

Outline of Lecture 1

Martin-Löf tests and martingales

- The Cantor space.
- Lebesgue measure on Cantor space.
- Martin-Löf tests.
- Basic properties of random sequences.
- Betting games and martingales.
- Equivalence of Martin-Löf tests and effective martingales.
- Alternative randomness concepts.

Cantor Space

We will study randomness for **infinite binary sequences**.

Cantor space: set of all such sequences, denoted by $2^{\mathbb{N}}$.

Ways to interpret sequences $X \in 2^{\mathbb{N}}$:

- sets of natural numbers, $S_X = \{n \in \mathbb{N} : X(n) = 1\}$,
- real numbers in $[0, 1]$, $\alpha_X = \sum_n X(n)2^{-n}$.

Metric

$$d(X, Y) = \begin{cases} 2^{-N(X, Y)} & \text{if } X \neq Y \\ 0 & \text{if } X = Y. \end{cases}$$

where $N(X, Y) = \min\{n : X(n) \neq Y(n)\}$.

Cantor Space

Topological properties of $2^{\mathbb{N}}$

- compact
- perfect
- totally disconnected

$2^{\mathbb{N}}$ has a countable basis of **clopen sets**, the so-called **cylinder sets**

$$[[\sigma]] = \{X : X \upharpoonright_n = \sigma\},$$

where σ is a finite binary sequence (**string**) and $X \upharpoonright_n$ denotes the first n bits of X .

The open subsets of $2^{\mathbb{N}}$ are unions of cylinder sets. They can be represented by a set $W \subseteq 2^{<\mathbb{N}}$. We write $[[W]]$ to denote the open set induced by W .

Lebesgue Measure on Cantor Space

Over \mathbb{R} : **Lebesgue measure** λ unique Borel measure that is translation invariant and assigns every interval (a, b) measure $|b - a|$.

Over $2^{\mathbb{N}}$:

- **Diameter** of a basic open cylinder $[[\sigma]]$ is $2^{-|\sigma|}$.

Hence we will set $\lambda[[\sigma]] = 2^{-|\sigma|}$.

Some basic results of measure theory ensure that λ can be uniquely extended to all Borel sets.

- *We will return to this in more detail in Lecture 4.*

Lebesgue Measure on Cantor Space

Alternative view of Lebesgue measure:

- $X \mapsto \alpha_X = \sum_n X(n)2^{-n}$ yields a **surjection** of $2^{\mathbb{N}}$ onto $[0, 1]$.
- The image of $[[\sigma]]$ is the **dyadic interval**

$$\left[\sum_{k=0}^{n-1} \sigma(k)/2^{k+1}, 2^n + \sum_{k=0}^{n-1} \sigma(k)/2^{k+1} \right].$$

- The Lebesgue measure (in \mathbb{R}) of this interval is 2^{-n} .

Lebesgue Measure on Cantor Space

Yet another view:

- $X \in 2^{\mathbb{N}}$ represents outcome of an infinite sequence of coin tosses – 0 is Heads, 1 is Tails.
- If the coin is fair, each outcome has probability $1/2$.
- A finite string σ represents the outcome of a finite number of independent coin tosses.
- The probability of outcome σ is $(1/2)^{|\sigma|}$.

Nullsets

Nullsets are sets that are measure theoretically small, just as countable sets are small with respect to cardinality.

Intuitively, a nullset is a set that can be covered by open sets of arbitrary small measure.

Definition

A subset $A \subseteq 2^{\mathbb{N}}$ is a **nullset** for Lebesgue measure (or has Lebesgue measure zero) if for every $\varepsilon > 0$ there exists an open set $U = \bigcup_{\sigma \in W} [\sigma]$ such that

$$A \subseteq U \quad \text{and} \quad \sum_{\sigma \in W} \lambda[\sigma] = \sum_{\sigma \in W} 2^{-|\sigma|} < \varepsilon.$$

Nullsets

To define Martin-Löf tests, it is convenient to reformulate this a little.

Proposition

A set $A \subseteq 2^{\mathbb{N}}$ is a nullset iff there exists a set $W \subseteq \mathbb{N} \times 2^{<\mathbb{N}}$ such that, if we let $W_n = \{\sigma : (n, \sigma) \in W\}$, for all $n \in \mathbb{N}$,

$$A \subseteq \bigcap_n [W_n] \quad \text{and} \quad \sum_{\sigma \in W_n} 2^{-|\sigma|} < 2^{-n}.$$

$\bigcap_n [W_n]$ is itself a nullset. It is an intersection of a sequence of open sets. Such sets are called G_δ or Π_2^0 -sets.

\Rightarrow Every nullset is contained in a G_δ nullset.

Nullsets

Remarks

- We can always assume the sequence (W_n) is **nested**.
(Why?)
- G_δ sets can be easily effectivized. What ‘codes’ a G_δ set in Cantor space is a subset of $\mathbb{N} \times 2^{<\mathbb{N}}$.
- On such sets we can easily impose **definability/effectivity conditions**, e.g. require that they are **recursively enumerable**.

Martin-Löf Tests and Randomness

Definition

- A **Martin-Löf (ML) test** (for Lebesgue measure) is a recursively enumerable set $W \subseteq \mathbb{N} \times 2^{<\mathbb{N}}$ such that, if we let $W_n = \{\sigma : (n, \sigma) \in W\}$, for all $n \in \mathbb{N}$,

$$\sum_{\sigma \in W_n} 2^{-|\sigma|} < 2^{-n}.$$

- A set $A \subseteq 2^{\mathbb{N}}$ is **Martin-Löf null** if it is covered by a Martin-Löf test, i.e. if there exists a Martin-Löf test W such that $A \subseteq \bigcap_n [W_n]$.
- A sequence $X \in 2^{\mathbb{N}}$ is **Martin-Löf random** if $\{X\}$ is not Martin-Löf null.

Existence of Random Sequences

- Every ML-test W describes a G_δ nullset, with the additional requirement that it is effectively presented (W is r.e.).
- There are only countably many r.e. sets, and hence only countably many ML-tests.
- Being random means not being contained in the union of all G_δ sets defined by any ML test.
- A basic result of measure theory says that a countable union of nullsets is again a nullset (the standard “ $\varepsilon/2^n$ -proof”).
- Therefore, the set of all non-random sequences is a nullset, and consequently, λ -almost every sequence is ML random.

Universal Tests

In the last argument, we used that a **countable union of nullsets is a nullset**.

It turns out that even more is true: The union of all ML-tests is again a ML-test, a **universal test**.

- There exists a ML-test (U_n) such that X is ML-random iff X is not covered by (U_n) .
- In other words, the ML-random sequences are precisely the ones in the complement of $\bigcap_n [U_n]$.
- The ML-random sequences form the **largest** effective (in the sense of Martin-Löf) set of measure 1.

Universal Tests

Construction of a universal test

- Start uniformly enumerating all r.e. subsets $W^{(e)}$ of $\mathbb{N} \times 2^{<\mathbb{N}}$.
- Once we see that the measure condition of some $W_n^{(e)}$ is violated, we stop enumerating it.
- Given a uniform enumeration of all tests $(\tilde{W}_n^{(e)})$ (with possible repetitions), we can define a universal test (U_n) by letting

$$U_n = \bigcup_e \tilde{W}_{n+e+1}^{(e)}$$

Note that this test has the nice property that it is **nested**, i.e. $\llbracket U_{n+1} \rrbracket \subseteq \llbracket U_n \rrbracket$. We will always assume this from now on.

Later we will encounter other ways to define universal tests.

Basic Properties of Random Sequences

- The set of Martin-Löf random reals is **invariant under prefix operations** (adding, deleting, replacing a finite prefix).
- If $Z \subseteq \mathbb{N}$ is **computably enumerable**, then the sequence given by the characteristic function of Z is **not Martin-Löf random**.
- **Any finite string appears** somewhere in a Martin-Löf random real, in fact it appears **infinitely often** in a Martin-Löf random real.
- For every Martin-Löf random sequence $X \in 2^{\mathbb{N}}$,

$$\lim_n \frac{\sum_{k=0}^{n-1} X(k)}{n} = \frac{1}{2}.$$

These assertions can be proved directly by defining a suitable test. (Exercise!) But we will prove different characterizations of random sequences which may make this easier.

Betting Games and Martingales

Betting strategies

A **betting strategy** b is a function $b : 2^{<\mathbb{N}} \rightarrow [0, 1] \times \{0, 1\}$.

Interpretation:

- A string σ represents the outcomes of a 0-1-valued (infinite) process (e.g. a coin toss).
- $b(\sigma) = (\alpha, i)$ then tells the gambler on which outcome to bet next, i , and what percentage of his current capital to bet on this outcome, α .
- When the next bit of the process is revealed and agrees with i , the capital is multiplied by $(1 + \alpha)$. If it is different from i , the gambler loses his bet, i.e. his capital is multiplied by $(1 - \alpha)$.

Betting Games and Martingales

We can keep track of the player's capital through a function $F : 2^{<\mathbb{N}} \rightarrow [0, \infty)$.

F satisfies

$$F(\sigma) = \frac{F(\sigma 0) + F(\sigma 1)}{2} \quad \text{for all } \sigma. \quad (*)$$

This reflects the property that the game is **fair** – the expected value of the capital after the next round is the same as the player's capital before he makes his bet.

Any function satisfying $(*)$ is called a **martingale**.

Given a martingale, we can reconstruct the accordant betting function from it.

Betting Games and Martingales

Successful martingales

A martingale is **successful** on an infinite sequence X if

$$\limsup_{n \rightarrow \infty} F(X \upharpoonright_n) = \infty,$$

We can actually replace \limsup by \lim :

- For every martingale F there exists a martingale G such that for all X ,

$$\limsup_n F(X \upharpoonright_n) = \infty \quad \text{implies} \quad \lim_n G(X \upharpoonright_n) = \infty.$$

(Set some money aside regularly.)

Betting Games and Martingales

A martingale succeeds only on very few sequences.

Martingale Convergence Theorem [Ville, Doob]

For any martingale F , the set of sequences $X \in 2^{\mathbb{N}}$ such that

$$\limsup_{n \rightarrow \infty} F(X \upharpoonright_n) = \infty \tag{1}$$

has λ -measure zero.

We will prove an effective version of this theorem.

From ML-tests to Martingales

Goal: Given a ML-test (\mathcal{U}_n) , define a martingale succeeding on the sequences covered by (\mathcal{U}_n) .

Basic Idea: Whenever a string appears at level n of the test, F reaches a value of at least n .

- For a single string σ , define the following martingale.

$$F_{\sigma}(\tau) = \begin{cases} 2^{-(|\sigma|-|\tau|)} & \text{if } \tau \subseteq \sigma, \\ 1 & \text{if } \tau \supseteq \sigma, \\ 0 & \text{otherwise.} \end{cases}$$

- F_{σ} starts out with a capital of $2^{-|\sigma|}$ and doubles its capital every step along σ , then stops betting.
- If an outcome is not compatible with σ , its capital is lost.

From ML-tests to Martingales

- Now, for one level \mathcal{U}_n of the ML-test, blend the individual “string”-martingales into one.
- If (F_n) is a sequence of martingales and $\sum_n F(\epsilon) < \infty$, then

$$F = \sum_n F_n$$

is a martingale.

- Hence define

$$F_n(\tau) = \sum_{\sigma \in \mathcal{U}_n} F_\sigma(\tau).$$

and check that the sum of the $F_\sigma(\epsilon)$ is finite.

- $F_\sigma(\epsilon) = 2^{-|\sigma|}$.
- Hence $F_n(\epsilon) = \sum_{\sigma \in \mathcal{U}_n} F_\sigma(\epsilon) = \sum_{\sigma \in \mathcal{U}_n} 2^{-|\sigma|} \leq 2^{-n}$.

From ML-tests to Martingales

- The inequality $F_n(\epsilon) \leq 2^{-n}$ further lets us combine the martingales for each \mathcal{U}_n into one martingale F ,

$$F(\sigma) = \sum_n F_n(\sigma).$$

- If $X \in \bigcap_n \llbracket \mathcal{U}_n \rrbracket$, there exists a sequence (σ_n) such that for all n , $\sigma_n \in \mathcal{U}_n$ and $\sigma_n \subset X$.
- It follows that $F_n(\sigma_n) \geq 1$.
- More importantly, by the definition of F_n , $F_n(\tau) \geq 1$ for all $\tau \supseteq \sigma$, hence in particular for all σ_m where $m \geq n$.
- It follows that for all n , $F(\sigma_n) \geq \sum_{k=1}^n F_k(\sigma_n) \geq n$, that is, F is unbounded along X .

Left-enumerable Martingales

What is the **computational complexity** of F ?

- A function $F : 2^{<\mathbb{N}} \rightarrow \mathbb{R}$ is **enumerable from below** or **left-enumerable** if there exists, uniformly in σ , a recursive nondecreasing sequence $(q_k^{(\sigma)})$ of rational numbers such that $q_k^{(\sigma)} \rightarrow F(\sigma)$.
- Equivalently, the **left cut** of $F(\sigma)$ is uniformly enumerable, i.e. the set

$$\{(q, \sigma) : q < F(\sigma)\}$$

is recursively enumerable.

It is not hard to see that F defined above is left-enumerable.

Left-enumerable Martingales

We have proved the following:

For any ML-test (U_n) there exists a left-enumerable martingale F such that if $X \in \bigcap_n [U_n]$, then F succeeds on X .

In other words, if X is not ML-random, we can find a left-enumerable martingale that succeeds on X .

From Martingales to ML-Tests

Does a converse of this hold? Can we transform a left-enumerable martingale F into a ML-test?

Basic idea: Whenever F first reaches a capital of 2^n on some string σ , enumerate σ into U_n .

- Since F is enumerable from below, this is an r.e. event.
- We only have to make sure that there are not too many such σ .
- This is guaranteed by **Kolmogorov's inequality** (actually due to Ville).
 - Suppose F is a martingale. For any string σ and any prefix-free set W of strings extending σ ,

$$F(\sigma) \geq \sum_{\tau \in W} 2^{|\sigma| - |\tau|} F(\tau)$$

- **Prefix-free:** No two strings are comparable by \subseteq .

From Martingales to ML-Tests

From Kolmogorov's inequality we get the desired result:

Given a martingale F , let $C_k(F) = \{\sigma : F(\sigma) \geq k\}$. Then

$$\lambda[C_k(F)] \leq F(\epsilon)/k.$$

- Let W be a prefix-free set such that $[[W]] = [[C_k(F)]]$. (This can be found effectively.)
- Then $\lambda[C_k(F)] = \lambda[W] = \sum_{\tau \in W} 2^{-|\tau|}$.
- By Kolmogorov's inequality,
 $F(\epsilon) \geq \sum_{\tau \in W} 2^{-|\tau|} F(\tau) \geq \sum_{\tau \in W} 2^{-|\tau|} k$.
- Hence $\lambda[C_k(F)] \leq F(\epsilon)/k$, as required.

From Martingales to ML-Tests

We have proved the **first main theorem** of algorithmic randomness, due to Schnorr and independently Levin.

Theorem

A sequence X is **ML-random** if and only if no **left-enumerable martingale succeeds** on it.

Alternative Randomness Concepts

Of course, ML-tests are not the only possible way to effectivize nullsets.

ML-randomness is the most prominent concept because it shows a rather strong **robustness with respect to the different approaches**.

We will briefly discuss a few other notions – some based on martingales, others based on tests.

Alternative Randomness Concepts

Test-based concepts

- Weak 2-randomness
- Schnorr randomness

Martingale-based concepts

- Computable randomness
- Resource-bounded randomness

Weak 2-Randomness

Martin-Löf test has to fulfill **two effectivity requirements**.

- uniform recursive enumerability of (W_n) ,
- measure of the W_n converges to 0 **effectively**, $\lambda[W_n] \leq 2^{-n}$.

For a **weak 2-test** we only require that (W_n) is uniformly r.e. and that $\lambda \bigcap_n [W_n] = 0$.

One can show that weak 2-randomness is **strictly stronger** than ML-randomness. There exists an X that is ML-random but not weak 2-random.

We will encounter such an example later.

Schnorr Randomness

On the other hand, one might argue that the effectivity requirement for ML-tests is **too weak**. Test should be **computable** in some form, not merely r.e.

Schnorr tests

A ML-test (W_n) is a **Schnorr test** if the real number

$$\sum_{\sigma \in W_n} 2^{-|\sigma|}$$

is computable uniformly in n .

A real number α is **computable** if there exists a computable function $g : \mathbb{N} \rightarrow \mathbb{Q}$ such that for all n , $|\alpha - g(n)| \leq 2^{-n}$.

Note: If (W_n) is a Schnorr test then the sets W_n are uniformly computable.

Computable Randomness

The same criticism applies to the martingale characterisation of randomness. Betting strategies should be computable [Schnorr].

Definition

A sequence X is **computably random** if no computable martingale succeeds on it.

A function $F : 2^{<\mathbb{N}} \rightarrow \mathbb{R}$ is **computable** if there exists a computable function $g : 2^{<\mathbb{N}} \times \mathbb{N} \rightarrow \mathbb{Q}$ such that for all σ, n ,
 $|F(\sigma) - g(\sigma, n)| \leq 2^{-n}$.

One can refine the computability requirement even further, by imposing a time-bound on F . This leads to the theory of **resource-bounded measure**, which has successfully been used in **computational complexity**.

Relations between Randomness Concepts

The following **strict implications** hold:

