

# Connecting the size of $\mathbf{RP}$ to the hardness of $\mathbf{ZPP}$

Philippe Moser\*

## Abstract

We use Lutz's resource bounded measure theory to prove that, either  $\mathbf{RP}$  is small, or  $\mathbf{ZPP}$  is hard. More precisely, we prove that if  $\mathbf{RP}$  has not p-measure zero, then  $\mathbf{ZPP}$  equals  $\mathbf{EXP}$  on almost half the input lengths. Second, we prove that if  $\mathbf{RP}$  has not p-measure zero, then for every  $k \geq 1$ ,  $\mathbf{ZPP}$  is not included in  $\mathbf{DTIME}(2^{O(n^k)})$  for almost half the input lengths  $n$ , i.e. on almost half the input lengths  $\mathbf{ZPP}$  is hard. Next, we prove that if  $\mathbf{NP}$  has not p-measure zero, then derandomization of  $\mathbf{AM}$  is possible on almost half the input lengths, i.e.  $\mathbf{NP} = \mathbf{AM}$  on almost half the input lengths. Finally, we prove easiness versus randomness tradeoffs for classes in the polynomial time hierarchy. We show that it appears to every strong adversary, that either, every  $\Sigma_i^{\mathbf{P}}$  algorithm can be simulated infinitely often by a subexponential co-nondeterministic time algorithm, having oracle access to  $\Sigma_{i-2}^{\mathbf{P}}$ , or  $\Sigma_i^{\mathbf{P}} = \mathbf{BP}\Sigma_i^{\mathbf{P}}$ .

## 1 Introduction

Not much is known about the relationship between  $\mathbf{ZPP}$ ,  $\mathbf{RP}$  and  $\mathbf{BPP}$ , except the trivial inclusions  $\mathbf{ZPP} \subseteq \mathbf{RP} \subseteq \mathbf{BPP}$ . For instance it is not known whether  $\mathbf{RP}$  being easy implies the easiness of  $\mathbf{BPP}$ . A similar relation between  $\mathbf{ZPP}$  and  $\mathbf{RP}$  is also unknown. In fact there are relativized worlds in which  $\mathbf{P} = \mathbf{RP}$ , but  $\mathbf{P} \neq \mathbf{BPP}$  [MV96]. As pointed out by V. Kabanets in [Kab00], the question whether assuming  $\mathbf{P} = \mathbf{ZPP}$  yields any non-trivial easiness result for  $\mathbf{RP}$  also remains open.

In this paper, we use Lutz measure theory [Lut97], to prove relationships between the easiness of  $\mathbf{ZPP}$  and  $\mathbf{RP}$ . First we show that either  $\mathbf{RP}$  is small, or  $\mathbf{ZPP}$  is hard. More precisely, we prove that if  $\mathbf{RP}$  has not p-measure zero, then  $\mathbf{ZPP}$  is equal to  $\mathbf{EXP}$  on at least half the input lengths; thus connecting the size of  $\mathbf{RP}$  to the hardness of  $\mathbf{ZPP}$ . Next, we show that if  $\mathbf{RP}$  has not p-measure zero, then for every integer  $k$ ,  $\mathbf{ZPP}$  is not contained in  $\mathbf{DTIME}(2^{O(n^k)})$ , for at least almost half the input lengths  $n$ ; i.e.  $\mathbf{RP}$  being large implies  $\mathbf{ZPP}$  being hard on at least almost half the input lengths. As a Corollary, we obtain that if  $\mathbf{P}$  equals  $\mathbf{ZPP}$ , then  $\mathbf{RP}$  has p-measure zero; thus obtaining a first step towards an easiness result for  $\mathbf{RP}$  under the assumption  $\mathbf{P} = \mathbf{ZPP}$ ; namely that under the assumption  $\mathbf{P} = \mathbf{ZPP}$ ,  $\mathbf{RP}$  must be a small subset of  $\mathbf{EXP}$ . In fact we obtain the stronger Corollary, stating that if  $\mathbf{ZPP} \subseteq \mathbf{DTIME}(2^{O(n^k)})$ , for some  $k \geq 1$ , then  $\mathbf{RP}$  must be a small subset of  $\mathbf{EXP}$ , i.e.  $\mathbf{RP}$  has p-measure zero.

One ingredient used in our proof is the easy witness technique from [Kab00]. We simulate a given  $\mathbf{RP}$  algorithm using truth table of easy functions instead of purely random strings. If the

---

\*Address: Computer Science Department, University of Geneva. Email: moser@cui.unige.ch

simulation works, we can construct a martingale that succeeds on **RP**, which implies that **RP** has p-measure zero. On the other hand, if the simulation fails, we get a hardness test, which can be used to guess a hard function, yielding a pseudorandom generator by using Impagliazzo and Wigderson's [IW97] result. This combined with Melkebeek's zero-one law for **BPP** [Mel00], yields that **ZPP** is hard.

The derandomization of complexity classes beyond **BPP**, such as **AM**, has just started to be studied. Klivans and Melkebeek [KvM99] showed that the Nissan-Wigderson [NW94] approach relativizes to any oracles, and gave conditional derandomization results for **AM**. Miltersen and Vinodchandran [MV99] used hitting sets to derandomize **AM**, under weaker assumptions than Klivans and Melkebeek [KvM99].

In [AK97] V. Arvind and J. Köbler proved that under the assumption that **NP** has not p-measure zero, partial derandomization of **AM** was possible. More precisely they proved that **NP** not having p-measure zero implies  $\mathbf{NP}/\log n = \mathbf{AM}$ . Using different techniques, we show that the assumption **NP** has not p-measure zero implies that  $\mathbf{NP} = \mathbf{AM}$ , on a fraction  $1/2 - \epsilon$  of lengths  $n$ , for any positive number  $\epsilon$ . Thus we get rid of the nonuniformity in V. Arvind and J. Köbler's result [AK97], but on the other hand we obtain derandomization only on half the input lengths.

Finally following Lu [Lu00], we show a similar result for classes in the polynomial time hierarchy. But instead of using pseudorandom generators as in [Lu00], we use the hitting set construction of Miltersen and Vinodchandran [MV99]. Thus we gain one level in the polynomial hierarchy, enabling us to prove that it appears to every nondeterministic adversary, having oracle access to  $\Sigma_{i-1}^P$ , that either every  $\Sigma_i^P$  algorithm can be simulated by a co-nondeterministic subexponential time algorithm, having oracle access to  $\Sigma_{i-2}^P$ , infinitely often, or  $\Sigma_i^P = \mathbf{BP}\Sigma_i^P$ .

## 2 Preliminaries

We use standard notation for traditional complexity classes; see for instance the books of Balczar, Diaz and Gabarro [BDG95], [BDG90], or the one from Papadimitriou [Pap94]. The polynomial hierarchy is the following sequence of classes: First,  $\Sigma_0^P = \Pi_0^P = \mathbf{P}$ , and for all  $i \geq 0$ ,  $\Sigma_{i+1}^P = \mathbf{NP}^{\Sigma_i^P}$  and  $\Pi_{i+1}^P = \mathbf{coNP}^{\Sigma_i^P}$ . We denote by  $\mathbf{QSAT}_i$  the standard  $\Sigma_i^P$ -complete language. For  $i \geq 2$ , the two-sided probabilistic version of  $\Sigma_i^P$  is equal to the one-sided error version. A proof of this result can be found in [BDG90].

**Proposition 1** ([BDG90]) *Let  $L \in \mathbf{BP}\Sigma_i^P$ , then there exists a relation  $M \in \Sigma_i^P$ , and a polynomial  $m(n)$  such that for all  $x \in \{0, 1\}^n$ ,*

$$\begin{aligned} x \in L &\Rightarrow \Pr_{y \in \{0,1\}^m} [(x, y) \in M] = 1, \text{ and} \\ x \notin L &\Rightarrow \Pr_{y \in \{0,1\}^m} [(x, y) \in M] \leq \frac{1}{4}. \end{aligned}$$

## 2.1 Partial Equalities

We will prove equalities for complexity classes, holding for a fraction of words lengths. To make this notion precise, we define the following measure over  $\mathbb{N}$ .

**Definition 1** *Let  $A \subseteq \mathbb{N}$  be any subset. We define its measure by:*

$$\mu(A) := \lim_{n \rightarrow \infty} \frac{|A^{\leq n}|}{n}$$

where  $A^{\leq n}$  denotes the set of numbers in  $A$ , that are smaller than  $n$ .

The following properties are very easy to show.

**Lemma 1** *Let  $A, B$  be any subsets of  $\mathbb{N}$ .*

1.  $\mu(A) \in [0, 1]$ .
2.  $\mu(\mathbb{N}) = 1$ , and  $\mu(F) = 0$  if  $F$  is finite.
3.  $\mu(\bar{A}) = 1 - \mu(A)$ , where  $\bar{A} = \mathbb{N} - A$
4.  $\mu(A \cup B) \leq \mu(A) + \mu(B)$ .

Let  $P(n)$  be any predicate, where  $n \in \mathbb{N}$ , and let  $\epsilon \in [0, 1]$ . We say that  $P(n)$  holds on a fraction  $\epsilon$  of lengths  $n$ , if the set of  $n$ 's where  $P(n)$  holds has measure at least  $\epsilon$ .

## 2.2 Refuters and Pseudo Classes

Let  $A$  be any language. We say that a multi-valued nondeterministic procedure, with oracle access to  $A$  produces some object, if there is a nondeterministic oracle Turing machine  $M^A$  with oracle access to  $A$  such that at the end of its computation, each nondeterministic branches either produces the desired object, or is marked with reject, and at least one of the branches produces the desired object.

Kabanets [Kab00] introduced the concept of refuters. A refuter is a length preserving Turing machine  $R$  such that on input  $1^n$ ,  $R$  outputs a string of length  $n$ . We will consider multi-valued nondeterministic refuters with oracle access to languages in the polynomial time hierarchy.

**Definition 2** *Let  $A$  be any language. A multi-valued nondeterministic refuter with oracle access to  $A$  (abbreviated **NPMV** <sup>$A$</sup> ) is a nondeterministic polynomial time Turing machine  $M$  with oracle access to  $A$ , that on input  $1^n$ , nondeterministically produces a string in  $\{0, 1\}^n$ .*

Kabanets [Kab00] introduced zero-error probabilistic refuters, i.e. refuters that halt with great probability, and which whenever they halt, output a string. The class of such refuters will be denoted by **FZPP**.

Let  $\epsilon \in [0, 1]$ . For a complexity class of languages  $\mathcal{C}$ , we will consider the class of languages indistinguishable (for a certain class of refuters) from languages of  $\mathcal{C}$ . We say that two languages  $L$  and  $M$  are  $\epsilon$ -distinguishable for a certain class of refuters  $\mathcal{R}$ , if, there exists a refuter  $R \in \mathcal{R}$ , such that every string  $y$  produced by  $R(1^n)$  satisfies  $y \in L \Delta M$  on a fraction at least  $1/2 - \epsilon$  of lengths  $n$ , where  $L \Delta M$  denotes the symmetric difference of  $L$  and  $M$ .

**Definition 3** Let  $\mathcal{C}$  be a complexity class of languages, let  $A$  be any language, and let  $\epsilon \in [0, 1]$ . We define :

$[\text{pseudo}_{\text{NPMV}^A}^\epsilon]\mathcal{C} = \{L \subseteq \{0, 1\}^* \mid \exists M \in \mathcal{C} \text{ such that any multi-valued nondeterministic refuter with oracle access to } A \text{ fails to } \epsilon\text{-distinguish } L \text{ from } M\}$

If the refuters of Definition 3 are replaced with zero-error probabilistic refuters, the corresponding pseudo class is denoted by  $[\text{pseudo}_{\text{FZPP}^A}^\epsilon]\mathcal{C}$ . If the refuters are required to distinguish  $L$  from  $M$  on almost all input lengths  $n$ , the corresponding pseudo class is denoted by  $[\text{pseudo}_{\text{NPMV}^A}]\mathcal{C}$ . Finally, if the refuters are required to distinguish  $L$  from  $M$  on infinitely many input lengths  $n$ , the corresponding pseudo class is denoted by i.o. $[\text{pseudo}_{\text{NPMV}^A}]\mathcal{C}$ .

### 2.3 Hitting Sets

Since sets in the polynomial time hierarchy can be computed by nondeterministic circuits, we will need the following definition of nondeterministic circuits.

**Definition 4** A nondeterministic Boolean circuit  $C$  contains, in addition to the usual AND, OR and NOT gates, choice gates of fan-in 0. The circuit evaluates to 1 on input  $x$ , and we say that  $C(x) = 1$ , if there is some assignment of truth values to the choice-gates that makes the circuit evaluate to 1. Otherwise  $C(x) = 0$ .

A co-nondeterministic circuit  $C$  is defined similarly: The circuit evaluates to 0 on input  $x$ , and we say that  $C(x) = 0$ , if there is some assignment of truth values to the choice-gates that makes the circuit evaluate to 0. Otherwise  $C(x) = 1$ .

Similarly, a single-valued (abbreviated SV) nondeterministic circuit  $C$  computing a function  $f$  has, in addition to its usual output, an extra output bit, called the flag. For any input  $x$ , and any setting of the choice-gates, if the flag is on, the circuit should output the correct value of  $f(x)$ . Furthermore, for any  $x$ , there should be some setting of the choice-gates that turn the flag on.

Pseudorandom generators are used to derandomize two-sided error algorithms. In order to derandomize one-sided error algorithms, hitting sets are used. In particular, we will need hitting sets for co-nondeterministic circuits.

**Definition 5** Let  $A$  be any language. A subset  $H \subseteq \{0, 1\}^n$  is a  $\frac{1}{2}$ -hitting set for Boolean co-nondeterministic circuits of size  $s(n)$ , with oracle gates to  $A$ , if for any such circuit  $C^A$  on  $n$  inputs, the following holds:

If  $\Pr_{x \in \{0, 1\}^n} [C^A(x) = 1] \geq \frac{1}{2}$ , then  $\exists h \in H$  such that  $C^A(h) = 1$ .

The following hardness-randomness tradeoffs are from [MV99]

**Theorem 1 (Miltersen, Vinodchandran)** Let  $A$  be any language. For any  $\epsilon > 0$ , there is a  $\gamma > 0$  so that the following holds. There is a deterministic polynomial time procedure which, given as input the truth table of a  $\log m$ -variables Boolean function  $f : \{0, 1\}^{\log m} \rightarrow \{0, 1\}$  with circuit complexity greater than  $m^\epsilon$  for nondeterministic oracle circuits with oracle gates for  $A$ , outputs a hitting set in  $\{0, 1\}^n$ , with threshold  $\frac{1}{2}$ , for co-nondeterministic oracle circuits having oracle access to  $A$  of size  $n$ , where  $n = m^\gamma$ .

Thanks to Proposition 1 producing a hitting set with a multi-valued nondeterministic procedure is enough to derandomize  $\mathbf{BP}\Sigma_i^{\mathbf{P}}$ , more precisely:

**Proposition 2** *If there is a multi-valued nondeterministic procedure with oracle access to  $\Sigma_{i-1}^{\mathbf{P}}$  which on input  $1^n$  outputs a hitting set in  $\{0, 1\}^n$ , with threshold  $\frac{1}{2}$ , for co-nondeterministic oracle circuits with oracle gates for  $\Sigma_{i-1}^{\mathbf{P}}$ , of size  $n$ , then  $\Sigma_i^{\mathbf{P}} = \mathbf{BP}\Sigma_i^{\mathbf{P}}$ .*

## 2.4 p-measure

In this section we describe the fragment of Lutz's measure theory for the class  $\mathbf{EXP}$  that we will need. For a more detailed presentation of this theory we refer the reader to the survey by Lutz [Lut97].

The measure on  $\mathbf{EXP}$  is obtained by imposing appropriate resource-bound on a game theoretical characterization of the classical Lebesgue measure.

A martingale is a function  $d : \{0, 1\}^* \rightarrow [0, \infty[$  such that,

$$d(w) = \frac{d(w0) + d(w1)}{2}$$

for every  $w \in \{0, 1\}^*$ .  $d$  is a p-martingale if  $d$  is computable in time polynomial in  $|w|$ .

This definition can be motivated by the following betting game in which a gambler puts bets on the successive membership bits of a hidden language  $A$ . Denote by  $s_0, s_1, \dots$  the enumeration of all Boolean strings in lexicographic order. The game proceeds in infinitely many rounds where at the end of round  $n$ , it is revealed to the gambler whether  $s_n \in A$  or not. The game starts with capital 1. Then, in round  $n$ , depending on the first  $n-1$  outcomes  $w = \chi_A[0 \dots n-1]$ , the gambler bets a certain fraction  $\epsilon_w d(w)$  of his current capital  $d(w)$ , that the  $n$ th word  $s_n \in A$ , and bets the remaining capital  $(1 - \epsilon_w)d(w)$  on the complementary event  $s_n \notin A$ . The game is fair, i.e. the amount put on the correct event is doubled, the one put on the wrong guess is lost. The value of  $d(w)$ , where  $w = \chi_A[0 \dots n]$  equals the capital of the gambler after round  $n$  on language  $A$ . The player wins on a language  $A$  if he manages to make his capital arbitrarily large during the game. We say that a martingale  $d$  succeeds on a language  $A$ , if  $d(A) := \limsup_{w \sqsubset A, w \rightarrow A} d(w) = \infty$ , where we identify language  $A$  with its characteristic sequence  $\chi_A$ .

**Definition 6** *A class  $\mathbf{C}$  has p-measure zero if there is a single p-martingale  $d$  that succeeds on every language  $A$  of  $\mathbf{C}$ .*

This property is monotone in the following sense: If class  $\mathbf{D}$  is contained in a class  $\mathbf{C}$  of p-measure zero, then  $\mathbf{D}$  also has p-measure zero.

**Definition 7** *A class  $\mathbf{C}$  has p-measure one if the complement of  $\mathbf{C}$  has p-measure zero.*

Lutz showed in [Lut92] that the class  $\mathbf{E}$  does not have p-measure zero, which he called the measure conservation property. Since finite unions of null classes is a null class, it's impossible for a class to have both measure zero and one.

Lutz also proved in [Lut92] that "easy" infinite union of null classes is null.

**Theorem 2 (Lutz)** *Suppose  $\{d_i\}_{i \geq 1}$  is a set of martingales, each covering class  $C_i$ ; where  $d(i, w) := d_i(w)$  is computable in time  $q = (i, |w|)$  for a certain polynomial  $q$ . Then  $\cup_{i \geq 1} C_i$  has  $p$ -measure zero.*

We will need the following zero-one law for **BPP**, stating that **BPP** has either  $p$ -measure zero or one.

**Theorem 3 (Melkebeek [Mel00])** ***BPP** has either  $p$ -measure zero or else has  $p$ -measure one.*

### 3 RP not being small implies ZPP being hard

The following result shows that no derandomization of **ZPP** is possible unless **RP** is small. More precisely it states that **ZPP** is as hard as **EXP** on almost half the input lengths, unless **RP** has  $p$ -measure zero.

**Theorem 4**  $\forall \epsilon > 0$ , **ZPP** = **EXP** on a fraction  $1/2 - \epsilon$  of input lengths  $n$ , unless **RP** has  $p$ -measure zero.

#### Proof

Let  $\epsilon \in [0, 1]$ . The results in [Kab00] can be modified to obtain the following Theorem.

**Theorem 5**  $\forall \epsilon > 0$ , at least one of the following statements holds.

1.  $\forall \delta > 0$ , **RP**  $\subseteq$  [pseudo $^\epsilon_{\mathbf{FZPP}}$ ]**DTIME**( $2^{n^\delta}$ ), or
2. **ZPP** = **BPP** for a fraction  $1/2 - \epsilon$  of input lengths  $n$ .

Suppose the first statement of Theorem 5 holds. Taking  $\delta = 1$ , we have that for every language  $A \in \mathbf{RP}$ , there exists a language  $B \in \mathbf{DTIME}(2^n)$ , such that every **FZPP** refuter fails to  $\epsilon$ -distinguish  $A$  from  $B$ ; i.e. for every **FZPP** refuter  $R$ , there is a  $y$  produced by  $R(1^n)$ , such that  $A(y) = B(y)$ , on a fraction  $1/2 + \epsilon$  of lengths  $n$ .

So let  $A$  be any language in **RP** and let  $B$  be as above. The refuter  $R(1^n) := 1^n$  is a **FZPP** refuter, therefore we have,

$$A(1^n) = B(1^n) \text{ on a fraction } 1/2 + \epsilon \text{ of lengths } n. \tag{1}$$

Let  $a > 0$  be any dyadic rational such that  $a < \epsilon/c$ , where  $c$  will be determined later. Consider the following martingale  $d_B$ , that for each length  $n$ , only bets on the membership bit of the string  $1^n$ , and for each of these strings bets a fraction  $a$  of its capital on the outcome that the membership is the same as for  $B$ . To compute  $d_B(w)$ , where  $w$  is the characteristic sequence of some language for words up to length  $t$ , one only needs to compute whether  $1^n \in B$  for  $n = 1, 2, \dots, t$ . Since  $B \in \mathbf{DTIME}(2^n)$ ,  $d_B$  is computable in time  $t2^t \leq |w|^2$ . Moreover if  $A$  is a language such that 1 holds, then  $d_B$  multiplies its capital by a factor  $1 + a$  on at least a fraction  $1/2 + \epsilon$  of all bets it makes, and loses a factor  $(1 - a)$  of its capital on at most a fraction

$1/2 - \epsilon$  of all bets. Consider  $g_A(n)$  the number of length  $l \leq n$  such that  $A(1^l) = B(1^l)$  divided by  $n$ , and  $p_A(n)$  the number of length  $l \leq n$  such that  $A(1^l) \neq B(1^l)$  divided by  $n$ . We have,

$$\limsup_{n \rightarrow \infty} g(n) = \frac{1}{2} + \epsilon \quad \text{and} \quad \limsup_{n \rightarrow \infty} p(n) = \frac{1}{2} - \epsilon.$$

Therefore,

$$\limsup_{m \rightarrow \infty} d_B(\chi_A \upharpoonright 2^m) = \limsup_{m \rightarrow \infty} [(1+a)^{g(m)}(1-a)^{p(m)}]^m = \limsup_{m \rightarrow \infty} [(1+a)^{1/2+\epsilon}(1-a)^{1/2-\epsilon}]^m.$$

Consider  $F(a) = (1+a)^{1/2+\epsilon}(1-a)^{1/2-\epsilon}$ . Let us prove that  $F(a) > 1$ , which implies  $\limsup_{m \rightarrow \infty} d_B(\chi_A \upharpoonright 2^m) = \infty$ . By taking the Taylor series of the logarithm function, we have for every number  $x$ ,  $x < 1$ ,

$$\log(1+x) = x + R(x) \quad \text{where } |R(x)| \leq cx^2 \text{ for some positive constant } c. \quad (2)$$

Thus,

$$F(a) > 1 \Leftrightarrow (1+a)^{1+2\epsilon}(1-a)^{1-2\epsilon} > 1 \Leftrightarrow (1+2\epsilon)\log(1+a) + (1-2\epsilon)\log(1-a) > 0.$$

Since

$$(1+2\epsilon)\log(1+a) + (1-2\epsilon)\log(1-a) \geq (1+2\epsilon)(a-ca^2) + (1-2\epsilon)(-a-ca^2)$$

by Equation 2,

$$F(a) > 1 \Leftrightarrow a < \epsilon/c,$$

which is true by the choice of  $a$ .

Denote by  $L_B$  the class of languages  $A$  such that equation 1 holds. Let  $M_1, M_2, \dots$  be a standard enumeration of Turing machines running in deterministic time  $2^n$ , where  $M_i$  runs in time polynomial in  $i+2^n$ , and denote by  $B_i$  the language decided by  $M_i$ . The martingale defined by  $d(i, w) = d_{B_i}(w)$  is computable in time polynomial in  $i+|w|$ . Therefore the class  $C = \bigcup_{i \geq 1} L_i$  has p-measure zero, by Theorem 2. Since  $\mathbf{RP} \subseteq C$ ,  $\mathbf{RP}$  has p-measure zero. Now if  $\mathbf{RP}$  has not p-measure zero, we have that  $\mathbf{ZPP} = \mathbf{BPP}$  on a fraction  $1/2 - \epsilon$  of lengths  $n$ . Moreover, since  $\mathbf{RP}$  has not p-measure zero,  $\mathbf{BPP}$  has not p-measure zero, therefore  $\mathbf{BPP} = \mathbf{EXP}$ , by Theorem 3. Therefore  $\mathbf{RP}$  not having p-measure zero implies  $\mathbf{ZPP} = \mathbf{EXP}$ , on a fraction  $1/2 - \epsilon$  of lengths  $n$ , which ends the proof.  $\square$

As a consequence, we obtain a relation between the size of  $\mathbf{RP}$ , and the easiness of  $\mathbf{ZPP}$ ; namely that if  $\mathbf{RP}$  has not p-measure zero, then derandomization of  $\mathbf{ZPP}$  is impossible on almost half the lengths  $n$ . In fact we prove the following stronger result, stating that if  $\mathbf{RP}$  has not p-measure zero, then for every  $k \geq 1$ ,  $\mathbf{ZPP}$  is not included in  $\mathbf{DTIME}(2^{O(n^k)})$ .

**Theorem 6** *If  $\mathbf{RP}$  has not p-measure zero, then  $\forall \epsilon > 0, \forall k \geq 1, \mathbf{ZPP} \not\subseteq \mathbf{DTIME}(2^{O(n^k)})$  on a fraction  $1/2 - \epsilon$  of lengths  $n$ ; i.e. there exists a language  $A \in \mathbf{ZPP}$  such that for every language  $B \in \mathbf{DTIME}(2^{O(n^k)})$ ,  $A \neq B$  on a fraction  $1/2 - \epsilon$  of lengths  $n$ .*

**Proof** Let  $\epsilon > 0$ . Let  $M_1, M_2, \dots$  be a standard enumeration of  $\mathbf{DTIME}(2^{O(n^k)})$ , where  $M_i$  runs in time polynomial in  $2^{in^k} + i$ . We construct a language  $L \in \mathbf{EXP}$ , which is different from every language in  $\mathbf{DTIME}(2^{O(n^k)})$  on almost every input lengths. Let  $n > 0$ ,  $L^{=n}$  is defined as follows, where  $L^{=n}$  denotes the set of  $n$ -sized words of  $L$ . Denote by  $s_1^n, s_2^n, \dots, s_{2^n}^n$  all words of size  $n$ , ordered lexicographically. Then,

$$L(s_i^n) := \begin{cases} 1 - M_i(s_i^n) & \text{if } i \leq n \\ 0 & \text{otherwise.} \end{cases}$$

$L \in \mathbf{EXP}$ , because on input  $x$ , where  $x$  is the  $i$ th word of length  $n$ , (with  $i \leq n$ ),  $L$ 's machine simply needs to simulate  $M_i$  on  $x$ , and invert its answer. This takes time polynomial in  $2^{in^k} + i$ , which is less than  $2^{O(n^{k+1})}$ .

**Claim** For every language  $B$  of  $\mathbf{DTIME}(2^{O(n^k)})$ ,  $L^{=n} \neq B^{=n}$  for almost every  $n$ .

Indeed let  $B$  be any language of  $\mathbf{DTIME}(2^{O(n^k)})$ . Then there is an index  $j$  such that  $\mathcal{L}(M_j) = B$ . Thus  $L(s_j^n) \neq B(s_j^n)$  for every  $n \geq j$ .

Since by hypothesis  $\mathbf{RP}$  has not  $p$ -measure zero, Theorem 4 implies that  $\mathbf{ZPP} = \mathbf{EXP}$  on a fraction  $1/2 - \epsilon$  of lengths  $n$ . Therefore there is a set  $A \in \mathbf{ZPP}$ , such that  $L = A$  on a fraction  $1/2 - \epsilon$  of lengths  $n$ , which ends the proof.

As a consequence, we obtain an easiness result for  $\mathbf{RP}$  under the assumption  $\mathbf{P} = \mathbf{ZPP}$ .

**Corollary 1** *If  $\mathbf{ZPP} \subseteq \mathbf{DTIME}(2^{O(n^k)})$ , then  $\mathbf{RP}$  has  $p$ -measure zero. In particular, if  $\mathbf{P} = \mathbf{ZPP}$ , then  $\mathbf{RP}$  has  $p$ -measure zero.*

## 4 NP being small implies derandomization of AM

We obtain an analogue of Theorem 4 for  $\mathbf{AM}$ ; namely that the assumption  $\mathbf{NP}$  has not  $p$ -measure zero, implies partial derandomization of  $\mathbf{AM}$ .

**Theorem 7**  $\forall \epsilon > 0$ ,  $\mathbf{NP} = \mathbf{AM}$  on a fraction  $1/2 - \epsilon$  of lengths  $n$ , unless  $\mathbf{NP}$  has  $p$ -measure zero.

**Proof**

Let  $\epsilon \in [0, 1]$ . The results in [Lu00] can be modified to obtain the following Theorem.

**Theorem 8**  $\forall \epsilon > 0$ , at least one of the following statements holds.

1.  $\forall \delta > 0$ ,  $\mathbf{NP} \subseteq [\text{pseudo}^\epsilon_{\mathbf{NPMV}}]\mathbf{DTIME}(2^{n^\delta})$ , or
2.  $\mathbf{NP} = \mathbf{AM}$  on a fraction  $1/2 - \epsilon$  of lengths  $n$ .

Suppose the first statement of Theorem 8 holds. Taking  $\delta = 1$ , we have that for every language  $A \in \mathbf{NP}$ , there exists a language  $B \in \mathbf{DTIME}(2^n)$ , such that every  $\mathbf{NPMV}$  refuter fails to  $\epsilon$ -distinguish  $A$  from  $B$ ; i.e. for every  $\mathbf{NPMV}$  refuter  $R$ , there is a  $y$  produced by  $R(1^n)$ , such that  $A(y) = B(y)$ , on a fraction  $1/2 + \epsilon$  of lengths  $n$ .



So let  $A$  be any language in **NP** and let  $B$  be as above. The refuter  $R(1^n) := 1^n$  is a **NPMV** refuter, therefore we have,

$$A(1^n) = B(1^n) \text{ on a fraction } 1/2 + \epsilon \text{ of lengths } n. \quad (3)$$

The end of the proof is similar to that of Theorem 4.

□

## 5 Derandomization of PH in a Uniform Setting

The following result is an easiness versus randomness tradeoff for classes in the polynomial hierarchy.

**Theorem 9** *For every  $i \in \mathbb{N}$ , at least one of the following statements holds.*

1.  $\forall \epsilon > 0, \Sigma_i^P \subseteq \text{i.o.}[\text{pseudo}_{\text{NPMV}_{\Sigma_{i-1}^P}}] \text{coNTIME}^{\Sigma_{i-2}^P}(2^{n^\epsilon})$ , or
2.  $\Sigma_i^P = \text{BP}\Sigma_i^P$ .

**Proof.**

Suppose inclusion 1 is false, i.e. there exists  $\epsilon_0 > 0$  and a language  $A \in \Sigma_i^P$ , such that  $A \notin \text{i.o.}[\text{pseudo}_{\text{NPMV}_{\Sigma_{i-1}^P}}] \text{coNTIME}^{\Sigma_{i-2}^P}(2^{n^{\epsilon_0}})$ .

Since  $A \in \Sigma_i^P$ , there exists a relation  $M \in \Pi_{i-1}^P$ , and a polynomial  $m(n)$ , such that for every  $x \in \{0, 1\}^n$ ,

$$x \in A \iff \exists y \in \{0, 1\}^m \text{ such that } (x, y) \in M.$$

For  $m \in \mathbb{N}$  and  $\delta > 0$ , let  $S_m^\delta$  be the set of truth tables of all log  $m$  variables Boolean functions with circuit complexity smaller than  $m^\delta$ , for nondeterministic oracle circuits with oracle gates for  $\text{QSAT}_{i-1}$ . We have :  $|S_m^\delta| \leq 2^{m^{O(\delta)}}$ .

Consider the following procedure  $B_m^\delta$ , which accepts  $x$  iff there exists a truth table  $y \in S_m^\delta$ , such that  $(x, y) \in M$ .

**Claim** The procedure  $B_m^\delta$  is in  $\text{coNTIME}^{\Sigma_{i-2}^P}(2^{n^{\epsilon_0}})$ .

Indeed here is a description of the procedure  $B_m^\delta$ .

1. Construct all  $2^{m^{O(\delta)}}$  single-valued nondeterministic oracle gates circuits with oracle gates for  $\text{QSAT}_{i-1}$  of size at most  $m^\delta$ .
2. Given one such circuit, compute its truth table.
3. Check for each  $y \in S_m^\delta$  whether  $(x, y) \in M$ .

Since simulating a  $\text{QSAT}_{i-1}$  gate of fan in at most  $m^\delta$  is in  $\mathbf{DTIME}^{\Sigma_{i-2}^P}(2^{O(m^\delta)})$ , and since there are at most  $m^\delta$  nondeterministic choice gates, the running time of step 2 is in  $\mathbf{DTIME}^{\Sigma_{i-2}^P}(2^{O(m^\delta)})$ . For step 3, since  $M \in \Pi_{i-1}$ , step 3 can be executed in  $\mathbf{coNTIME}^{\Sigma_{i-2}^P}(2^{m^{O(\delta)}})$ . Thus  $B_m^\delta$  is in  $\mathbf{coNTIME}^{\Sigma_{i-2}^P}(2^{m^{c\delta}})$ , for some constant  $c > 0$ . Choosing  $\delta$  such that  $m^{c\delta} \leq n^{\epsilon_0}$  proves the claim.

Now since  $A \notin \text{i.o.}[\text{pseudo}_{\mathbf{NPMV}^{\Sigma_{i-1}^P}}]\mathbf{coNTIME}^{\Sigma_{i-2}^P}(2^{n^{\epsilon_0}})$ , there is a multi-valued nondeterministic refuter  $R$  with oracle access to  $\text{QSAT}_{i-1}$ , such that for almost every  $n$ , a string  $x$  produced by  $R$  is in the symmetrical difference  $A\Delta L(B_m^\delta)$  (where  $L(B_m^\delta)$  is the language decided by procedure  $B_m^\delta$ ).

Since  $L(B_m^\delta) \subseteq A$ , we have  $A\Delta L(B_m^\delta) = A \setminus L(B_m^\delta)$ , therefore for each such string  $x$  we have: For every  $y \in S_m^\delta$  it holds that  $(x, y) \notin M$ , but there exists  $y \in \{0, 1\}^m \setminus S_m^\delta$  such that  $(x, y) \in M$ .

Now let  $\epsilon = \delta$ , and let  $\gamma > 0$  be as in Theorem 1, and let  $k = m^\gamma$ . The following multi-valued nondeterministic procedure with oracle access to  $\Sigma_{i-1}^P$  produces the truth table of a Boolean function with  $\log m$  variables with circuit complexity greater than  $m^\epsilon$  for nondeterministic oracle circuits with oracle gate for  $\text{QSAT}_{i-1}$ .

Procedure PRODUCE-FUNCTION:

1. Use  $R$  to nondeterministically produce a string  $x$  in  $A \setminus L(B_m^\delta)$ .
2. Nondeterministically guess a string  $y$  of length  $m$  and output it if  $(x, y) \in M$ .

Once we obtain the truth table of a hard function, we use Theorem 1 to produce in time polynomial in  $k$  a hitting set in  $\{0, 1\}^k$  with threshold  $\frac{1}{2}$ , for co-nondeterministic circuits with oracle gates for  $\text{QSAT}_{i-1}$  of size  $k$ . Applying Proposition 2, we get  $\Sigma_i^P = \mathbf{BP}\Sigma_i^P$ .

□

## 6 Final Remarks

It would be interesting to see whether Theorem 4 could be improved to prove a zero-one measure law for  $\mathbf{RP}$ . Since it is possible that  $\mathbf{ZPP}$  and  $\mathbf{EXP}$  are equal on almost half the input lengths and still  $\mathbf{ZPP}$  has not p-measure one, it seems that stronger refuters than  $\mathbf{FZPP}$  refuters are needed.

## References

- [AK97] V. Arvind and J. Köbler. On pseudorandomness and resource-bounded measure. *Proceedings 17th Conference of the Foundations of Software Technology and Theoretical Computer Science*, 1346:235–249, 1997.
- [BDG90] J. L. Balcazar, J. Diaz, and J. Gabarro. *Structural Complexity II*. EATCS Monographs on Theoretical Computer Science Volume 22, Springer Verlag, 1990.

- [BDG95] J. L. Balcazar, J. Diaz, and J. Gabarro. *Structural Complexity I*. EATCS Monographs on Theoretical Computer Science Volume 11, Springer Verlag, 1995.
- [IW97] R. Impagliazzo and A. Wigderson. P = BPP if E requires exponential circuits: derandomizing the XOR lemma. *Proceedings of the 29th Annual ACM Symposium on Theory of Computing*, pages 220–229, 1997.
- [Kab00] V. Kabanets. Easiness assumptions and hardness test: Trading time for zero error. *Proceedings of the Fifteenth Annual IEEE Conference on Computational Complexity*, pages 150–157, 2000.
- [KvM99] A. Klivans and D. van Melkebeek. Graph nonisomorphism has subexponential size proofs unless the polynomial hierarchy collapses. *Proceedings of the 31st Annual ACM Symposium on Theory of Computing*, pages 659–667, 1999.
- [Lu00] C.-J. Lu. Derandomizing arthur-merlin games under uniform assumptions. *Proceedings of the Eleventh Annual International Symposium on Algorithms and Computation*, 2000.
- [Lut92] J.H. Lutz. Almost everywhere high nonuniform complexity. *Journal of Computer and System Science*, 44:220–258, 1992.
- [Lut97] J.H. Lutz. The quantitative structure of exponential time. In L.A. Hemaspaandra and A.L. Selman, editors, *Complexity Theory Retrospective II*, pages 225–260. Springer, 1997.
- [Mel00] D. Melkebeek. The zero one law holds for BPP. *Theoretical Computer Science*, 244(1-2):283–288, 2000.
- [MV99] P. Miltersen and N. Vinodchandran. Derandomizing Arthur-Merlin games using hitting sets. *Proceedings of the 40th IEEE Symposium on Foundations of Computer Science*, pages 71–80, 1999.
- [MV96] A.A. Muchnik and N.K. Vereshchagin. A general method to construct oracles realizing given relationships between complexity classes. *Theoretical Computer Science*, 157:227–258, 1996.
- [NW94] N. Nisan and A. Wigderson. Hardness vs. randomness. *Journal of Computer and System Science*, 49:149–167, 1994.
- [Pap94] C. Papadimitriou. *Computational complexity*. Addison-Wesley, 1994.