

Luzin's (N) and randomness reflection

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We show that a computable function $f : \mathbb{R} \rightarrow \mathbb{R}$ has Luzin's property (N) if and only if it reflects Π_1^1 -randomness, if and only if it reflects $\Delta_1^1(\mathcal{O})$ -randomness, and if and only if it reflects \mathcal{O} -Kurtz randomness, but reflecting Martin-Löf randomness or weak-2-randomness does not suffice. Here a function f is said to reflect a randomness notion R if whenever $f(x)$ is R -random, then x is R -random as well. If additionally f is known to have bounded variation, then we show f has Luzin's (N) if and only if it reflects weak-2-randomness, and if and only if it reflects \emptyset' -Kurtz randomness. This links classical real analysis with algorithmic randomness.

1 Introduction

We revisit a notion from classic real analysis, namely Luzin's property (N), from the perspective of computability theory. A function $f : \mathbb{R} \rightarrow \mathbb{R}$ has Luzin's (N), if the image of any (Lebesgue) null set under f has again measure 0. This concept was studied extensively by Luzin in his thesis [13]. For functions with bounded variation, this notion is just equivalent to absolutely continuous functions – but already for general continuous functions, Luzin's (N) is a somewhat intricate property. A formal result amounting to this was obtained by Holický, Ponomarev, Zajíček and Zelený, showing that the set of functions with Luzin's (N) is Π_1^1 -complete in the space of continuous functions [9].

From a computability-theoretic perspective, Luzin's (N) is readily seen to be some kind of randomness reflection: By contraposition, it states that whenever $f[A]$ has positive measure (i.e. contains a random point for a suitable notion of randomness), then A has positive measure, too (i.e. contains a random point). It thus seems plausible that for some suitable randomness notion, Luzin's (N) for computable functions is equivalent to saying that whenever $f(x)$ is random, then so is x . Our main finding (Theorem 16) is that this is indeed the case, and that Π_1^1 -randomness is such a suitable randomness notion. An indication that this is a non-trivial result is that our proof uses ingredients such as Friedman's conjecture (turned into a theorem by Martin [8, 14, 28]).

While the exploration of how randomness interacts with function application, and the general links to real analysis, has a long tradition (see e.g. the survey by Rute [21]), the concepts of randomness preservation (if x is random, so is $f(x)$) and no-randomness-from-nothing (if y is random, then there is some random $x \in f^{-1}(y)$) have received far more attention than randomness reflection. Our results not only fill this gap, but may shed a light on why randomness reflection has been less popular: As the most natural notion of randomness reflection turns out to be Π_1^1 -randomness reflection, we see that studying higher randomness is essential for this endeavour.

Our theorems and proofs generally refer to computability. However, we stress that since the results relativize, one can obtain immediate consequences in classic real analysis. An example of this is Corollary 17, which recovers a theorem by Banach. More such examples can be found in Section 8, where, by applying relativized computability method, we are able to prove some results in classical analysis. While we are not aware of such consequences that would advance the state of the art in real analysis, it is plausible that future use of our techniques could accomplish this.

Overview of our paper In Section 2 we do not discuss randomness reflection at all, but rather prove a result in higher randomness of independent interest. Theorem 1 is of the form “if a somewhat random X is hyp-computed by a very random Y , then X is already very random”. It is the higher randomness analog of [15, Theorem 4.3] by Miller and the third author. This result is a core ingredient of our main theorem.

Section 3.1 contains the main theorem of our paper, the equivalence of Luzin’s (N) for computable functions with Π_1^1 -randomness reflection and with $\Delta_1^1(\mathcal{O})$ -randomness reflection. We consider higher Kurtz randomness in Section 3.3, and show that for continuous functions $f : \mathbb{R} \rightarrow \mathbb{R}$, Luzin’s (N) is equivalent to the reflection of \mathcal{O} -Kurtz randomness, and separate this from Δ_1^1 -Kurtz randomness reflection. In Subsection 3.4 we discuss the open questions raised by our main theorem: Just because Luzin’s (N) is equivalent to reflection of several higher randomness notions does not mean that it cannot be also equivalent to randomness reflection for some “lower” notions. For Martin-Löf-randomness reflection and weak-2-randomness reflection, we provide a separation from Luzin’s (N) in Section 4.

In Sections 5 and 6 we consider Luzin’s (N) for more restricted classes of functions, namely functions with bounded variation and strictly increasing functions. Here Luzin’s (N) turns out to be equivalent to weak-2-randomness reflection, but we can still separate it from several other randomness-reflection-notion. These investigations tie in to a project by Bienvenu and Merkle [2] regarding how two computable measures being mutually absolutely continuous (i.e. having the same null sets) relates to randomness notions for these measures coinciding.

In Section 7 we take a very generic look at the complexity of randomness reflection, and show that the Π_1^1 -hardness established for Luzin’s (N) in [9] applies to almost all other randomness reflection notions, too.

Section 8 contains a brief digression about functions where the image of null sets is small in some other sense (countable or meagre). We prove these classical analysis results via various classical and higher computability methods.

We then conclude in Section 9 with a discussion of how this line of investigation could be continued in the future.

2 Randomness and hyperarithmetic reductions

Throughout, we assume familiarity with the theory of algorithmic randomness and higher randomness in particular. A standard references for the former are [6] and [17]. For the latter, readers may refer to [5]. We use standard computability-theoretic notation. The Lebesgue measure is denoted by λ .

Our goal in this section is to establish the following:

Theorem 1. Let Y be $\Delta_1^1(Z)$ -random and Π_1^1 -random, let X be Δ_1^1 -random and let $X \leq_h Y$. Then X is $\Delta_1^1(Z)$ -random.

This is a higher-randomness counterpart to [15, Theorem 4.3], and the proof proceeds by adapting both this and [15, Lemma 4.2]. We will use the theorem in the following form:

Corollary 2. Let $Z \geq \mathcal{O}$. If X is Δ_1^1 -random, Y is $\Delta_1^1(Z)$ -random and $X \leq_h Y$, then X is $\Delta_1^1(Z)$ -random.

Lemma 3. Fix $\alpha \in \mathcal{O}$ and $e \in \mathbb{N}$. If X is Δ_1^1 -random, then:

$$\exists c \forall n \lambda(\{Y \mid \varphi_e(Y^{(\alpha)}) \in [X_{\upharpoonright n}]\}) < 2^{-n+c}$$

Proof. Analogous to the proof of [15, Lemma 4.2]. Let $\mathcal{H}_\sigma = \{Y \mid \varphi_e(Y^{(\alpha)}) \in [\sigma]\}$, and then let $F_i = \{\sigma \mid \lambda(\mathcal{H}_\sigma) > 2^{-|\sigma|+i}\}$. By construction, the \mathcal{H}_σ are uniformly $\Delta_{\alpha+1}^0$ (as subsets of $\{0, 1\}^{\mathbb{N}}$), and so the sets F_i are uniformly $\Delta_{\alpha+2}^0$ (as subsets of $2^{<\omega}$).

A counting argument shows that $\lambda([F_i]) < 2^{-i}$: Pick a prefix free set $D \subseteq F_i$ with $[D] = [F_i]$. Then:

$$1 \geq \lambda\left(\bigcup_{\sigma \in D} \mathcal{H}_\sigma\right) = \sum_{\sigma \in D} \lambda(\mathcal{H}_\sigma) \geq \sum_{\sigma \in D} 2^{-|\sigma|+i} = 2^i \lambda([F_i])$$

We see that $([F_i])_{i \in \mathbb{N}}$ is a Martin-Löf test relative to $\emptyset^{\alpha+2}$. Since X is Δ_1^1 -random, there has to be some $c \in \mathbb{N}$ with $X \notin [F_c]$. This in turn means that $\forall n \in \mathbb{N} X_{\upharpoonright n} \notin F_c$, which by definition of F_c is the desired claim. \square

Fact 4 (Sacks [22]). $\Delta_1^1(Z)$ -randomness (defined by being contained in no $\Delta_1^1(Z)$ -null sets) is equivalent to being \hat{Z} -random for every $\hat{Z} \in \Delta_1^1(Z)$.

Lemma 5. Fix $\alpha \in \mathcal{O}$ and $e \in \mathbb{N}$. If $X = \varphi_e(Y^{(\alpha)})$, X is Δ_1^1 -random and Y is $\Delta_1^1(Z)$ -random, then X is $\Delta_1^1(Z)$ -random.

Proof. We follow the proof of [15, Theorem 4.3]. Let c be the constant guaranteed for X by Lemma 3. As in the proof of Lemma 3, let $\mathcal{H}_\sigma = \{W \mid \varphi_e(W^{(\alpha)}) \in [\sigma]\}$. Let $\mathcal{G}_\sigma = \mathcal{H}_\sigma$ if $\lambda(\mathcal{H}_\sigma) < 2^{-|\sigma|+c}$ and $\mathcal{G}_\sigma = \emptyset$ else. Note that \mathcal{G}_σ is still uniformly Δ_1^1 . The choice of c in particular guarantees that $Y \in \mathcal{G}_{X_{\upharpoonright n}}$ for each $n \in \mathbb{N}$.

Let $\cap_n U_n$ denote a Martin-Löf test relative to \hat{Z} for some $\hat{Z} \in \Delta_1^1(Z)$. By Fact 4, it suffices to show that $X \notin \cap_n U_n$. We set $\mathcal{K}_i = \bigcup_{\sigma \in U_{c+i}} \mathcal{G}_\sigma$, and $\mathcal{K} = \bigcap_{i \in \mathbb{N}} \mathcal{K}_i$. We find that \mathcal{K} is $\Delta_1^1(Z)$. Moreover, we have that:

$$\lambda(\mathcal{K}_i) \leq \sum_{\sigma \in U_{c+i}} \lambda(\mathcal{G}_\sigma) \leq \sum_{\sigma \in U_{c+i}} 2^{-|\sigma|+c} \leq 2^{-i}$$

Hence, it follows that $\lambda(\mathcal{K}) = 0$, so for some i , $Y \notin \mathcal{K}_i$. Then $X \notin U_{c+i}$, because $Y \in \mathcal{G}_\sigma$ for all $\sigma \prec X$. \square

Fact 6 (Sacks [22]). If Y is Π_1^1 -random, then $\omega_1^{\text{CK}} = \omega_1^{\text{CK}, Y}$.

Proof of Theorem 1. Since Y is Π_1^1 -random, we know that $X \leq_h Y$ implies the existence of some $\alpha \in \mathcal{O}$ and $e \in \mathbb{N}$ such that $X = \varphi_e(Y^{(\alpha)})$ (rather than merely $\alpha \in \mathcal{O}^Y$). We can thus invoke Lemma 5 to conclude that X is $\Delta_1^1(Z)$ -random. \square

As an aside, the requirement in Theorem 1 that Y be Π_1^1 -random might be unexpected at first – it has no clear counterpart in [15, Theorem 4.3]. The following example, which is not needed for anything else in the paper, shows that this assumption is necessary.

Example 7. There are Δ_1^1 -random X and $\Delta_1^1(Z)$ -random Y with $X \leq_h Y$ but X is not $\Delta_1^1(Z)$ -random. In fact, we shall chose $X = Z$, and make X even Π_1^1 -random.

Proof. Let Y be a Δ_1^1 -random satisfying $Y \geq_h \mathcal{O}$. The existence of such a Y was shown in [4]. Let X be Martin-Löf random relative to $Y \oplus \mathcal{O}$ while satisfying $X \leq_h Y$. This choice ensures that X is Π_1^1 -random (so in particular Δ_1^1 -random).

By van Lambalgen's theorem relativized to $\emptyset^{(\alpha)}$, if both X and Y are Δ_1^1 -random, then for any $\alpha < \omega_1^{\text{CK}}$ it holds that X is $Y \oplus \emptyset^{(\alpha)}$ -random iff $X \oplus Y$ is $\emptyset^{(\alpha)}$ -random iff Y is $X \oplus \emptyset^{(\alpha)}$ -random. Since by choice of X , we know that in particular X is $Y \oplus \emptyset^{(\alpha)}$ -random, we conclude that Y is $X \oplus \emptyset^{(\alpha)}$ -random.

From ([3, Corollary 4.3]) it follows that for Π_1^1 -random X and $\beta < \omega_1^{\text{CK}}$ it holds that $X^{(\beta)} \leq_T X \oplus \emptyset^{(\beta)}$. (The conclusion above surely does not require full Π_1^1 -randomness of X , but too much precision would take us afield.) Together with the above, this shows that Y is $X^{(\alpha)}$ -random for every $\alpha < \omega_1^{\text{CK}}$. Since X is Π_1^1 -random, by Fact 6 we have that $\omega_1^{\text{CK}} = \omega_1^{\text{CK}, X}$, and thus that Y is Z -random for any $Z \in \Delta_1^1(X)$. By Fact 4 this establishes Y to be $\Delta_1^1(X)$ -random. But trivially, X cannot be $\Delta_1^1(X)$ -random. \square

3 Luzin's (N) and randomness reflection

Definition 8. A function satisfies Luzin's (N) iff the image of every null set is null.

Definition 9. For any randomness notion R and a function f , we say that f *reflects R -randomness* if $f(x)$ is R -random implies x is R -random for all x in the domain of f .

3.1 Luzin's (N) and higher Martin-Löf randomness reflection

By noting that the sets of points not Martin-Löf random relative to some oracle are canonical choices of null sets, we obtain the following:

Proposition 10. The following are equivalent for a computable function $f : \mathbb{R} \rightarrow \mathbb{R}$:

1. f satisfies Luzin's (N)
2. $\forall p \in \{0, 1\}^{\mathbb{N}} \exists q \in \{0, 1\}^{\mathbb{N}} f(x) \in \text{MLR}(q) \Rightarrow x \in \text{MLR}(p)$
3. $\forall p \in \{0, 1\}^{\mathbb{N}} f(x) \in \Delta_1^1\text{-random}(p) \Rightarrow x \in \text{MLR}(p)$
4. $\forall p \in \{0, 1\}^{\mathbb{N}} f(x) \in \Delta_1^1\text{-random}(p) \Rightarrow x \in \Delta_1^1\text{-random}(p)$

Proof. 1. \Leftrightarrow 2. Each null set is contained in a set of the form $\text{MLR}(q)^C$ for some oracle $q \in \{0, 1\}^{\mathbb{N}}$. Luzin's (N) is thus equivalent to saying that for any p there is a q with $f[\text{MLR}(p)^C] \subseteq \text{MLR}(q)^C$. Taking the contrapositive yields (2).

2. \Rightarrow 3. Any Σ_1^1 -null set is contained in a Δ_1^1 -null set ([22]). Thus it suffices to choose q as something hyperarithmetical in p .

3. \Rightarrow 1. Trivial.

3. \Rightarrow 4. Assume that (4) fails, i.e. that there is some $p \in \{0, 1\}^{\mathbb{N}}$ and some $x \notin \Delta_1^1\text{-random}(p)$ with $f(x) \in \Delta_1^1\text{-random}(p)$. But if $x \notin \Delta_1^1\text{-random}(p)$, then there is some $q \leq_h p$ with $x \notin \text{MLR}(q)$, but $f(x) \in \Delta_1^1\text{-random}(q) = \Delta_1^1\text{-random}(p)$, hence (3) is violated, too.

4. \Rightarrow 3. Trivial. □

Corollary 11. A computable function satisfying Luzin's (N) reflects Δ_1^1 -randomness relative to any oracle.

We can now ask whether reflecting Δ_1^1 -randomness relative to some specific oracle already suffices.

Fact 12 ([14, 28]). If A is an uncountable $\Delta_1^1(y)$ -class such that $y \leq_h z$ for every $z \in A$, then there is some $x \in A$ with $\mathcal{O}^y \leq_h x$.

Fact 13 (Sacks [22]). If $\mathcal{O} \leq_h x$, then x is not $\Delta_1^1(\mathcal{O})$ -random.

Corollary 14. If computable f reflects $\Delta_1^1(\mathcal{O})$ -randomness, then for any $\Delta_1^1(\mathcal{O})$ -random y we find that $f^{-1}(y)$ is countable.

Proof. Assume that y is $\Delta_1^1(\mathcal{O})$ -random and $f^{-1}(y)$ is uncountable. Then by Fact 12, there is some $x \in f^{-1}(y)$ with $\mathcal{O} \leq_h x$. By Fact 13, we find that x is not $\Delta_1^1(\mathcal{O})$ -random, contradicting that f reflects $\Delta_1^1(\mathcal{O})$ -randomness. □

Observation 15. The following are equivalent for computable $f : \mathbb{R} \rightarrow \mathbb{R}$:

1. For almost all y it holds that $f^{-1}(\{y\})$ is countable.
2. For every Δ_1^1 -random y it holds that $f^{-1}(\{y\})$ is countable.

Proof. The implication 2 \Rightarrow 1 is trivial. For the other direction, note that

$$\{y \mid f^{-1}(\{y\}) \text{ is uncountable}\}$$

is Σ_1^1 . This holds because for a Σ_1^1 -set A , being uncountable is equivalent to containing an element which is not hyperarithmetic relative to A . Due to Kleene's HYP-quantification theorem, an existential quantifier over non-hyperarithmetic elements is equivalent to an unrestricted existential quantifier. By assumption, it is a null set. Any Σ_1^1 -null set is contained in a Δ_1^1 -null set, so it is then contained in a Δ_1^1 -null set, and so cannot contain any Δ_1^1 -randoms. □

Theorem 16. The following are equivalent for computable $f : \mathbb{R} \rightarrow \mathbb{R}$:

1. f satisfies Luzin's (N)
2. f reflects $\Delta_1^1(\mathcal{O})$ -randomness.
3. f reflects Δ_1^1 -randomness and for almost all y , $f^{-1}(y)$ is countable.
4. f reflects Π_1^1 -randomness.

Proof. That (1) implies (2) follows from Proposition 10. To see that (2) implies (1), we show that if f reflects $\Delta_1^1(\mathcal{O})$ -randomness, then it reflects $\Delta_1^1(r)$ -randomness for all $r \geq_T \mathcal{O}$. This will be enough because if f reflects $\Delta_1^1(r)$ -randomness for all $r \geq_T \mathcal{O}$, then for any $p \in \{0, 1\}^{\mathbb{N}}$, if $f(x) \in \text{MLR}(\mathcal{O}^{p \oplus \mathcal{O}})$, then $f(x)$ is $\Delta_1^1(p \oplus \mathcal{O})$ -random, so x is $\Delta_1^1(p \oplus \mathcal{O})$ -random, so $x \in \text{MLR}(p)$. Thus f satisfies condition (2) of Proposition 10.

So let y be $\Delta_1^1(r)$ -random for some $r \geq_T \mathcal{O}$. By Corollary 14, $f^{-1}(y)$ is countable. So if $x \in f^{-1}(y)$, then $x \leq_h y$. Since f reflects $\Delta_1^1(\mathcal{O})$ -randomness, we know that x is Δ_1^1 -random. We can thus invoke Theorem 1 to conclude that x is $\Delta_1^1(r)$ -random, and have reached our goal. (Note that y is Π_1^1 -random because $r \geq_T \mathcal{O}$.)

To see that (1) implies (3) we use Proposition 10 and Corollary 14. The proof that (3) implies (1) proceeds analogously to the proof that (2) implies (1), except that we conclude that $f^{-1}(\{y\})$ is countable from Observation 15 rather than Corollary 14.

To see that (3) implies (4), by Observation 15, in fact the inverse image of every Δ_1^1 -random point is countable. In particular if y is Π_1^1 -random, then $f^{-1}(y)$ is countable. We use the following characterization of Π_1^1 -randomness due to Stern: y is Π_1^1 -random if and only if y is Δ_1^1 -random and $\omega_1^y = \omega_1^{\text{CK}}$ [26, 27, 4]. Recall also that $\omega_1^y = \omega_1^{\text{CK}}$ if and only if $y \not\leq_h \mathcal{O}$ ([23, Corollary 7.7]). Let $x \in f^{-1}(y)$. By Δ_1^1 -randomness reflection, x is Δ_1^1 -random. Since $f^{-1}(y)$ is countable, $y \geq_h x$. Thus $x \not\leq_h \mathcal{O}$. Therefore x is Π_1^1 -random.

The proof that (4) implies (1) is the same as how conditions (2) or (3) implied (1). Letting y be $\Delta_1^1(r)$ -random, since $r \geq_T \mathcal{O}$ then y is Π_1^1 -random. If $f^{-1}(y)$ were uncountable, by Fact 12, it would contain some x with $x \geq_h \mathcal{O}$, contradicting Π_1^1 -randomness reflection. \square

3.2 A note on the countability of fibers

We obtain as a corollary a reproof of a theorem by Banach [1], cf. [24, Chapter IX, Theorem 7.3]:

Corollary 17. If f is continuous and satisfies Luzin's (N), then for almost all y we find that $f^{-1}(y)$ is countable.

The following generalization to measurable functions was also known, but we give a new proof.

Corollary 18. 1. If f satisfies Luzin's (N) and there are a continuous function g and a Borel set A so that f agrees with g on A , then for almost every real $y \in f(A)$ we find that $f^{-1}(y) \cap A$ is countable.

2. If f satisfies Luzin's (N) and is measurable, then for almost every real y we find that $f^{-1}(y)$ is countable.

Proof. (1). Fix a real x so that g is computable in x and A is $\Delta_1^1(x)$. Assume that y is $\Delta_1^1(\mathcal{O}^x)$ -random and $f^{-1}(y) \cap A = g^{-1}(y) \cap A$ is uncountable. Since A is $\Delta_1^1(x)$, by Claim 12, then there is some $z \in g^{-1}(y) \cap A = f^{-1}(y) \cap A$ with $\mathcal{O}^x \leq_h z \oplus x$. By Fact 13, we find that z is not $\Delta_1^1(\mathcal{O}^x)$ -random, contradicting that g reflects $\Delta_1^1(\mathcal{O}^x)$ -randomness.

(2). By Luzin's theorem, there are a sequence Borel sets $\{A_n\}_{n \in \omega}$ and continuous functions $\{g_n\}_{n \in \omega}$ so that $\mathbb{R} \setminus \bigcup_n A_n$ is null and f agrees with g_n over A_n for every n . As f has Luzin's (N), also $f[\mathbb{R} \setminus \bigcup_n A_n]$ is null, and thus can be ignored for our argument. For $y \notin f[\mathbb{R} \setminus \bigcup_n A_n]$, we find that $f^{-1}(y) \subseteq \bigcup_{n \in \mathbb{N}} g_n^{-1}(y)$. Since each set in the right-hand union is countable for almost all y by Corollary 17, the union itself is countable for almost all y , proving the claim. \square

Note that the Borelness of the set A in the corollary above cannot be replaced by "arbitrary set", as it is consistent with ZFC that the corresponding statement is false. For example, assuming the continuum hypothesis (CH) or the even weaker Martin's axiom (MA) suffices to construct a counterexample. We do not know whether the following proposition can be proved within ZFC.

Proposition 19 (ZFC + MA). There is a function $f : [0, 1] \rightarrow [0, 1]$ having Luzin's (N) and a set $A \subseteq [0, 1]$ such that $f|_A$ is a computable, $f[A]$ is non-null, and for any $y \in f[A]$ the set $f^{-1}(\{y\}) \cap A$ is uncountable.

Proof. We actually need only a weaker condition than MA for our construction, namely the equality $\text{cof}(\mathcal{L}) = \text{cov}(\mathcal{L}) = \text{non}(\mathcal{L})$ in Cichoń's diagram. Recall that $\text{cof}(\mathcal{L})$ is the least cardinality of a set R of null sets such that any null set is a subset of an element of R ; $\text{cov}(\mathcal{L})$ is the least cardinal α such that $[0, 1]$ is a union of α -many null sets, and $\text{non}(\mathcal{L})$ is the least cardinal of a non-null set. It is a consequence of MA that all these cardinals are 2^{\aleph_0} . As they all are clearly uncountable and at most the continuum, CH trivially implies the same. Let κ denote the value of these three invariants.

First, we observe that $\kappa = \text{cof}(\mathcal{L})$ means that there exists a family $(z_\alpha)_{\alpha < \kappa}$ such that a set $A \subseteq \{0, 1\}^{\mathbb{N}}$ is null iff $A \subseteq \text{MLR}(z_\alpha)^C$ for some $\alpha < \kappa$. Next, we point out that $\kappa = \text{cov}(\mathcal{L})$ means that for any $\alpha < \kappa$ and family $(w_\beta)_{\beta < \alpha}$ there exists some u which is Martin-Löf random relative to all w_β .

We start with a family $(z_\alpha)_{\alpha < \kappa}$ as above, and then choose $(x_\alpha)_{\alpha < \kappa}$ such that each x_α is Martin-Löf random relative to any z_β for $\beta \leq \alpha$. We then choose another sequence $(y_\alpha)_{\alpha < \kappa}$ such that y_α is Martin-Löf random relative to any $x_\beta \oplus z_\gamma$ for $\beta, \gamma \leq \alpha$. We identify $\{0, 1\}^{\mathbb{N}}$ with a positive measure subset of $[0, 1]$ (a fat Cantor set), and then define $A = \{y_\beta \oplus x_\alpha \mid \alpha \leq \beta < \kappa\}$, and $f : [0, 1] \rightarrow [0, 1]$ as $f(y_\beta \oplus x_\alpha) = x_\alpha$ and $f(w) = 0$ for $w \notin A$.

As $f|_A$ is just the projection, it is clear that it is computable. To see that $f[A]$ is non-null, note that if it were null, it would need to be contained in $\text{MLR}(z_\alpha)^C$ for some $\alpha < \kappa$. But $x_\alpha \in f[A]$ is explicitly chosen to prevent this. For any $x_\alpha \in f[A]$, we find that $f^{-1}(\{x_\alpha\}) \cap A = \{y_\beta \oplus x_\alpha \mid \alpha \leq \beta < \kappa\}$ has cardinality κ , and κ is uncountable.

It remains to argue that f has Luzin's (N). As f is constant outside of A , we only need to consider null sets $B \subseteq A$. Again invoking van Lambalgen's theorem, we see that any $y_\beta \oplus x_\alpha$ is Martin-Löf random relative to z_γ whenever $\gamma \leq \alpha \leq \beta$. As such, each null $B \subseteq A$ is contained in some $\{y_\beta \oplus x_\alpha \mid \alpha \leq \beta < \gamma\}$ for $\gamma < \kappa$. It follows that $f[B] \subseteq \{x_\alpha \mid \alpha < \gamma\}$ has cardinality strictly below κ , and hence is null due to $\kappa = \text{non}(\mathcal{L})$. \square

That all fibers are countable is just preservation of h -degrees:

Observation 20. For computable $f : \mathbb{R} \rightarrow \mathbb{R}$ the following are equivalent:

1. f preserves h -degrees, i.e. $\forall x \in [0, 1] \ x \equiv_h f(x)$.
2. For all $y \in \mathbb{R}$, $f^{-1}(y)$ is countable.

3.3 Luzin's (N) and higher Kurtz randomness reflection

In his thesis [13], Luzin showed that if a continuous function $f : \mathbb{R} \rightarrow \mathbb{R}$ fails to have property (N), then in fact there is a compact witness to this failure. For the reader's convenience, we give a proof of this fact below.

Proposition 21. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be continuous and map some null set to a non-null set. Then there is a compact subset $A \subseteq \mathbb{R}$ with $\lambda(A) = 0$ and $\lambda(f(A)) > 0$.

Proof. Observe that a function $f : \mathbb{R} \rightarrow \mathbb{R}$ satisfies Luzin's (N) if and only if its restriction $f \upharpoonright [a, b]$ satisfies Luzin's (N) for every closed interval $[a, b]$. So without loss of generality, we assume that f fails Luzin's (N) because $\mu(A) = 0$ but $\mu(f(A)) = d > 0$ for some $A \subseteq [a, b]$. As

every null set is contained in a $\mathbf{\Pi}_2^0$ -null set, without loss of generality we can assume $A = \bigcap_n U_n$ for some decreasing sequence of open sets U_n . Each U_n is itself equal to an increasing union of closed sets. The idea is by picking big enough closed $F_n \subseteq U_n$, we can find a closed set $\bigcap_{n \in \mathbb{N}} F_n \subseteq A$ whose image still has positive measure. How large to pick the F_n ? Let $F_0 \subseteq U_0$ be large enough that $\mu(f(F_0 \cap A)) > d/2$. In general, if we have found $(F_i)_{i < n}$ such that $\mu(f(\bigcap_{i < n} F_i \cap A)) > d/2$, then since $A \subseteq U_n$, we can find closed $F_n \subseteq U_n$ large enough that $\mu(f(\bigcap_{i < n} F_i \cap F_n \cap A)) > d/2$ as well. Therefore for all n , we have $\mu(f(\bigcap_{i < n} F_i)) > d/2$, and therefore $\mu(\bigcap_n f(\bigcap_{i < n} F_i)) \geq d/2$. Claim: $\bigcap_n f(\bigcap_{i < n} F_i) = f(\bigcap_n F_n)$. One direction is clear. In the other, suppose that $y \in \bigcap_n f(\bigcap_{i < n} F_i)$. Then $\bigcap_{i < n} F_n \cap f^{-1}(y) \neq \emptyset$ for all n . By compactness, $\bigcap_n F_n \cap f^{-1}(y) \neq \emptyset$. \square

As the image of as images of compact sets under continuous functions are uniformly compact, this shows that Luzin's (N) implies the reflection of all kinds of Kurtz randomness. This is in contrast to the situation for Martin-Löf randomness, because the image of a $\mathbf{\Pi}_2^0$ set under a continuous f is not even $\mathbf{\Pi}_2^0$ in general, let alone with the same oracle.

Proposition 22. If $f : \mathbb{R} \rightarrow \mathbb{R}$ satisfies Luzin's (N), then f reflects Kurtz randomness relative to every oracle.

Proof. Given oracle Z , suppose that x is not Z -Kurtz random because $x \in F$, where F is a Z -computable compact set of measure 0. Then $f(F)$ is also a Z -computable compact set, which has measure 0 because f has Luzin's (N). Therefore, $f(x)$ is also not Z -Kurtz random. \square

An immediate consequence is that any f with Luzin's (N) also reflects Δ_1^1 -Kurtz randomness relative to every oracle. In general, Kurtz randomness reflection for stronger oracles implies it for weaker ones.

Proposition 23. If a continuous function $f : \mathbb{R} \rightarrow \mathbb{R}$ reflects Z -Kurtz randomness, then f reflects X -Kurtz randomness for every $X \leq_T Z$.

Proof. Assume that f reflects Z -Kurtz randomness. Suppose x is not X -Kurtz random. Let P be an X -computable compact null set with $x \in P$. Then $f(P)$ is an X -computable compact set, which is null because P is also Z -computable. Therefore, $f(x)$ is not X -Kurtz random. \square

Additionally, any witness to the failure of Luzin's (N) also provides an oracle relative to which Kurtz randomness reflection fails.

Proposition 24. Suppose that $f : \mathbb{R} \rightarrow \mathbb{R}$ and $A \subseteq \mathbb{R}$ is a Z -computable compact null set with $\lambda(f(A)) > 0$. Then f does not reflect Z -Kurtz randomness.

Proof. Since $f(A)$ is also Z -computable and has positive measure, it must contain some Z -Kurtz random y . There is some $x \in A$ with $f(x) = y$, but since A is a Z -computable compact null set, it cannot contain any Z -Kurtz randoms. Hence, f does not reflect Z -Kurtz randomness. \square

We can thus characterize Luzin's (N) in terms of Kurtz randomness reflection.

Theorem 25. The following are equivalent for computable $f : \mathbb{R} \rightarrow \mathbb{R}$:

1. f has Luzin's (N).
2. For every \mathcal{O} -computable compact set A with $\lambda(A) = 0$ also $\lambda(f(A)) = 0$.
3. f reflects \mathcal{O} -Kurtz randomness.

Proof. 1. \Rightarrow 2. Trivial.

2. \Rightarrow 1. We observe that given computable $f : \mathbb{R} \rightarrow \mathbb{R}$ and number n , the set

$$\{A \subseteq [-n, n] \text{ compact} \mid \lambda(A) = 0 \wedge \lambda(f(A)) \geq 2^{-n}\}$$

is a Π_2^0 -subset of the Polish space of compact subsets of $[-n, n]$. By Proposition 21, if f fails Luzin's (N), this set is non-empty for some n . If it is non-empty, it must have an \mathcal{O} -computable element by Kleene's basis theorem.

1. \Rightarrow 3. By Proposition 22.

3. \Rightarrow 2. By Proposition 24. □

We also see that Δ_1^1 -Kurtz randomness reflection does not suffice for a characterization.

Lemma 26. Reflecting Δ_1^1 -Kurtz randomness is a Σ_1^1 -property of continuous $f : \mathbb{R} \rightarrow \mathbb{R}$.

Proof. By Proposition 24, reflecting Δ_1^1 -Kurtz randomness is equivalent to the statement that for any Δ_1^1 compact set A with $\lambda(A) = 0$ we have $\lambda(f(A)) = 0$. By Kleene's HYP quantification theorem [11, 12], a universal quantification over Δ_1^1 can be replaced by an existential quantification over Baire space. That $\lambda(A) = 0$ implies $\lambda(f(A)) = 0$ is a Δ_3^0 -statement for given f and A . □

Corollary 27. Reflecting Δ_1^1 -Kurtz randomness is strictly weaker than Luzin's (N).

Proof. By Proposition 22, Luzin's (N) implies Δ_1^1 -Kurtz randomness reflection. By Lemma 26 reflecting Δ_1^1 -Kurtz randomness is a Σ_1^1 -property. But it was shown in [9] that Luzin's (N) is Π_1^1 -complete for continuous functions. Thus the two notions cannot coincide. □

3.4 Open questions

Theorems 16 and 25 tell us that $\Delta_1^1(\mathcal{O})$ -randomness reflection and \mathcal{O} -Kurtz randomness reflection each characterize Luzin's (N) for computable functions. This does not rule out that other kinds of randomness reflection could also characterize Luzin's (N). In the next section we shall see that none of MLR-reflection, W2R-reflection, or $\text{MLR}(\emptyset')$ -reflection imply Luzin's (N) for arbitrary computable functions (Corollary 31). Because reflection asks for the same level of randomness on both sides, there are no completely trivial implications between the Π_2^0 -type randomness reflection notions. Indeed, results in [2] suggest that the implication structure between Π_2^0 -type randomness reflection notions may have little relation to the implication structure between notions of randomness. However, the most interesting open question seems to be:

Open Question 28. Can a computable function reflect Δ_1^1 -randomness but fail Luzin's (N)?

By Theorem 16 any such example would need to have a positive measure of fibers being uncountable, which is incompatible with most niceness conditions. We also do not know the answer to the above question if Δ_1^1 -randomness is replaced with Martin-Löf randomness relative to $\emptyset^{(\alpha)}$ for any $\alpha \geq 2$.

Related questions concern basis theorems for failures of Luzin's (N). We have already seen in Theorem 25 that any computable f which fails Luzin's (N) must see that failure witnessed

by a \mathcal{O} -computable compact set. The proof shows that such a set can also be chosen hyperarithmetically low, by applying Gandy basis theorem in place of the Kleene. On the other hand, Corollary 27 shows that a function which fails Luzin's (N) need not have a hyperarithmetically compact witness. Indeed, one can obtain specific examples of this separation by feeding pseudo-well-orders into the Π_1^1 -completeness construction of [9]. Thus the results for compact witnesses are rather tight overall.

The situation for the minimum complexity of Π_2^0 witnesses is less well understood. The proof of Corollary 31 shows that a computable function may fail Luzin's (N) while still mapping all rapidly null $\Pi_2^0(\emptyset')$ sets to null sets. That is, the set $\text{MLR}(\emptyset')^C$ is mapped to a null set.

Open Question 29. Can a computable function map $W3R^C$ to a null set but fail Luzin's (N)? Equivalently, can a computable function map all null $\Pi_2^0(\emptyset')$ sets to null sets while failing Luzin's (N)?

We note that the functions produced by the Π_1^1 -completeness construction of [9] are of no help because the failure of Luzin's (N), when it occurs, is witnessed by an effectively null Π_2^0 set.

4 Separating Luzin's (N) from MLR-reflection

We present a construction of a computable function that violates Luzin's (N), and yet is piecewise-linear in a neighborhood of every point that is not $\text{MLR}(\emptyset')$. Here, we say that f is piecewise-linear in a neighborhood of x , if there are rationals $a < x < b$ such that $f|_{[a,x]}$ and $f|_{[x,b]}$ are linear functions. Computable piecewise-linear functions reflect essentially all kinds of randomness.

Theorem 30. For each $\Pi_1^0(\emptyset')$ -set $A \subseteq [0, 1]$ there is a computable function $f : [0, 1] \rightarrow [0, 1]$ such that:

1. For every $x \in [0, 1] \setminus A$, f is piecewise-linear on a neighbourhood of x .
2. For every $\varepsilon > 0$, there is a null $\Pi_1^0(\emptyset'')$ set $B \subseteq A$ such that $\lambda(f[B]) \geq \lambda(A) - \varepsilon$.

Corollary 31. There is a computable function $f : [0, 1] \rightarrow [0, 1]$ that reflects ML-randomness, weak-2-randomness and $\text{ML}(\emptyset')$ -randomness, yet does not have Luzin's (N), nor reflects weak-3-randomness.

Proof. Let A be the complement of the first component of a universal $\text{ML}(\emptyset')$ -test. Then $\lambda(A) > \frac{1}{2}$. We invoke Theorem 30 on A and $\varepsilon = \frac{1}{4}$. The resulting function is the desired one: If $x \in [0, 1]$ is not $\text{ML}(\emptyset')$ -random, then $x \notin A$, f is piecewise-linear on a neighborhood of x , and thus $f(x)$ is not $\text{ML}(\emptyset')$ -random. As such, we conclude that whenever $f(x)$ is $\text{ML}(\emptyset')$ -random, then so is x (same for the other notions).

Since we can choose the witness B as being $\Pi_1^0(\emptyset'')$, it is also $\Pi_2^0(\emptyset')$, and thus contains only elements which are not weak-3-random. Since $f[B]$ has positive measure, it contains a weak-3-random – hence f does not reflect weak-3-randomness. \square

We remark that this is the strongest result possible for the strategy we are using. We are making sure that f reflects $\text{MLR}(\emptyset')$ -randomness by making f piecewise-linear in a neighborhood of every non- $\text{MLR}(\emptyset')$ point. However, the following proposition shows that the set of points where f can be this simple has a descriptive complexity of $\Sigma_1^0(\emptyset')$. But the weak-3-non-randoms are not contained in any $\Sigma_1^0(\emptyset')$ set except $[0, 1]$.

Proposition 32. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be computable. The set of points where f is locally piecewise-linear is $\Sigma_1^0(\emptyset')$.

Proof. We consider the property 2L of a function f and an interval $[a, b]$ that there is some $x \in [a, b]$ such that both $f|_{[a, x]}$ and $f|_{[x, b]}$ are linear. We first claim that this is a Π_1^0 -property. To this, we observe that 2L is equivalent to:

$$\forall n \in \mathbb{N} \exists i \leq n \left(\forall j \in \{0, \dots, n-2\} \setminus \{i-1, i, i+1\} f(a + \frac{j}{n}) - f(a + \frac{j+1}{n}) = f(a + \frac{j+1}{n}) - f(a + \frac{j+2}{n}) \right)$$

Next, we observe that f is locally piecewise-linear in x iff there is some rational interval $(a, b) \ni x$ such that f has property 2L on $[a, b]$. Using \emptyset' , we can enumerate all these intervals. \square

Generalizing this idea slightly, recall that a function is bi-Lipschitz, if both the function and its inverse are Lipschitz functions, i.e. if there exists some constant L such that $d(f(x), f(y)) \leq Ld(x, y) \leq L^2d(f(x), f(y))$ for all x, y in the domain. Since computable locally bi-Lipschitz functions preserve and reflect all kinds of randomness, Another way for f to ensure a given notion of randomness reflection is by being locally bi-Lipschitz on the non-random points for that notion. However, we still get a $\Sigma_1^0(\emptyset')$ -set of suitable points.

Proposition 33. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be computable. The set of points where f is locally bi-Lipschitz is $\Sigma_1^0(\emptyset')$.

Proof. The following is a co-c.e. property in $a, b \in \mathbb{Q}$ and $L \in \mathbb{N}$ and $f \in \mathcal{C}(\mathbb{R}, \mathbb{R})$:

$$\forall x, y \in [a, b] \quad d(x, y) \leq Ld(f(x), f(y)) \leq L^2d(x, y)$$

We obtain the set of points where f is locally bi-Lipschitz by taking the union of all (a, b) having the property above for some $L \in \mathbb{N}$ – access to \emptyset' suffices to get such an enumeration. \square

4.1 High-level proof sketch

Before diving into the details of the proof of Theorem 30 we give a high-level sketch of what is going on. Consider first the case where A is Π_1^0 . Then $A = \bigcap_n A_n$, where each A_n is a finite union of closed intervals and $A_{n+1} \subseteq A_n$. We iteratively define a sequence of piecewise linear functions f_0, f_1, \dots , where $f_0 : [0, 1] \rightarrow [0, 1]$ is the identity, and f_n is obtained from f_{n-1} by performing a “tripling” operation on those line segments of f_{n-1} which are contained in A_n . In order to “triple” a line segment, we replace it by a zig-zag of three line segments each of which has triple the slope of the original. (See Figure 1) We want to make the sequence (f_n) converge in the supremum norm, so before tripling we add invisible break points to f_{n-1} so that none of its linear pieces are more than 2^{-n} tall. Letting f be the limit function, observe that if $x \notin A_n$, then f coincides with f_n on a neighborhood of x , and thus f is linear on a neighborhood of x . On the other hand, A is then exactly the set of points where we tripled infinitely often.

Next we describe how to find a closed set $B \subseteq A$ such that $\mu(f(B)) \geq \mu(A)$. (The ε in the statement of the theorem exists in order to bring down the descriptive complexity of B , but we can ignore it for now.) We want $B \subseteq A$, so of course we throw out of B any interval that leaves A . Also, every time we perform a tripling, we choose two-thirds of the tripled interval to throw out of B . We do this so that the one-third which we keep has maximal measure of intersection with A . Observe that $\mu(B) = 0$.

Here is why $\mu(f(B)) \geq \mu(A)$. Let $B_0 = [0, 1]$ and let B_n denote the set of points that remain in B at the end of stage n . By induction, $f_n \upharpoonright B_n$ is injective (except possibly at break points) and $\mu(f_n(B_n \cap A)) \geq \mu(A)$. The key to the induction is that by the choice of thirds, we always have $3^n \mu(B_n \cap A) \geq \mu(A)$, and since f_n has slope $\pm 3^n$ on all of B_n and is essentially injective, $\mu(f_n(B_n \cap A)) = 3^n \mu(B_n \cap A)$. It now follows that $\mu(f_n(B_n)) \geq \mu(A)$ for all n . Furthermore, since the continuous image of a compact set is uniformly compact, we cannot have $\mu(f(B)) < \mu(A)$, for this would have been witnessed already for some $\mu(f_n(B_n))$. This completes the sketch for the case where A is Π_1^0 .

If A is $\Pi_1^0(\emptyset')$, we can do essentially the same construction, tripling on the stage- n approximation to A_n instead of A_n itself. Any interval which is going to leave A eventually leaves the approximations for good, so the key features of the above argument are maintained even as the structure of the triplings gets more complicated.

4.2 Proof of Theorem 30

The remainder of this section is devoted to the preparation for the proof of Theorem 30, and the proof itself.

Lemma 34. Given a $\Pi_1^0(\emptyset')$ -set $A \subseteq [0, 1]$ and some open $U \supseteq A$ we can compute some open V with $A \subseteq V \subseteq U$ such that $d(V, U^C) > 0$.

Proof. Since U^C and A are disjoint closed sets, there is some $N \in \mathbb{N}$ such that $d(U^C, A) > 2^{-N}$. If we actually had access to A , we could compute a suitable N . Since A is computable from \emptyset' , we can compute N with finitely many mindchanges. The monotonicity of correctness here means we can actually obtain suitable $N \in \mathbb{N}_{<}$. We now obtain V by enumerating an interval (a, b) into V once we have learned that U covers $[a - 2^{-N}, b + 2^{-N}]$ (which is semidecidable in $U \in \mathcal{O}(\mathbb{R})$ and $N \in \mathbb{N}_{<}$). \square

For an interval $[a, b]$, let $T_0([a, b]) = [a, a + \frac{b-a}{3}]$, $T_1([a, b]) = [a + \frac{b-a}{3}, a + 2\frac{b-a}{3}]$ and $T_2([a, b]) = [a + 2\frac{b-a}{3}, b]$.

Lemma 35. Let A be a $\Pi_1^0(\emptyset')$ set. Then there is a computable double-sequence $(I_n^k)_{k,n \in \mathbb{N}}$ of closed intervals with the following properties:

1. $A = \bigcap_{n \in \mathbb{N}} \bigcup_{k \in \mathbb{N}} I_n^k$.
2. I_n^k and I_n^ℓ intersect in at most one point.
3. For $m < n$, we find that $\bigcup_{k \in \mathbb{N}} I_n^k$ has positive distance to the complement of $\bigcup_{k \in \mathbb{N}} I_m^k$.
4. $\forall k, n \in \mathbb{N} \quad |I_n^k| \geq |I_n^{k+1}|$
5. Fix $n > 0$. For each k there are ℓ, i such that $|I_n^k| < 3^{-2}|I_{n-1}^\ell|$ and $I_n^k \subseteq T_i(I_{n-1}^\ell)$.

Proof. Any $\Pi_1^0(\emptyset')$ is in particular Π_2^0 , and thus has Π_2^0 -approximation $A = \bigcap_{n \in \mathbb{N}} U_n$. We invoke Lemma 34 inductively first on A and U_0 to obtain V_0 , then on A and $U_1 \cap V_0$ to obtain V_1 , and so on. This will ensure Condition (3). We can effectively write any open set $V_n \subseteq [0, 1]$ as a union of closed intervals such that the pairwise intersections contain at most one point. To make Conditions (4,5) work it suffices to subdivide intervals sufficiently much. \square

Definition 36. An interval J is *well-located* relative to $(I_n^k)_{k,n \in \mathbb{N}}$, if for all k, n one of the following hold:

1. $|J \cap I_n^k| \leq 1$
2. $J \supseteq I_n^k$
3. $J \subseteq T_i(I_n^k)$ for some $i \in \{0, 1, 2\}$

For well-located J , let its depth be the greatest n such that $J \subseteq I_n^k$ for some k . We call two well-located intervals J_0, J_1 peers, if whenever $J_b \subsetneq I_n^k$ for both $b \in \{0, 1\}$ and some k, n , then there is one $i \in \{0, 1, 2\}$ such that $J_b \subseteq T_i(I_n^k)$ for both $b \in \{0, 1\}$.

Note that our requirements for the $(I_n^k)_{k, n \in \mathbb{N}}$ in Lemma 35 in particular ensure that each $I_{n_0}^{k_0}$ is well-located relative to $(I_n^k)_{n, k \in \mathbb{N}}$.

Definition 37. We are given a double-sequence $(I_n^k)_{k, n \in \mathbb{N}}$ for a set A as in Lemma 35 and $\varepsilon > 0$. Let $N_n \in \mathbb{N}$ be chosen sufficiently large such that $\lambda(\bigcup_{k > N_n} I_n^k) < 3^{-n}2^{-n-2}\varepsilon$. Let $b_{k, n} \in \{0, 1, 2\}$ be chosen such that $\lambda(T_{b_{k, n}}(I_n^k) \cap A) + 3^{-n}2^{-n-3-k}\varepsilon \geq \lambda(T_c(I_n^k) \cap A)$ for all $k, n \in \mathbb{N}$ and $c \in \{0, 1, 2\}$. Let $\ell_{n, k}$ and $i_{n, k}$ be the witnesses for Condition 5 in Lemma 35. We then inductively define $\mathfrak{J}_0^\varepsilon = \{I_0^k \mid k \leq N_0\}$ and:

$$\mathfrak{J}_n^\varepsilon = \{I_n^k \mid k \leq N_n \wedge I_{n-1}^{\ell_{n, k}} \in \mathfrak{J}_{n-1} \wedge b_{\ell_{n, k}, (n-1)} = i_{n, k}\}$$

In words, the intervals in $\mathfrak{J}_n^\varepsilon$ are those on the n -th level which occur inside a particular third of their parent intervals on the $n-1$ -st level which has the maximal measure of its intersection with the set A . By construction, the intervals in \mathfrak{J}_n are pairwise peers.

Lemma 38. Starting with a $\Pi_1^0(\emptyset')$ -set A , we can compute the sets $\mathfrak{J}_n^\varepsilon$ relative to \emptyset'' .

Proof. As the double-sequence $(I_n^k)_{k, n \in \mathbb{N}}$ is computable, we can obtain the sufficiently large N_n by using \emptyset' . We have $T_c(I_n^k) \cap A$ available to us as $\Pi_1^0(\emptyset')$ -sets, so \emptyset'' lets us compute $\lambda(T_c(I_n^k) \cap A) \in \mathbb{R}$. Then getting the choices for the $b_{k, n}$ right can be done computably. The witnesses $\ell_{n, k}, i_{n, k}$ can also be found computably. \square

Lemma 39. $3^{-n} \geq \lambda(\bigcup \mathfrak{J}_n^\varepsilon) \geq 3^{-n}(\lambda(A) - \varepsilon)$

Proof. We prove both inequalities by induction. For the first, the base case is trivial. For the induction step, we note that $(\bigcup \mathfrak{J}_{n+1}^\varepsilon) \subseteq \bigcup_{I_n^k \in \mathfrak{J}_n^\varepsilon} T_{b_{k, n}}(I_n^k)$, and that $\lambda\left(\bigcup_{I_n^k \in \mathfrak{J}_n^\varepsilon} T_{b_{k, n}}(I_n^k)\right) = \frac{1}{3}\lambda(\bigcup \mathfrak{J}_n^\varepsilon)$.

For the second inequality, we prove a stronger claim, namely that $\lambda(A \cap \bigcup \mathfrak{J}_n^\varepsilon) \geq 3^{-n}(\lambda(A) - (1 - 2^{-n-1})\varepsilon)$. The base case follows from $\bigcup_{k \in \mathbb{N}} I_0^k \supseteq A$ together with the choice of N_0 . We then observe that $A \cap (\bigcup \mathfrak{J}_{n+1}^\varepsilon) = A \cap \bigcup \{I_{n+1}^k \mid \exists I_n^\ell \in \mathfrak{J}_n^\varepsilon \exists i \in \{0, 1, 2\} I_{n+1}^k \subseteq T_i(I_n^\ell)\}$. By definition of $b_{\ell, n}$ in Definition 37, this also means that $\lambda(A \cap (\bigcup \mathfrak{J}_{n+1}^\varepsilon)) + 3^{-n}2^{-n-2}\varepsilon \geq 3^{-1}\lambda(A \cap \bigcup \{I_{n+1}^k \mid \exists I_n^\ell \in \mathfrak{J}_n^\varepsilon I_{n+1}^k \subseteq T_{b_{\ell, n}}(I_n^\ell)\})$. The set on the right hand side differs from $\bigcup \mathfrak{J}_n^\varepsilon$ only by the fact that in the latter, we restrict to $\ell \leq N_n$. By the induction hypothesis together with the guarantee that $\lambda(\bigcup_{k > N_n} I_n^k) < 3^{-n}2^{-n-2}\varepsilon$ we get the desired claim. \square

Lemma 40. For a sequence $(I_n^k)_{k, n \in \mathbb{N}}$ as in Lemma 35 and $x \notin A$, it holds that there exists some $a < x < b$ and some $N \in \mathbb{N}$ such that $I_n^k \cap (a, b) = \emptyset$ for every $n > N$.

Proof. If $x \notin A$, then there is some N with $x \notin \bigcup_{k \in \mathbb{N}} I_N^k$. By Condition (3) of Lemma 35, we have that x has positive distance to $\bigcup_{k \in \mathbb{N}} I_{N+1}^k$, hence there exists an interval (a, b) around x disjoint from $\bigcup_{k \in \mathbb{N}} I_{N+1}^k$, and by monotonicity, also from $\bigcup_{k \in \mathbb{N}} I_n^k$ for every $n > N$. \square

Proof of Theorem 30. Preparation: We note that for each Π_2^0 -set A and each $n \in \mathbb{N}$, there is a $\Pi_1^0(\emptyset')$ -set C with $C \subseteq A$ and $\lambda(C) \geq \lambda(A) - 2^{-n}$. We can assume w.l.o.g. that A is already $\Pi_1^0(\emptyset')$. We then obtain a computable double-sequence $(I_n^k)_{k,n \in \mathbb{N}}$ as in Lemma 35.

Construction: We obtain our function f as the limit of functions $f_{N,K}$ for $N, K \in \mathbb{N}$. $f_{0,0}$ is the identity on $[0, 1]$. The construction of $f_{N,K}$ takes into account only intervals I_n^k with $n \leq N$ and $|I_n^k| > 2^{-K}$, of which there are only finitely many (and by monotonicity of the enumerations, we can be sure that we have found them all). We process intervals with smaller n first, and replace the linear growth f currently has on I_n^k by a triple as shown in Figure 1. Property 5 from Lemma 35 ensures that through the process, the function is linear on each interval I_n^k yet to be processed. We then define $f_N := \lim_{K \rightarrow \infty} f_{N,K}$ and $f := \lim_{N \rightarrow \infty} f_N$.

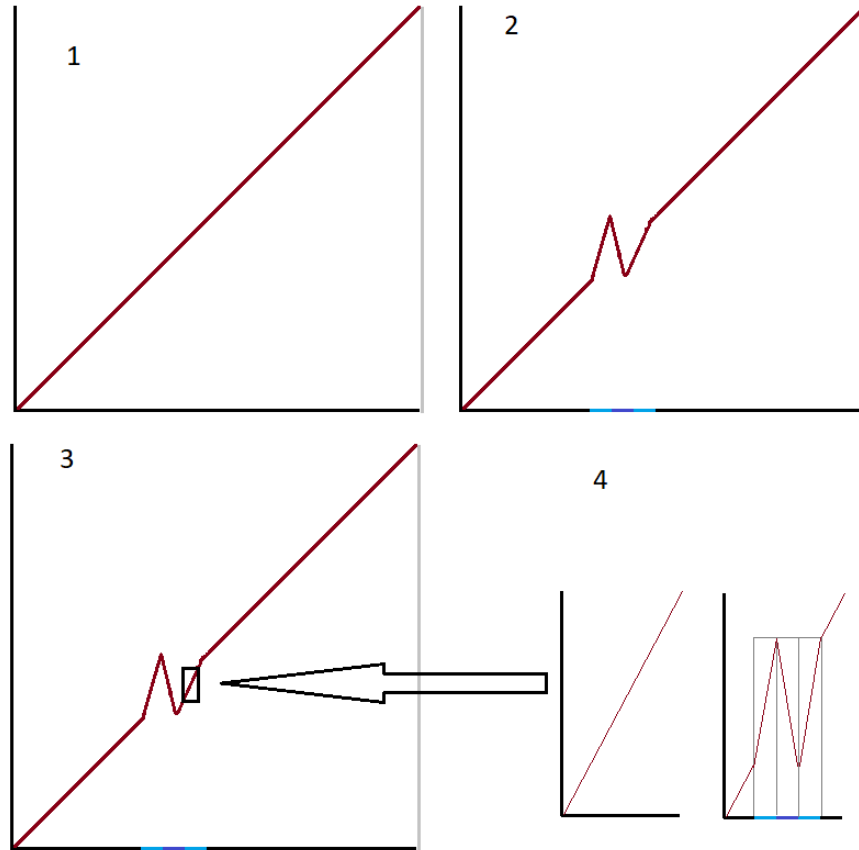


Figure 1: Demonstrating the iterative construction of the function f in the proof of Theorem 30

That the first limit has a computable rate of convergence follows from the monotonicity of $|I_n^k|$ in k . Since the size of the intervals shrinks sufficiently fast compared to the potential growth rates of f_N , we see that we also do have a computable rate of convergence of $(f_N)_{N \in \mathbb{N}}$.

Property 1: If $x \notin A$, we can invoke Lemma 40 to obtain a neighbourhood U of x that is disjoint from any I_n^k for $n > N$. But that ensures that $f|_U$ is 3^N -Lipschitz, and by potentially restricting the interval further we can make f locally bi-Lipschitz.

- Lemma 41.**
1. Let J be well-located at depth N . Then $f[J] = f_N[J]$.
 2. Let J be well-located at depth n . Then $\lambda(f[J]) = 3^n \lambda(J)$.
 3. Let J_0, J_1 be peer well-located intervals. Then $|f[J_0] \cap f[J_1]| \leq 1$.

Proof. 1. First, we observe that for any $M > N$ it holds that $f_M[J] = f_N[J]$, since all modifications based on intervals I_n^k with $n > N$ will affect $f|_J$ at most by locally replacing the shape of the graph with a different shape having the same range. It remains to argue that the identity of the image $f_M[J]$ is preserved by limits. Since $[0, 1]$ is compact and Hausdorff, we find that $\mathcal{A}([0, 1]) \cong \mathcal{K}([0, 1])$, hence we can compute $g[J] \in \mathcal{A}([0, 1])$ from g and $A \in \mathcal{A}([0, 1])$. We have access to $g[A] \in \mathcal{V}([0, 1])$ given $A \in \mathcal{V}([0, 1])$ anyway. Since $\mathcal{A}([0, 1]) \wedge \mathcal{V}([0, 1])$ is Hausdorff [19], this yields the claim.

2. By (1.), it suffices to show $\lambda(f_n[J]) = 3^n \lambda(J)$ instead. Now $(f_n)|_J$ is just a linear function with slope 3^n , which yields the claim.
3. If J_0, J_1 are peers and well-located, and $J_0 \subseteq I_n^k$ but $J_1 \not\subseteq I_n^k$, then I_n^k and J_1 are also peers. It thus suffices to prove the claim for the case where $J_0 = I_{n+1}^{k_0}$ and $J_1 = I_{n+1}^{k_1}$. These are both contained in the same $T_i(I_n^\ell)$, and $(f_n)|_{T_i(I_n^\ell)}$ is a linear function. Since $|J_0 \cap J_1| \leq 1$ it follows that $|f_n[J_0] \cap f_n[J_1]| \leq 1$. By (1.), this already yields the claim. \square

Property 2: We obtain the desired set B as $B = \bigcap_{n \in \mathbb{N}} (\bigcup \mathfrak{J}_n^\varepsilon)$. Since each $\mathfrak{J}_n^\varepsilon$ is a finite collection of closed intervals, B is indeed closed. Since the intersection is nested and $\lambda(\bigcup \mathfrak{J}_n^\varepsilon) \leq 3^{-n}$ by the first part of Lemma 39, we conclude that $\lambda(B) = 0$. Since the intervals in $\mathfrak{J}_n^\varepsilon$ are well-located and pairwise peers, we know that $\lambda(f(\bigcup \mathfrak{J}_n^\varepsilon)) = 3^n \lambda(\bigcup \mathfrak{J}_n^\varepsilon)$ by Lemma 41 2&3. Invoking the second inequality from Lemma 39 then lets us conclude $\lambda(f(\bigcup \mathfrak{J}_n^\varepsilon)) \geq (\lambda(A) - \varepsilon)$. Since this estimate holds for every stage of a nested intersection of compact sets, it follows that $\lambda(f[B]) \geq \lambda(A) - \varepsilon$ as desired. That B is obtainable by an oracle of the claimed strength follows from Lemma 38. \square

5 Luzin's (N), absolute continuity and bounded variation

We recall the definitions of absolute continuity and bounded variation:

Definition 42. A function $f : [0, 1] \rightarrow \mathbb{R}$ is *absolutely continuous*, if for every $\varepsilon > 0$ and every $x_0 < y_0 < x_1 < y_1 \dots < x_k < y_k$ there is a $\delta > 0$ such that:

$$\sum_{i \leq k} |y_i - x_i| < \delta \quad \Rightarrow \quad \sum_{i \leq k} |f(y_i) - f(x_i)| < \varepsilon.$$

Definition 43. A function $f : [0, 1] \rightarrow \mathbb{R}$ has bounded variation, if there is some bound $M \in \mathbb{N}$ such that for any $k \in \mathbb{N}$ and any $x_0 < x_1 < \dots < x_k$ it holds that

$$\sum_{i < k} |f(x_{i+1}) - f(x_i)| < M$$

Being absolutely continuous implies having bounded variation. These notions are related to Luzin's (N) by the following classical fact:

Fact 44 (see [24], Theorem VII.6.7). A continuous function is absolutely continuous iff it has both bounded variation and Luzin's (N).

We observe that being absolutely continuous is a Π_3^0 -property, and recall that Luzin's (N) is Π_1^1 -complete [9]. As such, restricting our attention to functions of bounded variation should alter the situation significantly.

Proposition 45. If $f : [0, 1] \rightarrow \mathbb{R}$ is computable and absolutely continuous, then f reflects weak-2-randomness.

Proof. First, we consider how we can exploit connectedness of \mathbb{R} to say something about the images of open sets under computable functions. We are given open sets in the form $U = \bigcup_{i \in \mathbb{N}} I_i$, where each I_i is an open interval with rational endpoints. We can then compute $\sup f(I_i)$ and $\inf f(I_i)$ (as these are equal to $\max f(\overline{I_{n,i}})$ and $\min f(\overline{I_{n,i}})$, and we can compute minima and maxima of continuous functions on compact sets). Let $V = \bigcup_{i \in \mathbb{N}} (\inf f(I_i), \sup f(I_i))$. We note that we can compute V from U , that $V \subseteq f[U]$, and that $f[U] \setminus V$ can only contain computable points. In particular, $\lambda(V) = \lambda(f[U])$.

Now we assume that f additionally is absolutely continuous, and that we are dealing with a Π_2^0 -null set $A = \bigcap_{n \in \mathbb{N}} U_n$ witnessing that some $x \in A$ is not weak-2-random. We assume that $U_{n+1} \subseteq U_n$. As A is null, we know that $\lim_{n \rightarrow \infty} \lambda(U_n) = 0$. Since f is absolutely continuous, we also have $\lim_{n \rightarrow \infty} \lambda(f[U_n]) = 0$. Let V_n be obtained from U_n as in the first paragraph, and $B = \bigcap_{n \in \mathbb{N}} V_n$. It follows that $\lambda(B) = 0$, and moreover, $f[A]$ is contained in B with the potential exception of some computable points. So we can conclude that $f(x)$ is not weak-2-random, either because $f(x)$ is computable, or because $f(x) \in B$. \square

Note that if we had started with a Martin-Löf test in the argument above, we would have no guarantee of ending up with one, because the modulus of absolute continuity is not computable in general. Indeed, absolute continuity does not imply MLR reflection. See Corollary 53.

Lemma 46. If $f : [0, 1] \rightarrow \mathbb{R}$ is computable, has bounded variation, and reflects \emptyset' -Kurtz randomness, then f has property (N).

Proof. Suppose that f does not have (N). Since f has bounded variation, it must fail absolute continuity. Let $\varepsilon > 0$ be such that for all $\delta > 0$, there is a finite union of intervals $A_\delta \subseteq [0, 1]$ with $\mu(A_\delta) < \delta$ and $\mu(f(A_\delta)) > \varepsilon$. Computably, given δ we can find such A_δ by searching. Let $A = \bigcap_n U_n$, where $U_n = \bigcup_{m > n} A_{2^{-m}}$. Then A is Π_2^0 , and $\mu(A) = 0$, and $\mu(\bigcap_n f(U_n)) \geq \varepsilon$. We claim that its subset $f(A)$ also has $\mu(f(A)) \geq \varepsilon$. Let $\text{Var}_f : [0, 1] \rightarrow \mathbb{R}$ denote the cumulative variation function of f , defined by setting $\text{Var}_f(x)$ to be equal to the variation of f on $[0, x]$. Since f has bounded variation and $U_{n+1} \subseteq U_n$, $\sum_n \text{Var}_f(U_n \setminus U_{n+1})$ is finite, so by choosing N large enough, we can make $\sum_{n > N} \mu(f(U_n \setminus U_{n+1}))$ as small as we like. Now observe that no matter how large N we choose,

$$\left(\bigcap_n f(U_n) \setminus f(A) \right) \subseteq \bigcup_{n > N} f(U_n \setminus U_{n+1}).$$

This proves the claim. We have found a Π_2^0 set $A = \bigcap_n U_n$ which witnesses the failure of (N).

Observe that for any c.e. open set U , $\mu(f(U))$ is c.e.. Therefore, since f has bounded variation, \emptyset' can search around to find, for each n , a closed set $F_n \subseteq U_n$ such that $\mu(f(U_n \setminus F_n)) < 2^{-n-2}\varepsilon$. The existence of such a closed set is guaranteed by f having bounded variation.

Let $F = \bigcap_n F_n$. Then $F \subseteq A$ and $A \setminus F = \bigcup_n (A \setminus F_n)$. So

$$\mu(f(A \setminus F)) \leq \sum_n \mu(f(A \setminus F_n)) \leq \sum_n \mu(f(U_n \setminus F_n)) \leq \sum_n 2^{-n-2}\varepsilon < \varepsilon.$$

The positive measure of $f(F)$ then follows as

$$\varepsilon \leq \mu(f(A)) \leq \mu(f(A \setminus F)) + \mu(f(F)).$$

Therefore, F is an \emptyset' -computable closed set of measure zero whose image has positive measure. So f does not reflect \emptyset' -Kurtz randomness. \square

Theorem 47. The following are equivalent for computable functions $f : [0, 1] \rightarrow \mathbb{R}$ having bounded variation:

1. f has Luzin's (N).
2. f reflects weak-2-randomness.
3. f reflects \emptyset' -Kurtz randomness.
4. f reflects $\Delta_1^1(\mathcal{O})$ -randomness.
5. f reflects Z -Kurtz randomness for any $Z \geq \emptyset'$.

Proof. The implication from (1) to (2) is given by Proposition 45. To see that (2) implies (3), first observe that weak-2-randomness reflection implies that $\mu(f(A)) = 0$ for any null Π_2^0 set A , for if $f(A)$ had positive measure then it would certainly contain weak-2-random elements. A $\Pi_1^0(\emptyset')$ set is in particular Π_2^0 , so the image of any \emptyset' -Kurtz test has measure 0, and is thus also an \emptyset' -Kurtz test because the continuous image of a compact set is uniformly compact. The implication (3) \Rightarrow (1) is in Lemma 46.

Finally, the equivalence of (1) and (4) is just Theorem 16, the implication from (1) to (5) is Proposition 22, and the implication from (5) to (3) is Proposition 23. \square

Corollary 48. If a computable function $f : [0, 1] \rightarrow \mathbb{R}$ of bounded variation reflects ML-randomness, then it has Luzin's (N).

The converse is false; see Corollary 53.

Proof. The same argument works as for the implication (2) \Rightarrow (3) in Theorem 47. \square

In this section we have stated all results for $f : [0, 1] \rightarrow \mathbb{R}$ because this is a natural setting in which to consider functions of bounded variation. Of course, our pointwise results are equally true for any computable $f : \mathbb{R} \rightarrow \mathbb{R}$ which is locally of bounded variation.

An often useful result about continuous functions of bounded variation is that they can be obtained as difference between two strictly increasing continuous functions. In light of our investigation of Luzin's (N) for strictly increasing functions, one could wonder why we are not exploiting this property here. There are two obstacles: One the one hand, the computable counterpart of the decomposition result is false: There is a computable function of bounded variation, which cannot be written as the difference between any two strictly increasing computable functions [29]. On the other hand, Luzin's (N) is very badly behaved for sums. For example, for every continuous function f having Luzin's (N) there exists another continuous function g having Luzin's (N) such that $f + g$ fails (N) [20].

6 The relationship to absolute continuity of measures

For increasing functions we see a connection to absolute continuity of measures. Recall that a measure μ is absolutely continuous w.r.t. a measure ν (in symbols $\mu \ll \nu$), if $\nu(A) = 0$ implies that $\mu(A) = 0$. The notions are related through the following observations:

Observation 49. If continuous surjective $f : [0, 1] \rightarrow [0, 1]$ is increasing, then the probability measure μ defined as $\mu(A) = \lambda(f(A))$ is non-atomic, and $\mu \ll \lambda$ iff f has Luzin's (N).

Observation 50. If μ is a non-atomic measure on $[0, 1]$, then its cumulative distribution function $\text{cdf}_\mu(x) := \mu([0, x])$ is a continuous increasing function which has Luzin's (N) iff $\mu \ll \lambda$.

In [2], Bienvenu and Merkle have done an extensive survey of the conditions under which two computable measures μ and ν share the same randoms for a variety of notions of randomness (Kurtz, computable, Schnorr, MLR, and weak-2-random). Two trivial situations where μ -randomness and λ -randomness fail to coincide is if μ has an atom or if $\mu(J) = 0$ for some open interval J . When discussing the connections among Luzin's (N), randomness reflection, and coincidence of randomness notions, we will restrict our attention to computable measures μ which avoid these two degenerate situations. When μ is atomless, cdf_μ is continuous and computable. To say $\mu(J) > 0$ for all open intervals J , it is equivalent to say that cdf_μ is strictly increasing. When the degenerate situations are avoided, cdf_μ is a computable homeomorphism of $[0, 1]$, so cdf_μ^{-1} is also a computable homeomorphism. In this situation, randomness reflection for cdf_μ is exactly randomness preservation for cdf_μ^{-1} .

Proposition 51. Let μ be a non-atomic computable probability measure on $[0, 1]$ with cdf_μ strictly increasing. Then x is μ -MLR (μ -Schnorr random, μ -Kurtz random, μ - Δ_1^1 -random) iff $\text{cdf}_\mu(x)$ is Martin-Löf random (Schnorr random, Kurtz random, Δ_1^1 -random) w.r.t. the Lebesgue measure.

Proof. For any set A , we have $\mu(A) = \lambda(\text{cdf}_\mu(A))$, and cdf_μ and cdf_μ^{-1} are both computable homeomorphisms. We can thus move any relevant test from domain to codomain and vice versa. \square

Therefore, cdf_μ reflects a given notion of randomness exactly when the μ -randoms are contained in the λ -randoms for that notion of randomness. Similarly, cdf_μ^{-1} reflects a given notion of randomness exactly when the λ -randoms are contained in the μ -randoms.

Using our previous results, we obtain the following corollary. The equivalence of (1) and (4) was proved in ([2, Proposition 58]), but the others are new.

Corollary 52. The following are equivalent for a computable probability measure μ .

1. μ is mutually absolutely continuous with the Lebesgue measure.
2. cdf_μ is a homeomorphism and both cdf_μ and cdf_μ^{-1} have Luzin's (N).
3. μ - $\Delta_1^1(\mathcal{O})$ -randomness and $\Delta_1^1(\mathcal{O})$ -randomness coincide.
4. μ -weak-2-randomness and weak-2-randomness coincide.
5. μ -Kurtz(\emptyset')-randomness and Kurtz(\emptyset')-randomness coincide.

Proof. First observe that in all cases above, cdf_μ is a homeomorphism. That is because none of the cases is compatible with μ having an atom or assigning measure 0 to an interval.

Then (1) \iff (2) follows from Observation 50 for the case of cdf_μ , and by similar reasoning for the case of cdf_μ^{-1} .

Since cdf_μ and cdf_μ^{-1} are computable functions of bounded variation, by Theorems 16, and 47, they have Luzin's (N) if and only if they reflect each kind of randomness mentioned in (3)-(6). So the implications (2) \iff (3), (2) \iff (4), and (2) \iff (5) now follow from Proposition 51. \square

Bienvenu and Merkle also give some separations. In particular, they show as [2, Proposition 51 a)] that there exists a computable probability measure μ which is mutually absolutely continuous with Lebesgue measure, but μ -MLR does not coincide with λ -MLR, μ -Schnorr random does not coincide with λ -Schnorr random, and μ -computably random does not coincide with λ -computably random. Essentially, μ is obtained by thinning out the Lebesgue measure around Chaitin's Ω in a way that derandomizes Ω without introducing new null sets.

Corollary 53. Luzin's (N) does not imply any of Martin-Löf randomness reflection, Schnorr randomness reflection nor computable-randomness reflection; even for strictly increasing computable functions.

Proof. If Luzin's (N) were to imply reflection for any of these kinds of randomness, they could be included in the list in Corollary 52 by the same reasoning, but this would contradict Bienvenu and Merkle's result above. \square

We still need to discuss reflection of (unrelativized) Kurtz randomness. In [2, Proposition 56], Bienvenu and Merkle construct a non-atomic computable probability measure μ such that μ -Kurtz random and Kurtz random coincide, yet makes the Lebesgue measure not absolutely continuous relative to μ . The construction is based on an involved characterization of 2-randomness in terms of Kolmogorov complexity obtained by Nies, Stephan and Terwijn [18]. We could already conclude that Kurtz randomness reflection does not imply Luzin's (N) from this, but instead we will provide a direct, more elementary construction in the following. Our separation works "the other way around", that is we obtain a probability measure μ which is not absolutely continuous w.r.t. the Lebesgue measure. This shows that the Lebesgue measure has no extremal position for relative absolute continuity inside the class of measures having the same Kurtz randoms. For comparison, a measure satisfies Steinhaus theorem iff it is absolutely continuous w.r.t. Lebesgue measure [16].

Theorem 54. There is an increasing surjective computable function $f : [0, 1] \rightarrow [0, 1]$ which is not absolutely continuous, yet for any Π_1^0 set A with $\lambda(A) = 0$, it holds that $\lambda(f(A)) = 0$.

Corollary 55. There is a non-atomic probability measure μ such that μ -Kurtz random and Kurtz random coincide, yet $\mu \not\ll \lambda$.

Proof. Let $\hat{\mu}$ be the probability measure whose cumulative distribution function is f , equivalently $\hat{\mu}(B) := \lambda(f(B))$. Since f does not have Luzin's (N), there is some set B with $\lambda(B) = 0$ and $\hat{\mu}(B) > 0$. Let $\mu = \frac{1}{2}\hat{\mu} + \frac{1}{2}\lambda$. Then using the same B , we see that $\mu \not\ll \lambda$. On the other hand, if A is a Π_1^0 set, then $\lambda(A) = 0$ implies $\hat{\mu}(A) = 0$, and thus $\lambda(A) = 0$ if and only if $\mu(A) = 0$. \square

Corollary 56. For increasing computable functions $f : [0, 1] \rightarrow [0, 1]$, reflecting Kurtz randomness does not imply Luzin's (N).

We prepare our construction. Suppose $h : [0, 1] \rightarrow [0, 1]$ is a piecewise linear increasing function, $B \subseteq [0, 1]$ is a finite union of intervals with rational endpoints, and $\delta > 0$. We define a new function

$$\text{Concentrate}(h, B, \delta) : [0, 1] \rightarrow [0, 1]$$

which concentrates $\lambda(B)$ -much measure onto a set of Lebesgue measure at most δ , as follows.

Definition 57 (Definition of Concentrate). Given h, B, δ as above, write $B = \cup_{k < n} I_k$ where I_k are almost disjoint intervals and $h \upharpoonright h^{-1}(I_k)$ is linear (contained in a single piece of the piecewise function). Modify h on each interval $h^{-1}(I_k)$ by substituting a piecewise linear function which alternates between a slope of 0 and a large positive slope. The modification is chosen in a canonical computable way to obtain the following outcomes. Below, \hat{h} denotes $\text{Concentrate}(h, B, \delta)$.

1. $h = \hat{h}$ outside of $h^{-1}(B)$.
2. Letting F denote the union of the pieces of $\hat{h}^{-1}(B)$ which have positive slope, we have $\lambda(F) < \delta$ and $f(F) = B$, and
3. For all x , $|h(x) - \hat{h}(x)| < \delta$.

Lemma 58. Suppose that $(B_n)_{n \in \mathbb{N}}$ is a computable sequence of finite unions of intervals in $[0, 1]$. Define a sequence of functions $(f_n)_{n \in \mathbb{N}}$ inductively by setting $f_0(x) = x$ and

$$f_{n+1} = \text{Concentrate}(f_n, B_n, 2^{-n}).$$

Then $(f_n)_{n \in \mathbb{N}}$ converges uniformly to a computable increasing function f . Furthermore, if there is some $\varepsilon > 0$ such that $\lambda(B_n) > \varepsilon$ for all n , then f fails Lusin's (N).

Proof. The uniform convergence to a computable f follows from the third property in the definition of Concentrate, and f is increasing because each f_n is. Observe that Concentrate never changes the value of h at a break point of h . Therefore, the second property in the definition of Concentrate, which tells us that $f_n(F) = B$ for some F with $\lambda(F) < 2^{-n}$, implies that $f(F) = B$ as well (here we also used the fact that f is continuous and increasing). It follows that f is not absolutely continuous, and thus fails Lusin's (N). \square

Proof of Theorem 54. We construct a computable sequence $(B_n)_{n \in \mathbb{N}}$ such that $\lambda(B_n) > 1/2$ for all n , and argue that the function f constructed as in Lemma 58 satisfies $\lambda(f(P)) = 0$ whenever $P \in \Pi_1^0$ and $\lambda(P) = 0$.

The strategy for a single Π_1^0 class P_e is as follows. Let $C_{e,0}$ be some interval of length ε_e . Let $B_s = [0, 1] \setminus C_{e,s}$. As long as $f_s(P_{e,s}) \cap C_{e,s}$ has measure at least $\varepsilon_e/2$, define $C_{e,s+1} = C_{e,s}$. If $f_s(P_{e,s}) \cap C_{e,s}$ has measure less than $\varepsilon_e/2$, define $C_{e,s+1} = (f_s(P_{e,s}) \cap C_{e,s}) \cup C$, where C is a new interval or finite union of intervals almost disjoint from $\cup_{t \leq s} C_{e,t}$. Choose C so that that $C_{e,s+1}$ has measure ε_e , if possible; if this is not possible, choose C so that $\cup_{t \leq s+1} C_{e,t} = [0, 1]$. In the latter case the measure of $C_{e,s+1}$ may be less than ε_e and this is also fine. If we reach this degenerate situation, we also stop checking the measures and simply let $C_{e,t} = C_{e,s+1}$ for all $t > s$.

We claim that if $\lambda(P_e) = 0$, then $\lambda(f(P_e)) = 0$. Suppose at some stage s we have that the measure of $f_s(P_{e,s}) \cap C_{e,s}$ is greater than $\varepsilon_e/2$. If this continues for all $t > s$, then f and f_s coincide on the set $J := f^{-1}(C_{e,s})$. It follows that f is piecewise linear on J , but $f(P_e \cap J)$ has positive measure; this is impossible since P_e has measure 0. We conclude that nothing lasts forever; eventually we do reach a stage s where $\cup_{t \leq s} C_{e,t} = [0, 1]$. Since $C_{e,s}$ never changes again,

f and f_s again coincide on $J := f^{-1}(C_{e,s})$. Observe also that $P_e \subseteq J$. Since f_s is piecewise linear and $\lambda(P_e) = 0$, we also have $\lambda(f_s(P_e)) = 0$, and thus $\lambda(f(P_e)) = 0$.

The above strategy works purely with negative requirements, specifically freezing f on $f_s^{-1}(C_{e,s})$. If other requirements also freeze f on other places, it has no effect on the proof above. The only thing to consider when combining requirements is that we need to make sure $\lambda(B_s) > 1/2$ for all s , where we now define

$$B_s = [0, 1] \setminus \bigcup_{e < s} C_{e,s}.$$

Since we always have $\lambda(C_{e,s}) \leq \varepsilon_e$, we can keep the sets B_s large by choosing the values of ε_e to satisfy $\sum_e \varepsilon_e < 1/2$. \square

7 Π_1^1 -hardness of randomness reflection

If we do not restrict the domain of the functions to (locally) compact spaces, then essentially any form of randomness reflection is Π_1^1 -hard. We show a construction which yields a function having either null range, or is surjective when restricted to a specific null subset of its domain. In particular, our construction is independent of the randomness notions involved.

Theorem 59. Let $K, L \subseteq [0, 1]^2$ be non-empty sets containing only Kurtz randoms. Then “whenever $f(x) \in K$, then already $x \in L$ ” is a Π_1^1 -hard property of continuous functions $f : ([0, 1] \setminus \mathbb{Q}) \times [0, 1] \rightarrow [0, 1]^2$.

Proof. It is well-known that $[0, 1] \setminus \mathbb{Q}$ and $\mathbb{N}^{\mathbb{N}}$ are homeomorphic, and even computably so. We identify the spaces in such a way that the Lebesgue measure induced on $\mathbb{N}^{\mathbb{N}}$ satisfies $\lambda(\{p \in \mathbb{N}^{\mathbb{N}} \mid \forall n \ p_{2n} = p_{2n+1}\}) = 0$.

We construct a function $f_T : \mathbb{N}^{\mathbb{N}} \times [0, 1] \rightarrow [0, 1]^2$ from a countably-branching tree T . First, we modify T to obtain $\hat{T} = \{w_0 w_0 w_1 w_1 \dots w_{n-1} w_{n-1} w_n \mid w \in T\} \cup \{w_0 w_0 w_1 w_1 \dots w_{n-1} w_{n-1} w_n w_n \mid w \in T\}$. Clearly, T is well-founded iff \hat{T} is, and $[\hat{T}]$ contains no Kurtz-randoms (so in particular, $[\hat{T}] \times [0, 1] \cap L = \emptyset$). For any $p \in \mathbb{N}^{\mathbb{N}}$, let $|T, p| = n$ iff n is minimal such that $p \upharpoonright n \notin \hat{T}$, and $|T, p| = \infty$ if $p \in [\hat{T}]$.

Let $s_\infty : [0, 1] \rightarrow [0, 1]^2$ be a computable space-filling curve, and let $(s_n)_{n \in \mathbb{N}}$ be a computable fast Cauchy sequence converging to s_∞ such that any $s_n([0, 1])$ is a finite union of line segments. We then define $f_T(p, x) = s_{|T, p|}(x)$. This construction is computable in T . We claim that f_T has our reflection property iff T is well-founded.

If T is well-founded, then the range of f_T is $\bigcup_{n \in \mathbb{N}} s_n([0, 1])$. Since any $s_n([0, 1])$ is a null Π_1^0 -set, we see that f_T never takes any Kurtz random values (in particular, none in K), and thus vacuously, if $f(x) \in K$ then $x \in L$. The argument in fact establishes that for arbitrary T , whenever $p \notin [\hat{T}]$ then $f_T(p, x)$ is not Kurtz random regardless of x .

Now assume that T is ill-founded and that $y \in K$. We find that $f_T^{-1}(\{y\}) = [\hat{T}] \times s_\infty^{-1}(\{y\})$. Since \hat{T} is illfounded and s_∞ is space-filling, this set is non-empty. But by construction of \hat{T} , it cannot contain any elements of L . Hence, f_T does not have our reflection property. \square

8 A glimpse at related notions

As a slight digression, we have a look at related properties of functions, namely those where the image of null sets are required to belong to some other ideals of *small* sets, such as being

countable or being meager. These properties were investigated by Sierpinski [25] and Erdős [7], amongst others. Our results are formulated in some generality, but as a consequence, we do see that we do not get any “regular” functions with these properties. In contrast, Erdős showed that under CH there is a bijection $f : \mathbb{R} \rightarrow \mathbb{R}$ mapping meager sets to null sets with f^{-1} mapping null sets to meager sets.

- Theorem 60.** (1) If A is a nonnull Σ_1^1 set and f is a continuous function mapping any null subset of A to a countable set, then the range of f restricted to A is countable. In particular, if A is an interval, then range of f restricted to A is a constant function.
- (2) Assume CH, there is a function f mapping any null set to a countable set such that the range of f is \mathbb{R} , and $f(A)$ is uncountable for any nonnull set A but for every y , $f^{-1}(y)$ is an uncountable Borel null set.
- (3) If A is a nonnull set and f is a continuous function mapping any null subset of A to a meager set, then the range of f restricted to A is meager. In particular, if A is an interval, then range of f restricted to A is a constant function.
- (4) If f is a measurable function and maps a null set to a meager set, then the range of f is meager. In particular, if f is continuous with the property, then f is constant.
- (5) If f has the Baire property and maps a meager set to a null set, then the range of f is null. In particular, if f is continuous with the property, then f is constant.

Proof. (1). Fix a real x so that f is computable in x and A is $\Sigma_1^1(x)$. Now for any real $z \in A$, let g be a $\Delta_1^1(\mathcal{O}^{x \oplus z})$ -generic real. Then z cannot be Martin-Löf random relative to g and so there must be a $\Delta_1^1(g)$ -null set G so that $z \in G \cap A$. By the assumption, $f(G \cap A)$ is a $\Sigma_1^1(x \oplus g)$ -countable set and so every real in $f(G \cap A)$ is hyperarithmetically below $x \oplus g$. In particular, $f(z) \leq_h x \oplus g$. Since f is computable in x , we also have that $f(z) \leq_h x \oplus z$. Then $f(z) \leq_h x$ since g is $\Delta_1^1(\mathcal{O}^{x \oplus z})$ -generic real. By the arbitrariness of z , the range of f restricted to A is countable. So if A is an interval, then range of f restricted to A is a constant function.

(2). Fix an enumeration of nonempty G_δ -null sets $\{G_\alpha\}_{\alpha < \aleph_1}$ and all the reals $\{y_\alpha\}_{\alpha < \aleph_1}$. We define f and $\{\beta_\alpha\}_{\alpha < \aleph_1}$ by induction on α .

At stage 0, define $f(x) = y_0$ for any $x \in G_0$. Define $\beta_0 = 0$.

At stage $\alpha < \aleph_1$, let β_α be the least ordinal γ so that $G_\gamma \setminus \bigcup_{\alpha' < \alpha} (\bigcup_{\gamma' \leq \beta_{\alpha'}} G_{\gamma'})$ is uncountable. Define $f(x) = y_\alpha$ for any $x \in G_\gamma \setminus \bigcup_{\alpha' < \alpha} (\bigcup_{\gamma' \leq \beta_{\alpha'}} G_{\gamma'})$.

Clearly the range of f is \mathbb{R} . Moreover, for any $\alpha < \aleph_1$, $f^{-1}(y_\alpha) = G_{\beta_\alpha} \setminus \bigcup_{\alpha' < \alpha} (\bigcup_{\gamma' \leq \beta_{\alpha'}} G_{\gamma'})$ is an uncountable Borel null set. Now for any null set A , there must be some $\alpha < \aleph_1$ so that $A \subseteq G_\alpha$. By the construction, $f(A) \subseteq f(G_\alpha) \subseteq \{y_\beta \mid \beta \leq \alpha\}$ is a countable set.

(3). Fix a real x so that f is computable in x restricted to A . Fix a $2-x$ -random real $r \in A$. Then $f(r) \leq_T x \oplus r$. But $f(r)$ cannot be $2-x$ -generic (see [18]). So the range f restricted to $A \cap \{r \mid r \text{ is a } 2-x\text{-random}\}$ is meager. But $A \setminus \{r \mid r \text{ is a } 2-x\text{-random}\}$ is a null set. So, by the assumption on f , the range of f restricted to A is meager.

(4). Suppose that f is measurable function and maps a null set to a meager set. Without loss of generality, we may assume that the domain of f is $[0, 1]$. Then there is a sequence closed sets $\{F_n\}_{n \in \omega}$ so that $[0, 1] \setminus \bigcup_{n \in \omega} F_n$ is null and f restricted to F_n is continuous for every n . By (3), the range of f restricted to F_n is a meager set. So the range of f restricted to $\bigcup_{n \in \omega} F_n$ is

also a meager set. Note that $[0, 1] \setminus \bigcup_{n \in \omega} F_n$ is null. So the range of f is meager. In particular, if f is continuous with the property, then f is constant.

(5). This is dual to (4). □

9 Outlook

The most prominent avenue of future research seems to be the resolution of Question 28, asking for a separation (or equivalence proof) of Δ_1^1 -randomness reflection and $\Delta_1^1(\mathcal{O})$ -randomness reflection. There are a few further aspects that merit further investigation, though.

Topological properties While we have not been systematic in exploring the impact of topological properties of the domain (and maybe codomain) of the functions we explore, we observe that our proofs differ in the requirements they put on the spaces involved. For example, the majority of the arguments presented in Sections 3.1 and 3.2 are relying just on the theory of randomness, and are thus applicable to any space where randomness works as usual (see [10]). In Section 3.3, (local) compactness of the domain is a core ingredient in our arguments. In Section 5 we do use particular properties of the reals, in particular connectedness. Further investigation of how topological properties of spaces relate to how randomness reflection behaves for functions on them seems warranted.

Formalizing randomness reflection With the exception of Theorem 59, we have only considered symmetric notions of randomness reflection: Whenever $f(x)$ is random in some sense, we demand that x is random in the very same sense. While this seems natural, a downside is that we do not get trivial implications between different notions of $\mathbf{\Pi}_2^0$ -type randomness reflection. We could consider the full *square* of reflection notions, (K, L) -randomness reflection being that whenever $f(x)$ is K -random, then x is L -random for randomness notions K, L . An extremal version also makes sense, where we just ask for when the image of all non-randoms under f has positive measure. Whenever the latter property holds for some randomness notion L , then f cannot have (K, L) -randomness reflection for any randomness notion K at all. We typically prove non-randomness reflection in this manner.

It seems too early to pass judgement on what precise formulations of randomness reflection will ultimately be the most fruitful.

Functions beyond measurability So far, the most general class of functions we considered for Luzin's (N) were the measurable functions. If we consider unrestricted functions in full generality, it is unsurprising that we quickly move beyond the confines of ZFC. For example, we are wondering whether Corollary 17 holds for all functions having Luzin's (N)? An investigation into such questions is on its way by Yinhe Peng and the third author.

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