MEASURES AND THEIR RANDOM REALS

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ABSTRACT. We study the randomness properties of reals with respect to arbitrary probability measures on Cantor space. We show that every noncomputable real is non-trivially random with respect to some measure. The probability measures constructed in the proof may have atoms. If one rules out the existence of atoms, i.e. considers only continuous measures, it turns out that every non-hyperarithmetical real is random for a continuous measure. On the other hand, examples of reals not random for any continuous measure can be found throughout the hyperarithmetical Turing degrees.

1. Introduction

Over the past decade, the study of algorithmic randomness has produced an impressive number of results. The theory of Martin-Löf random reals, with all its ramifications (e.g. computable or Schnorr randomness, lowness and triviality) has found deep and significant applications in computability theory, many of which are covered in recent books by Downey and Hirschfeldt [5] and Nies [23]. Usually, the measure for which randomness is considered in these studies is the uniform (1/2, 1/2)-measure on Cantor space, which is measure theoretically isomorphic to Lebesgue measure on the unit interval.

However, one may ask what happens if one changes the underlying measure. It is easy to define a generalization of Martin-Löf tests which allows for a definition of randomness with respect to arbitrary computable measures. For arbitrary measures, the situation is more complicated. Martin-Löf [20] studied randomness for arbitrary Bernoulli measures, Levin [16, 17, 18] studied arbitrary measures on 2^{ω} , while Gács [7] generalized Levin's approach to a large class of computable metric spaces. Most recently, Hoyrup and Rojas [10] showed that Levin's approach can be extended to any computable metric space. It also turned out that much of the theory (i.e. existence of a universal test, the connection with descriptive complexity etc.) can be preserved under reasonable assumptions for the underlying space.

In this paper, we study the following question: Given a real $x \in 2^{\omega}$, with respect to which measures is x random? This can be seen as a dual to the usual investigations in algorithmic randomness: Given a (computable) measure μ (where μ usually is the uniform distribution), what are the properties of a μ -random real?

Of course, every real x is trivially random with respect to a measure μ which assigns some positive mass to it (as a singleton set), i.e. $\mu\{x\} > 0$. But one can ask if there is a measure for which this is not the case and x is still random. It turns out that this is possible precisely for the non-computable reals. Furthermore, one

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could ask whether there exists a measure μ such that x is μ -random and μ does not have atoms at all, i.e. $\mu\{y\} = 0$ for all y. The answer to this question shows an unexpected correspondence between the randomness properties of a real and its complexity in terms of effective descriptive set theory: If x is not Δ_1^1 , then there exists a non-atomic measure with respect to which x is random.

These results motivate a further investigation. What is the exact classification of all reals (inside Δ_1^1) which are not random with respect to any continuous measure? If we look at n-randomness ($n \geq 2$), what is the size and structure of the reals that are not n-random for some continuous measure? The latter question will be studied in separate paper [26].

2. Transformations and measures on Cantor Space

In this section we first quickly review the basic notions of Turing functionals and of measures on Cantor space 2^{ω} . The space of probability measures on 2^{ω} is compact Polish, and we can devise a suitable effective representation of measures in terms of Cauchy sequences with respect to a certain metric. This enables us to code measures as binary sequences, so we can use them as oracles in Turing machine computations. This way we can extend Martin-Löf's notion of randomness to arbitrary measures by requiring that a Martin-Löf test for a measure is uniformly enumerable in a representation of the measure. This will be done in Section 3.

2.1. The Cantor space as a metric space. The Cantor space 2^{ω} is the set of all infinite binary sequences, also called *reals*. The usual metric on 2^{ω} is defined as follows: Given $x, y \in 2^{\omega}$, $x \neq y$, let $x \cap y$ be the longest common initial segment of x and y (possibly the empty string ϵ). Define

$$d(x,y) = \begin{cases} 2^{-|x \cap y|} & \text{if } x \neq y, \\ 0 & \text{if } x = y. \end{cases}$$

Endowed with this metric, 2^{ω} is a compact Polish space. A countable basis is given by the *cylinder sets*

$$\llbracket \sigma \rrbracket = \{x : x \upharpoonright |\sigma| = \sigma\},\$$

where σ is a finite binary sequence (string), and $|\sigma|$ denotes the length of σ . We use $2^{<\omega}$ to denote the set of all finite binary sequences, and we use \sqsubseteq to denote the usual prefix partial ordering between finite strings. This partial ordering extends in a natural way to $2^{<\omega} \cup 2^{\omega}$. Thus, $x \in \llbracket \sigma \rrbracket$ if and only if $\sigma \sqsubseteq x$. Finally, given $U \subseteq 2^{<\omega}$, we write $\llbracket U \rrbracket$ to denote the open set induced by U, i.e. $\llbracket U \rrbracket = \bigcup_{\sigma \in U} \llbracket \sigma \rrbracket$.

2.2. Turing functionals. The notion of a Turing functional will be important in this paper, so we give a formal definition and explain how functionals give rise to partial, continuous mappings from 2^{ω} to 2^{ω} .

A Turing functional Φ is a computably enumerable set of triples (m, k, σ) such that m is a natural number, k is either 0 or 1, and σ is a finite binary sequence. Further, for all m, for all k_1 and k_2 , and for all compatible σ_1 and σ_2 , if $(m, k_1, \sigma_1) \in \Phi$ and $(m, k_2, \sigma_2) \in \Phi$, then $k_1 = k_2$ and $\sigma_1 = \sigma_2$.

In the following, we will also assume that Turing functionals Φ are *use-monotone*, which means the following hold.

(1) For all (m_1, k_1, σ_1) and (m_2, k_2, σ_2) in Φ , if σ_1 is a proper initial segment of σ_2 , then m_1 is less than m_2 .

(2) For all m_1 and m_2 , k_2 and σ_2 , if $m_2 > m_1$ and $(m_2, k_2, \sigma_2) \in \Phi$, then there are k_1 and σ_1 such that $\sigma_1 \sqsubseteq \sigma_2$ and $(m_1, k_1, \sigma_1) \in \Phi$.

We write $\Phi^{\sigma}(m) = k$ to indicate that there is a τ such that τ is an initial segment of σ , possibly equal to σ , and $(m, k, \tau) \in \Phi$. In this case, we also write $\Phi^{\sigma}(m) \downarrow$, as opposed to $\Phi^{\sigma}(m) \uparrow$, indicating that for all k and all $\tau \sqsubseteq \sigma$, $(m, k, \tau) \notin \Phi$.

If $x \in 2^{\omega}$, we write $\Phi^{x}(m) = k$ to indicate that there is an l such that $\Phi^{x \upharpoonright l}(m) = k$. This way, for given $x \in 2^{\omega}$, Φ^{x} defines a partial function from ω to $\{0,1\}$ (identifying reals with sets of natural numbers). If this function is total, it defines a real y, and in this case we write $\Phi(x) = y$ and say that y is Turing reducible to x via Φ , $y \leq_{\mathbf{T}} x$.

By use-monotonicity, if $\Phi^{\sigma}(m) \downarrow$, then $\Phi^{\sigma}(n) \downarrow$ for all n < m. This way, Φ^{σ} gives rise to a (possibly infinite) string α . If $\Phi^{\sigma}(n) \uparrow$ for all n, we put $\alpha = \epsilon$. We write $\Phi(\sigma) = \alpha$. This way a Turing functional induces a function from $2^{<\omega}$ to $2^{<\omega} \cup 2^{\omega}$ that is monotone, that is, $\sigma \sqsubseteq \tau$ implies $\Phi(\sigma) \sqsubseteq \Phi(\tau)$. We can effectively approximate it by prefixes. More precisely, there exists a computable mapping $(\sigma,s) \mapsto \Phi_s(\sigma) \in 2^{<\omega}$ so that $\Phi_s(\sigma) \sqsubseteq \Phi_{s+1}(\sigma)$, $\Phi_s(\sigma) \sqsubseteq \Phi_s(\sigma^{\frown}i)$ ($i \in \{0,1\}$), and $\lim_s \Phi_s(\sigma) = \Phi(\sigma)$. For technical reasons that will become clear in Section 4, we also require $\Phi_s(\sigma)$ to have the following properties.

- (a) $|\Phi_s(\sigma)| \leq s$ (the approximation does not grow too quickly in length), and
- (b) $|\Phi_{s+1}(\sigma^{\hat{}}i)| \leq |\Phi_s(\sigma)| + 1$, for any $i \in \{0,1\}$ (one more unit of time and one more bit of information will yield at most one additional bit of output).

If, for a real x, $\lim_n |\Phi(x \upharpoonright n)| = \infty$, then $\Phi(x) = y$, where y is the unique real that extends all $\Phi(x \upharpoonright n)$. In this way, Φ also induces a partial, continuous function from 2^{ω} to 2^{ω} . We will use the same symbol Φ for the Turing functional, the monotone function from $2^{<\omega}$ to $2^{<\omega}$, and the partial, continuous function from 2^{ω} to 2^{ω} . It will be clear from the context which Φ is meant.

A Turing functional Φ has computably bounded use if there exists a computable function $g: \mathbb{N} \to \mathbb{N}$ so that $(m, k, \sigma) \in \Phi$ implies that $|\sigma| \leq g(m)$. If $\Phi(x) = y$ for such a functional, we say that y is bounded Turing or weak truth-table reducible to $x, y \leq_{\text{wtt}} x$.

Turing functionals can be relativized with respect to a parameter z, by requiring that Φ is c.e. in z. We call such functionals $Turing\ z$ -functionals. This way we can consider relativized Turing reductions. A real x is Turing reducible to a real y relative to a real z, written $x \leq_{T(z)} y$, if there exists a Turing z-functional Φ such that $\Phi(x) = y$.

2.3. Probability measures. A measure on 2^{ω} is a countably additive, monotone function $\mu: \mathcal{F} \to [0,1]$, where $\mathcal{F} \subseteq \mathcal{P}(2^{\omega})$ is a σ -algebra. If μ is normalized, i.e. if $\mu(2^{\omega}) = 1$, then μ is called a probability measure. A measure μ is a Borel measure if \mathcal{F} is the Borel σ -algebra on 2^{ω} . It is a basic result of measure theory that a measure with domain \mathcal{F} is uniquely determined by the values it takes on an algebra $\mathcal{A} \subseteq \mathcal{F}$ that generates \mathcal{F} . It is not hard to see that in 2^{ω} , the Borel sets are generated by the algebra of clopen sets, i.e. finite unions of basic open cylinders. Normalized, countably additive, monotone set functions on the algebra of clopen sets are induced by any function $\rho: 2^{<\omega} \to [0,1]$ satisfying

(2.1)
$$\rho(\epsilon) = 1 \text{ and for all } \sigma \in 2^{<\omega}, \ \rho(\sigma) = \rho(\sigma^{\smallfrown} 0) + \rho(\sigma^{\smallfrown} 1),$$

where ϵ denotes the empty string. If ρ is as in (2.1), then putting $\mu(\llbracket \sigma \rrbracket) = \rho(\sigma)$ induces a monotone, additive function on the clopen sets, which in turn uniquely

extends to a Borel probability measure on 2^{ω} . In the following, we will deal exclusively with Borel measures. Thus, when we speak of *measures*, we will always mean Borel measures. If convenient, we write μA for $\mu(A)$ to improve readability.

The Lebesgue measure λ on 2^{ω} is obtained by distributing a unit mass uniformly along the paths of 2^{ω} , i.e. by setting $\lambda(\sigma) = 2^{-|\sigma|}$. A Dirac measure, on the other hand, is defined by putting a unit mass on a single real, i.e. for $x \in 2^{\omega}$, let

$$\delta_x \llbracket \sigma \rrbracket = \begin{cases} 1 & \text{if } \sigma \sqsubset x, \\ 0 & \text{otherwise.} \end{cases}$$

If, for a measure μ and $x \in 2^{\omega}$, $\mu\{x\} > 0$, then x is called an *atom* of μ . Obviously, x is an atom of δ_x . A measure that does not have any atoms is called *continuous*.

2.4. The space of probability measures on Cantor space. We denote by $\mathcal{P}(2^{\omega})$ the set of all probability measures on 2^{ω} . $\mathcal{P}(2^{\omega})$ can be given a topology (the so-called weak-* topology) by letting $\mu_n \to \mu$ if $\int f d\mu_n \to \int f d\mu$ for all continuous real-valued functions f on 2^{ω} .

It is known that if X is compact metrizable, then so is the space of all probability measures on X (see for instance [12]). Therefore, $\mathcal{P}(2^{\omega})$ is compact metrizable. In particular, it is Polish. A compatible metric (see for example [8]) is given by

$$d_{\text{meas}}(\mu, \nu) = \sum_{n=1}^{\infty} 2^{-n} d_n(\mu, \nu),$$

where

$$d_n(\mu, \nu) = \frac{1}{2} \sum_{|\sigma|=n} |\mu[\![\sigma]\!] - \nu[\![\sigma]\!]|.$$

A countable, dense subset $\mathcal{D} \subseteq \mathcal{P}(2^{\omega})$ is given by the set of measures which assume positive, dyadic rational values (of the form $m/2^n$ with $m, n \geq 0$) on a finite number of dyadic rationals, i.e. \mathcal{D} is the set of measures of the form

$$\nu_{\Delta,Q} = \sum_{\sigma \in \Delta} Q(\sigma) \delta_{\sigma \frown 0^{\omega}},$$

where Δ is a finite set of finite strings (representing dyadic rational numbers) and $Q: \Delta \to [0,1]$ such that $\sum_{\sigma \in \Delta} Q(\sigma) = 1$ and for all $\sigma \in \Delta$, $Q(\sigma)$ is a dyadic rational number.

It is straightforward to show that in case μ, ν are restricted to measures in \mathcal{D} , the following relations are computable:

$$d_{\text{meas}}(\mu, \nu) < q$$
 and $d_{\text{meas}}(\mu, \nu) \le q$ $(q \in \mathbb{Q} \cap [0, 1]).$

By effectively enumerating all possible combinations (Δ, Q) , we can also effectively enumerate the set \mathcal{D} . In the following, we fix such an enumeration $\mathcal{D} = \{\nu_0, \nu_1, \nu_2, \dots\}$. The triple $(\mathcal{P}(2^{\omega}), \mathcal{D}, d_{\text{meas}})$ forms a *computable metric space* (see e.g. [31], or [10], [7] in this particular context).

We can represent measures in $\mathcal{P}(2^{\omega})$ through Cauchy sequences of measures in \mathcal{D} , which in turn can be encoded as reals using the enumeration of \mathcal{D} . Furthermore, the fact that $\mathcal{P}(2^{\omega})$ is a compact, computable metric space can be used to devise a coding scheme that captures the topology of $\mathcal{P}(2^{\omega})$.

Proposition 2.1. There exists a Turing functional Γ so that for all $x \in 2^{\omega}$ and all $n \in \mathbb{N}$, $\Gamma^{x}(n)$ is defined and

$$d_{\text{meas}}(\nu_{\Gamma^x(n)}, \nu_{\Gamma^x(n+1)}) \le 2^{-n}$$
.

Furthermore, Γ induces a continuous surjection $\rho: 2^{\omega} \to \mathcal{P}(2^{\omega})$ by letting

$$\rho(x) = \lim_{n} \nu_{\Gamma^x(n)},$$

where the limit is taken with respect to the weak-* topology. Finally, ρ is such that for any $x \in 2^{\omega}$, the set

$$\rho^{-1}(\{\rho(x)\})$$

is $\Pi_1^0(x)$.

For a proof, see [3]. ([25] has a similar development of representations of measures.) If $\mu \in \mathcal{P}(2^{\omega})$, any real r with $\rho(r) = \mu$ is called a *representation* of μ . Note that the representation of a measure is not unique. In fact, a measure may have uncountably many distinct representations.

Proposition 2.2. Let $r \in 2^{\omega}$ be a representation of a measure $\mu \in \mathcal{P}(2^{\omega})$. Then the relations

$$\mu[\![\sigma]\!] < q \quad and \quad \mu[\![\sigma]\!] > q \quad (\sigma \in 2^{<\omega}, q \in \mathbb{Q})$$

are c.e. in r.

Proof. Let $\gamma(k) = \Gamma^r(k)$, where Γ is the Turing functional from Proposition 2.1. Since $d_{\text{meas}}(\nu_{\gamma(k)}, \mu) \leq 2^{-n+1}$, it follows from the definition of d_{meas} that

$$|\nu_{\gamma(k)}[\![\sigma]\!] - \mu[\![\sigma]\!]| \le 2^{-k+|\sigma|+2}.$$

This in turn implies

$$\mu[\![\sigma]\!] \le \nu_{\gamma(k)}[\![\sigma]\!] + 2^{-k+|\sigma|+2} \le \mu[\![\sigma]\!] + 2^{-k+|\sigma|+3}.$$

Hence $\mu \llbracket \sigma \rrbracket < q$ if and only if

$$\exists k \ \nu_{\gamma(k)}[\![\sigma]\!] + 2^{-k+|\sigma|+2} < q.$$

But $\nu_{\gamma(k)}[\![\sigma]\!]$ is a dyadic rational uniformly computable in r, and hence $\nu_{\gamma(k)}[\![\sigma]\!] + 2^{-k+|\sigma|+2} < q$ is decidable given r as an oracle. The proof for $\mu[\![\sigma]\!] > q$ is symmetrical.

Proposition 2.2 easily implies the following

Proposition 2.3. Let $r \in 2^{\omega}$ be a representation of a measure $\mu \in \mathcal{P}(2^{\omega})$. Then r computes a function $g_{\mu}: 2^{<\omega} \times \mathbb{N} \to \mathbb{Q}$ such that for all $\sigma \in 2^{<\omega}$, $n \in \mathbb{N}$,

$$|g_{\mu}(\sigma, n) - \mu[\![\sigma]\!]| \le 2^{-n}.$$

¹The way we defined Turing functionals in Section 2.2, they induce partial mappings from 2^{ω} to 2^{ω} . Here we obviously assume that Γ induces a mapping from 2^{ω} to ω^{ω} . The definition given in Section 2.2 is easily adjustable to this case.

3. Randomness and Transformations of Measures

3.1. Random Reals. We define (relative) randomness of reals for arbitrary measures as a straightforward extension of Martin-Löf's test notion. The basic idea is to require the test to be *enumerable* (Σ_1^0) in a representation of the measure.

Definition 3.1. Let r_{μ} be a representation of a measure μ , and let $z \in 2^{\omega}$.

(a) An (r_{μ}, z) -test is given by a sequence $(V_n : n \in \mathbb{N})$ of uniformly $\Sigma_1^0(r_{\mu} \oplus z)$ -sets $V_n \subseteq 2^{<\omega}$ such that for all n,

$$\sum_{\sigma \in V_n} \mu \llbracket \sigma \rrbracket \le 2^{-n}.$$

- (b) A real $x \in 2^{\omega}$ passes an (r_{μ}, z) -test (V_n) if $x \notin \bigcap_n \llbracket V_n \rrbracket$. Otherwise we say the test (V_n) covers x.
- (c) A real $x \in 2^{\omega}$ is (r_{μ}, z) -random if it passes all (r_{μ}, z) -tests.

If, in the previous definition, $z = \emptyset$ (where we identify reals with subsets of natural numbers via the characteristic sequence), we simply speak of an r_{μ} -test and of x being r_{μ} -random.

The previous definition defines randomness with respect to a specific representation. If x is random for one representation, it is not necessarily random for other representations. On the other hand, we can ask whether a real exhibits randomness with respect to some representation, so the following definition makes sense.

Definition 3.2. A real $x \in 2^{\omega}$ is μ -random relative to $z \in 2^{\omega}$, or simply μ -z-random, if there exists a representation r_{μ} of μ so that x is (r_{μ}, z) -random.

One might argue that this definition of randomness is subject to a certain arbitrariness, as it depends on a particular representation of a measure. However, it has recently been shown by Day and Miller [3] that the definition of randomness given in Definition 3.2 is equivalent to the representation-independent approach via uniform tests due to Levin [16, 17, 18] and Gács [7].

Moreover, in this paper we are interested in results of the type "For which reals does there exist a measure for which x looks non-trivially random?" – i.e. can x look random at all? If there exists such a (representation of a) measure, there is good reason to say that x has some random content. It is not the aim of this paper to provide a most general solution to the problem of defining randomness for arbitrary measures. The goal is to exhibit an interesting connection between the randomness properties of reals (with respect to some representation) and its logical complexity.

Of course, every real x is trivially μ -random if it is a μ -atom. The question is under what circumstances x is non-trivially μ -random, i.e. when does there exist a measure μ so that x is μ -random and $\mu\{x\} = 0$.

A most useful property of the theory of Martin-Löf randomness is the existence of universal tests. Universal tests subsume all other tests. Furthermore, they can be defined uniformly with respect to any parameter. More precisely, for any representation r_{μ} of a measure μ , there exists a uniformly c.e. in r_{μ} sequence $(U_n : n \in \mathbb{N})$ of sets $U_n \subseteq 2^{<\omega}$ such that, if we set for $z \in 2^{\omega}$,

$$U_n^z = \{ \sigma \colon \langle \sigma, \tau \rangle \in U_n, \ \tau \sqsubseteq z \},$$

then (U_n^z) is an (r_μ, z) -test and $x \in 2^\omega$ is (r_μ, z) -random if and only if x passes (U_n^z) . We call (U_n) a universal oracle test for r_μ . For details on the existence of universal tests, see [2]. One can also construct universal tests that have a larger number of parameters. In Section 4 we will need a test that is universal for two parameters. Such a test is a uniformly c.e. in r_μ sequence $(U_n^{(2)}: n \in \mathbb{N})$ so that, if we set for $z_0, z_1 \in 2^\omega$,

$$U_n^{z_0,z_1} = \{ \sigma \colon \langle \sigma, \tau_0, \tau_1 \rangle \in U_n^{(2)}, \ \tau_0 \sqsubseteq z_0, \tau_1 \sqsubseteq z_1 \},$$

then $(U_n^{z_0,z_1})$ is an $(r_\mu, z_0 \oplus z_1)$ -test and $x \in 2^\omega$ is $(r_\mu, z_0 \oplus z_1)$ -random if and only if x passes $(U_n^{z_0,z_1})$.

3.2. Computable measures. A measure is *computable* if there exists a computable function $g: 2^{<\omega} \times \mathbb{N} \to \mathbb{Q}$ such that for all $\sigma \in 2^{<\omega}$, $n \in \mathbb{N}$,

$$|g(\sigma, n) - \mu[\![\sigma]\!]| \le 2^{-n}.$$

By Proposition 2.3, any measure with a computable representation is computable. The converse holds, too (see [10]). For a computable measure μ , not being μ -z-random is equivalent to the existence of a sequence (V_n) uniformly c.e. in z so that for all n, $\mu[V_n] \leq 2^{-n}$ and so that (V_n) covers x. We call the latter a (μ, z) -test.

Lebesgue measure λ is computable, and it is arguably the most prominent measure on 2^{ω} . λ -random reals are the most studied ones by far, and we will, in consistency with the literature, use the name $Martin-L\"{o}f$ random reals for them.

Reals that are random with respect to some computable probability measure have been called proper [33] or natural [22].

Obviously no computable real can be non-trivially random with respect to any measure. The following observation yields that the trivially random reals with respect to computable measures are precisely the computable reals.

Proposition 3.3 (Levin [33]). If μ is a computable measure and $\mu\{x\} > 0$ for some $x \in 2^{\omega}$, then x is computable.

Proof. Suppose $\mu\{x\} > 2^{-m} > 0$ for some computable μ and $m \geq 1$. Let g be a computation function for μ , i.e. g is computable and for all σ and n, $|g(\sigma,n) - \mu[\![\sigma]\!]| \leq 2^{-n}$. Define a computable tree $T \subseteq 2^{<\omega}$ by letting $\sigma \in T$ if and only if $g(\sigma, |\sigma|) \geq 2^{-m} - 2^{-|\sigma|}$. x is an infinite path through T. We claim that x is a fully isolated path, i.e. there exists a string σ such that for all $\tau \supseteq \sigma$, $\tau \in T$ implies $\tau \sqsubseteq x$. Clearly any fully isolated path through T is computable.

Suppose x were not fully isolated. Then there exist infinitely many $\sigma_n \sqsubset x$, $\sigma_n \sqsubset \sigma_{n+1}$, such that $\sigma_n^{\vee} \in T$, where σ_n^{\vee} is obtained from σ_n by switching the last bit. It follows that the σ_n^{\vee} are pairwise incompatible. Since $\sigma_n^{\vee} \in T$, we have that for sufficiently large n,

$$\mu[\![\sigma_n^\vee]\!] \ge \frac{1}{2^{m+1}},$$

which is impossible since the σ_n^{\vee} are pairwise incompatible and μ is a probability measure. \Box

As regards non-computable reals, Levin proved that, from a computability theoretic point of view, randomness with respect to a computable probability measure is computationally as powerful as Martin-Löf randomness. This was independently shown by Kautz [11].

Theorem 3.4 (Levin [33], Kautz [11]). A non-computable real which is random with respect to some computable measure is Turing equivalent to a Martin-Löf random real.

The proof of Theorem 3.4 uses the fact that reductions induce continuous (partial) mappings from 2^{ω} to 2^{ω} . Such mappings transform measures.

3.3. Transformation of Measures. Let μ be a Borel measure on 2^{ω} , and let $f: 2^{\omega} \to 2^{\omega}$ be a μ -measurable function, that is, for all measurable $A \subseteq 2^{\omega}$, $f^{-1}(A)$ is μ -measurable, too. Such f induces a new measure μ_f , often referred to as the *image measure* or *push-forward* of μ , on 2^{ω} by letting

$$\mu_f(A) = \mu(f^{-1}(A)).$$

Any non-atomic measure can be transformed into Lebesgue measure λ this way. Recall that a function $f: 2^{\omega} \to 2^{\omega}$ is a *Borel automorphism* if it is bijective and for any $A \subseteq 2^{\omega}$, A is a Borel set if and only if f(A) is.

Theorem 3.5 (see [9]). Let μ be a non-atomic probability measure on 2^{ω} .

- (1) There exists a continuous mapping $f: 2^{\omega} \to 2^{\omega}$ such that $\mu_f = \lambda$.
- (2) There exists a Borel automorphism h of 2^{ω} such that $\mu_h = \lambda$.

The simple idea to prove (1) is to identify Cantor space with the unit interval [0,1] and define $f(x) = \mu([0,x])$, that is, to let f equal the distribution function of μ .

To prove Theorem 3.4, Levin showed that this idea works in the effective case, too. Furthermore, one can even deal with the presence of atoms, as long as the measure is computable. Roughly speaking, a computable transformation transforms a computable measure into another computable measure. At the same time, a computable transformation will preserve randomness in the sense that a real random with respect to μ , when transformed by a computable mapping, is random with respect to the image measure of μ .

In the following, we will use a (relativized) transformation of Lebesgue measure to render a given real random.

4. Randomness with Respect to Arbitrary Measures

The Levin-Kautz result implies that the set of reals that are non-trivially random with respect to a computable measure are contained in the set of Martin-Löf random Turing degrees. But what about the reals non-trivially random with respect to an arbitrary measure? In this section we will show that these coincide with the non-computable reals. Every non-computable real is non-trivially random with respect to some measure.

The proof of this result uses two important results from computability theory and algorithmic randomness, the Kučera-Gács Theorem and the Posner-Robinson Theorem.

We will need the Kučera-Gács Theorem in the following, relativized form.

Theorem 4.1 (Kučera [15], Gács [6]). Let $x, z \in 2^{\omega}$. There exists a real y that is Martin-Löf random relative to z such that

$$x \leq_{\text{wtt}(z)} y \leq_{\text{wtt}(z)} x \oplus z'.$$

Theorem 4.2 (Posner and Robinson [24]). If $x \in 2^{\omega}$ is non-computable, then there is a $z \in 2^{\omega}$ such that $x \oplus z \geq_{\mathbf{T}} z'$.

Corollary 4.3. For every non-computable real x, there exist reals $y, z \in 2^{\omega}$ so that y is Martin-Löf random relative to z and

$$x \leq_{\text{wtt}(z)} y$$
 and $y \leq_{\text{T}(z)} x$.

The (relative) Turing equivalence to a random real allows for transforming Lebesgue measure λ in a sufficiently controlled manner. More precisely, we can obtain a Π^0_1 class M of representations of measures. Each measure represented in this class is a good candidate for a measure that renders x random. We will use a compactness argument to show that at least one member r_μ of M has the property that the Martin-Löf random real y is still λ -random relative to r_μ (here r_μ is viewed as a real, not a measure). Then, x has to be r_μ -random, since otherwise an r_μ -test could be effectively transformed into a Martin-Löf λ -test relative to r_μ which y would fail.

We now state and prove the main result of this section.

Theorem 4.4. For any real $x \in 2^{\omega}$, the following are equivalent:

- (i) There exists a probability measure μ such that x is not a μ -atom and x is μ -random.
- (ii) x is not computable.

Proof. (i) \Rightarrow (ii): If x is computable and μ is a measure with $\mu\{x\} = 0$, then we can construct a μ -test that covers x by using x and any representation of μ to search for an initial segments of x whose measure is sufficiently small. More formally, given n, compute, using any representation r_{μ} of μ as an oracle, a length l_n for which $\mu[x \upharpoonright l_n] < 2^{-n}$. Define a μ -test (V_n) by letting $V_n = \{x \upharpoonright l_n\}$.

(ii) \Rightarrow (i): Let x be a non-computable real. Using Corollary 4.3, we obtain a real y which is Martin-Löf random relative to some $z \in 2^{\omega}$ and which is T(z)-equivalent to x.

There are Turing z-functionals Φ and Ψ such that

$$\Phi(y) = x$$
 and $\Psi(x) = y$.

We will use the functionals Φ and Ψ to define a set of measures M. If Φ were total and invertible, there would be no problem to define the desired measure, as one could simply 'push forward' Lebesgue measure using Φ . In our case we have to use Φ and Ψ to control the measure. We are guaranteed that this will work *locally*, since Φ and Ψ are mutual inverses when restricted to x and y. Therefore, given a string σ (a possible initial segment of x) we will single out strings which appear to be candidates for initial segments of an inverse real.

For any σ , let $\operatorname{Pre}^*(\sigma) \subseteq 2^{<\omega}$ be defined as

$$\operatorname{Pre}^*(\sigma) = \{ \tau \in 2^{<\omega} : \Phi(\tau) \supseteq \sigma \& \Psi_{|\sigma|}(\sigma) \sqsubseteq \tau \}.$$

We will need only the elements of Pre^* that are minimal with respect to the prefix relation. Let

$$\operatorname{Pre}(\sigma) = \{ \tau \in \operatorname{Pre}^*(\sigma) : \forall \tau' \in \operatorname{Pre}^*(\sigma) \ (\tau, \tau' \text{ compatible } \to \tau \sqsubseteq \tau') \}$$

Note that $\operatorname{Pre}(\sigma)$ is uniformly c.e. in z, since we can approximate $\Phi(\tau)$ by longer and longer prefixes (the strings $\Phi_s(\tau)$), and we assume the reductions Φ, Ψ to be use monotone.

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

- (1) The measure μ dominates the partial push-forward of Lebesgue measure induced by Φ . This will help ensure that any Martin-Löf random real is mapped by Φ to a μ -random real.
- (2) The measure μ must not be atomic on x.

To meet these requirements, we restrict the values of μ in the following way:

(4.1)
$$\lambda \| \operatorname{Pre}(\sigma) \| \le \mu \| \sigma \| \le \lambda \| \Psi_{|\sigma|}(\sigma) \|$$

The first inequality ensures that (1) is met, whereas the second guarantees that μ is non-atomic on the domain of Ψ (since if $\Psi(z)$ is defined, then $\lim_s |\Psi_s(z \upharpoonright s)| = \infty$ and thus $\lambda[\![\Psi_s(z \upharpoonright s))\!]\!] \to 0$). If $\Psi(z)$ is undefined, then $\Psi_s(z \upharpoonright s)$ is constant from some point on and hence imposes a constant positive upper bound on all $\mu[\![z \upharpoonright s]\!]$ from that point on.

Let $M \subseteq 2^{\omega}$ be the set of all representations of measures that satisfy (4.1). We show that M is non-empty and $\Pi_1^0(z)$.

Claim. The set M is not empty.

Proof. We exhibit a measure μ that respects all upper and lower bounds given by (4.1). Since every measure has a representation, this implies that M is non-empty. We construct μ inductively on the basic open cylinders. Put $\mu[\![\epsilon]\!] = 1$. Suppose $\mu[\![\sigma]\!]$ is given such that

$$\lambda \| \operatorname{Pre}(\sigma) \| \le \mu \| \sigma \| \le \lambda \| \Psi_{|\sigma|}(\sigma) \|$$

It follows from the definition of Pre that

$$\operatorname{Pre}(\sigma^{\frown}0), \operatorname{Pre}(\sigma^{\frown}1) \subseteq \operatorname{Pre}(\sigma)$$
 and $\operatorname{Pre}(\sigma^{\frown}0) \cap \operatorname{Pre}(\sigma^{\frown}1) = \emptyset$.

Hence, since λ is a measure,

$$\lambda \| \operatorname{Pre}(\sigma^{\widehat{}} 0) \| + \lambda \| \operatorname{Pre}(\sigma^{\widehat{}} 1) \| \le \lambda \| \operatorname{Pre}(\sigma) \|$$

Furthermore, by the properties of the approximation Ψ_s stated in Section 2.2, we have

$$\lambda \llbracket \Psi_{|\sigma^{\frown}0|}(\sigma^{\frown}0) \rrbracket + \lambda \llbracket \Psi_{|\sigma^{\frown}1|}(\sigma^{\frown}1) \rrbracket \geq 2^{-|\Psi_{|\sigma|}(\sigma)|-1} + 2^{-|\Psi_{|\sigma|}(\sigma)|-1} = \lambda \llbracket \Psi_{|\sigma|}(\sigma) \rrbracket$$
 Thus,

$$\lambda \llbracket \operatorname{Pre}(\sigma^{\smallfrown} 0) \rrbracket + \lambda \llbracket \operatorname{Pre}(\sigma^{\smallfrown} 1) \rrbracket \le \mu \llbracket \sigma \rrbracket \le \lambda \llbracket \Psi_{|\sigma^{\smallfrown} 0|}(\sigma^{\smallfrown} 0) \rrbracket + \lambda \llbracket \Psi_{|\sigma^{\smallfrown} 1|}(\sigma^{\smallfrown} 1) \rrbracket.$$

Since the mapping $\theta: [0,1] \times [0,1] \to \mathbb{R}$ given by

$$\begin{split} \theta(s,t) &= \lambda \llbracket \operatorname{Pre}(\sigma ^{\frown} 0) \rrbracket + s \left(\lambda \llbracket \Psi_{|\sigma ^{\frown} 0|}(\sigma ^{\frown} 0) \rrbracket - \lambda \llbracket \operatorname{Pre}(\sigma ^{\frown} 0) \rrbracket \right) + \\ & \lambda \llbracket \operatorname{Pre}(\sigma ^{\frown} 1) \rrbracket + t \left(\lambda \llbracket \Psi_{|\sigma ^{\frown} 1|}(\sigma ^{\frown} 1) \rrbracket - \lambda \llbracket \operatorname{Pre}(\sigma ^{\frown} 1) \rrbracket \right) \end{split}$$

is continuous, it follows from the intermediate value theorem that there exist s_0, t_0 such that $\theta(s_0, t_0) = \mu \llbracket \sigma \rrbracket$. Put

$$\mu[\![\sigma \cap 0]\!] = \lambda[\![\operatorname{Pre}(\sigma \cap 0)]\!] + s_0 (\lambda[\![\Psi_{|\sigma \cap 0|}(\sigma \cap 0)]\!] - \lambda[\![\operatorname{Pre}(\sigma \cap 0)]\!])$$
$$\mu[\![\sigma \cap 1]\!] = \lambda[\![\operatorname{Pre}(\sigma \cap 1)]\!] + t_0 (\lambda[\![\Psi_{|\sigma \cap 1|}(\sigma \cap 1)]\!] - \lambda[\![\operatorname{Pre}(\sigma \cap 1)]\!])$$

Claim. The set M is $\Pi_1^0(z)$.

Proof. A representation r is in M if and only if

$$\forall \sigma \ \lambda \llbracket \operatorname{Pre}(\sigma) \rrbracket \leq \rho(r) \llbracket \sigma \rrbracket \leq \lambda \llbracket \Psi_{|\sigma|}(\sigma) \rrbracket.$$

It suffices to show that the relation $\lambda[\![\operatorname{Pre}(\sigma)]\!] \leq \rho(r)[\![\sigma]\!] \leq \lambda[\![\Psi_{|\sigma|}(\sigma)]\!]$ is uniformly $\Pi_1^0(z)$.

The set $\operatorname{Pre}(\sigma)$ is c.e. in z (uniformly in σ), and thus the measure $\lambda[\operatorname{Pre}(\sigma)]$ is left-enumerable in z. There exists a strictly increasing, computable sequence of dyadic rationals (q_n) so that $q_n \to \lambda[\operatorname{Pre}(\sigma)]$. We have that

$$\rho(r) \llbracket \sigma \rrbracket < \lambda \llbracket \operatorname{Pre}(\sigma) \rrbracket \quad \Leftrightarrow \quad \exists n \ \rho(r) \llbracket \sigma \rrbracket < q_n.$$

By Proposition 2.2, $\rho(r)\llbracket \sigma \rrbracket < q_n$ is $\Sigma_1^0(z)$. Hence $\rho(r)\llbracket \sigma \rrbracket < \operatorname{Pre}(\sigma)$ is $\Sigma_1^0(z)$, too. Proposition 2.2 also yields that the relation $\rho(r)\llbracket \sigma \rrbracket > \lambda \llbracket \Psi_{|\sigma|}(\sigma) \rrbracket$ is $\Sigma_1^0(z)$. Hence the relation $\lambda \llbracket \operatorname{Pre}(\sigma) \rrbracket \leq \rho(r) \llbracket \sigma \rrbracket \leq \lambda \llbracket \Psi_{|\sigma|}(\sigma) \rrbracket$ is $\Pi_1^0(z)$, as desired, and the argument above is uniform in σ .

Let $r_{\mu} \in M$. It follows from the definition of M that x is not an atom of μ . Suppose (V_n) is an (r_{μ}, z) -test that covers x. For each n, let

$$W_n = \{ \operatorname{Pre}(\sigma) \colon \sigma \in V_n \}.$$

Then W_n is uniformly $\Sigma_1^0(r_\mu \oplus z)$ since $\operatorname{Pre}(\sigma)$ is uniformly c.e. in z. Furthermore,

$$\lambda \llbracket W_n \rrbracket \leq \sum_{\sigma \in V_n} \lambda \llbracket \operatorname{Pre}(\sigma) \rrbracket \leq \sum_{\sigma \in V_n} \mu \llbracket \sigma \rrbracket \leq 2^{-n}.$$

Hence (W_n) is a $(\lambda, r_\mu \oplus z)$ -test. Moreover, since $y \in [\![\operatorname{Pre}(x \upharpoonright k)]\!]$ for all k and (V_n) covers x, (W_n) covers y.

At this point, we are not yet quite able to derive a contradiction, since we only have that y is Martin-Löf random relative to z, and not necessarily relative to $r_{\mu} \oplus z$. However, we can use the following basis theorem for Π^0_1 -classes to infer the existence of an element $r \in M$ so that y is $(\lambda, r \oplus z)$ -random. Lemma 4.5 was independently obtained by Downey, Hirschfeldt, Miller, and Nies [4].

Lemma 4.5. Let $z \in 2^{\omega}$, and let $T \subseteq 2^{<\omega}$ be a tree computable in z such that T has an infinite path. Then, for every real y that is Martin-Löf random relative to z, there is an infinite path v through T such that y is Martin-Löf random relative to $z \oplus v$.

Proof. We use the universal oracle test $(U_n^{(2)})$ introduced in Section 3. Given $z \in 2^{\omega}$ and $\tau \in 2^{<\omega}$, let

$$U_n^{z,\tau} = \{ \sigma \colon \langle \sigma, \tau_0, \tau_1 \rangle \in U_n^{(2)}, \ \tau_0 \sqsubset z, \tau_1 \sqsubseteq \tau \},$$

where $(U_n^{(2)})$ is a universal λ -test for two parameters, as introduced in Section 3.1. Note that the sequence $(U_n^{z,\sigma})$ is uniformly c.e. in z and forms a (λ,z) -test.

We enumerate a (λ, z) -test (V_n) as follows: enumerate a string σ into V_n if $\llbracket \sigma \rrbracket$ is contained in $\llbracket U_n^{z,\tau} \rrbracket$ for all $\tau \in T$ with $|\tau| = |\sigma|$ (note that there are only finitely many such τ).

If y is Martin-Löf random relative to z, there has to be some n such that $y \notin \llbracket V_n \rrbracket$. Let

$$T_0 = \{ \tau \in T : \llbracket y \upharpoonright |\tau| \rrbracket \text{ is not contained in } \llbracket U_n^{z,\tau} \rrbracket \}.$$

 T_0 is clearly closed under initial segments, hence it is a subtree of T. For every m, $y \upharpoonright m$ is not enumerated in V_n , which implies that T_0 has nodes of every length, in

particular it is infinite. Applying König's Lemma yields an infinite path v through T_0 .

Now, y is Martin-Löf random relative to $z \oplus v$. For suppose not, then, since $(U_n^{(2)})$ is a universal oracle test for λ , $y \in \bigcap_k \llbracket U_k^{z,v} \rrbracket$. In particular, $y \in \llbracket U_n^{z,v} \rrbracket$. By the use principle, there is an initial segment $\tau \sqsubset v$ such that $y \in \llbracket U_n^{z,\tau} \rrbracket$, contradicting the fact that $\tau \in T_0$.

Using Lemma 4.5, we obtain a representation $r_{\mu} \in M$ so that x is (r_{μ}, z) -random. To complete the proof of Theorem 4.4, note that any (r_{μ}, z) -random real is also r_{μ} -random, and hence μ -random.

How hard is it to find a measure that makes a given real x random? The tree determining M is computable in the parameter z, stemming from the Posner-Robinson Theorem. z can be found recursively in $x \oplus \emptyset'$. It takes another jump to find a path through the tree which conserves randomness for some y Turing z-equivalent to x. Hence, there is a representation $r_{\mu} \leq_{\mathrm{T}} x''$ of a measure μ so that x is r_{μ} -random.

5. Randomness with Respect to Continuous Measures

In this section we investigate what happens if one replaces the property "random with respect to an arbitrary probability measure" (which turned out to hold for any non-computable real) with being random for a continuous probability measure.

First, we give an explicit construction of a non-computable real which does not have the latter property.

5.1. A real not continuously random.

Theorem 5.1. There exists a non-computable real which is not random with respect to any continuous measure.

Proof. Consider the halting problem \emptyset' . Denote by \emptyset'_t the approximation to \emptyset' enumerated after t steps. We define a set of markers $\gamma_t(n)$ that capture when the first n bits of \emptyset'_t have settled. For each t, let $\gamma_t(0) = 0$ and define

$$\gamma_t(n+1) := \max\{\min\{s \leq t: \, \emptyset_s' \upharpoonright n+1 = \emptyset_t' \upharpoonright n+1\}, \gamma_t(n)+1\}.$$

Each $\gamma_t(n)$ (as a function of t) will be constant from some point on as computations of \emptyset' settle. We denote this value by $\gamma(n)$. Let y_t the real given by the characteristic sequence of $\{\gamma_t(n): n \geq 0\}$, and let y be the real given by the characteristic sequence of $\{\gamma(n): n \geq 0\}$. We claim that y is not random with respect to any continuous measure.

Let $\mu \in \mathcal{P}(2^{\omega})$ be continuous, and let r_{μ} be any representation of μ . Since 2^{ω} is compact and μ is continuous, for every rational $\varepsilon > 0$ there exists a number $l(\varepsilon)$ such that

$$\forall \sigma \in \{0,1\}^{l(\varepsilon)} \ \mu[\![\sigma]\!] < \varepsilon.$$

By Proposition 2.2, the relation $\mu\llbracket\sigma\rrbracket < \varepsilon$ is $\Sigma^0_1(r_\mu)$. Therefore, r_μ can compute such a function l. We use this to uniformly enumerate the nth level of an r_μ -test $(V_n)_{n\in\mathbb{N}}$ that covers y.

First, compute $n_0 = l(2^{-n-1})$ and $n_1 = l(2^{-n-1}/n_0)$. Enumerate $y_{n_1} \upharpoonright n_0$ into V_n . Furthermore, for all k such that $\gamma_{n_1}(k) < n_0$, enumerate the string

$$(y_{n_1} \upharpoonright \gamma_{n_1}(k)) \cap 0^{n_1 - \gamma_{n_1}(k)}$$

into V_n .

Note that $y_t \upharpoonright n_0$ can change at most n_0 many times. Observe further that, if the approximation to \emptyset' changes at a position m at time t, it will move the marker $\gamma_t(m)$ and with it all the markers $\gamma_t(k)$, k > m to a position $\geq t$. Hence, for the maximum $k \leq n_0$ so that $\gamma_{n_1}(k) = \gamma(k)$,

$$y_{n_1} \upharpoonright \gamma_{n_1}(k) = y \upharpoonright \gamma_{n_1}(k) = y \upharpoonright \gamma(k),$$

and between $\gamma_{n_1}(k)$ and n_1 , the characteristic sequence of y has only 0s. Therefore, y is covered by V_n .

Finally, note that the measure of V_n is at most 2^{-n} , since

$$\sum_{v \in V_n} \mu[\![v]\!] = \mu[\![y_{n_1} \upharpoonright n_0]\!] + \sum_{\{k \colon \gamma_{n_1}(k) < n_0\}} \mu[\![(y_{n_1} \upharpoonright \gamma_{n_1}(k))^{\frown} 0^{n_1 - \gamma_{n_1}(k)}]\!]$$

$$\leq 2^{-n-1} + n_0 \frac{2^{-n-1}}{n_0} = 2^{-n}.$$

5.2. Classifying the not continuously random reals. Theorem 5.1 suggests the following question: Is it possible to classify the reals which are not random with respect to a continuous measure? Denote by NCR the set of all such reals. Can we obtain bounds on the complexity of NCR?

An observation by Kjos-Hanssen and Montalbán [13] shows that NCR is cofinal the hyperarithmetical Turing degrees, i.e. for any hyperarithmetical x there exists a hyperarithmetical $y \in \text{NCR}$ so that $x \leq_T y$.

It follows directly from the countable additivity of measures that, for every continuous measure μ , any countable subset of 2^{ω} has μ -measure zero.

For countable Π_1^0 classes, we can strengthen this to effective μ -measure zero, and hence no countable Π_1^0 class contains a real in NCR.

Theorem 5.2 (Kjos-Hanssen and Montalbán [13]). If $A \subseteq 2^{\omega}$ is countable and Π_1^0 , then no member of A can be random with respect to a continuous measure.

Proof. Let A = [T] for some computable tree T, and suppose μ is a continuous measure. Let r_{μ} be any representation of μ . Let $T^{=n}$ denote the strings in T of length n. It holds that

$$\mu[T^{=n}] \to \mu[T] = 0 \qquad (n \to \infty).$$

Using Proposition 2.2, we can browse the tree T level by level till we see that the measure of $[T^{=n}]$ falls below 2^{-n} . When this happens, we enumerate all strings in $T^{=n}$ into the n-th level of an r_{μ} -test.

Kreisel [14] showed that every member of a countable Π_1^0 class is hyperarithmetic, i.e. contained in Δ_1^1 . Furthermore, he showed that members of countable Π_1^0 classes (also called *ranked points*) can be found cofinally the hyperarithmetical Turing degrees. Later, Cenzer, Clote, Smith, Soare, and Wainer [1] showed that ranked points appear at each hyperarithmetical level of the Turing jump, i.e. can be found in Turing degrees obtained by iterating the Turing jump along a computable ordinal.

Corollary 5.3. For every computable ordinal β there exists an $x \in NCR$ such that $x \equiv_T \emptyset^{(\beta)}$.

We would like to obtain an upper bound on the complexity of reals in NCR; in particular, we would like to know whether NCR is countable. We start with the following simple observation.

Proposition 5.4. The set NCR of all reals not random with respect to any continuous measure is Π_1^1 .

Proof. We have

 $x \in NCR \Leftrightarrow (\forall r)[r \text{ represents a measure } \mu \text{ and } \mu \text{ is continuous}]$

 \rightarrow some r-test covers x].

The property 'r represents a measure' is obviously arithmetic, and due to compactness of 2^{ω} , a measure is μ is continuous if and only if

$$(\forall n)(\exists l)(\forall \sigma)[|\sigma| = l \to \mu \llbracket \sigma \rrbracket \le 2^{-n}].$$

Hence ' μ is continuous' is arithmetic in r. Furthermore 'some test covers x' can be expressed as

$$(\exists e)(\forall n) \left[\left[(\forall s) \sum_{\sigma \in W_{\{e\}_s^r(n)}} \rho(r) \llbracket \sigma \rrbracket \leq 2^{-n} \right] \wedge (\exists \sigma) \left[\sigma \in W_{\{e\}_s^r(n)} \wedge \sigma \sqsubset x \right] \right],$$

which is arithmetic in x and r.

Furthermore, it is not hard to see that NCR does not have a perfect subset.

Proposition 5.5. NCR does not have a perfect subset.

Proof. Assume $X \subseteq NCR$ is a perfect subset represented by a perfect tree T with [T] = X. We devise a measure μ by setting $\mu[\![\epsilon]\!] = 1$, and define inductively

$$\mu\llbracket\sigma^{\smallfrown}i\rrbracket = \begin{cases} \mu\llbracket\sigma\rrbracket & \text{if } \sigma^{\smallfrown}(1-i) \notin T, \\ \frac{1}{2}\mu\llbracket\sigma\rrbracket & \text{otherwise.} \end{cases}$$

i.e. we distribute the measure uniformly over the infinite paths through T.

Obviously, μ is continuous, and since $\mu(X) = 1$, X must contain a μ -random real. (The set of μ -random reals is always a set of μ -measure 1.)

The Perfect Subset Property refers to the principle that every set in a point class is either countable or contains a perfect subset. The Perfect Subset Property for Π_1^1 is not provable in ZFC. Gödel showed that if V=L, then there exists an uncountable Π_1^1 set without a perfect subset. Mansfield [19] and Solovay [30] showed that any Σ_2^1 set without a perfect subset is contained in the constructible universe L.

Corollary 5.6. NCR is contained in Gödel's constructible universe L.

The upper bound L appears indeed very crude, and an analysis of the proof technique of Theorem 4.4 together with a recent result by Woodin [32] will yield that NCR is countable. Even more, Corollary 5.3 is in certain sense optimal: Every real outside Δ_1^1 is random with respect to some continuous measure.

In the proof of Theorem 4.4, the decisive property which ensured the non-trivial μ -randomness of x was (4.1):

$$\lambda \llbracket \operatorname{Pre}(\sigma) \rrbracket \leq \mu \llbracket \sigma \rrbracket \leq \lambda \llbracket \Psi_{|\sigma|}(\sigma) \rrbracket.$$

Here, the second inequality guarantees that x is not a μ -atom. Although we know $\mu\llbracket\sigma\rrbracket$ will converge to 0 as we consider longer and longer initial segments $\sigma \sqsubseteq x$, this may not to be the case for reals other than x, as the reduction Ψ from x to y is a Turing reduction.

If, however, Ψ is a wtt-reduction, we can modify the construction in the proof of Theorem 4.4 to obtain a continuous measure with respect to which x is random.

Theorem 5.7. Let $x \in 2^{\omega}$. Suppose there exist reals $y, z \in 2^{\omega}$ so that y is Martin-Löf random relative to z and

$$x \equiv_{\text{wtt}(z)} y$$
,

then x is random with respect to a continuous measure.

Proof. The proof is similar to the proof of Theorem 4.4, with one important modification. Suppose Φ and Ψ are with z-functionals such that

$$\Phi(y) = x$$
 and $\Psi(x) = y$.

(In fact, for the proof to work it suffices that Φ is a Turing z-functional.) Let $g: \mathbb{N} \to \mathbb{N}$ be a computable bound on the use of Ψ . Since both x, y are non-computable, we have that g is unbounded. We may assume that g is strictly decreasing. Define a computable function h as

$$h(k) = \max\{m \colon g(m) \le k\}.$$

If, for some $v \in 2^{\omega}$, $\Psi(v)$ is a real, then $\Psi(v \upharpoonright k)$ is a string of length at least h(k) (for all k). Since g is strictly increasing, we have that $h(k+1) \leq h(k) + 1$.

Define the set $Pre^*(\sigma)$ now as

$$\operatorname{Pre}^*(\sigma) = \{ \tau \in 2^{<\omega} : \Phi(\tau) \supseteq \sigma \& \exists s \ (|\Psi_s(\sigma)| \ge h(|\sigma|) \& \Psi_s(\sigma) \sqsubseteq \tau).$$

As before, let Pre be the set of minimal elements of Pre^* with respect to the prefix relation. Note that Pre is c.e. in z, as before.

Condition (4.1) is replaced by

$$\lambda \llbracket \operatorname{Pre}(\sigma) \rrbracket \leq \mu \llbracket \sigma \rrbracket \leq 2^{-h(|\sigma|)},$$

and M is the set of all reals r so that $\rho(r)$ satisfies (5.1) for all $\sigma \in 2^{<\omega}$.

As $h(k) \to \infty$ as $k \to \infty$, it follows that every measure represented in M is continuous. The condition $h(k+1) \le h(k) + 1$ ensures that the proof showing M is non-empty still goes through. A similar argument as in the proof of Theorem 4.4 shows that M is $\Pi_1^0(z)$.

Finally, since $\Psi(x) = y$, $y \in [Pre(x \mid n)]$ for all n. This in turn implies that the final part of the argument, transforming a possible test for x into a test for y (with respect to the accordant measures), goes through as well.

Woodin showed that outside the hyperarithmetical sets, the Posner-Robinson Theorem holds with truth-table equivalence.

Theorem 5.8 (Woodin [32]). If $x \in 2^{\omega}$ is not hyperarithmetic, then there is a $z \in 2^{\omega}$ such that $x \equiv_{\text{tt}(z)} z'$.

Combining Theorems 4.1, 5.8, and 5.7 now yields the desired upper bound for NCR.

Theorem 5.9. If a real x is not hyperarithmetic, then there exists a continuous measure μ such that x is μ -random.

Theorem 5.9 yields an interesting measure-theoretic characterization of Δ_1^1 . The result can also be obtained via a game-theoretic argument using Borel determinacy, along with a generalization of the Posner-Robinson Theorem via Kumabe-Slaman forcing. This is part of a more general argument which shows that for all n, the set NCR_n of reals which are not n-random for some continuous measure is countable. Here n-random means that a test has access to the (n-1)st jump of a representation of the measure. The countability result for NCR_n has an interesting metamathematical twist. This work will be presented in a separate paper [26].

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