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# Complements of Multivalued Functions

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19 March, 1999

## Abstract

We study the class  $\text{coNPMV}$  of complements of  $\text{NPMV}$  functions. Though defined symmetrically to  $\text{NPMV}$ , this class exhibits very different properties. We clarify the complexity of  $\text{coNPMV}$  by showing that it is essentially the same as that of  $\text{NPMV}^{\text{NP}}$ . Complete functions for  $\text{coNPMV}$  are exhibited and central complexity-theoretic properties of this class are studied. We show that computing maximum satisfying assignments can be done in  $\text{coNPMV}$ , which leads us to a comparison of  $\text{NPMV}$  and  $\text{coNPMV}$  with Krentel's classes  $\text{MaxP}$  and  $\text{MinP}$ . The difference hierarchy for  $\text{NPMV}$  is related to the query hierarchy for  $\text{coNPMV}$ . Finally, we examine a functional analogue of Chang and Kadin's relationship between a collapse of the Boolean hierarchy over  $\text{NP}$  and a collapse of the polynomial-time hierarchy.

## 1 Introduction

<sup>1-1</sup> Consider the complexity class  $\text{NPMV}$  of partial multivalued functions that are computed nondeterministically in polynomial time. As this class captures the complexity of computing witnesses of sets in  $\text{NP}$ , by studying this class, and more generally, by studying relations among complexity classes of partial multivalued functions, we directly contribute to understanding the complexity of computing witnesses. It is well known that a partial multivalued function  $f$  belongs to  $\text{NPMV}$  if and only if it is polynomial length-bounded and  $\text{graph}(f) = \{\langle x, y \rangle : y \text{ is a value of } f(x)\}$  belongs to  $\text{NP}$ .

<sup>1-2</sup> Now consider the class  $\text{coNPMV}$ . We will give a formal definition in the preliminaries section below. It will follow from the definition that a partial multivalued

function  $f$  belongs to  $\text{coNPMV}$  if and only if it is polynomial length-bounded and  $\text{graph}(f)$  belongs to  $\text{coNP}$ . Given this symmetry, graphs of functions in  $\text{NPMV}$  are in  $\text{NP}$  while graphs of functions in  $\text{coNPMV}$  are in  $\text{coNP}$ , and given what we know about  $\text{NP}$  and  $\text{coNP}$ , one might expect that  $\text{coNPMV}$  has essentially the same complexity as  $\text{NPMV}$ . Indeed, it is easy to see that  $\text{coNPMV} = \text{NPMV}$  if and only if  $\text{NP} = \text{coNP}$ . However, the point of this paper is to show that in many ways  $\text{coNPMV}$  is a more powerful class than is  $\text{NPMV}$ . One can derive more information from computing the complement of a function in  $\text{NPMV}$  than from computing the function. For one example of this phenomenon, we prove here that  $\text{coNPMV}$  is not included in  $\text{FP}^{\text{NPMV}}$  unless the polynomial hierarchy collapses. (This is an extension of a result of Fenner et al [FHOS97].) Thus, a  $\text{coNPMV}$  oracle provides more information than an  $\text{NPMV}$  oracle. This is surprising since function oracles, just as set oracles, provide knowledge about both their domains and their co-domains.

<sup>1-3</sup> We will define many-one reductions between multivalued functions. This will be a straightforward adaptation of the many-one metric reducibility of Krentel [Kre88]. In Section 3, we will consider many-one complete functions for  $\text{coNPMV}$ .

<sup>1-4</sup> Consider the partial multivalued function  $\text{sat}$ , defined so that  $y$  is a value of  $\text{sat}(\varphi)$  if and only if  $y$  is a satisfying assignment of Boolean formula  $\varphi$ . The function  $\text{sat}$  is complete for  $\text{NPMV}$ . Nevertheless, in Section 4 we will see that  $\text{sat}$  and similar functions belong to  $\text{coNPMV}$ . Even the seemingly more powerful  $\text{FP}^{\text{NP}}$ -complete function  $\text{maxsat}$ , that gives the maximum satisfying assignment of a formula, is contained in  $\text{coNPMV}$ . However, we will see that *neither*  $\text{NPMV}$  nor  $\text{FP}^{\text{NP}}$  are contained in  $\text{coNPMV}$ , and hence  $\text{coNPMV}$  is not closed under metric many-one reductions, unless the polynomial-time hierarchy collapses. Clearly, these function classes have strange closure properties, which we describe below.

<sup>1-5</sup> As an upper bound on the complexity of  $\text{coNPMV}$ , we show that, for any  $k \geq 2$ ,

$$\text{coNPMV} \subseteq \text{NPMV}(2) \subseteq \text{NPMV}(k) \subseteq$$

$$\text{NPMV}(k+1) \subseteq \text{NPMV}(\text{Poly}) \subseteq \text{NPMV}^{\text{NP}},$$

where  $\text{NPMV}(k)$  is the  $k$ -th level of the difference hierarchy for  $\text{NPMV}$  as defined by Fenner et al. [FHOS97].

<sup>1-6</sup> On the other hand, even though there is an infinite hierarchy of complexity classes between  $\text{coNPMV}$  and  $\text{NPMV}^{\text{NP}}$  (the difference hierarchy over  $\text{NPMV}$  does not collapse unless the polynomial-time hierarchy collapses [FHOS97]), our results suggest that the complexity of  $\text{coNPMV}$  is essentially the same as the complexity

of  $\text{NPMV}^{\text{NP}}$ : We prove in Section 5 that  $\text{NPMV}^{\text{NP}} = \pi_2^1 \circ \text{coNPMV}$  (where  $\pi_2^1$  is the projection function that maps a pair of strings to its first component). It follows that  $\text{NPMV}^{\text{NP}}$  is the closure of  $\text{coNPMV}$  under metric many-one reductions.

<sup>1-7</sup> In Section 6, we show that if the difference hierarchy for  $\text{NPMV}$  collapses, then the  $\text{NPMV}$  oracle hierarchy collapses. This is the functional analogue of the well-known result by Chang and Kadin relating a collapse of the Boolean hierarchy over  $\text{NP}$  to a collapse of the polynomial-time hierarchy.

<sup>1-8</sup> Finally, we remark that the phenomenon that universal quantification seems to lead to larger function classes was previously observed by Toda. We show in Section 7 how this observation follows from our results.

## 2 Preliminaries

<sup>2-1</sup> We fix  $\Sigma$  to be the finite alphabet  $\{0, 1\}$ . Let  $<$  denote the standard lexicographic order on  $\Sigma^*$ . For  $n \geq 0$  we define  $\Sigma^n = \{x \in \Sigma^* \mid |x| = n\}$ . By  $\langle \cdot, \cdot \rangle$  we denote a standard pairing function on  $\Sigma^* \times \Sigma^*$ .

<sup>2-2</sup> We use the standard complexity classes  $\text{P}$  and  $\text{NP}$  for (nondeterministic) polynomial time,  $\Sigma_k^p$  and  $\Delta_k^p = \text{P}^{\Sigma_{k-1}^p}$  for the levels of the polynomial-time hierarchy, and  $\text{NP}(k)$  for the levels of the Boolean hierarchy, for  $k \geq 1$ .

<sup>2-3</sup> Let  $f$  be a relation on  $\Sigma^* \times \Sigma^*$ . We will call  $f$  a (*partial*) *multivalued function* from  $\Sigma^*$  to  $\Sigma^*$ . By  $f(x) \mapsto y$  we denote that  $(x, y) \in f$  and say that  $f$  maps  $x$  to  $y$ . By  $\text{set-}f(x)$  we denote the set of outcomes of  $f$  on  $x$ ,  $\text{set-}f(x) = \{y : f(x) \mapsto y\}$ . The *graph of*  $f$  is  $\text{graph}(f) = \{\langle x, y \rangle : f(x) \mapsto y\}$ . The *domain of*  $f$ ,  $\text{dom}(f)$ , is the set of  $x$  where  $\text{set-}f(x)$  is nonempty. We will say that  $f$  is undefined at  $x$  if  $x \notin \text{dom}(f)$ . The domain of a class  $\mathcal{F}$  of functions is  $\text{dom}(\mathcal{F}) = \{\text{dom}(f) \mid f \in \mathcal{F}\}$ .

<sup>2-4</sup>

<sup>2-5</sup> Given partial multivalued functions  $f$  and  $g$ , define  $g$  to be a *refinement of*  $f$  if  $\text{dom}(g) = \text{dom}(f)$  and  $\text{graph}(g) \subseteq \text{graph}(f)$ . Let  $\mathcal{F}$  and  $\mathcal{G}$  be classes of partial multivalued functions. Purely as a convention, if  $f$  is a partial multivalued function, we define  $f \in_c \mathcal{G}$  if  $\mathcal{G}$  contains a refinement of  $f$ , and we define  $\mathcal{F} \subseteq_c \mathcal{G}$  if, for every  $f \in \mathcal{F}$ ,  $f \in_c \mathcal{G}$ . This notation is consistent with our intuition that  $\mathcal{F} \subseteq_c \mathcal{G}$  should entail that the complexity of computing values of functions in  $\mathcal{F}$  is not greater than the complexity of computing values of functions in  $\mathcal{G}$ .

<sup>2-6</sup> A transducer  $T$  is a nondeterministic Turing machine with a read-only input tape, a write-only output tape, read-write work tapes, and accepting states in the usual manner.  $T$  computes a value  $y$  on an input string  $x$  if there is an accepting computation of  $T$  on  $x$  for which  $y$  is the final content of  $T$ 's output tape. (In

this case, we will write  $T(x) \mapsto y$ .) Such transducers compute partial, multivalued functions. (As transducers do not typically accept all input strings, when we write “function,” “partial function” is always intended. If a function  $f$  is total, it will always be explicitly noted.)

2-7 The following classes of partial functions were first defined by Book, Long, and Selman [BLS84].

2-8

- NPMV is the set of all partial, multivalued functions computed by nondeterministic polynomial time-bounded transducers;
- NPSV is the set of all  $f \in \text{NPMV}$  that are single-valued;
- FP is the set of all partial functions computed by deterministic polynomial time-bounded transducers.

2-9

A function  $f$  belongs to NPMV if and only if it is polynomially length-bounded and  $\text{graph}(f)$  belongs to NP. In this paper we will adopt the convention, different from other papers on the subject, that all outputs of a function  $f \in \text{NPMV}$  on input  $x$  are of the same length, namely,  $p(|x|)$ , where  $p$  is some polynomial. This convention is merely for convenience and can easily be removed in all our results by using a padding argument.

2-10

The domain of every function in NPMV belongs to NP. An example is *sat*, which maps Boolean formulas to their satisfying assignments.

2-11

Fenner et al. [FHOS97] define the *difference hierarchy over* NPMV as follows. Let  $\mathcal{F}$  be a class of partial multivalued functions. A partial multivalued function  $f$  is in  $\text{co}\mathcal{F}$  if there exist  $g \in \mathcal{F}$  and a polynomial  $p$  such that for every  $x$ ,

$$\text{set-}f(x) = \Sigma^{p(|x|)} - \text{set-}g(x).$$

Let  $\mathcal{F}$  and  $\mathcal{G}$  be two classes of partial multivalued functions. A partial multivalued function  $h$  is in  $\mathcal{F} \wedge \mathcal{G}$ , respectively  $\mathcal{F} \vee \mathcal{G}$ , if there exist partial multivalued functions  $f \in \mathcal{F}$  and  $g \in \mathcal{G}$  such that for every  $x$ ,

$$\begin{aligned} \text{set-}h(x) &= \text{set-}f(x) \cap \text{set-}g(x), \text{ respectively} \\ \text{set-}h(x) &= \text{set-}f(x) \cup \text{set-}g(x). \end{aligned}$$

Let  $\mathcal{F} - \mathcal{G}$  denote  $\mathcal{F} \wedge \text{co}\mathcal{G}$ . Then,  $\text{NPMV}(k)$  is the class of partial multivalued functions defined in the following way:

$$\text{NPMV}(1) = \text{NPMV},$$

and, for  $k \geq 2$ ,

$$\text{NPMV}(k) = \text{NPMV} - \text{NPMV}(k-1).$$

Fenner et al. prove that, for every  $k \geq 1$ ,  $f \in \text{NPMV}(k)$  if and only if  $f$  is polynomially length-bounded and  $\text{graph}(f) \in \text{NP}(k)$ .

<sup>2-12</sup> In particular, we are interested in the class  $\text{coNPMV}$ . It follows that a function  $f$  belongs to  $\text{coNPMV}$  if and only if it is polynomially length-bounded and  $\text{graph}(f)$  belongs to  $\text{coNP}$ . Observe that the classes  $\text{NPMV}$  and  $\text{coNPMV}$  satisfy the nice symmetry that graphs of functions in the former class are in  $\text{NP}$  and those in the latter class are in  $\text{coNP}$ .

<sup>2-13</sup> Just as the definition of the Boolean hierarchy over  $\text{NP}$  leads to the class  $\text{NP}(\text{Poly})$  (see [Wag90]), we now introduce the class  $\text{NPMV}(\text{Poly})$ . It can be shown that a function  $h$  belongs to  $\text{NPMV}(k)$  if and only if there is a 2-ary function  $f \in \text{NPMV}$  such that

$$\text{set-}h(x) = \text{set-}f(x, k) - \left( \text{set-}f(x, k-1) - \left( \text{set-}f(x, k-2) - \left( \dots - \text{set-}f(x, 1) \dots \right) \right) \right).$$

We say that  $h \in \text{NPMV}(\text{Poly})$  if and only if there is a function  $f \in \text{NPMV}$  and a polynomial  $p$  such that

$$\text{set-}h(x) = \text{set-}f(x, p(|x|)) - \left( \text{set-}f(x, p(|x|) - 1) - \left( \text{set-}f(x, p(|x|) - 2) - \left( \dots - \text{set-}f(x, 1) \dots \right) \right) \right).$$

The above-mentioned result by Fenner et al. can be extended to show that  $f \in \text{NPMV}(\text{Poly})$  if and only if  $f$  is polynomially length-bounded and  $\text{graph}(f) \in \text{NP}(\text{Poly})$ .

<sup>2-14</sup> The primary new contribution of Fenner et al. is the development of hierarchies of classes of functions that access classes of partial functions as oracles. This development is based on the following description of oracle Turing machines with oracles that compute partial functions. Assume first that the oracle is a single valued partial function  $g$ . Let  $\perp$  be a symbol not belonging to the finite alphabet  $\Sigma$ . In order for a machine  $M$  to access a partial function oracle,  $M$  has a write-only input oracle tape, a separate read-only output oracle tape, and a special oracle call

state  $q$ . To query  $g$  on a string  $x$ ,  $M$  enters state  $q$  with  $x$  on the oracle input tape in the usual fashion. The oracle then returns the value  $g(x)$  on the oracle output tape if the value exists, and writes  $\perp$  on the tape otherwise. (It is possible that  $M$  may read only a portion of the oracle's output if the oracle's output is too long to read with the resources of  $M$ .) We shall assume, without loss of generality, that  $M$  never makes the same oracle query more than once on any possible computation path.

2-15 If  $g$  is a single-valued partial function and  $M$  is a deterministic oracle transducer as just described, then we let  $M[g]$  denote the single-valued partial function computed by  $M$  with oracle  $g$ .

**Definition 1** [FHOS97] *Let  $f$  and  $g$  be multivalued partial functions.  $f$  is Turing reducible to  $g$  in polynomial time,  $f \leq_T^P g$ , if for some deterministic polynomial-time oracle transducer  $M$ , for every single-valued refinement  $g'$  of  $g$ ,  $M[g']$  is a single-valued refinement of  $f$ .*

2-16 Fenner et al. prove that  $\leq_T^P$  is a reflexive and transitive relation over the class of all partial multivalued functions.

2-17 Let  $\mathcal{F}$  be a class of partial multivalued functions.  $\text{FP}^{\mathcal{F}}$  denotes the class of partial multivalued functions  $f$  that are  $\leq_T^P$ -reducible to some  $g \in \mathcal{F}$ .  $\text{FP}^{\mathcal{F}[k]}$  (respectively,  $\text{FP}^{\mathcal{F}[\log]}$ ) denotes the class of partial multivalued functions  $f$  that are  $\leq_T^P$ -reducible to some  $g \in \mathcal{F}$  via a machine that, on input  $x$ , makes  $k$  adaptive queries (respectively,  $\mathcal{O}(\log |x|)$  adaptive queries) to its oracle.

2-18 This definition template defines classes of multivalued partial functions such as  $\text{FP}^{\text{NPMV}}$  and can easily be extended to define  $\text{NPMV}^{\text{NPMV}}$ . If  $\mathcal{K}$  is a class of sets, then  $\text{FP}^{\mathcal{K}}$  is defined as usual, except that we allow it to compute partial functions (at the discretion of the oracle machine).

2-19 We will use the following generalization of the many-one metric reducibility of Krentel [Kre88] in order to discuss complete functions for classes of multivalued functions.

*Definition 2-1* **Definition 2** *Given partial multivalued functions  $f, g : \Sigma^* \mapsto \Sigma^*$ , we say  $f$  is metric many-one reducible to  $g$ , or symbolically,  $f \leq_m^P g$ , if there are functions  $t_1, t_2 \in \text{FP}$  such that the multivalued partial function  $h$  defined by*

$$h(x) = t_2(x, (g \circ t_1)(x))$$

*is a refinement of  $f$ , where  $\text{set-}h(x)$  is defined as*

$$\{t_2(x, y) : g(t_1(x)) \mapsto y\}.$$

Definition 2-2

If, in addition, we have  $\text{set-}h(x) = \text{set-}f(x)$  for all  $x$ , we call it a strong metric many-one reduction, denoted by  $f \leq_{sm}^P g$ .

2-20

The motivation underlying this definition is that, given a value of  $g(x)$ , one can compute in polynomial time a value of  $f(x)$ . In the case of a strong reduction, one gets all values of  $f(x)$  when varying over all values of  $g(x)$ . Obviously,  $f \leq_m^P g$  implies  $f \leq_T^P g$ .

2-22

The classes that we have been considering relate in interesting ways to studies of the complexity of optimization problems. In order to capture the complexity of optimization problems, Krentel [Kre88] defined the complexity classes MaxP and MinP as the functions computable by taking the maximum, respectively minimum, over sets of feasible solutions of problems in NP. Further, Krentel extended these classes to hierarchies of classes of optimization functions [Kre92]. Krentel defined these functions using the notion of a *metric Turing machine*, which we now review. Consider nondeterministic polynomial-time Turing machines that print an output value on every path. We associate with every inner node of the computation tree either the function min or the function max (for the classes MinP and MaxP, all nodes are associated with the same function). Thus, metric Turing machines define (total) functions from input words to integers via the usual bottom-up evaluation of the machine's computation tree. Since all the function classes considered in this paper are partial, we extend the metric Turing machine just defined by allowing the machine to output a special symbol  $\perp$  that denotes that the computation on the corresponding path ends with an undefined result. We extend the min and max functions in the obvious way: define  $\max(x, \perp) = \max(\perp, x) = x$  and  $\min(x, \perp) = \min(\perp, x) = x$ , for all  $x$  (including  $\perp$  itself). Vollmer and Wagner [VW93, VW95] gave a detailed structural examination of Krentel's hierarchy. Here, we just define the class MaxP using an operator-characterization from [VW95]. MinP is defined analogously.

2-23

$$h \in \text{MaxP} \iff \exists f, g \in \text{FP} : h(x) = \max_{0 \leq y \leq g(x)} f(x, y).$$

### 3 Functions Complete for coNPMV

3-1

NPMV is precisely the class of functions that compute witnesses for NP sets in the following sense: For any set  $L \in \text{NP}$  there exist a set  $A \in \text{P}$  and a polynomial  $p$  such that for all  $x$ , we have

$$x \in L \iff \exists y \in \Sigma^{p(|x|)} : (x, y) \in A.$$

Any  $y$  such that  $(x, y) \in A$  is called a *witness for  $x$*  (with respect to  $A$ ). Clearly, there is a function  $f_A \in \text{NPMV}$  such that  $\text{set-}f(x)$  is exactly the set of witnesses for  $x$ . On the other hand, any NPMV function  $f$  defines a set in NP, namely  $\text{dom}(f)$ . As a consequence of this discussion, we see that  $\text{dom}(\text{NPMV}) = \text{NP}$ .

3-2 Next, we extend the notion of a witness to  $\Sigma_2^p$ . For any  $\Sigma_2^p$  set  $L$  there exist a set  $B \in \text{coNP}$  and a polynomial  $p$  such that for all  $x$ , we have  $x \in L \iff \exists y \in \Sigma^{p(|x|)} : (x, y) \in B$ . A  $y$  such that  $(x, y) \in B$  is called a *witness for  $x$*  (with respect to  $B$ ). What function class captures the computation of witnesses for  $\Sigma_2^p$  sets? Since  $\Sigma_2^p = \text{NP}^{\text{NP}}$ , certainly witnesses can be computed in  $\text{NPMV}^{\text{NP}}$ . However, we will see below that the seemingly weaker class  $\text{coNPMV}$  already suffices to do so.

3-3 Let us consider set  $L$  again. We may safely assume that, for all  $(x, y) \in B$ , we have  $y \in \Sigma^{p(|x|)}$ . Since  $B \in \text{coNP}$ , it is then the graph of a  $\text{coNPMV}$  function  $f$ , so that  $\text{set-}f(x)$  is exactly the set of witnesses for  $x$ . Hence,  $\text{coNPMV}$  can compute witnesses for sets in  $\Sigma_2^p$ . Conversely, for any  $\text{coNPMV}$  function  $f$ , we have  $\text{dom}(f) \in \Sigma_2^p$ . This is because, for any  $x$ ,  $x \in \text{dom}(f) \iff \exists y \in \Sigma^{p(|x|)} : y \in \text{set-}f(x)$ . Thus,  $\text{coNPMV}$  is *precisely* the class of functions that computes witnesses for  $\Sigma_2^p$  sets. As a consequence, we have the following proposition.

Proposition 1  $\text{dom}(\text{coNPMV}) = \Sigma_2^p$ .

3-4 Witnesses of  $\Sigma_2^p$ -complete sets can give rise to complete functions for  $\text{coNPMV}$ . Consider, for example, the satisfiability problem  $\text{QBF}_2$  for Boolean formulas with two quantifiers. Let  $\varphi$  be a Boolean formula in the variables  $x = (x_1, \dots, x_k)$  and  $y = (y_1, \dots, y_l)$ . Then we define

$$\varphi(x, y) \in \text{QBF}_2 \iff \exists x \forall y : \varphi(x, y) = 1.$$

Let  $F_2$  be the multivalued function that computes witnesses, i.e., partial assignments  $x = (x_1, \dots, x_k)$ , for  $\text{QBF}_2$  formulas  $\varphi$  as above.

Theorem 1  $F_2$  is  $\leq_{sm}^P$ -complete for  $\text{coNPMV}$ .

Proof of Theorem 1 We have argued already that  $F_2 \in \text{coNPMV}$ . Let  $f$  be any  $\text{coNPMV}$  function. There is an NP transducer  $M$  and a polynomial  $p$  such that for all  $x$ , we have  $\text{set-}f(x) = \Sigma^{p(|x|)} - \text{set-}M(x)$ . We show how to compute a  $y \in \text{set-}f(x)$  from  $F_2(\varphi_x)$ , for an appropriately constructed formula  $\varphi_x$ .

3-5 Define a machine  $M'$  on input  $x$  as follows. First,  $M'$  guesses a  $y \in \Sigma^{p(|x|)}$ . Then,  $M'$  simulates  $M$  on input  $x$ . If  $M$  outputs  $y$  on the simulated path, then  $M'$  rejects. Otherwise,  $M'$  accepts.

3-6 We have to define the reduction functions  $t_1$  and  $t_2$  as required in Definition 2. Function  $t_1$  is the Cook-Levin reduction<sup>1</sup> applied to  $x$  with  $M'$  as the underlying machine. This will give a Boolean formula  $\varphi_x$  that, intuitively, describes the work of  $M'$  on input  $x$ . The variables of  $\varphi_x$  can be partitioned into two parts, for example,

- $y_1, \dots, y_k$ , that are used to describe that  $M'$  guesses a  $y \in \Sigma^{p(|x|)}$ , and
- $z_1, \dots, z_l$ , that are used to describe the subsequent simulation of  $M$ .

Furthermore, from any setting of the variables  $y_1, \dots, y_k$  of  $\varphi_x$ , we can reconstruct in polynomial time the  $y \in \Sigma^{p(|x|)}$  guessed by  $M'$ . This is done by function  $t_2$ .

3-7 Let us fix a setting of the variables  $y_1, \dots, y_k$  and let  $y \in \Sigma^{p(|x|)}$  be the corresponding string guessed by  $M'$ . Then we have

$$\begin{aligned} & \forall z_1, \dots, z_l : \varphi_x(y_1, \dots, y_k, z_1, \dots, z_l) = 1 \\ \iff & M' \text{ accepts on all paths following } y \\ \iff & y \notin \text{set-}M(x) \\ \iff & f(x) \mapsto y, \end{aligned}$$

and hence,  $\text{set-}t_2(x, F_2 \circ t_1(x)) = \text{set-}f(x)$ , where  $t_1(x) = \varphi_x$ .

Proof of Theorem 1  $\square$

3-8 A crucial point in the above proof is that the Cook-Levin reduction maintains witnesses. That is, from a given assignment for the constructed formula  $\varphi_x$ , one can recover a corresponding path of the nondeterministic machine. Thus, any  $\Sigma_2^p$ -complete set sharing this property with  $\text{QBF}_2$  defines a  $\text{coNPMV}$ -complete function in an analogous way.

3-9 As an example, consider the following set  $L_f$ . For any  $\text{NPMV}$  function  $f$  and any even-valued polynomial  $p$  such that  $f$  maps strings of length  $n$  to strings of length  $p(n)$ ,  $x \in L_f$  if and only if

$$\exists y \in \Sigma^{p(|x|)/2} \forall z \in \Sigma^{p(|x|)/2} : f(x) \not\mapsto yz.$$

In other words, string  $y$  is not a prefix of an output of  $f(x)$ .

3-10 Clearly, for every  $f \in \text{NPMV}$ , we have that  $L_f$  is in  $\Sigma_2$ . Thus, in particular, taking  $f = \text{sat}$ ,  $L_{\text{sat}}$  is  $\Sigma_2^p$ -complete and has the above-mentioned property. We

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<sup>1</sup>This is the well-known reduction that transforms the computation of an NP machine to a Boolean formula that is satisfiable iff the machine accepts; see, e.g., [Coo71].

conclude that the corresponding witness function, *not-pre-sat*, is complete for coNPMV, where (by definition)  $\text{not-pre-sat}(\varphi) \mapsto y \iff y$  is a truth assignment of the first half of  $\varphi$ 's variables that is not a prefix of a satisfying assignment of  $\varphi$ .

Theorem 2 *not-pre-sat* is  $\leq_{sm}^P$ -complete for the class coNPMV.

3-11 *Not-pre-sat* is a trivial transformation of  $F_2$ , so Theorem 2 can be seen directly via a straightforward metric reduction from  $F_2$ .

## 4 Properties of coNPMV

4-1 NPMV is closed under  $\leq_{sm}^P$ -reductions, but not under  $\leq_m^P$ -reductions; in fact, it is possible to have  $g \in \text{NPMV}$  and  $f \leq_m^P g$  but  $\text{graph}(f)$  be noncomputable. (For example, define  $f$  to map  $x$  to two values, the first of which is either 0 or 1 and solves the halting problem on  $x$  and the second of which is the constant 10. Then clearly  $\text{graph}(f)$  is not computable, but the constant function 10 is a refinement of  $f$  in NPMV.) However, NPMV is closed under this reduction in a weaker sense, defined below.

Definition 3 A class  $\mathcal{F}$  is *c-closed under reducibility*  $\leq_r$ , if  $g \in \mathcal{F}$  and  $f \leq_r g$  implies  $f \in_c \mathcal{F}$ .

4-2 It is immediate from this definition that NPMV is c-closed under  $\leq_m^P$ -reductions. One might suspect that this same fact holds for coNPMV. However, it is quite unlikely that coNPMV is c-closed under this reducibility: otherwise, since  $\text{sat} \in \text{coNPMV}$  and  $\text{sat}$  is complete for NPMV, we would get that  $\text{NPMV} \subseteq_c \text{coNPMV}$ . But this seems to be very unlikely, as the following extension of a result of Fenner et al. [FHOS97] shows.

Theorem 3  $\text{NPMV} \subseteq \text{coNPMV} \iff \text{NPMV} \subseteq_c \text{coNPMV} \iff \text{NP} = \text{coNP}$ .

Proof of Theorem 3 We cycle through the implications. The first implication is trivial. For the second, let  $L \in \text{NP}$ . Define function

$$\chi_L(x) = \begin{cases} 1 & \text{if } x \in L \\ \perp & \text{otherwise.} \end{cases}$$

Then we have  $\chi_L \in \text{NPMV}$ , and hence, by assumption,  $\chi_L \in \text{coNPMV}$ . Therefore,  $\text{graph}(\chi_L) \in \text{coNP}$ , which implies that  $L \in \text{coNP}$  since  $x \in L \iff (x, 1) \in \text{graph}(\chi_L)$ .

4.3 Now suppose that  $\text{NP} = \text{coNP}$  and let  $f \in \text{NPMV}$ . Then  $\text{graph}(f) \in \text{NP}$  and, therefore, in  $\text{coNP}$  by assumption. Thus  $f \in \text{coNPMV}$ .

Proof of Theorem 3  $\square$

Corollary 1  $\text{coNPMV}$  is  $c$ -closed under  $\leq_m^P$ -reducibility if and only if  $\text{NP} = \text{coNP}$ .

4.4 We observe that the proof of Theorem 3 shows also that  $\text{NPSV} \subseteq \text{coNPMV} \iff \text{NP} = \text{coNP}$ , even though it is fairly easy to see that  $\text{NPSV}_t$ , the class of all total NPSV functions, is contained in  $\text{coNPMV}$ . We also note that Theorem 3 extends to higher levels of the difference hierarchies over NPMV and NP, that is,  $\text{NPMV}(k) \subseteq \text{coNPMV}(k) \iff \text{NPMV}(k) \subseteq_c \text{coNPMV}(k) \iff \text{NP}(k) = \text{coNP}(k)$ . By a result of Kadin [Kad88], a collapse of the Boolean hierarchy implies a collapse of the polynomial-time hierarchy. Hence, there is likely to be a whole hierarchy between  $\text{coNPMV}$  and  $\text{NPMV}^{\text{NP}}$ .

Theorem 4 For all  $k \geq 2$ , we have

$$\begin{aligned} \text{coNPMV} \subseteq \text{NPMV}(2) \subseteq \text{NPMV}(k) \subseteq \\ \text{NPMV}(k+1) \subseteq \text{NPMV}(\text{Poly}) \subseteq \text{NPMV}^{\text{NP}}. \end{aligned}$$

Furthermore, all of the inclusions are strict unless the polynomial-time hierarchy collapses.

Proof of Theorem 4 It remains to show the last inclusion. Let  $f \in \text{NPMV}(\text{Poly})$ . Then the graph of  $f$  is in  $\text{NP}(\text{Poly})$ , which is known to be equal to  $\text{P}^{\text{NP}[\log]}$  [Wag90]. Obviously,  $f$  can be computed by an NPMV algorithm with access to a  $\text{P}^{\text{NP}[\log]}$  oracle: simply guess an output of  $f$  and, querying its graph, check that the guess is correct. Thus,  $\text{NPMV}(\text{Poly}) \subseteq \text{NPMV}^{\text{P}^{\text{NP}[\log]}} \subseteq \text{NPMV}^{\text{NP}}$ .

Proof of Theorem 4  $\square$

4.5 Under the likely assumption that  $\text{NP} \neq \text{coNP}$ , we see, by Theorem 3, that the class NPMV is not included in  $\text{coNPMV}$ , even though the function  $\text{sat}$ , which is complete for NPMV, belongs to  $\text{coNPMV}$ . This phenomenon happens again for  $\text{maxsat}$ , the function that maps a Boolean formula to its lexicographically largest satisfying assignment. Fenner et al. [FHOS97] show that  $\text{maxsat} \in \text{NPMV}(2)$ . In fact, it is even in  $\text{coNPMV}$ . However, we will show that the corresponding classes, namely  $\text{MaxP}$  or  $\text{FP}^{\text{NP}}$ , are included in  $\text{coNPMV}$  if and only if  $\text{NP} = \text{coNP}$ .

4.6

Theorem 5  $\text{maxsat} \in \text{coNPMV}$ .

Proof of Theorem 5 Consider an NPMV machine  $M$  that, on input of a formula  $\varphi$ , guesses an assignment  $y$  for  $\varphi$ . If  $y$  does not satisfy  $\varphi$ , then  $M$  accepts and outputs  $y$ . Otherwise, if  $y$  does satisfy  $\varphi$ ,  $M$  guesses another assignment  $y' > y$ . If  $y'$  also satisfies  $\varphi$ ,  $M$  outputs  $y$ ; otherwise,  $M$  rejects (and outputs nothing).

4-7  $M$  outputs every assignment except the maximum satisfying one (if there is one). Hence,  $\text{maxsat} \in \text{coNPMV}$ .

Proof of Theorem 5  $\square$

4-8 Krentel [Kre92] showed that  $\text{FP}^{\text{NP}} = \text{FP}^{\text{MaxP}[1]}$ . Since  $\text{FP}^{\text{NPMV}} = \text{FP}^{\text{NP}}$  [FHOS97] and  $\text{maxsat}$  is complete for MaxP, we have that  $\text{FP}^{\text{NPMV}} \subseteq \text{FP}^{\text{coNPMV}[1]}$ . That is, polynomially many queries of a FP function to NPMV can be replaced by one query to coNPMV. Hence, as we have mentioned, coNPMV seems to be a more powerful class than NPMV. We will give more evidence for this in the next section.

4-9

Corollary 2  $\text{MaxP} \subseteq \text{coNPMV} \iff \text{MinP} \subseteq \text{coNPMV} \iff \text{NP} = \text{coNP}$ .

Proof of Corollary 2 If  $\text{MaxP} \subseteq \text{coNPMV}$ , then  $\text{NPMV} \subseteq_c \text{MaxP} \subseteq \text{coNPMV}$ , and, therefore,  $\text{NPMV} \subseteq_c \text{coNPMV}$ . But by Theorem 3, this implies  $\text{NP} = \text{coNP}$ . Conversely, if  $\text{NP} = \text{coNP}$ , then  $\text{NPMV}^{\text{NP}} = \text{NPMV}^{\text{NP} \cap \text{coNP}} = \text{NPMV}$ . This implies  $\text{MaxP} \subseteq \text{NPMV}$  and, since the hypothesis also implies  $\text{NPMV} = \text{coNPMV}$ , that  $\text{MaxP} \subseteq \text{coNPMV}$ .

Proof of Corollary 2  $\square$

4-10 We conclude this section with an observation regarding the relationship between MaxP and NPMV. First, note that trivially  $\text{NPSV} \subseteq \text{MaxP} \cap \text{MinP}$  since *the* output of an NPSV function is both the minimum and the maximum. Similarly,  $\text{NPMV} \subseteq_c \text{MaxP} \cap \text{MinP}$ . The more interesting question is whether these inclusions are strict. This is quite likely.

Theorem 6  $\text{MaxP} \subseteq \text{NPMV} \iff \text{MinP} \subseteq \text{NPMV} \iff \text{NP} = \text{coNP}$ .

Proof of Theorem 6 If  $\text{NP} = \text{coNP}$ , then  $\text{FP}^{\text{NP}} \subseteq \text{NPMV}$  [Sel94]; thus, especially  $\text{MaxP} \cup \text{MinP} \subseteq \text{NPMV}$ .

4-11 Now suppose  $\text{MaxP} \subseteq \text{NPMV}$  (the case for  $\text{MinP}$  is analogous). Let  $L \in \text{coNP}$ . Define

$$f(x) = \begin{cases} 0 & \text{if } x \in L \\ 1 & \text{otherwise.} \end{cases}$$

Then  $f \in \text{MaxP}$  and hence, by assumption, is in  $\text{NPMV}$ . Since  $x \in L$  if and only if  $f(x) = 0$ , we have  $L \in \text{NP}$ .

Proof of Theorem 6  $\square$

4-12 The last two results relativize; analogous results hold for higher levels of the  $\text{NPMV}$  hierarchy and Krentel's min/max hierarchy [FHOS97, VW95]. For the relativized version of Theorem 6 one has to use techniques from Krentel [Kre92] and Vollmer and Wagner [VW95].

## 5 A Characterization of $\text{coNPMV}$

5-1 As we have already seen in the preceding section,  $\text{coNPMV}$  seems to be a more powerful class than  $\text{NPMV}$ . This is somewhat surprising in light of the aforementioned symmetry in the definitions of  $\text{coNPMV}$  and  $\text{NPMV}$  by their graphs.

5-2 The following theorem shows that  $\text{coNPMV}$  is in fact very close to  $\text{NPMV}^{\text{NP}}$ . This is surprising as well, as we have already seen in Theorem 4 that there is a hierarchy of function classes between  $\text{coNPMV}$  and  $\text{NPMV}^{\text{NP}}$ .

5-3 If  $f$  is a multivalued function and  $g$  is a single-valued function, then  $g \circ f$  is defined by  $\text{graph}(g \circ f) = \{ \langle x, g(y) \rangle : f(x) \mapsto y \}$ . Let  $\pi_2^1$  denote the projection function that maps a pair of strings to its first component. By  $\pi_2^1 \circ \text{coNPMV}$  we denote  $\{ \pi_2^1 \circ f : f \in \text{coNPMV} \}$ .

Theorem 7  $\text{NPMV}^{\text{NP}} = \pi_2^1 \circ \text{coNPMV}$ .

Proof of Theorem 7 The right-to-left containment follows from Theorem 4 and the fact that the projection of any  $\text{NPMV}^{\text{NP}}$  function is still in  $\text{NPMV}^{\text{NP}}$ ; hence,  $\pi_2^1 \circ \text{coNPMV} \subseteq \text{NPMV}^{\text{NP}}$ .

5-4 For the other direction, let  $f \in \text{NPMV}^{\text{NP}}$ . By a standard argument,  $\text{graph}(f) \in \Sigma_2^p$ , and thus there is a polynomial  $q$  and a predicate  $R \in \text{coNP}$  such that for any  $x$  and  $y \in \Sigma^{q(|x|)}$

$$f(x) \mapsto y \iff \exists z \in \Sigma^{q(|x|)} : R(x, y, z).$$

Define  $f'$  such that for any  $x$  and any  $y, z \in \Sigma^{q(|x|)}$

$$f'(x) \mapsto \langle y, z \rangle \iff R(x, y, z).$$

5-5 So  $R$  witnesses that  $f' \in \text{coNPMV}$ . But  $f(x) = \pi_2^1 \circ f'(x)$ , which shows that  $f \in \pi_2^1 \circ \text{coNPMV}$ .

Proof of Theorem 7  $\square$

5-6 The reason why it is likely that  $\text{coNPMV}$  is a proper subclass of  $\text{NPMV}^{\text{NP}}$  is not because outputs of  $\text{coNPMV}$  functions give too little information, but rather that they give too much. We can compute an arbitrary  $\text{NPMV}^{\text{NP}}$  function simply by throwing away part of the output of a  $\text{coNPMV}$  function. This is what the projection operator accomplishes, and it is most likely necessary.

5-7 If we apply Theorem 7, many properties of  $\text{NPMV}^{\text{NP}}$  now carry over to  $\text{coNPMV}$ . In Section 3 we have shown functions  $F_2$  and *not-pre-sat* complete for  $\text{coNPMV}$ . Since the projection function is in  $\text{FP}$ , we get that those functions are complete for  $\text{NPMV}^{\text{NP}}$  as well.

Corollary 3  $\text{NPMV}^{\text{NP}}$  is the  $c$ -closure of  $\text{coNPMV}$  under  $\leq_m^{\text{P}}$ -reducibility and the closure of  $\text{coNPMV}$  under  $\leq_{sm}^{\text{P}}$ -reducibility.

5-8 In particular, we get the following corollary.

Corollary 4  $\text{FP}^{\text{coNPMV}[k]} = \text{FP}^{\text{NPMV}^{\text{NP}}[k]}$  for all  $k \geq 1$ , and  $\text{FP}^{\text{coNPMV}} = \text{FP}^{\text{NPMV}^{\text{NP}}} = \text{FP}^{\text{NP}^{\text{NP}}}$ .

5-9 Observe, by contrast, that  $\text{FP}^{\text{NPMV}} = \text{FP}^{\text{NP}} = \text{FP}^{\text{MinP}} = \text{FP}^{\text{MaxP}}$ , so  $\text{coNPMV}$  and  $\text{NPMV}$  define different  $\Delta$ -levels of the functional polynomial hierarchy.

5-10 Fenner et al. [FHOS97] have shown that  $\text{NPMV}(2) \subseteq \text{FP}^{\text{NPMV}} \iff \Sigma_2^{\text{P}} = \Delta_2^{\text{P}}$ . Note that, in contrast for the corresponding language classes, we have  $\text{NP}(k) \subseteq \text{P}^{\text{NP}}$  for all  $k$ . We can now improve the result of Fenner et al.

Corollary 5  $\text{coNPMV} \subseteq \text{FP}^{\text{NPMV}} \iff \Sigma_2^{\text{P}} = \Delta_2^{\text{P}}$ .

Proof of Corollary 5 If  $\Sigma_2^{\text{P}} = \Delta_2^{\text{P}}$ , then

$$\begin{aligned} \text{coNPMV} &\subseteq \text{FP}^{\text{coNPMV}} = \text{FP}^{\text{NPMV}^{\text{NP}}} = \text{FP}^{\Sigma_2^{\text{P}}} \\ &= \text{FP}^{\Delta_2^{\text{P}}} = \text{FP}^{\text{NP}} = \text{FP}^{\text{NPMV}}, \end{aligned}$$

where the last equality is Theorem 1 in [FHOS97] and the second equally follows from the relativized version of the same theorem. Conversely, if  $\text{coNPMV} \subseteq \text{FP}^{\text{NPMV}}$ , then  $\text{dom}(\text{coNPMV}) \subseteq \text{P}^{\text{NP}} = \Delta_2^p$ , so that  $\Sigma_2^p \subseteq \Delta_2^p$ .

Proof of Corollary 5  $\square$

Corollary 6 *For any  $k \geq 1$ , we have*

$$\begin{aligned} \text{NPMV}^{\text{NP}} \subseteq \text{FP}^{\text{coNPMV}[1]} \subseteq \text{FP}^{\text{coNPMV}[k]} \subseteq \\ \text{FP}^{\text{coNPMV}[k+1]} \subseteq \text{FP}^{\text{coNPMV}} = \text{FP}^{\text{NP}^{\text{NP}}}. \end{aligned}$$

*Furthermore, all inclusions are strict unless the polynomial-time hierarchy collapses.*

**Proof of Corollary 6** It remains to show the strictness of the inclusions. Suppose  $\text{FP}^{\text{coNPMV}[1]} \subseteq \text{NPMV}^{\text{NP}}$ . This is equivalent to  $\text{FP}^{\text{NPMV}^{\text{NP}}[1]} \subseteq \text{NPMV}^{\text{NP}}$  by Corollary 4, which implies  $\text{P}^{\Sigma_2^p[1]} \subseteq \Sigma_2^p$ . But then  $\Pi_2^p = \Sigma_2^p = \text{PH}$ . For the other inclusions, suppose  $\text{FP}^{\text{coNPMV}[k]} = \text{FP}^{\text{coNPMV}[k+1]}$ . Then  $\text{FP}^{\text{NPMV}^{\text{NP}}[k]} = \text{FP}^{\text{NPMV}^{\text{NP}}[k+1]}$ . By a theorem of Fenner et al. [FHOS97], this implies that  $\text{FP}^{\Sigma_2^p[k]} = \text{FP}^{\Sigma_2^p[k+1]}$ , which, by a relativization of Kadin's theorem [Kad88], implies that the polynomial hierarchy collapses.

Proof of Corollary 6  $\square$

5-11

Thus we see, combining Theorem 4 and Corollary 6, that all classes of the difference hierarchy over NPMV are included in the query hierarchy over coNPMV, in fact, already in its first level. There are (under reasonable assumptions) no inclusions in the opposite direction. Concerning the relationship between the query hierarchy over NPMV and the difference hierarchy over NPMV, we know from Fenner et al. [FHOS97] that all classes of the first hierarchy are included in certain classes of the second hierarchy. Any inclusion in the opposite direction implies  $\text{coNPMV} \subseteq \text{FP}^{\text{NPMV}}$ , which again implies a collapse of the polynomial hierarchy, by Corollary 5.

## 6 Relationships between the Functional Difference and Polynomial-Time Hierarchies

6-1

Chang and Kadin [CK96] showed that if the Boolean hierarchy over NP collapses to the  $k$ -th level, then the polynomial hierarchy collapses to the  $k$ -th level of the

Boolean hierarchy over  $\text{NP}^{\text{NP}}$ : if  $\text{NP}(k+1) = \text{NP}(k)$ , then  $\text{PH} = \text{NP}^{\text{NP}}(k)$ . It is a simple consequence of known results that a similar connection exists for the corresponding functional hierarchies, namely  $\text{NPMV}(k)$  and  $\Sigma\text{MV}_k = \text{NPMV}^{\Sigma_k^p}$ .

**Theorem 8** *For any  $k \geq 1$ , if  $\text{NPMV}(k+1) = \text{NPMV}(k)$ , then  $\Sigma\text{MV}_3 = \text{NPMV}^{\text{NP}}(k)$ .*

*Proof of Theorem 8*  $\text{NPMV}(k+1) = \text{NPMV}(k)$  is equivalent to  $\text{NP}(k+1) = \text{NP}(k)$  [FHOS97], which implies  $\Sigma_3^p = \text{NP}^{\text{NP}}(k)$  [CK96] (relativized). By considering the graphs of functions [FHOS97], we immediately get that  $\Sigma\text{MV}_3 = \text{NPMV}^{\text{NP}}(k)$ .

Proof of Theorem 8  $\square$

6-5 Since  $\text{NP}^{\text{NP}}(k) \subseteq \text{P}^{\text{NP}^{\text{NP}}[k]}$ , a consequence of Chang and Kadin's theorem is that, if  $\text{NP}(k+1) = \text{NP}(k)$ , then  $\Sigma_3^p = \text{P}^{\text{NP}^{\text{NP}}[k]}$  (indeed, they prove this directly in their paper before treating the stronger result). The functional analogue of such a collapse would be  $\Sigma\text{MV}_3 = \text{FP}^{\text{NPMV}^{\text{NP}}[k]}$  or, equivalently,  $\Sigma\text{MV}_3 = \text{FP}^{\text{coNPMV}[k]}$ . We cannot expect this as a direct consequence of Theorem 8, since the difference and query hierarchies are not intertwined in this context. Nevertheless, such an analogous result does hold. To see this, we have to modify the proof of the Chang and Kadin theorem.

**Theorem 9** *If  $\text{NPMV}(k+1) = \text{NPMV}(k)$ , then  $\Sigma\text{MV}_3 = \text{NPMV} \circ \text{FP}^{\text{coNPMV}[k-1]}$ .*

*Proof of Theorem 9* In order to explain how Chang and Kadin's proof gives this result, we recall some of their definitions, with some minor modifications in notation (for greater detail, we refer the reader to their paper [CK96]). Denote the  $\leq_m^p$ -complete language for  $\text{NP}(k)$  (respectively,  $\text{coNP}(k)$ ) as  $L_{\text{NP}(k)}$  (respectively,  $L_{\text{coNP}(k)}$ ). For example,  $L_{\text{NP}(1)} = \text{SAT}$  and  $L_{\text{NP}(2)} = \{ \langle x_1, x_2 \rangle : x_1 \in \text{SAT} \text{ and } x_2 \in \overline{\text{SAT}} \}$ . Since, by hypothesis,  $\text{NP}(k) = \text{coNP}(k)$ , it follows that  $L_{\text{NP}(k)} \leq_m^p L_{\text{coNP}(k)}$ . The basic idea underlying the Chang and Kadin proof is that such a reduction induces a reduction from an initial segment of  $\text{SAT}$  to an initial segment of  $\overline{\text{SAT}}$ . This is done inductively via the notion of a "hard sequence," which is a  $j$ -tuple that, together with a  $\leq_m^p$ -reduction from  $\text{NP}(k)$  to  $\text{coNP}(k)$ , can be used to find a  $\leq_m^p$ -reduction from  $\text{NP}(k-j)$  to  $\text{coNP}(k-j)$ .

**Definition 4** *Let  $L_{\text{NP}(k)} \leq_m^p L_{\text{coNP}(k)}$  via some polynomial time function  $h$ . Then we call  $\langle 1^m, x_1, \dots, x_j \rangle$  a hard sequence with respect to  $h$  for length  $m$  of order  $j$ , if either  $j = 0$  or the following conditions hold:*

1.  $1 \leq j \leq k - 1$ ,
2.  $|x_j| \leq m$ ,
3.  $x_j \in \overline{\text{SAT}}$ ,
4.  $\langle 1^m, x_1, \dots, x_{j-1} \rangle$  is a hard sequence with respect to  $h$ , and
5. for all  $y_1, \dots, y_\ell \in \Sigma^*$  where  $\ell = k - j$ , and for all  $1 \leq i \leq \ell$ ,  $|y_i| \leq m$ ,  
 $\pi_{\ell+1} \circ h(\langle y_1, \dots, y_\ell, x_j, \dots, x_1 \rangle) \in \overline{\text{SAT}}$ .

<sup>6-3</sup> A hard sequence is called *maximal* if it cannot be extended to a hard sequence of a higher order. In this case, the order of the sequence  $j$  is said to be maximal.

<sup>6-4</sup> We can now outline the proof. Chang and Kadin's Lemma 3 [CK96] states that, given a maximal hard sequence for an appropriate (polynomially bounded) length, an NP machine can recognize an initial segment of the canonical complete language for  $\text{NP}^{\text{NP}}$ . That is, with the aid of such a sequence, we can replace a  $\Sigma_2^p$  machine with an NP machine. Thus, it suffices to find a maximal hard sequence to collapse the NP's of a  $\Sigma\text{MV}_3 = \text{NPMV}^{\Sigma_2^p}$  machine.

<sup>6-5</sup> Our principle observation is this: *Hard sequences of any given order can be obtained by a single query to a coNPMV oracle.* This result can easily be seen as follows. Define the function  $H: 1^+ \times \mathbb{N} \mapsto \Sigma^*$  such that  $H(1^m, j) \mapsto \langle 1^m, x_1, \dots, x_j \rangle$  if and only if  $\langle 1^m, x_1, \dots, x_j \rangle$  is a hard sequence for length  $m$  of order  $j$ . It follows from Definition 4 that the set of hard sequences is in coNP [CK96]; hence,  $\text{graph}(H) \in \text{coNP}$ , so that  $H \in \text{coNPMV}$ . Therefore, we can obtain a maximal hard sequence for the appropriate polynomial length  $m = p(|x|)$  by querying a coNPMV oracle for a value of  $H(1^m, j)$  for  $j$  varying from 1 to  $k - 1$ . We then feed the resulting maximal hard sequence, along with the original input  $x$ , to an NPMV machine that can, via the induced reduction from coNP to NP, collapse the NP oracles in an  $\text{NPMV}^{\text{NP}^{\text{NP}}}$  computation.

Proof of Theorem 9  $\square$

## 7 A Remark on Counting Classes

<sup>7-1</sup> The results of our paper show that in the context of relational structures computed by polynomial-time machines, in a sense the universal mode is more powerful than the existential one. In the context of counting classes, a similar phenomenon has

been observed by Seinosuke Toda [Tod91]. In this section, we briefly show that Toda's result is a special case of one of our observations.

7-2 Recall the following general definition of counting classes from [Tod91, VW93]:

7-3 Let  $\mathcal{K}$  be a class of sets. Then,  $\#\cdot\mathcal{K}$  consists of those functions  $f$  for which there exist a set  $A \in \mathcal{K}$  and a polynomial  $p$  such that, for all  $x$ ,

$$f(x) = |\{y \mid |y| \leq p(|x|) \wedge \langle x, y \rangle \in A\}|.$$

It is obvious that  $\#\cdot\mathbf{P} = \#\mathbf{P}$ . Moreover it can be shown that  $\#\cdot\mathbf{NP} = \text{span}\mathbf{P}$ , where  $\text{span}\mathbf{P}$  is the class of functions that count the number of distinct outputs of a nondeterministic polynomial-time transducer.

7-4 We have the following relationship to our classes of functions:

Proposition 2

1.  $\#\cdot\mathbf{NP}$  consists of exactly those functions  $h$  for which there exists a function  $f \in \mathbf{NPMV}$  such that for all  $x$ ,  $h(x) = |\text{set-}f(x)|$ .
2.  $\#\cdot\text{coNP}$  consists of exactly those functions  $h$  for which there exists a function  $f \in \text{coNPMV}$  such that for all  $x$ ,  $h(x) = |\text{set-}f(x)|$ .

7-5 Now we have the following surprising result, which was already proved by Toda [Tod91, Theorem 4.1.6]:

Corollary 7  $\#\cdot\text{coNP} = \#\mathbf{P}^{\mathbf{NP}}$ .

Proof of Corollary 7 Proof is immediate by the preceding proposition and the fact that  $\text{coNPMV}$  is the class of functions that compute witnesses for  $\Sigma_2^p$  computations. See the discussion before Proposition 1.

Proof of Corollary 7  $\square$

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