$ sets

By requiring that the covering nullset is **effectively**  $G_\delta$ , 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  $z$ .
- ▶ A real  $x$  **passes** a test  $W$  if  $x \notin \bigcap_n N(W_n)$ , where  $W_n = \{\sigma : (n, \sigma) \in W\}$ .

Hence a real passes a test  $W$  if it is not in the  $G_\delta$ -set represented by  $W$ .

# Randomness for Outer Measures

## Martin-Löf tests

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

### Definition

Suppose  $\mu$  is a measure on  $2^\omega$ . A test  $W$  is **correct for  $\mu$**  if for all  $n$ ,

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

Any test which is correct for  $\mu$  will be called a **test for  $\mu$** .



# Randomness for Outer Measures

## 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.
- ▶ Outer measures are determined by the underlying premeasure  $\rho$ . It seems reasonable to represent these values via approximation by rational intervals.

### Definition

Given a premeasure  $\rho$ , define its rational representation  $r_\rho$  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  $\mathcal{P}(X)$  of all probability measures on  $X$  (under the weak topology). Also, if  $X$  is compact metrizable, so is  $\mathcal{P}(X)$ .
- ▶ This yields various ways to represent a measure: Cauchy sequences, list of basic open balls it is contained in, etc.
- ▶ We can obtain a nice effective representation (e.g. by following the framework in **Moschovakis'** book).

## Theorem

*There is a recursive surjection  $\pi: 2^\omega \rightarrow \mathcal{P}(2^\omega)$  and a  $\Pi_1^0$  subset  $P$  of  $2^\omega$  such that  $\pi \upharpoonright_P$  is one-to-one and  $\pi(P) = \mathcal{P}(2^\omega)$ .*

# Randomness for Outer Measures

## Tests for Arbitrary Measures

### Definition

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

Hence, a real  $x$  is random with respect to an arbitrary measure  $\mu_\rho$  if and only if it passes all tests which are enumerable in the representation  $r_\rho$  of the underlying premeasure  $\rho$ .

- ▶  $n$ -randomness: tests r.e. in  $\emptyset^{(n-1)}$ .
- ▶ Accordingly, arithmetical randomness.

# Making Reals Random

## Image measures

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

Define the **image measure**  $\mu_f$  by setting

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

## Conservation of randomness

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

# Making Reals Random

## Computable measures

Trivially, every **atom** of a measure is random with respect to it.

- ▶ For **recursive reals**, this is the only way to become random.

If  $\mu$  is a computable measure, then an atom of  $\mu$  is  $\mu$ -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  $\lambda$ -random real.*

- ▶ The proof works by showing that the distribution function can be computed effectively.

# Hausdorff Randomness

## The dimension problem

Let  $h : \mathbb{N} \rightarrow \mathbb{R}^{\geq 0}$  be a computable, nondecreasing, unbounded function (effective dimension function).

- ▶ Interesting connection with Kolmogorov complexity: A real  $x$  is Hausdorff  $2^{-h}$ -random if and only if for some  $c$ ,

$$K(x \upharpoonright_n) \geq h(n) - c \quad \text{for all } n.$$

## The dimension problem

If  $x$  is Hausdorff  $2^{-h}$ -random, does it compute a  $\lambda$ -random real?

# Non-trivial Randomness

## The basic question

Unfortunately, not every Hausdorff  $2^{-h}$ -random real is random for some computable probability measure (for arbitrary  $h$ ).

- ▶ Join a  $\lambda$ -random and a  $1$ -generic with appropriate density.

### Question

Are such reals at least non-trivially random with respect to some measure? What reals are in general?

# Non-trivial Randomness

Too coarse

It turns out every non-recursive real is random.

## Theorem (Reimann and Slaman)

*For any real  $x$ , the following are equivalent.*

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



# Non-trivial Randomness

## Making reals random

Features of the proof:

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

# Non-Trivial Randomness

## Making reals random

The Turing equivalence to a  $\lambda$ -random real translates into **effectively closed consistency conditions** for a probability measure.

- ▶ The following basis theorem (Downey, Hirschfeldt, Miller, and Nies; Reimann and Slaman) ensures that one of the measures will not affect the randomness of  $z$ .

### Theorem

*If  $B \subseteq 2^\omega$  is nonempty and  $\Pi_1^0$ , then, for every  $y$  which is  $\lambda$ -random there is  $z \in B$  such that  $y$  is  $\lambda$ -random relative to  $z$ .*

- ▶ This argument seems to be applicable in more generality, proving **existence of measures**.

# 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  $\Pi_1^0$  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  $\lambda$ - $z$ -random real.
- (ii)  $x$  is random for a continuous (dyadic) measure recursive in  $z$ .
- (iii) There exists a functional  $\Phi$  recursive in  $z$  which is an order-preserving homeomorphism of  $2^\omega$  such that  $\Phi(x)$  is  $\lambda$ - $z$ -random.

This is an effective version of the **classical isomorphism theorem** for continuous probability measures.

# The Class $\text{NCR}$

## Question

Which level of logical complexity guarantees continuous randomness?

Let  $\text{NCR}_n$  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  $\Pi_1^0$  class is contained in  $\text{NCR}_1$ . (It follows that elements of  $\text{NCR}_1$  is cofinal in the hyperarithmetical Turing degrees.)
- ▶ **Woodin:** outside  $\Delta_1^1$  the Posner-Robinson theorem holds with  $\text{tt}$ -equivalence.
- ▶ Conclude that  $\text{NCR}_1 \subseteq \Delta_1^1$ .

# The Class $\text{NCR}$

## Examples of higher order

### Theorem

*Kleene's  $\mathcal{O}$  is an element of  $\text{NCR}_3$ .*

Based on this, one can use the theory of **jump operators** (Jockusch and Shore) to obtain a whole class of examples.

Proof:

- ▶ Tree representation  
 $\mathcal{O} = \{e : \text{the } e\text{th recursive tree } T_e \subseteq \omega^{<\omega} \text{ is well-founded}\}.$
- ▶ Suppose  $\mathcal{O}$  is 3-random for some  $\mu$ .
- ▶ We want to use **domination properties** of random reals.

# The Class NCR

## Examples of higher order

- ▶ **Well-known** (Kurtz and others): If  $X$  is  $n$ -random for  $\mu$ ,  $n > 1$ , then every function  $f \leq_T X$  is dominated by a function recursive in  $\mu'$ .
- ▶ Therefore,  $\mu'$  computes a uniform family  $\{g_e\}$  of functions dominating the leftmost infinite path of  $T_e$ .
- ▶ Infer: For every  $e$ , the following are equivalent.
  - (i)  $T_e$  is well-founded.
  - (ii) The subtree of  $T_e$  to the left of  $g_e$  is finite.
- ▶ The latter condition is  $\Pi_1^0(\mu')$ , hence  $\mathcal{O}$  is  $\Pi_2^0(\mu)$ .
- ▶ But this is impossible if  $\mathcal{O}$  is 3-random for  $\mu$ .

# The Class NCR

## The non-helpfulness lemma

The domination property of higher randomness implies that random reals are **not helpful** when adding them as oracles/parameters.

### Lemma

*Suppose that  $n \geq 2$ ,  $y \in 2^\omega$ , and  $R$  is  $\lambda$ - $n$ -random relative to  $\mu$ . If  $i < n$ ,  $y$  is recursive in  $(R \oplus \mu)$  and recursive in  $\mu^{(i)}$ , then  $y$  is recursive in  $\mu$ .*

**Corollary:** For all  $k$ ,  $\emptyset^{(k)}$  is not  $n$ -random relative to any  $\mu$ ,  $n \geq 2$ .

- ▶ Suppose  $\emptyset^{(k)}$  is  $n$ -random relative to  $\mu$ .
- ▶  $\emptyset'$  is recursively enumerable relative to  $\mu$  and recursive in the supposedly  $n$ -random  $\emptyset^{(k)}$ . Hence,  $\emptyset'$  is recursive in  $\mu$  and so  $\emptyset''$  is recursively enumerable relative to  $\mu$ .
- ▶ Use induction to conclude  $\emptyset^{(k)}$  is recursive in  $\mu$ , a contradiction.



# Upper Bounds for Continuous Randomness

In general, can we give a distinct bound on  $\text{NCR}_n$  like in the case  $n = 1$ ?

- ▶ There is some evidence that  $\text{NCR}_n$  grows very quickly with  $n$ .
- ▶ Can we give an upper bound?

Theorem (Reimann and Slaman)

*For all  $n$ ,  $\text{NCR}_n$  is countable.*

# $\text{NCR}_n$ is Countable

## Main Features of the Proof

Show that the complement of  $\text{NCR}_n$  contains an upper Turing cone.

- ▶ Show that the complement of  $\text{NCR}_n$  contains a Turing invariant and cofinal (in the Turing degrees) Borel set.
- ▶ We can use the set of all  $\mathbf{y}$  that are Turing equivalent to some  $\mathbf{z} \oplus \mathbf{R}$ , where  $\mathbf{R}$  is  $(n+1)$ -random relative to a given  $\mathbf{z}$ .
- ▶ These  $\mathbf{y}$  will be  $n$ -random relative to some continuous measure and are  $T$ -above  $\mathbf{z}$ .
- ▶ Use **Martin's result on Borel Turing sets** to infer that the complement of  $\text{NCR}_n$  contains a cone.
- ▶ The cone is given by the **Turing degree of a winning strategy** in the corresponding game.

# $\text{NCR}_n$ is Countable

## Main Features of the Proof

Go on to show that the elements of  $\text{NCR}_n$  show up at a rather low level of the constructible universe.

►  $\text{NCR}_n \subseteq L_{\beta_n}$ , where  $\beta_n$  is the least ordinal such that

$L_{\beta_n} \models \text{ZFC}^- +$  there exist  $n$  many iterates of the power set of  $\omega$ ,

where  $\text{ZFC}^-$  is Zermelo-Fraenkel set theory without the Power Set Axiom.

# $NCR_n$ is Countable

## Main Features of the Proof

Given  $x \notin L_{\beta_n}$ , construct a set  $G$  such that

- (i)  $L_{\beta_n}[G]$  is a model of  $ZFC_n^-$ .
- (ii) For all  $y \in L_{\beta_n}[G] \cap 2^\omega$ ,  $y \leq_T x \oplus G$ .

$G$  is constructed by **Kumabe-Slaman forcing**.

The existence of  $G$  allows to conclude:

- ▶ If  $x$  is not in  $L_{\beta_n}$ , it will belong to every cone with base in  $L_{\beta_n}[G]$ .
- ▶ In particular, it will belong to the cone given by Martin's argument (relativized to  $G$  – use absoluteness), i.e. the cone avoiding  $NCR_n$ .
- ▶ Hence  $x$  is random relative to  $G$  for some continuous  $\mu$ , hence in particular  $\mu$ -random.

# $\text{NCR}_n$ is Countable

Is the metamathematics necessary?

## Question

Do we need to use metamathematical methods to prove the countability of  $\text{NCR}_n$ ?

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  $\Sigma_5^0$ -games on countable trees are determined.

Martin improved this to  $\Sigma_4^0$ .

# 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  $\Sigma_4^0$ -determinacy does not hold.
- ▶ This model is  $L_{\beta_0}$ .

# NCR and Iterates of the Power Set

We can work along similar lines to obtain a result concerning the countability of  $\text{NCR}_n$ .

## Theorem

*For every  $k$ , the statement*

*For every  $n$ ,  $\text{NCR}_n$  is countable.*

*cannot be proven in*

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



# NCR and Iterates of the Power Set

## Features of the proof

The proof (for  $k = 0$ ) shows that there is an  $n$  such that  $\text{NCR}_n$  is cofinal in the Turing degrees of  $L_{\beta_0}$ . Hence,  $\text{NCR}_n$  is not countable in  $L_{\beta_0}$ .

- ▶ The witnesses for  $\text{NCR}_n$  are Jensen's master codes of models  $L_\alpha$  for limit ordinals  $\alpha < \beta_0$ .

We choose  $n$  large enough to capture recognition and comparison (of well-foundedness) of models they code.

# NCR and Iterates of the Power Set

## Features of the proof

Suppose some  $M_\lambda$ ,  $\lambda < \beta_0$ , were  $\mathfrak{n}$ -random relative to  $\mu$ .

- ▶ Let  $\mathfrak{M}$  be the sequence of possible master codes which are recursive in  $\mu$  (satisfying some arithmetical formula).
  - ▶ The well-founded part of  $\mathfrak{M}$  is of the form  $\mathfrak{M}_{<\gamma} = (M_\alpha : \alpha < \gamma)$  for some  $\gamma \leq \lambda$ .
  - ▶  $\mathfrak{M}_{<\gamma}$  is uniformly arithmetically definable from  $M_\lambda$  and hence from  $\mu$ .
- ▶  $M_\gamma$  is obtained by iterating uniformly arithmetically definable operations on  $\mathfrak{M}_{<\gamma}$ .
- ▶ The results at each step and  $M_\gamma$  itself are recursive in  $M_\lambda$ .
- ▶ The results at each step and  $M_\gamma$  itself are recursive in  $\mu$ , by the non-helpfulness lemma.
- ▶  $M_\gamma$  is in the well-founded part of  $\mathfrak{M}$ . Contradiction.