## LOWNESS FOR BOUNDED RANDOMNESS

ROD DOWNEY AND KENG MENG NG

ABSTRACT. In [3], Brodhead, Downey and Ng introduced some new variations of the notions of being Martin-Löf random where the tests are all clopen sets. We explore the lowness notions associated with these randomness notions.

## 1. INTRODUCTION

The underlying idea behind algorithmic randomness is that *randomness* can only be understood through *computational* considerations. This interpretation of randomness means that the object in question *avoids* simpler algorithmic descriptions, either through effective betting, effective regularities or effective compression.

This idea meas that we avoid any idea that there is "true" randomness, and work with the idea that we only have calibrations of randomness according to the sensitivity of the measuring tool. That is, exactly what we mean here by "effective" delineates notions of algorithmic randomness. A major theme in the area of algorithmic randomness seeks to calibrate notions of randomness by varying the notion of effectivity. For example, classical Martin-Löf randomness<sup>1</sup> uses tests, shrinking connections of c.e. open sets whose measure is bounded by effective bounds, whereas Schnorr randomness has the tests of some precise effective measure. We then see that Schnorr and Martin-Löf randomness are related but can have very different properties; for example outside the high degrees they coincide, but the lowness concepts are completely disjoint.

Given that we *define* randomness via computation it is natural to ascertain how randomness so defined relates to computational power. To do this we relate algorithmic randomness to measures of relative computability, like the Turing degrees. If something is random, can it have high computational power, for instance? A classic result in this area is Stephan's theorem [18] that if a Martin-Löf real is random and has enough computational power to be able to compute a  $\{0, 1\}$ -valued fixed point free function then it must be Turing complete.

The goal of the present paper is to study some some new variations of algorithmic randomness introduced by the authors and Paul Brodhead in [3] exploring both of the themes above. In particular, we study what we call bounded variations of the notion of Martin-Löf randomness where the tests

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<sup>&</sup>lt;sup>1</sup>We assume that the reader is familiar with the basic notions of algorithmic randomness as found in the early chapters of either Downey-Hirschfeldt [9] or Nies [17].

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are all finite. These notions generalize the notion of Kurtz (or weak) randomness but are incomparable with both Schnorr and computable randomness. As more precisely defined in the next section, together with Brodhead we defined what we called computably bounded (CB) and finitely bounded (FB) notions of finite Martin-Löf randomness. The [3] paper showed that the bounded notions of randomness we meet in the next section were strongly related to degree classes like the totally  $\omega$ -c.e. degrees, and notions of initial segment complexity.

The goal of the present paper is to explores the associated lowness notions. Aside from the intrinsic interest in this, one good reason for such study is to give insight into other studied lowness and randomness notions. And indeed this turns out to be the case. In the case of CB randomness we construct continuum many reals low for the concept, and disciver that that they are a subclass of the recently discovered reals low for Demuth randomness. (as per Bienvenu, Downey, Greenberg, Nies and Turetsky [2]) We do this using an apparent extension of what is called in [2], BLR-traceability. It remains an open question if our lowness class coincides with lowness for Demuth randomness. In the case of FB randomness (which is simply that the ML tests are all finite) the lowness class coincides with the K-trivials. In some sense, this last result is somwhat unexpected, and somehow says something deep about the nature of K-triviality. That is because it means we can characterize lowness for ML-randomness via a much weaker notion of randomness which seems strange.

We give the basic definitions in Section 2 and review some elementary properties from [3]. In Section 3 we include a proof of a basic result from [3] as its proof and statement are on importance for the present paper. In Section 4 we construct reals low for CB randomness, and finally in sectionktriv we show that lowness for FB randomness and K-triviality coincide.

### 2. NOTATION

If W is a finite set then #W denotes the cardinality of W.  $|\sigma|$  denotes the length of a finite string  $\sigma$ . We work in the Cantor space  $2^{\omega}$  with the usual clopen topology. The basic open sets are of the form  $[\sigma]$  where  $\sigma$  is a finite string, and  $[\sigma] = \{X \in 2^{\omega} \mid X \supset \sigma\}$ . We fix some effective coding of the set of finite strings, and we freely identify finite strings with their code numbers. We denote  $[W] = \cup \{[\sigma] : \sigma \in W\}$  as the  $\Sigma_1$  open set associated with the c.e. set W.  $\mu([W])$  denotes Lebesgue measure, and we write  $\mu(W)$ instead of  $\mu([W])$ . We let \* be the string concatenation symbol. We let  $D_n$  be the  $n^{th}$  canonical finite set. If W is an open set and  $\sigma \in 2^{<\omega}$  we let  $\mu(D \mid [\sigma]) = \frac{\mu(D \cap [\sigma])}{\mu([\sigma])}$ , i.e. the relative measure of D in  $[\sigma]$ .

**Definition 2.1** (Brodhead, Downey and Ng [3]). (a) A Martin-Löf (ML) test is a uniform c.e. sequence  $\{U_n\}_{n\in\omega}$  of sets  $U_n$  such that  $\mu(U_n) < 2^{-n}$ .

- (b) A Martin-Löf test  $\{U_n\}_{n\in\omega}$  is finitely bounded (FB) if  $\#U_n < \infty$  for every n.
- (c) A Martin-Löf test  $\{U_n\}_{n\in\omega}$  is computably bounded (CB) if there is some total computable function f such that  $\#U_n \leq f(n)$  for every n.

(d) A real  $X \in 2^{\omega}$  passes a *CB*-test (*FB*-test)  $\{U_n\}_{n \in \omega}$  if  $X \notin \bigcap_n [U_n]$ . A real  $X \in 2^{\omega}$  is computably bounded random if X passes every *CB*-test. X is finitely bounded random if it passes every *FB*-test. We always assume that in any *FB*-test  $\{U_n\}$ , each  $U_n$  contains only pairwise incomparable strings. For a *CB*-test we are unable to make such a convention.

These two notions of randomness are weaker than Martin-Löf randomness, although they imply Kurtz randomness. The obvious implications are:

$$ML$$
-random  $\rightarrow$   $FB$ -random  $\rightarrow$   $CB$ -random  $\rightarrow$  Kurtz random  $\checkmark$  Schnorr random

In [3], with Brodhead we proved that no implications hold other than those stated in the diagram. We did this as follows. First, we showed that there is a  $\Delta_3^0$  1-generic real which is *FB*-random while no Schnorr random is weakly 1-generic. No incomplete c.e. degree can compute a *FB*-random (Proposition 3.1(i)). However, some incomplete c.e. degree bounds a *CB*random since in [3] we prove the following.

**Theorem 2.2** (Brodhead, Downey and Ng [3]). Suppose A is a c.e. real. The following are equivalent.

- (i)  $deg_T(A)$  is not totally  $\omega$ -c.a.,
- (ii)  $deg_T(A)$  contains a CB-random,
- (iii) There is some c.e. real  $B \leq_T A$  which is CB-random,
- (iv) There is some  $B \leq_T A$  which is CB-random.

Here, a degree **a** is called *totally c.e.* iff every function  $f \leq_T \mathbf{a}$  has a limit lemma approximation  $f(x) = \lim_s g(x, s)$  where there is a computable h with  $|\{s: g(x, s+1) \neq g(x, s)\}| < h(x)$  for all x. (See Downey, Greenberg, Weber [8], Downey and Greenberg [6, 7] and Barmpalias, Downey and Greenberg [1]. It is known that there are low c.e. degrees that are not totally  $\omega$ -c.a..

Finally in [3], we showed that

**Proposition 2.3** (Brodhead, Downey and Ng [3]). Every CB-random is of effective packing dimension 1.

Proof. For completeness we give a proof of Proposition 2.3 as it is quite short. Suppose  $K(\alpha \upharpoonright n) \leq cn$  for all  $n \geq N$  for some  $N \in \mathbb{N}$  and c < 1 is rational. Fix a computable increasing sequence of natural numbers  $\{n_i\}$  al l larger than N, such that  $n_i > \frac{i}{1-c}$  for all i. Now define a CB- test  $\{V_i\}$  by the following:  $V_i := \{\sigma \in 2^{n_i} \mid K(\sigma) \leq cn_i\}$ . Here we have  $\#V_i \leq 2^{cn_i}$ .  $\Box$ 

Lathrop and Lutz [16] showed that there is a computably random set X such that for every order function g,  $K(X \upharpoonright n) \leq K(n) + g(n)$  for almost every n. Hence X cannot be CB-random, by Proposition 2.3. This gives the last separation for the diagram.

As we noted in [3], these finitary notion of randomness these notions of randomness turn out to have strong relationships with degrees classes hitherto unrelated to algorithmic randomness. We will show that FBrandomness and Martin-Löf -randomness coincide on the  $\Delta_2^0$  sets but are distinct on the  $\Delta_3^0$  sets (Theorem 3.1). There is one other known restriction on such reals.

# **Proposition 2.4** (Brodhead, Downey and Ng [3]). *No CB-random is c.e. traceable.*

Kurtz showed that every nonzero c.e. degree contains a Kurtz random real, but (by Theorem 2.2 above) the degrees containing CB-random reals is a subclass of the c.e. degrees : those that are not totally  $\omega$ -c.a.. This is a class of c.e. degress introduced by Downey, Greenberg and Weber [8] to explain certain "multiple permitting" phenomena in degree constructions such as "critical triples" in the c.e. degrees, and a number of other constructions as witnessed in the subsequent papers Barmpalias, Downey and Greenberg [1] and Downey and Greenberg [7]. This class extends the notion of array noncomputable reals, and correlates to the fact that all CB random reals have effective packing dimension 1 (Theorem 2.3). Downey and Greenberg [6] having previously showed that the c.e. degrees containing reals of packing dimension 1 are exactly the array noncomputable reals. Brodhead, Downey and Ng [3] also show that if a c.e. degree **a** contains a CB random then every (not necessarily c.e.) degree above **a** contains a CB random as well. From all of this, we see that there remain a lot to understand for this class.

# 3. Basic results

One of the basic results shown in [3] is that the notion of *FB*-randomness and Martin-Löf -randomness coincide on the  $\Delta_2^0$  sets, and they differ on the  $\Delta_3^0$  sets. Since the proof is relevant to this paper, we include it here.

**Proposition 3.1** (Brodhead, Downey and Ng [3]). (i) Suppose  $Z \leq_T \emptyset'$ . Then Z is ML-random iff Z is FB-random.

(ii) There is some  $Z \leq_T \emptyset''$  such that Z is 1-generic, FB-random and not ML-random.

Proof. (i): Given an approximation  $Z_s$  of Z, and suppose  $\{U_x\}$  is the universal ML-test where  $Z \in \bigcap_x [U_x]$ . Enumerate an FB-test  $\{V_x\}$  by the following: at stage s, enumerate into  $V_x$ , the string  $Z_s \upharpoonright n$  for the least n such that  $Z_s \upharpoonright n \in U_x[s]$ . Then,  $\{V_x\}$  is uniformly c.e., where  $\mu(V_x) \leq \mu(U_x) < 2^{-x}$  for all x. Clearly  $Z \in [V_x]$  for all x. We know  $Z \upharpoonright n \in U_x$  for some least n, and let s be a stage such that  $Z_s \upharpoonright n$  is correct and  $Z \upharpoonright n$  has appeared in  $U_x[s]$ . Then,  $Z \upharpoonright n$  will be in  $V_x$  by stage s, and we will never enumerate again into  $V_x$  after stage s.

(ii): We build  $Z = \bigcup_s \sigma_s$  by finite extension. Let  $\{U_x\}$  be the universal ML-test, and  $\{U_x^e\}_x$  be the  $e^{th}$  ML-test. Assume we have defined  $\sigma_s$ , where for all e < s, we have

- all infinite extensions of  $\sigma_s$  are in  $U_e$ ,
- if  $\#U_x^e < \infty$  for all x, then  $\exists k$  such that no infinite extension of  $\sigma_s$  can be in  $U_k^e$ .

Now we define  $\sigma_{s+1} \supset \sigma_s$ . Firstly, find some  $\tau \supseteq \sigma_s$  such that all infinite extensions of  $\tau$  are in  $U_s$ ; such  $\tau$  exists because  $\{U_e\}$  is universal. Let  $k = |\tau|$ . Next, ask if  $\#U_k^s < \infty$ . If not, let  $\sigma_{s+1} = \tau * 0$  and we are done. If yes, then figure out exactly the strings  $\rho_i$  such that  $[U_k^s] = \cup \{[\rho_1], [\rho_2], \cdots, [\rho_n]\}$ . We

cannot have  $[U_k^s] \supseteq [\tau]$  since  $\mu(U_k^s) < 2^{-k}$ , so there has to be some  $\sigma_{s+1} \supset \tau$ such that  $[\sigma_{s+1}] \cap [U_k^s] = \emptyset$ , by the finiteness of  $U_k^s$ . We can figure  $\sigma_{s+1}$  out effectively from  $\rho_1, \rho_2, \cdots, \rho_n$ . Next extend  $\sigma_{s+1}$  (if necessary) to meet  $W_s$ . Clearly the properties above continue to hold for  $\sigma_{s+1}$ . All questions asked can be answered by the oracle  $\emptyset''$ .

A set A is  $\Delta_2^0$ -jump dominated if for every partial A-computable function  $\Phi^A$  there is a  $g \leq_T \emptyset'$  such that  $g(x) > \Phi^A(x)$  for every x. This notion has also been called "weakly jump traceable", and implies that the set is  $GL_1$ . The proof of (i) above also shows that ML- and FB-randomness coincides over the reals which are  $\Delta_2^0$ -jump dominated.

**Corollary 3.2.** Suppose that A is  $\Delta_2^0$ -jump dominated. Then A is Martin-Löf random iff A is FB-random.

There exists Martin-Löf randoms which are  $\Delta_2^0$ -jump dominated. For example it is easy to see that each Demuth random is  $\Delta_2^0$ -jump dominated.

## 4. CB Lowness and traceability

We investigate the lowness notions associated with FB- and CB-randomness. We call a real A low for FB-randomness if every FB-random real is FBrandom relative to A. A is low for FB-tests if for every A-relative FB-test  $\{U_x^A\}_{x\in\omega}$  there is an FB-test  $\{E_x\}_{x\in\omega}$  such that  $\cap_x[U_x^A] \subseteq \cap_x[E_x]$ . We can make similar definitions for CB-randomness.

Recall that Cole and Simpson [4] defined a function  $f : \omega \mapsto \omega$  to be BLR(A) for an oracle A to mean that there exists  $g \leq_T A$  and a computable function h such that for every  $x, f(x) = \lim_{s\to\infty} g(x,s)$  and  $chg_g(x) < h(x)$ . Here we denote  $\#\{s \mid g(x,s) \neq g(x,s+1)\} = chg_g(x)$ , the mind change function of g. Hence BLR(A) is the class of all functions which are  $\omega$ -c.e. relative to A, where the relativization is partial. Bienvenu, Downey, Greenberg, Nies and Turetsky [2] then called a set A BLR-traceable if there is a computable function h such that for every  $x, \#W_{k(x)} < h(x)$  and  $f(x) \in W_{k(x)}$ .

BLR-traceability was shown in [2] to be crucial in understanding lowness for Demuth randomness. Being BLR-traceable allows us to obtain a c.e. trace (with few values) for the canonical index of a clopen Demuth test relative to A. We can then use this to build a Demuth test covering a given A-Demuth test. Unfortunately BLR-traceability appears too weak to imply lowness for CB-randomness, because in order to approximate a given CBtest relative to A, we have to approximate the individual neighbourhoods enumerated into the test. This calls for a strengthening of BLR-traceability which we call faithfully BLR-traceable.

**Definition 4.1.** Given functions  $f, g : \omega^2 \mapsto \omega$  such that  $\lim_{s\to\infty} f(x,s)$  and  $\lim_{s\to\infty} g(x,s)$  both exist, we say that g is a *faithful BLR witness* for f if for every x,  $\lim_{s\to\infty} f(x,s) = \lim_{s\to\infty} g(x,s)$  and  $\cup \{f(x,s) \mid s \in \omega\} \supseteq \cup \{g(x,s) \mid s \in \omega\}$ . That is, g does not approximate f wastefully, and introduces no noise into the approximation.

We say that A is faithfully BLR-traceable if for every computable order h and every BLR(A)-approximation for a function  $f \in BLR(A)$ , there is

an  $\omega$ -c.e. function g and a function  $n : \omega \mapsto \omega$  such that for every x,  $\sum_{y \leq x} h(y) \leq n(x) < \sum_{y \leq x} h(y)$  and f has faithful BLR withness g(n).

Note that there are no requirement on the complexity of n. It is easy to see that in the definition of faithful *BLR*-traceability we may replace "for every computable order h" with "for some computable order h".

Fact 4.2. Faithfully BLR-traceable implies BLR-traceable.

Being faithfully BLR-traceable means more than having an  $\omega$ -c.e. trace for  $f \in BLR(A)$  – it means that we can identify h(x) many distinct attempts at approximating f(x) faithfully.

**Fact 4.3.** If A is faithfully BLR-traceable with respect to a constant function h(x) = c then A is computable.

*Proof.* Assume c = 1. Define the A-computable function  $f(x, s) = A \upharpoonright x$ . Fix the  $\omega$ -c.e. function g which is a faithful BLR witness for f. For every  $x, g(x, 0) = A \upharpoonright x$  and hence A is computable. If c > 1 we use g to construct a  $\Pi_1^0$  class containing A with at most c many paths.

**Lemma 4.4.** If A is faithfully BLR-traceable and of hyperimmune-free degree then A is low for CB-tests.

Proof. Fix a CB-test  $\{U_x^A\}$  relative to A. We may assume that  $\mu(U_x^A) < 2^{-2x}$  for each x. Let f(x,s) approximate the canonical index for  $U_x^A$ . Since A is HIF,  $f \in BLR(A)$ , with computable bound h on  $\#U_x^A$ . Let g and n be given such that g is  $\omega$ -c.e. and g(n) is a faithful BLR-witness for f with respect to the identity order function. For each x there are x many possibilities for n(x) in  $I_x = \{\sum_{y < x} y, \cdots, \sum_{y \le x} y - 1\}$ . Build  $T_x$  by copying, for each  $z \in I_x$ , the sets with canonical indices  $g(z,0), g(z,1), \cdots$  until we find some s such that  $\mu(\cup_{t \le s}[D_{g(z,t)}]) \ge 2^{-2x}$ , or  $\# \cup_{t \le s} D_{g(z,t)} > h(x)$ . Then  $T_x$  is clearly a CB-test, and since g(n) faithfully BLR-trace  $f, [T_x] \supseteq [U_x^A]$ . □

**Proposition 4.5.** There is a perfect class of sets which are all low for CB-randomness.

*Proof.* In [2] it was shown that there is a perfect  $\Pi_1^0$  class of sets which are all *BLR*-traceable. It is easy to see that the same proof shows that there is a perfect  $\Pi_1^0$  class of sets which are all faithfully *BLR*-traceable. We can then take a perfect subclass of reals which are all of hyperimmune-free degree. The proposition then follows from Lemma 4.4.

**Theorem 4.6.** If A is low for CB-randomness then A is of hyperimmunefree degree.

Proof. Suppose A is of hyperimmune degree. Fix a strictly increasing total function  $f \leq_T A$  such that f escapes domination by every computable function. We build  $Z \leq_T A'$  such that Z is CB-random but not CB-random relative to A. The construction will be computable in A. We let  $\{E_x^k\}_{x\in\omega}$  be the  $k^{th}$  possible CB-test with cardinality bound  $g_k(x)$ , where  $g_k$  is a partial computable function with domain an initial segment of  $\omega$ . We assume that for every  $k, x, \mu(E_x^k) < 2^{-x}$ . We also assume that  $E_x^k = \emptyset$  if  $g_k(x)$  has not

yet converged, and that  $\#E_x^k < g_k(x)$  otherwise. The construction maintains global parameters  $\{U_x\}_{x\in\omega}$ ,  $\sigma_i$  and  $k_i$ .  $\{U_x\}_{x\in\omega}$  will be the *CB*-test relative to *A* which *Z* fails, and  $\sigma_i \in 2^{<\omega}$  and  $k_i \in \omega$  are the parameters of requirement  $P_i$ .

Requirement  $P_i$  will ensure that Z passes  $\{E_x^i\}_{x\in\omega}$  if the latter is a CB-test. If E is a c.e. open set,  $\tau$  is a string and s is a stage we let  $Survivor(E,\tau,s)$  be the lex-least string  $\eta \supseteq \tau$  such that  $|\eta| = s$  and  $[\eta] \cap [E_s] = \emptyset$ . We adopt the convention that  $E_s$  only contains strings of length  $\langle s, \text{ so } Survivor(E,\tau,s)$  is undefined iff  $[E_s] \supseteq [\tau]$ .

Construction of  $\{U_x\}$ : At stage s = 0 set  $\sigma_0 = 00$  and  $k_0 = 0$ . At stage s > 0 we say that  $P_i$  requires attention if  $P_i$  has been started and there is some least  $j_0 > 0$  such that  $g_i(|\sigma_i| + j_0) \downarrow < f(k_i + j_0)$ , and a new element has entered  $E^i_{|\sigma_i|+j_0}$ . If this is the first time we discover  $j_0$  then we also say that  $P_i$  requires attention, regardless of the enumeration of  $E^i_{|\sigma_i|+j_0}$ .

At stage s pick the least i < s such that  $P_i$  requires attention. If i exists we act for  $P_i$  by doing the following. For each  $j < j_0$  enumerate  $\sigma_i * 0^j$  into  $U_{k_i+j}$ . Initialize  $P_m$  for every m > i (meaning that we cancel the values of  $\sigma_i, k_i$ ). Set  $\sigma_{i+1} = Survivor(E^i_{|\sigma_i|+j_0}, \sigma_i * 0^{j_0}, s) * 0^t$  where t is a fresh number. Set  $k_{i+1} = k_i + j_0$ .

If no  $P_i$  requires attention find the largest *i* such that  $P_i$  is not yet started. For this  $P_i$ ,  $\sigma_i$ ,  $k_i \downarrow$ . Declare the requirement to be started. Set  $\sigma_{i+1} = \sigma_i * 0^t$  for a fresh number *t* and  $k_{i+1} = k_i + 1$ .

Finally for every  $P_i$  we put  $\sigma_i$  into  $U_{k_i}$  if these parameters are defined.

Verification: It is clear that every call for Survivor during the construction returns a value. We verify that  $\{U_x\}$  is an A-relative CB-test. The sequence  $\{U_x\}$  is clearly uniformly c.e. in oracle f. Checking the construction reveals that at every stage  $|\sigma_i| > k_i$  holds whenever these parameters are defined. In fact every time  $P_i$  is initialized  $|\sigma_i|$  is picked to have a length (never seen before by the construction) larger than  $k_i$ . Now fix an x and consider  $U_x$ . At every stage there is at most one  $P_i$  contributing to  $U_x$ . In fact if  $P_i$  contributes to  $U_x$  then  $k_i \leq x < k_{i+1}$  and  $i \leq x$ . Each requirement  $P_i$  can contribute at most one string of the form  $\sigma_i * 0^j$  into  $U_{k_i+j}$  before it is initialized. Since  $|\sigma_i| > k_i$  this means that  $|\sigma_i * 0^j| > x$ . If  $P_i$  gets initialized and later enumerates another string  $\sigma'_i * 0^{j'}$  into  $U_x$  again then we must have  $|\sigma'_i * 0^{j'}| > |\sigma'_i| > |\sigma_i * 0^j| > x$ , as lengths are always chosen fresh. The total measure of  $P_i$ 's contribution is at most  $2^{-x}$ . Hence  $\mu(U_x) \leq (x+1)2^{-x}$ .

Before we verify that  $\#U_x$  is computably bounded in A, we need a technical lemma.

**Lemma 4.7.** There is  $F(i, x) \leq_T A$  such that for every x and every i, the number of stages where  $P_i$  is initialized when  $k_i \leq x$  is at most F(i, x).

*Proof.* Fix x. We define F(i, x) recursively in i. Clearly F(0, x) = 0. Assume  $F(0, x), \dots, F(i-1, x)$  has been defined. Fix j < i. If  $P_j$  initializes  $P_i$  when  $k_i \leq x$  then  $k_j < k_i \leq x$ . If  $P_j$  does this then  $P_j$  has required attention which means that a new element has entered  $E_{|\sigma_j|+z}^j$  where  $k_j + z \leq k_i$  (unless this is the first time  $P_j$  is receiving attention). Hence  $P_j$  can do this at most  $1 + g_j(|\sigma_j| + z) < 1 + f(k_j + z) \leq 1 + f(k_i) \leq 1 + f(x)$  times before  $P_j$  itself is initialized. When  $P_j$  is finally initialized  $k_j < x$ . So  $P_j$  initializes

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 $P_i$  at most (1 + F(j, x))(1 + f(x)) many times. We can use this to compute F(i, x).

Next we verify that  $\#U_x \leq f(x)$ . Now fix  $i \leq x$ . How many times can  $P_i$  contribute to  $U_x$ ? Before  $P_i$  can contribute a second element to  $U_x$  it has to be first initialized by  $P_j$ , j < i and at the point of initialization  $k_i \leq x$ . By Lemma 4.7  $\#U_x \leq \sum_{i \leq x} 1 + F(i, x)$ , and so  $\{U_x\}$  is a *CB*-test relative to *A*.

It is easy to check that every requirement is initialized finitely often, and that for every  $i, \sigma_{i+1} \supset \sigma_i$  holds at every stage. Each  $\sigma_i$  must eventually get defined, so we let  $\tilde{\sigma}_i$  be the limit value of  $\sigma_i$ . Let  $Z = \bigcup_i \tilde{\sigma}_i \leq_T A'$ . Clearly for each  $i, \tilde{\sigma}_{i+1} \supset \tilde{\sigma}_i * 0^{\tilde{k}_{i+1}-\tilde{k}_i}$ , and that  $\tilde{\sigma}_i * 0^j \in U_{\tilde{k}_i+j}$  for every  $j < \tilde{k}_{i+1} - \tilde{k}_i$ . Thus  $Z \in \bigcap_x [U_x]$ . Finally fix i such that  $g_i$  is total. We argue that  $Z \notin [E^i_{|\tilde{\sigma}_i|+\tilde{k}_{i+1}-\tilde{k}_i}]$ . After  $P_i$  is never initialized,  $P_i$  must find  $j_0$ , because otherwise  $g_i(|\tilde{\sigma}_i|+x) \geq f(\tilde{k}_i+x)$  holds for every x, contrary to the hyperimmunity of A. Once this  $j_0 = \tilde{k}_{i+1} - \tilde{k}_i$  is found by  $P_i$ , it will ensure that  $[\tilde{\sigma}_{i+1}] \cap [E^i_{|\tilde{\sigma}_i|+j_0}] = \emptyset$ . Hence A is not low for CB-randomness.  $\Box$ 

## 5. Lowness for FB-randomness and K-triviality coincide

We now turn to analyzing the class of reals which are low for FB-randomness. Since  $\Omega$  is a  $\Delta_2^0$  real, by the relativized form of Theorem 3.1(i), each low for FB-randomness is low for  $\Omega$ . We prove that in fact low-ness for FB-randomness and several related lowness notions coincide with K-triviality.

## **Theorem 5.1.** Let A be a real. The following are equivalent.

- (i) A is K-trivial.
- (ii) A is low for FB-tests.
- (iii) A is low for FB-randomness.
- (iv) Every Martin-Löf random is FB-random relative to A.
- (v) Every FB- not ML-random is FB-random relative to A.

*Proof.* (i)  $\Rightarrow$  (ii): Suppose A is K-trivial, hence A is low for K and  $\Delta_2^0(A) = \Delta_2^0$ . Fix an A-relative FB-test  $\{U_x^A\}_{x\in\omega}$ . We show how to build  $\{V_x\}_{x\in\omega}$  covering  $\{U_x^A\}$ . Let  $f(x) = \lim_s f_s(x)$  for a computable approximation  $\{f_s\}$ , such that  $D_{f(x)} = U_x^A$  for every x. Effectively in A, for each  $\sigma$  and x, if we see  $\sigma \in U_x^A$  we issue a description of  $\langle \sigma, x \rangle$  of length  $|\sigma|$ . Since the total weight of all descriptions is less than 1, and since A is low for K, let d be a constant such that  $K(\langle \sigma, x \rangle) \leq |\sigma| + d$  for every  $\sigma \in U_x^A$ .

Now for each x, enumerate  $\sigma$  into  $V_x$  if we find a stage s such that  $\sigma \in D_{f(x,s)}$  and for every  $y \leq x$  there is  $\tau \subseteq \sigma$  such that  $K(\langle \tau, y \rangle) \leq |\tau| + d$ . Then  $\#V_x$  is clearly finite. We argue that  $\mu(V_x) < \frac{2^d}{x}$ . For each  $y \leq x$ , let  $E_y = \{\tau : K(\langle \tau, y \rangle) \leq |\tau| + d \text{ and } \exists \sigma \in V_x(\tau \subseteq \sigma)\}$ . Suppose  $\mu(V_x) \geq \frac{2^d}{x}$ . Since each  $\sigma \in V_x$  is an extension of some string in  $E_y$ , we have  $\sum_{\tau \in E_y} 2^{-|\tau|} \geq \sum_{\sigma \in V_x} 2^{-|\sigma|} \geq \frac{2^d}{x}$ . Each  $\tau \in E_y$  corresponds to a description of length  $|\tau| + d$ . These descriptions are all for different numbers so the halting probability is at least  $\sum_{y \leq x} \sum_{\tau \in E_y} 2^{-|\tau|-d} \geq \frac{2^d}{x}(x+1)2^{-d} > 1$ , a contradiction. It is easy to check that  $\cap_x U_x^A \subseteq \cap_x V_x$ . Hence  $\{V_{2^{x+d}}\}$  is our required *FB*-test.

(ii)  $\Rightarrow$  (iii)  $\Rightarrow$  (iv) and (iii)  $\Rightarrow$  (v) are obvious.

(iv)  $\Rightarrow$  (i): Suppose every *ML*-random is *FB*-random relative to *A*. By the Kučera-Gács theorem there exists a *ML*-random real *Z* such that  $A \leq_T Z \leq_T A \oplus \emptyset' \leq_T A'$ . Hence *Z* is *FB*-random relative to *A*. By Theorem 3.1(i) relativized to *A*, *Z* is *ML*-random relative to *A*. Hence *A* is a base for *ML*-randomness.

 $(v) \Rightarrow (i)$ : This follows from Theorem 5.3.

Lemma 5.2. The following are equivalent for a real A.

- (i) A is not low for ML-randomness.
- (ii) There is a uniform sequence of A-c.e. open sets  $\{Q_p \mid p \in \omega\}$  such that for every p,  $\mu(Q_p) < 2^{-p}$  and for each c.e. open set E such that  $\mu(E) < 1$ , we have  $Q_p \not\subseteq E$ .
- (iii) There is a uniform sequence of A-c.e. open sets  $\{S_{\sigma,p} \mid \sigma \in 2^{<\omega}, p \in \omega\}$  such that for every  $\sigma$  and p,  $\mu(S_{\sigma,p} \mid [\sigma]) < 2^{-p}$ ,  $S_{\sigma,p} \subseteq [\sigma]$  and for each c.e. open set E such that  $\mu(E \mid [\sigma]) < 1$ , we have  $S_{\sigma,p} \not\subseteq E \cap [\sigma]$ .

Proof. We use the characterization of non-*ML*-randomness by Kjos-Hanssen [15]. Thus we replace (i) by "there exists an *A*-c.e. open set *G* such that  $\mu(G) < 1$  and for each c.e. open set *E* where  $\mu(E) < 1$ , we have  $G \not\subseteq E$ ". Clearly (iii)  $\Rightarrow$  (ii)  $\Rightarrow$  (i) are trivial. (ii)  $\Rightarrow$  (iii) is also trivial, by letting  $S_{\sigma,p} = \sigma * Q_p$ . We now prove (i)  $\Rightarrow$  (ii). Fix *G* of measure at most  $r \in \mathbb{Q} \cap (0, 1)$  where *G* is not covered by any c.e. open set of measure less than 1. We construct an *A*-c.e. open set *Q* where  $\mu(Q) \leq \frac{r^2 + r}{2}$  and which avoids being covered by c.e. open sets of measure less than 1. Intuitively *Q* is a slightly expanded version of G \* G. This construction is effective. Since  $r > \frac{r^2 + r}{2} > \frac{\left(\frac{r^2 + r}{2}\right)^2 + \left(\frac{r^2 + r}{2}\right)}{2} > \cdots$  converges to 0, we can iterate this construction to get the sequence  $\{Q_p\}$ .

We let  $\{E_i\}$  be the  $i^{th}$  c.e. open set. Clearly the predicate " $\mu(E_i) > q$ " is uniformly c.e. in i, q. For each  $i, \sigma$  we let  $Q^{i,\sigma} = [\sigma] - (E_{i,s} \cap [\sigma])$  if a stage s is found such that  $\mu(E_{i,s} \mid [\sigma]) > 1 - \varepsilon 2^{-i-2|\sigma|}$ , where  $\varepsilon = \frac{r-r^2}{2}$ . If no such s is found we let  $Q^{i,\sigma} = \emptyset$ . Let  $Q = (G * G) \cup \{Q^{i,\sigma} \mid i \in \omega, \sigma \in G\}$ . Since  $\mu(Q^{i,\sigma}) \leq \varepsilon 2^{-i-2|\sigma|}$ , this means that  $\mu(Q) \leq r^2 + \varepsilon \sum_{i,\sigma} 2^{-i-2|\sigma|} \leq r^2 + \varepsilon = \frac{r+r^2}{2}$ .

Suppose that  $Q \subseteq E_i$  where  $\mu(E_i) < 1$ . If  $\sigma \in G$  then  $\mu(E \mid [\sigma]) = 1$ , because otherwise  $\{\tau \mid \sigma * \tau \in E\}$  is a c.e. open set of measure less than 1 covering G. Hence for all  $\sigma \in G$  and all  $i, Q^{i,\sigma} = 2^{\omega} - E_{i,s}$  for some s. Since  $E_i \supseteq Q \supseteq Q^{i,\sigma}$  this means that  $[\sigma] \subseteq [E_i]$ . Hence G is covered by  $E_i$ , a contradiction.  $\Box$ 

**Theorem 5.3.** Suppose that A is not low for ML-randomness. Then there is a FB-random real Z which is 1-generic relative to A and not FB-random relative to A.

*Proof.* Assume that A is not low for ML-randomness. By Lemma 5.2(iii) fix the sequence  $\{S_{\sigma,p}\}$ . Since  $[S_{\sigma,p}] \subseteq [\sigma]$ , for ease of notation, we consider

 $\{\tau \mid \sigma * \tau \in S_{\sigma,p}\}$  instead. Henceforth  $S_{\sigma,p}$  refers to this set of truncated strings. We first describe a construction C which we will later use as a black box to build Z. C takes in parameters  $\eta \in 2^{<\omega}, I, \varepsilon \in \omega$ . It effectively outputs a FB-test relative to A,  $\{U_x\}$ , where  $\mu(U_x) < 2^{-(x+1)\varepsilon-x}$ , and an A-computable sequence  $\{\sigma_s\}$  of finite strings. The construction  $C(\eta, I, \varepsilon)$ ensures that if  $\{E_x^I\}$  is a FB-test then  $\sigma = \lim_s \sigma_s \supset \eta$  exists,  $[E_x^I] \cap [\sigma] = \emptyset$  for some x and  $[\sigma] \subseteq \bigcap_{k \leq y} [U_k]$ , where y is the least stage such that  $\sigma_y$  is stable. Here  $\{E_x^I\}$  is the  $I^{th}$  ML-test. Intuitively construction Csearches, relative to the input parameters, for a safe spot for lower priority requirements to act.

Description of strategy: The construction C is an effective version of the proof of Proposition 3.1(ii).  $\eta$  is the environment in which C is called to work in, and is handed to us by higher priority requirements. For simplicity we assume that  $\eta = \emptyset$ , and that we are trying to avoid some c.e. open set  $E = E_x^i$  of measure less than 1. We build the approximation  $\sigma_s$  which attempts to locate a neighbourhood from which  $E_x^i$  is disjoint. We know this exists if E is finite, but since we have to diagonalize against every test, we have to deal with the possibility of E being infinite.

Since  $\sigma_s$  will be an initial segment of our real Z (provided E is finite), we have to enumerate  $\sigma_s$  into a FB-test  $\{U_x\}$  relative to A. The main difficulty here is that we have to keep each  $U_x$  finite even when E is infinite. We use the fact that A is not low for random. Let  $U_0 \operatorname{copy} S_{\emptyset,p}$  for some sufficiently large p. More specifically we let  $\sigma_s$  be the first string enumerated in  $S_{\emptyset,p}$ . We enumerate  $\sigma_s$  in  $U_0$ . We wait until  $[E] \supseteq [\sigma_s]$ , and we move on to the second string enumerated in  $S_{\emptyset,p}$ , and so on. Clearly if  $U_0$  is infinite then  $[E] \supseteq [S_{\emptyset,p}]$ , which by the properties of S implies that  $\mu(E) = 1$ , a contradiction. However this naive plan does not work well for us because E can cover extensions of each  $\sigma \in S_{\emptyset,p}$  without covering  $\sigma$  itself. We need to find a safe spot  $\lim_s \sigma_s$  for lower priority genericity requirements to work in, and thus  $[\lim_s \sigma_s]$  has to be disjoint from [E].

Thus we need to iterate the above strategy. While we are waiting for [E] to cover the current  $[\sigma^0]$  in  $S_{\emptyset,p}$ , we need to consider  $S_{\sigma^0,p^2}$ . For simplicity we drop  $p^2$  from the notation, with the understanding that each iteration of S is called with a sufficiently large  $p^n$ . We take the first string  $\sigma^1 \in S_{\sigma^0}$ , enumerate  $\sigma^1$  in  $U_1$  and wait for [E] to cover  $[\sigma^1]$ . In the meantime we call on  $S_{\sigma^1}$  and so on. If [E] does end up covering  $[\sigma^n]$  then by compactness we see this at a finite stage, and we can set  $\sigma^n$  to be the next element enumerated in  $S_{\sigma^{n-1}}$ . We let  $\sigma_s$  be  $\sigma^n$  for the largest n where  $\sigma^n$  is defined at stage s. Clearly if E is finite then  $[\lim_s \sigma^s]$  can be chosen to be disjoint from [E]. The problem with this approach is when  $\#E = \infty$ . In this case we may return to  $\sigma^n$  infinitely often, i.e. [E] covers  $[\sigma^n]$  for every value of  $\sigma^n$  we pick from  $S_{\sigma^{n-1}}$ . In this case  $\mu(E \mid [\sigma^{n-1}])$  must be 1, but if  $[E] \not\supseteq [\sigma^{n-1}]$  we will never abandon  $\sigma^{n-1}$  during the construction. This means that we will end up copying the infinite set  $S_{\sigma^{n-1}}$  in  $U_n$ . However we are committed to making  $\{U_x\}$  an FB-test relative to A, even when E is infinite.

The way around the problem above is to force [E] to cover  $[\sigma^{n-1}]$ . Above each string  $\sigma^{n-1}$  that we are currently guessing is an initial segment of Z, we pick  $\sigma^n$  from the set  $S_{\sigma^{n-1}}$  until  $\mu(E_s \mid [\sigma^{n-1}])$  is very close to 1 (say larger than  $1-\varepsilon$ ). When this happens, say at stage s, we switch and pick  $\sigma^n$  from the clopen set  $[\sigma^{n-1}] - [E_s]$ . This forces us to add a small amount of additional measure to  $U_n$  not accounted for by  $\mu(S_{\sigma^{n-1}})$ , but we can choose the threshold  $\varepsilon$  to be as small as we like to keep the measure of  $U_n$  small. Additionally we will be able to ensure that we consider only finitely many different strings  $\sigma^n$  above each value for  $\sigma^{n-1}$ .

Clearly if  $\#E < \infty$  then  $\lim_s \sigma_s$  exists and we will be able to choose  $[\lim_s \sigma_s]$  disjoint from [E]. We then allow a lower priority requirement to act above  $[\lim_s \sigma_s]$ , by calling C above  $[\lim_s \sigma_s]$ . On the other hand if  $\#E = \infty$  then  $\liminf\{n : \sigma^n \text{ is abandoned}\}$  must be  $\infty$ . To see this suppose that the lim inf is some number n. Then  $\mu(E \mid [\sigma^{n-1}]) = 1$  and so we must switch to pick  $\sigma^n$  from  $[\sigma^{n-1}] - [E_s]$ . Hence  $[E] \supseteq [\sigma^{n-1}]$  and so  $\sigma^{n-1}$  must be abandoned infinitely often, a contradiction. Since  $\liminf\{n : \sigma^n \text{ is abandoned}\} = \infty$ , this particular run of C contributes only finitely many elements to each  $U_x$ . At each stage where the output  $\sigma_s$  changes we allow the next requirement to call C above the node 0. Then  $Z \supseteq 0$  and may possibly be in [E], but it does not matter as  $\#E = \infty$ .

The formal proof is organized as follow. We first specify the working of the basic module  $C(\eta, I, \varepsilon)$ . The actual construction of Z is then carried out on a priority tree. Each node on the priority tree is allowed to call C with certain parameters. Z can then be read off the true path of the construction.

Construction  $C(\eta, I, \varepsilon)$ : Let  $E = E_{|\eta|}^I$ . We define, for each  $\sigma, p$ , the set  $\hat{S}_{\sigma,p}$  to copy  $S_{\sigma,p}$  until the first stage s is found such that  $\mu(E_s \mid [\sigma]) > 1 - 2^{-\varepsilon - p}$ . If this s is found we say that  $\hat{S}_{\sigma,p}$  has switched, and we let  $\hat{S}_{\sigma,p} = S_{\sigma,p,s} \cup ([\sigma] - E_s)$ . Otherwise we say that  $\hat{S}_{\sigma,p}$  remains unswitched and in this case  $\hat{S}_{\sigma,p} = S_{\sigma,p}$ .

We fix a 1-1 enumeration of the infinite set  $S_{\sigma,p}$ . This gives rise in the obvious way to a 1-1 (possibly finite) enumeration of  $\hat{S}_{\sigma,p}$ . We define  $\hat{S}^i$  inductively on *i*: Set  $\hat{S}^0 = \hat{S}_{\eta,\varepsilon+1}$ , and  $\hat{S}^{i+1} = \{\sigma * \tau \mid \sigma \in \hat{S}^i, \tau \in \hat{S}_{\sigma,\varepsilon+i+2}\}$ . For each  $\sigma \in \hat{S}^i$ , and  $\tau \in \hat{S}_{\sigma,\varepsilon+i+2}$  we define  $\alpha(\sigma * \tau) = \alpha(\sigma) * i$  where  $\tau$  is the  $i + 1^{th}$  element to be enumerated in  $\hat{S}_{\sigma,\varepsilon+i+2}$ . In this way we associate each  $\sigma \in \cup_i \hat{S}^i$  with  $\alpha(\sigma)$ . We order  $\omega^{<\omega}$  first lexicographically then by length  $(\alpha \subset \beta \text{ means } \alpha < \beta)$ .

We now define the sequence  $\alpha_0 \leq \alpha_1 \leq \cdots$ . At stage s = 0 set  $\alpha_s = 0$ . At stage s+1 we see if E has a new element. If so we set  $\alpha_{s+1}$  to be the <-least string of length s+3 such that  $\forall i \leq s \ ([\alpha_{s+1} \upharpoonright i+1] \not\subseteq [E_{s+1}])$ . Otherwise let  $\alpha_{s+1} = \alpha_s$ .

This produces an A-computable sequence of finite strings  $\{\alpha_s\}$ . It is easy to verify that  $\alpha_s \leq \alpha_{s+1}$  for every s.

#### **Claim 5.4.** Each time we need to pick $\alpha_{s+1}$ we can do so.

*Proof.* Suppose we are unable to pick  $\alpha_{s+1}$  at some stage s+1. This means there is some least i+1 and some  $\beta$  of length i+1 such that  $[\beta] \not\subseteq [E_{s+1}]$  but for every  $j, [\beta * j] \subseteq [E_{s+1}]$ . (If i = -1 then  $\beta$  is associated with the string  $\eta$ , where it is clear that  $[E] = [E_{[\eta]}^I] \not\supseteq [\eta]$ ).

If  $\hat{S}_{\beta,\varepsilon+i+2}$  is never switched, then we must be able to avoid the finite set  $E_{s+1}$ , since  $S_{\beta,\varepsilon+i+2}$  must be dense above  $\beta$ . Therefore  $\hat{S}_{\beta,\varepsilon+i+2}$  must

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switched at some stage t. If t > s + 1 then  $[\beta] - [E_t]$  must be disjoint from  $[E_{s+1}]$  and so we can pick  $\beta * j$  appropriately. On the other hand if  $t \leq s + 1$  and if  $[E_{s+1}] \supseteq [\hat{S}_{\beta,\varepsilon+i+2}]$  then this means that  $[E_{s+1}] \supseteq [\beta]$ , a contradiction.

**Claim 5.5.** If  $\#E = \infty$  then for each *i*,  $\lim_{s} \alpha_{s}(i)$  exists.

Proof. Suppose that  $\beta = \langle \alpha_s(j) \rangle_{j < i+1}$  is constant for all large enough s. Hence by compactness  $[E] \not\supseteq [\beta]$ . If  $\hat{S}_{\beta,\varepsilon+i+2}$  is ever switched then we are done, because  $\lim_s \alpha_s(i+1)$  is one of finitely many choices. Suppose  $\hat{S}_{\beta,\varepsilon+i+2}$ is never switched. This means that  $\mu(E \mid [\beta]) < 1$  and so  $[\beta * \hat{S}_{\beta,\varepsilon+i+2}] =$  $[\beta * S_{\beta,\varepsilon+i+2}] \not\subseteq [E]$ . Thus there exists some  $\tau \in \hat{S}_{\beta,\varepsilon+i+2}$  such that  $[\beta * \tau] \not\subseteq$ [E].

Claim 5.6. For each i,  $\mu(\hat{S}^i) < 2^{-\varepsilon i - i}$ .

*Proof.* If i = 0 then  $\mu(\hat{S}^0) = \mu(\hat{S}_{\eta,\varepsilon+1}) \le \mu(S_{\eta,\varepsilon+1}) + 2^{-2\varepsilon-1} < 1$ . If i+1 > 0 then similarly

$$\mu(\hat{S}^{i+1}) = \sum_{\sigma \in \hat{S}^{i}} \left\{ 2^{-|\sigma| - |\tau|} \mid \sigma \in \hat{S}^{i}, \tau \in \hat{S}_{\sigma,\varepsilon+i+2} \right\}$$
$$= \sum_{\sigma \in \hat{S}^{i}} 2^{-|\sigma|} \sum_{\tau \in \hat{S}_{\sigma,\varepsilon+i+2}} 2^{-|\tau|}$$

 $< \sum_{\sigma \in \hat{S}^{i}} 2^{-|\sigma|} \left( \mu(S_{\sigma,\varepsilon+i+2}) + \text{conditional measure added due to a switch} \right)$  $\leq \sum_{\sigma \in \hat{S}^{i}} 2^{-|\sigma|} \left( 2^{-\varepsilon-i-2} + 2^{-\varepsilon-i-2} \right) \leq 2^{-\varepsilon(i+1)-(i+1)}. \quad \Box$ 

$$\leq \sum_{\sigma \in \hat{S}^{i}} 2^{-|\sigma|} (2^{-\varepsilon - i - 2} + 2^{-\varepsilon - i - 2}) \leq 2^{-\varepsilon(i+1) - (i+1)}.$$

Finally let  $U_k = \{\alpha_s \mid k+2 \mid s \in \omega\}$ . If  $\#E < \infty$  then  $\alpha_s$  is eventually stable, and so  $\#U_k < \infty$ . If  $\#E = \infty$  then by Claim 5.5  $\#U_k < \infty$ . Since  $[U_k] \subseteq [\hat{S}^{k+1}]$ , by Claim 5.6 this means that  $\mu(U_k) \leq \mu(\hat{S}^{k+1}) < 2^{-(k+1)\varepsilon-k}$ . Set  $\sigma_s =$  string coded by  $\alpha_s$ . If  $\#E < \infty$  then clearly  $\eta \subset \lim_s \sigma_s$  exists. Let  $\alpha = \lim_s \alpha_s$ . Then for each  $2 \leq i \leq |\alpha|, \alpha \upharpoonright i \in U_{i-2}$ , where  $|\alpha| - 2$  is the least stage where  $\alpha_s$  is stable. Clearly  $[E] \cap [\alpha] = \emptyset$ , since for every t,  $E_t$  contains no string of length longer than t.

Now we use C as a procedure to build Z. We define the priority tree to be the full binary tree, with labels  $\infty$  instead of 0 and f in place of 1.  $\eta <_L \gamma$ denotes the usual left to right lexicographic ordering. If  $\eta \subset \gamma$  then we say  $\eta <_L \gamma$ . Each node is assigned parameters  $head(\eta) \in 2^{<\omega}$  and  $m(\eta) \in \omega$ . As usual all parameters retain their assigned values until they are initialized or reassigned. A node  $\eta$  of length i will attempt to diagonalize against  $\{E_x^i\}$ and meet the  $i^{th}$  genericity requirement.

At stage s = 0 we set  $head(\emptyset) = \emptyset$ , and do nothing else. At stage s we define  $\delta_s$  of length s, the stage s approximation to the true path. Assume that  $\eta = \delta_s \upharpoonright i$  and  $head(\eta)$  have been defined. We now act for  $\eta$ . If  $m(\eta) \uparrow$  or if  $head(\eta)$  has changed since the last visit to  $\eta$ , we pick a fresh number for  $m(\eta)$ . Run construction  $C(head(\eta), |\eta|, m(\eta))$  for one more step, say step t. If the output of this construction,  $\sigma_t$ , does not change (i.e.  $\sigma_t = \sigma_{t-1}$ ) then let  $\delta_s(i) = f$ , otherwise let  $\delta_s(i) = \infty$ . We now update  $head(\eta * \delta_s(i))$ . If

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 $\delta_s(i) = \infty$  let  $head(\eta * \infty)$  be the first string  $\tau \supseteq head(\eta) * 0^{m(\eta)}$  found such that  $\tau \in W^A_{|\eta|,s}$ . If no such  $\tau$  exists let  $head(\eta * \infty) = head(\eta) * 0^{m(\eta)}$ . If  $\delta_s(i) = f$  let  $head(\eta * f)$  be the first string  $\tau \supseteq \sigma_t$  found such that  $\tau \in W^A_{|\eta|,s}$ . If no such  $\tau$  exists let  $head(\eta * f) = \sigma_t$ . Finally initialize every node to the right of  $\eta * \delta_s(i)$ .

Now let  $TP = \liminf_{s} \delta_s$  be the true path of construction. Clearly if i < j < s then  $head(\delta_s \upharpoonright i) \subset head(\delta_s \upharpoonright j)$ . It is easy to verify that the following claim holds.

# **Claim 5.7.** For every *i*, $head(TP \upharpoonright i)$ and $m(TP \upharpoonright i)$ are eventually stable.

Now let  $Z = \bigcup_i \lim_s head(TP \upharpoonright i)$ .

Claim 5.8. Z is FB-random.

Proof. Fix i such that  $\{E_x^i\}$  is a FB-test, and let  $\eta = TP \upharpoonright i$ . Let h and m be the final values of  $head(\eta)$  and  $m(\eta)$  respectively. At almost every visit to  $\eta$  we will run C(h, i, m). By properties of C we have  $\lim_t \sigma_t$  exists, so  $\eta * f \subset TP$ . This means that  $Z \supset \lim_s head(\eta * f) \supset \lim_t \sigma_t$  and so  $Z \notin \bigcap_x [E_x^i]$ .

Claim 5.9. Z is 1-generic relative to A.

*Proof.* Fix *i* and look at  $\eta = TP \upharpoonright i + 1$ . We have  $Z \supset \lim_{s} head(\eta)$ . By the construction either  $head(\eta) \in W^{A}_{|\eta|-1}$  or else no extension of  $head(\eta)$  is in  $W^{A}_{|\eta|-1}$ .

Claim 5.10. Z fails to be FB-random relative to A.

Proof. The construction is effective in oracle A. Let  $t(\emptyset, s) = 0$ . For each node  $\eta$  and each stage s we let  $t(\eta, s) \uparrow$  if  $\delta_s \not\supseteq \eta$ . Otherwise  $\eta$  is visited at stage s. If  $\eta(|\eta|-1) = \infty$  let  $t(\eta, s) = t(\eta^-, s) + 1$ . Otherwise  $\eta(|\eta|-1) = f$  and we let  $t(\eta, s) = t(\eta^-, s) + 1$ + the largest  $w \leq t$  such that the output  $\sigma_w$  of the construction C called by  $\eta^-$  (with the current parameters) at stage s has changed. Here t is the number of steps C (with the current parameters) has been run by  $\eta^-$ . For instance each time  $C(head(\emptyset), 0, 1)$  changes its output  $\sigma_w$  we increase t(f, s) to match w.

We define the following FB-test  $\{V_x\}$  relative to A. For every node  $\eta$  and every s in which  $\eta$  is visited, we look at the output  $\{U_x\}$  of C run by  $\eta$  at s. For each  $x \in \omega$  add  $U_x$  to  $V_{x+t(\eta,s)}$ . If  $\eta * \infty$  is visited at s add  $head(\eta * \infty)$ to  $V_{t(n,s)}$ . We verify that  $\{V_x\}$  is an FB-test relative to A.

First observe that  $t(\eta, s) < t(\gamma, s)$  for every  $\eta \subset \gamma$ . It is also easy to see that  $m(\eta) > t(\eta)$  whenever  $\eta$  is visited. Fix an x. Only a node  $\eta$  of length  $\leq x$  can contribute to  $V_x$ . Each time the parameters of  $\eta$  is changed or  $t(\eta)$  is changed we will pick a fresh value for  $m(\eta)$ . At each visit to  $\eta$  we add  $U_y$  to  $V_{y+t(\eta,s)}$ , and possibly a string of length at least  $m(\eta)$  to  $V_{t(\eta,s)}$ . Since  $\mu(U_y) < 2^{-(y+1)m(\eta)-y} \leq 2^{-m(\eta)}2^{-t(\eta,s)-y}$ , it follows that the sum total of all contributions is at most  $2^{-x}$ . Hence  $\mu(V_x) < 2^{-x+2}$ .

Now let us argue that  $\eta$  enumerates only finitely many elements into  $V_x$ . If  $\eta$  is to the left of TP then  $\eta$  is visited only finitely often, so we only add finitely many different versions of its output  $\{U_y\}$ . If  $\eta$  is on TP then the parameters of  $\eta$  and  $t(\eta)$  are eventually stable, so  $\eta$  too, adds finitely many different versions of its output  $\{U_y\}$ . Assume  $\eta >_L TP$ . Let  $\gamma = TP \cap \eta$ . It is easy to see that  $\lim_s t(\gamma * f, s) = \infty$ . Since  $t(\eta) \ge t(\gamma * f)$  at every stage, this means that  $\eta$  will eventually stop adding elements to  $V_x$ . Hence  $\{V_x\}$ is an *FB*-test relative to *A*.

Finally we show that  $Z \in [V_x]$  for every x. Fix  $\eta$  on TP. If  $\eta$  has true outcome f, let  $t_0$  be the last step in C where the output  $\sigma_{t_0}$  changes. Then  $Z \supset \lim_s head(\eta * f) \supseteq \sigma_{t_0}$  and by the properties of C,  $[\sigma_{t_0}] \subseteq \cap_{w \le t_0}[U_w]$ . Since  $\lim_s t(\eta * f, s) = \lim_s t(\eta, s) + t_0 + 1$ , we copy  $\{U_w\}$  in  $V_{w+\lim_s t(\eta,s)}$ , it follows that  $Z \in [V_{\lim_s t(\eta,s)}] \cap \cdots \cap [V_{\lim_s t(\eta * f,s)-1}]$ . On the other hand if  $\eta$  has true outcome  $\infty$  then  $Z \supset \lim_s head(\eta * \infty)$ .

On the other hand if  $\eta$  has true outcome  $\infty$  then  $Z \supset \lim_s head(\eta * \infty)$ . Since  $\lim_s t(\eta * \infty, s) = \lim_s t(\eta, s) + 1$  and  $\lim_s head(\eta * \infty) \in V_{\lim_s t(\eta, s)}$ , it follows similarly that  $Z \in [V_{\lim_s t(\eta, s)}] \cap \cdots \cap [V_{\lim_s t(\eta * f, s) - 1}]$ .  $\Box$ 

This ends the proof of Theorem 5.3.

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School of Math, Statistics, & Operations Research, Victoria University, Cotton Building, Room 358, Gate 7, Kelburn Parade, Wellington, New Zealand *E-mail address*: Rod.Downey@vuw.ac.nz

School of Physical & Mathematical Sciences, Nanyang Technological University, 21 Nanyang Link, Singapore

*E-mail address*: kmng@ntu.edu.sg