

Martin-Löf randomness and Schnorr randomness

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1 Martin-Löf randomness

Remark 1.1. If $X \leq \emptyset'$, then X is not Martin-Löf random. First attempt: The characteristic sequence of the Halting language χ_H is compressible. (Exercise) To compute the first n bits of X , hence, use the oracle machine which computes X together with the use of χ_H which computes $X \upharpoonright n$ (oh but the use of $X \upharpoonright n$ can be much more than 2^n , hence the compressibility of χ_H to logarithmic levels may not compress X .) Second attempt: is there some sequence which, when used as an oracle, gives the first n instances of X within the first $\log n$ bits of the oracle? Maybe Chaitin's Ω ? Can we show that a constructive martingale can win on X ? Mayordomo's idea: $d(X \upharpoonright n)$ is distributed only among those strings which are in the language. Of course, there is this issue of prefix-freeness. X may not be prefix-free.

Definition 1.2. A set $\mathcal{A} \subseteq 2^\omega$ has MEASURE 0 iff $\forall \varepsilon > 0$ there is an open set G_ε sets such that $\mu(G_\varepsilon) < \varepsilon$, and $\mathcal{A} \subseteq \bigcap_\varepsilon G_\varepsilon$.

Example 1.3. A singleton set $\{r\}$ has measure 0, since for every $\varepsilon > 0$, we can construct the sequence $(r - \frac{\varepsilon}{2}, r + \frac{\varepsilon}{2})$. It is possible to extend the idea to show that countably infinite sets have measure 0. Consequently, rationals form a measure 0 set.

Example 1.4. We want to show that the set of all binary sequences in which *some* finite string is absent, has probability 0. This is also called the set of *non-disjunctive sequences*.

Consider the set of all binary sequences which does not contain a specific pattern, say 01. Define, for $i \geq 0$, $S_i : 2^\omega \rightarrow \{0, 1\}$ by

$$S_i^{01}(X) = \begin{cases} 1 & \text{if } X_i X_{i+1} \neq 01 \\ 0 & \text{otherwise.} \end{cases}$$

Example 1.5. which attains 1 exactly on the set of sequences which do not have 01 at the i^{th} position. We want $P[\bigcap_{i \in \mathbb{N}} 1^{-1}(S_i^{01})]$. This is not easy to compute exactly, since S_i s are dependent - for example, 01 occurring in the first position forbids it occurring in the second position. However, it is easy to see that $S_0^{01}, S_2^{01}, S_4^{01}, \dots$ are independent random variables, since $P[S_{2\ell}^{01} = 1 | S_{2j_1}^{01}, \dots, S_{2j_k}^{01}] = P[S_{2\ell}^{01} = 1]$, for any finite collection of even indexed random variables $S_{2j_1}^{01}, \dots, S_{2j_k}^{01}$, where $S_{2\ell}^{01} \notin \{S_{2j_1}^{01}, \dots, S_{2j_k}^{01}\}$. Moreover, $P[\bigcap_i 1^{-1}(S_i^{01})] \leq P[\bigcap_i 1^{-1}(S_{2i}^{01})]$. We have, by independence,

$$P[\bigcap_i 1^{-1}(S_{2i}^{01})] = \prod_i P(1^{-1}(S_{2i}^{01})) = \lim_{n \rightarrow \infty} \prod_{i=0}^n P(1^{-1}(S_{2i}^{01})) = \lim_{n \rightarrow \infty} \left(\frac{3}{4}\right)^n = 0,$$

hence $P[\bigcap_{i \in \mathbb{N}} 1^{-1}(S_i^{01})] = 0$.

Since the number of finite strings is countable, taking a countable union over all such specific finite strings $w \in 2^{<\omega}$, (replacing appropriately, $3/4$ in the above calculation with $(1 - \frac{1}{2^{|w|}})$), we conclude that the set of non-disjunctive sequences, $\bigcup_{w \in 2^{<\omega}} \bigcap_{i \in \mathbb{N}} S_i^w$, has measure 0.

Martin-Löf effectivized the notion of a measure 0 set to define a *constructive* measure 0 set, by requiring first, that there is a uniform enumeration of the open sets G_m . Additionally, the measure of the open sets also decreases in an "effective" manner - we require that the measures of the sets have upper bounds uniformly computable in m , the index of the open set in the sequence.

Definition 1.6. A sequence of open sets $\langle G_m \rangle_{m \in \mathbb{N}}$ is called a MARTIN-LÖF TEST if the sequence is uniformly c.e. and for every $m \in \mathbb{N}$, we have $\mu(G_m) \leq 2^{-m}$. An infinite binary sequence $Z \in 2^\omega$ FAILS the test if $Z \in \bigcap_m G_m$.

Theorem 1.7. *There is a universal Martin-Löf test.*

Proof. Let $\langle G_m^e \rangle_{e, m \in \mathbb{N}}$ be an uniform c.e. enumeration of open sets such that for every $e, m \in \mathbb{N}$, $\mu(G_m^e) \leq 2^{-m}$. Then define, for $b \in \mathbb{N}$, the set $U_b = \bigcup_{e \in \mathbb{N}} G_{e+b+1}^e$. Since it is a c.e. union of uniformly c.e. open sets, each U_b is c.e. open. It is also easy to see that U_b s are uniformly computably enumerable in b . Further, we have

$$\mu(U_b) \leq \sum_{e \in \mathbb{N}} \mu(G_{e+b+1}^e) \leq \sum_{e \in \mathbb{N}} 2^{-(e+b+1)} = 2^{-b}.$$

Now, suppose that $Z \in 2^\omega$ is not MLrandom. Then for some $e \in \mathbb{N}$, $Z \in \bigcap_m G_m^e$. By definition, for each $b \in \mathbb{N}$, $Z \in U_b$, i.e. $Z \in \bigcap_b U_b$. \square

The universal Martin-Löf test defines the *largest* constructive measure 0 set, say \mathcal{S} . Since each test captures some “randomness deficiency”, the set \mathcal{S} is the set of sequences which have randomness deficiency identifiable by a “constructive” test, as realized by a Martin-Löf test. Hence, the complement of \mathcal{S} , the smallest constructive measure 1 set, is the set of all “random” sequences. This set of sequences are now called *Martin-Löf random sequences*.

Definition 1.8. A sequence is MARTIN-LÖF RANDOM if it is an element of the smallest constructive measure 1 set.

Remark 1.9. If $\langle G_m \rangle_{m \in \mathbb{N}}$ is a uniformly c.e. sequence of open sets, then $\bigcap_m G_m$ is a Π_2^0 class. Since there is a universal constructive measure 0 set, it corresponds to a Π_2^0 class as well. Hence the set of Martin-Löf randoms is Σ_2^0 .

Example 1.10. Adapting the estimate in Example 1.4, we can show that the set of disjunctive sequences has constructive measure 0. [Exercise] This shows that every Martin-Löf random is disjunctive - i.e. every finite string appears in every Martin-Löf random.

Example 1.11. A computable $Z \in 2^\omega$ is not MLR. Consider $G_m = [Z \upharpoonright m]$. Then $\langle G_m \rangle_{m \in \mathbb{N}}$ is a Martin-Löf test: it is clear that G_m s are uniformly c.e. Moreover, for every $m \in \mathbb{N}$, $\mu(G_m) \leq 2^{-m}$.

The natural next question is whether we can show that every c.e. sequence is non-random. We introduce the notion of a LEFT-C.E. REAL. A real number r is left-c.e. if its left cut, the set of rationals less than r , is c.e. - i.e. $\{q \in \mathbb{Q} \mid q < r\}$ is c.e.

Let S be an infinite language. Then its characteristic sequence χ_S is defined by $\chi_S[i] = 1$ if the i^{th} string in the standard enumeration is an element of S , otherwise $\chi_S[i] = 0$. If S is computably enumerable, then χ_S is a left-c.e. real - consider, for every $k \in \mathbb{N}$, the set $S \upharpoonright k$ of the set of all strings of length at most k which is accepted within k steps by a fixed machine accepting S . Since S is infinite, it follows that $\chi_{S \upharpoonright k}$ is a rational which is strictly less than χ_S . The sequence $\chi_{S \upharpoonright k}$, $k \in \mathbb{N}$, can be used to show that χ_S is left-c.e.

Now we pose the question: is every left-c.e. real non-random? We expect the answer to be yes, since such sequences are approximable from below by Turing machines, even though the rate of convergence to the limit may not be computable. Surprisingly, however, there are random left-c.e. reals. The most famous such example is Chaitin’s Ω , described in the following example.

Example 1.12. It is not true that every left c.e. real is random. Consider the following sequence, called Chaitin’s Ω :

$$\Omega = \sum_{\substack{p \in \mathcal{P} \\ U(p) \downarrow}} \frac{1}{2^{|p|}}.$$

This is at most 1 by Kraft’s inequality. Moreover, it is left c.e. by a series of approximations $\langle \Omega_s \rangle_{s \in \mathbb{N}}$ which consider the summands corresponding to programs that halt by the s^{th} step.

$$\Omega_s = \sum_{\substack{p \in \mathcal{P} \\ U(p) \downarrow[s]}} \frac{1}{2^{|p|}}.$$

Why is Ω incompressible? It is possible to show that given the first n bits of Ω , we can decide whether all programs of length $\leq n$ halt. [Homework] This then makes it possible to compute the first string x in the standard ordering of strings with $K(x) > n$. But this is possible only for finitely many strings (see below, the discussion on K).

Example 1.12 shows perhaps that the notion of Martin-Lof randomness is not a very “strong” notion of randomness. We will see when we study the interaction of Turing reducibility and Martin-Lof randomness, another sense in which this notion has some weakness.

2 Equivalent characterizations of Martin-Löf randoms

2.1 Characterization using martingales

We defined ML non-randoms using constructive measure 0 sets. Another very useful way to look at ML non-randoms is that there are betting strategies which can succeed in making unbounded amounts of money by betting on them. This approach uses the notion of “martingales”, which are fair betting strategies.

In the following exposition, we will not give the general definition of a martingale, which requires measure theory, and the notion of filtrations of σ -algebras. We will, instead, use the definition specialized for 2^ω , as defined by Schnorr, and independently developed in the works of J. Lutz and others.

Definition 2.1. A MARTINGALE $m : 2^{<\omega} \rightarrow [0, \infty)$ is a function which satisfies the following conditions:

1. [finite initial capital] $m(\lambda) \leq 1$
2. [fairness] for every $w \in 2^{<\omega}$, we have $m(w) = \frac{m(w0) + m(w1)}{2}$.

If condition 2 is replaced by $m(w) \geq \frac{m(w0) + m(w1)}{2}$, then m is called a SUPERMARTINGALE.

The intuition is that a martingale is a “fair betting” strategy, betting on the binary tree. The martingale m starts with a finite initial capital, $m(\lambda)$, which is upper bounded by 1. At any string $w \in 2^{<\omega}$, the martingale m bets on its extensions $w0$ and $w1$. The fairness condition (condition 2) says that the expected amount of money after the next bet, i.e. $\frac{m(w0) + m(w1)}{2}$, is equal to the present capital, $m(w)$. Thus on an average, m neither loses nor wins money.

Of course, this does not prevent m from making unbounded amounts of money on some specific paths along the binary tree, as long as the set of those paths have 0 measure. This observation establishes a connection between measure 0 sets and the success of martingales. We formally define the notion of a martingale succeeding, as follows.

Definition 2.2. A martingale $m : 2^{<\omega} \rightarrow [0, \infty)$ SUCCEEDS on $Z \in 2^\omega$ if

$$\limsup_{n \rightarrow \infty} m(Z \upharpoonright n) = \infty.$$

The following inequality bounds the probability of success of a martingale.

Lemma 2.3. (Kolmogorov inequality) Let $m : 2^{<\omega} \rightarrow [0, \infty)$ be a martingale. Then $\mu(Z \in 2^\omega \mid \exists n m(Z \upharpoonright n) > N) < \frac{1}{N}$.

Definition 2.4. The above notion is classical, and we now impose computability restrictions on it. Contrary perhaps to our expectation, we do not insist that the martingale is computable, but only that there are computable approximations to the value from below.

Definition 2.5. A martingale $m : 2^{<\omega} \rightarrow [0, \infty)$ is called LOWER SEMICOMPUTABLE (or CONSTRUCTIVE) if there is a total computable function $\hat{m} : 2^{<\omega} \times \mathbb{N} \rightarrow [0, \infty) \cap \mathbb{Q}$ such that the following conditions hold.

1. [monotonicity from below] For every $w \in 2^{<\omega}$ and every $n \in \mathbb{N}$, we have $\hat{m}(w, n) \leq \hat{m}(w, n+1) \leq m(w)$.
2. [(non-effective) convergence] For every $w \in 2^{<\omega}$, we have $\lim_{n \rightarrow \infty} \hat{m}(w, n) = m(w)$.

Lemma 2.6. For every lower semicomputable martingale $m : 2^\omega \rightarrow [0, \infty)$, there is a Martin-Löf test $\langle G_i \rangle_{i \in \mathbb{N}}$ such that every infinite binary sequence on which m succeeds, is in $\bigcap_{i \in \mathbb{N}} G_i$. Conversely, for every Martin-Löf test $\langle G_i \rangle_{i \in \mathbb{N}}$, there is a lower semicomputable martingale $m : 2^\omega \rightarrow [0, \infty)$ which succeeds on every infinite sequence in $\bigcap_{i \in \mathbb{N}} G_i$.

Proof. Denote, for every $x \in 2^{<\omega}$, the set of all infinite binary sequences with x as a prefix, by $[x]$.

Let $\langle G_i \rangle_{i \in \mathbb{N}}$ be the universal Martin-Löf test. For each $i \in \mathbb{N}$, define $m_i : 2^{<\omega} \rightarrow [0, \infty)$ by

$$m_i(x) = \mu(G_i \cap [x])2^{|x|}.$$

Then $m_i(\lambda) \leq 2^{-i}$, and for every $x \in 2^{<\omega}$, we have

$$\frac{m(x0) + m(x1)}{2} = \frac{\mu(G_i \cap [x0])2^{|x0|} + \mu(G_i \cap [x1])2^{|x1|}}{2} = \frac{\mu(G \cap [x0]) + \mu(G \cap [x1])}{2} 2^{|x|+1} = \frac{\mu(G \cap [x])}{2} 2^{|x|+1} = m(x),$$

where the third equality follows since μ is a probability measure. Hence m_i is a martingale.

Note that for every $X \in 2^\omega$, for every prefix x of X inside G_i , we have $m_i(x) = \mu(G_i \cap [x])2^{|x|} = \mu([x])2^{|x|} = 1$.

Now, define the function $m : 2^\omega \rightarrow [0, \infty)$ by $m = \sum_{i \in \mathbb{N}} m_i$. We can easily verify that m is a lower semicomputable martingale. Also, if $X \in \cap_{i \in \mathbb{N}} G_i$. Then, for each $i \in \mathbb{N}$, we conclude that $\limsup_{n \rightarrow \infty} m(x) = \infty$.

Conversely, let $m : 2^\omega \rightarrow [0, \infty)$ be a lower semicomputable martingale. Then, for every $i \in \mathbb{N}$ consider the set $G_i = \{Z \in 2^\omega \mid \exists n m(Z \upharpoonright n) > 2^i\}$. It is easy to verify that G_i s are uniformly c.e. open in i . Moreover, by the Kolmogorov inequality, $\mu(G_i) \leq 2^{-i}$. If $\limsup_n m(Z \upharpoonright n) = \infty$, then $Z \in \cap_{i \in \mathbb{N}} G_i$. \square

Corollary 2.7. *There is a universal semicomputable martingale.*

Proof. This is the martingale which corresponds to the universal MLtest. \square

Since a sequence is Martin-Löf random if and only if it fails the universal MLtest, we have the following.

Theorem 2.8. *X is Martin-Löf random if and only if the universal constructive martingale fails on it.*

2.2 Characterization using incompressibility

Let $\mathcal{P} \subseteq 2^{<\omega}$ be a prefix-free set, i.e. if a string x is in \mathcal{P} , then no proper prefix of x , or a proper extension of x can be in \mathcal{P} .

Example 2.9. The set $\{0^n 1 \mid n \in \mathbb{N}\}$ is an infinite prefix-free set.

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Prefix-free sets are quite sparse. The following lemma is an important property of prefix-free sets, and captures a sense in which they are sparse.

Lemma 2.10. (*Kraft inequality*) *If $\mathcal{P} \subseteq 2^\omega$ is a prefix-free set, then we have $\sum_{x \in \mathcal{P}} 2^{-|x|} \leq 1$.*

Proof. Consider the following experiment: we toss an unbiased fair coin multiple times, marking the outcome as 1 if Heads, and 0 if tails until we either hit an element in \mathcal{P} and stop, or we toss forever. Then the probability of hitting an $x \in \mathcal{P}$ is exactly $2^{-|x|}$. By the prefix-free property, along any one sequence of trials, we can hit at most one element of \mathcal{P} . Thus, the probability that our experiment halts and produces some element of \mathcal{P} is exactly $\sum_{x \in \mathcal{P}} 2^{-|x|}$. Since this is the probability of an event in a well-defined probability space, it is at most 1. \square

Fix a prefix-free machine M . Consider the following cylinder sets:

$$R_b^M = [\{x \in 2^{<\omega} \mid K_M(x) \leq |x| - b\}].$$

This is the set of all infinite binary sequences with some b -compressible prefix.

Lemma 2.11. $\langle R_b^M \rangle_{b \in \mathbb{N}}$ is a Martin-Löf test.

Proof. The condition $K_M(x) \leq |x| - b$ is equivalent to checking that there is some prefix-free program σ such that $M(\sigma) = x$, and $|\sigma| \leq |x| - b$. Hence the sets R_b^M are c.e. uniformly in b .

Now we show that $\mu(R_b^M) \leq 2^{-b}$. Consider $S_b^M = \{x \in 2^{<\omega} \mid K_M(x) \leq |x| - b\}$, and let $V_b^M \subset S_b^M$ be the subset of strings which are minimal under the prefix ordering - that is, if x, y are both in S_b^M and y extends x , then $y \notin V_b^M$. Then

$$\sum_{x \in V_b^M} \frac{1}{2^{|x|}} = \mu(R_b^M).$$

Let $x \in V_b^M$ and let σ_x be a shortest M -description of x . Since $|\sigma_x| \leq |x| - b$, we have $2^{-|\sigma_x|} \geq 2^b 2^{-|x|}$. Then

$$1 \geq \sum_{x \in V_b^M} \frac{1}{2^{|\sigma_x|}} \geq 2^b \sum_{x \in V_b^M} \frac{1}{2^{|x|}} = 2^b \mu(R_b^M),$$

from which it follows that $\mu(R_b^M) \leq 2^{-b}$. □

The following result relies on a technique called Levin's coding theorem. See Nies 09 for an exposition. We skip the proof of Levin's coding theorem since it is technical.

Theorem 2.12. *A sequence $X \in 2^\omega$ is Martin-Löf random iff $\exists b \forall n K(X \upharpoonright n) > n - b$, equivalently, $X \notin R_b^M$.*

Proof. By the previous lemma, $\langle R_b^M \rangle_{b \in \mathbb{N}}$ is a Martin-Löf test. Hence, if there is a b such that $X \notin R_b^M$, then X is MLrandom.

Now, suppose $\langle G_m \rangle_{m \in \mathbb{N}}$ is a Martin-Löf test and $X \in \bigcap_m G_m$. We can assume that $\mu(G_m) \leq 2^{-2m}$.

We form a BOUNDED REQUEST SET L .

We obtain, uniformly in m , an antichain $\langle x_i^m \rangle_{i < N_m}$, such that $G_m = [\{x_i^m \mid i < N_m\}]$.

Let

$$L = \{(x_i^m, |x_i^m| - m + 1) \mid m \in \mathbb{N}, i < N_m\}.$$

Since $\mu(G_m) \leq 2^{-2m}$, the contribution of G_m to L is at most $2^{-2m+m-1} = 2^{-m-1}$, hence L is a bounded request set.

Let M_d be the prefix-free machine for L given by Levin's coding theorem. Fix $b \in \mathbb{N}$ and let $m = b + d + 1$. Since $X \in G_m$, we have a prefix x_i^m of X for some i . Thus, $K(x_i^m) \leq |x_i^m| - m + 1 + d = |x| - b$. □