

# A Note on a Syntactical Measure of the Complexity of Programs

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**Abstract**—We introduce a programming language operating on stacks and a syntactical measure  $\sigma$ , such that a natural number  $\sigma(\mathbf{P})$  is assigned to each program  $\mathbf{P}$ . The measure considers how the presence of loops defined over size-increasing (and non-size-increasing) subprograms influences the complexity of the program itself. Functions computed by a program of  $\sigma$ -measure  $n$  are exactly those computable by a Turing machine with running time in  $\mathcal{E}^{n+2}$  (the  $n+2$ -th Grzegorzcyk class). Programs of  $\sigma$ -measure 0 compute the polynomial-time computable functions. Thus, we have a syntactical characterization of the functions belonging to the Grzegorzcyk hierarchy; this result represents an improvement with respect to previous similar results.

**Index Terms**—polynomial-time complexity, Grzegorzcyk hierarchy, imperative programming languages, stack programs

## I. INTRODUCTION

The definition of a class of functions with a given computational complexity is usually given by introducing an explicit bound on time and/or space resources used by a Turing Machine during the computation of the functions. Other approaches capture complexity classes by means of some form of *limited recursion*; the first characterization of this type has been given by Cobham [3], who has shown that the polynomial-time computable functions are exactly those that are definable by *bounded recursion on notation*, starting from a set of simple basic functions.

In the recent years, a number of characterizations of complexity classes has been given, showing that they can be captured by means of various forms of *ramified recursion*, without any explicitly bounded scheme of recursion. Initiated by Simmons [23], Bellantoni and Cook [1] and Leivant [13] - [15], one can find functional characterization of linear-space/time computable functions Linspace and Logspace [9], polynomial time [16], polynomial space [18] [21], the elementary functions [21] [4], non-size-increasing computations [6], among the others.

A different approach can be found in [7] [8] [10] [11]; more recently, in [12] [17]. The properties of *imperative programs* (such as complexity, resource utilization, termination) are now investigated by analyzing their syntax only. In particular, the properties of a programming language

operating on stacks are studied in [10]; the language supports loops over stacks, conditionals and concatenation, besides the usual pop and push operations (see Section II for the detailed semantics). The natural concept of  $\mu$ -measure is then introduced; it is a syntactical method by which one is able to assign to each program  $\mathbf{P}$  a number  $\mu(\mathbf{P})$ . It is proved the following *bounding theorem*: functions computed by stack programs of  $\mu$  measure  $n$  have a length bound  $b \in \mathcal{E}^{n+2}$  (the  $n+2$ -th Grzegorzcyk class), that is  $|f(\vec{w})| \leq b(|\vec{w}|)$ ; as a consequence, stack programs of measure 0 have polynomial running time, and programs of measure  $n$  compute functions whose time complexity is in the  $n+2$ -th finite level of the Grzegorzcyk hierarchy. This result provides a measure of the impact of nesting loops on computational complexity; if a stack  $Z$  is updated into a loop controlled by a stack  $Y$  and, afterwards,  $Y$  is updated into a loop controlled by  $Z$ , we have a so called *top circle* in the program; when this circular reference occurs into an external loop, a blow up in the complexity of the program is produced. The  $\mu$ -measure is a syntactical way to detect top circles; each time one of them occurs in the body of a loop, the  $\mu$  measure is increased by 1 (see below, Section III and definition 3.1).

There are various ways of improving the measure  $\mu$  (for instance, see [11]), since it is an undecidable problem whether or not a function computed by a given stack program lies in a given complexity class. In this paper, we provide a refinement of  $\mu$ , starting from the following consideration: a program whose structure leads the  $\mu$ -measure to be equal to  $n$  contains  $n$  nested top circles, and this implies, by the bounding theorem, that the program has a length bound  $b \in \mathcal{E}^{n+2}$ . Suppose now that some of the sequences of pop and push (or, in general, some of the subprograms) iterated into the main program leave unchanged the overall space used; since not increasing programs can be iterated without leading to any growth in space, the effective space bound is lower than the bound obtained via the  $\mu$ -measure, and it can be evaluated by an alternative measure  $\sigma$ . While  $\mu$  grows each time a top circle appears in the body of a loop,  $\sigma$  grows only for *increasing* top circles. In other words, the new measure doesn't consider those situations in which some (potentially harmful) operations are performed, but

their overall balance is negative. We prove a new bounding theorem using the  $\sigma$ -measure, providing a more appropriate bound to the complexity of stacks programs.

In Section II, we recall concepts and definitions of stack programs and of the Grzegorzcyk hierarchy. In Section III, we recall the definition of  $\mu$ -measure. In Section IV, we introduce the definition of the new  $\sigma$ -measure and the new bounding theorem. Conclusions and future work can be found in Section V.

## II. PRELIMINARIES ON THE GRZEGORZCYK HIERARCHY AND ON STACK PROGRAMS

In this section, we recall the definition of the Grzegorzcyk hierarchy, and fundamentals on stack programs and their computations; the reader is referred to [22] [5] [2] [10] for complete definitions and properties.

*Definition 2.1:* Given a unary function  $f$ , the  $k$ -th iterate of  $f$  (denoted with  $f^k$ ) is  $f^0(x) = x$  and  $f^{k+1}(x) = f(f^k(x))$ .

*Definition 2.2:* The principal functions  $E_1, E_2, E_3, \dots$  are  $E_1(x) = x^2 + 2$  and  $E_{n+2}(x) = E_{n+1}^x(2)$  (the  $x$ -th iterate of  $E_{n+1}$ ).

*Definition 2.3:* A function  $f$  is defined by bounded recursion from functions  $g, h$ , and  $b$  if for all  $\vec{x}, y$  one has  $f(\vec{x}, 0) = g(\vec{x})$ ,  $f(\vec{x}, y) = h(\vec{x}, y, f(\vec{x}))$  and  $f(\vec{x}, y) \leq g(\vec{x}, y)$ .

*Definition 2.4:* The  $n$ -th Grzegorzcyk class  $\mathcal{E}^n$  is the least class of functions containing the initial functions zero, successor, projections, maximum and  $E_{n-1}$ , and closed under composition and bounded recursion.

Stack programs operate on variables serving as stacks; they contain arbitrary words over a fixed alphabet  $\Sigma$ , and they are manipulated by running a program built from imperatives  $\text{push}(a, X)$ ,  $\text{pop}(X)$ , and  $\text{nil}(X)$  combined by sequencing, conditional, and loop statements (respectively,  $P; Q$ ,  $\text{if } \text{top}(X) \equiv a$  then  $[P]$ , and  $\text{foreach } X [P]$ ).

*Definition 2.5:* The operational semantics of stack programs are defined as follows:

- 1)  $\text{push}(a, X)$  pushes a letter  $a$  on the top of the stack  $X$ ;
- 2)  $\text{pop}(X)$  removes the top of  $X$ , if any; it leaves  $X$  unchanged, otherwise;
- 3)  $\text{nil}(X)$  empties the stack  $X$ ;
- 4) if  $\text{top}(X) \equiv a$   $[P]$  executes  $P$  if the top of the stack  $X$  is equal to  $a$ ;
- 5)  $P_1; \dots; P_k$  are executed from left to right;
- 6)  $\text{foreach } X [P]$  executes  $P$  for  $|X|$  times

with the restriction that no imperatives over  $X$  may occur in the body of a loop  $\text{foreach } X [P]$  (i.e., in  $P$ ), and that the loop is executed call-by-value;  $X$  is the *control stack* of the loop.  $|X|$  stands for the length of the word stored in  $X$ .

The notation  $\{A\}P\{B\}$  means that if the condition expressed by the sentence  $A$  holds before the execution of  $P$ , then the condition expressed by the sentence  $B$  holds after the execution of  $P$ .

*Definition 2.6:* A stack program  $P$  computes a function  $f : (\Sigma^*)^n \rightarrow \Sigma^*$  if  $P$  has an output variable  $O$  and  $n$  input variables  $\vec{X} = X_{i_1}, \dots, X_{i_n}$  among stacks  $X_1, \dots, X_m$  such that  $\{\vec{X} = \vec{w}\}P\{O = f(\vec{w})\}$ , for all  $\vec{w} = w_1, \dots, w_n \in (\Sigma^*)^n$ .

For a fixed program  $P$ , two sets of variables are defined:  $\mathcal{U}(P) = \{X \mid P \text{ contains an imperative } \text{push}(a, X)\}$  and  $\mathcal{C}(P) = \{X \mid P \text{ contains a loop } \text{foreach } X [Q], \text{ and } \mathcal{U}(Q) \neq \emptyset\}$ . Variables in  $\mathcal{U}(P)$  can be altered by a  $\text{push}$  during a run of  $P$ , while variables in  $\mathcal{C}(P)$  control a loop occurring in  $P$ . The two sets are not disjoint.

*Definition 2.7:*  $X$  controls  $Y$  in the program  $P$  (denoted with  $X \prec_P Y$ ) if  $P$  contains a loop  $\text{foreach } X [Q]$ , and  $Y \in \mathcal{U}(Q)$ ; the transitive closure of  $\prec_P$  is denoted by  $\xrightarrow{P}$ .

Consider the following program:

$P_1 := \text{foreach } X_1 [\dots \text{foreach } X_l [\text{push}(a, Y)]]$

If words  $v_1 \dots v_l, w$  are stored in  $X_1 \dots X_l, Y$ , respectively, before  $P_1$  is executed, then  $Y$  holds the word  $w a^{|v_1| \dots |v_l|}$  after the execution of  $P_1$ . The depth of loop-nesting is a necessary condition for high computational complexity, but it is not a sufficient condition. Now, consider the following two programs:

$P_2 := \text{nil}(Y); \text{push}(a, Y); \text{nil}(Z); \text{push}(a, Z);$   
 $\text{foreach } X [\text{nil}(Z); \text{foreach } Y [\text{push}(a, Z); \text{push}(a, Z)];$   
 $\text{nil}(Y); \text{foreach } Z [\text{push}(a, Y)]]$

$P_3 := \text{nil}(Y); \text{push}(a, Y); \text{nil}(Z);$   
 $\text{foreach } X [$   
 $\text{foreach } Y [\text{push}(a, Z); \text{push}(a, Z); \text{push}(a, Y)]]$

Even if both  $P_2$  and  $P_3$  have nesting depth 2, if  $w$  is initially stored in  $X$ , then  $Z$  holds the word  $a^{2^{|w|}}$  after  $P_2$  is executed, while  $a^{|w|(|w|+1)}$  is stored in  $Z$  after the execution of  $P_3$ . Thus, we see that  $P_3$  runs in polynomial time, whereas  $P_2$  has exponential running time. This happens because of the (control) circle contained inside the outermost loop in  $P_2$ : inside the loop governed by  $X$ , first  $Y$  controls  $Z$  (in that  $Z$  is updated via  $\text{push}(a, Z)$  inside a loop governed by  $Y$ ), and then  $Z$  controls  $Y$  in the same sense. In contrast, there is no such circle in  $P_3$ . Stack programs where each body of a loop statement is circle-free compute exactly the functions computable within polynomial time, and must be separated from those programs with loops that cause a blow up in running time.

## III. THE $\mu$ -MEASURE ON STACK PROGRAMS

Starting from the previous relation  $\xrightarrow{P}$ , a measure over the set of stack programs is introduced in [10].

*Definition 3.1:* Let  $P$  be a stack program. The  $\mu$ -measure of  $P$  (denoted with  $\mu(P)$ ) is defined as follows, inductively:

- 1)  $\mu(\text{pop}) = \mu(\text{push}) = \mu(\text{nil}) := 0$ ;
- 2)  $\mu(\text{if } \text{top}(X) \equiv a [Q]) := \mu(Q)$ ;

- 3)  $\mu(\mathbf{P}; \mathbf{Q}) := \max(\mu(\mathbf{P}), \mu(\mathbf{Q}))$ ;
- 4)  $\mu(\text{foreach } \mathbf{X} [\mathbf{Q}]) := \mu(\mathbf{Q}) + 1$ , if  $\mathbf{Q}$  is a sequence  $\mathbf{Q}_1; \dots; \mathbf{Q}_l$  with a *top circle* (that is, if there exists  $\mathbf{Q}_i$  such that  $\mu(\mathbf{Q}_i) = \mu(\mathbf{Q})$ , some  $\mathbf{Y}$  controls some  $\mathbf{Z}$  in  $\mathbf{Q}_i$ , and  $\mathbf{Z}$  controls  $\mathbf{Y}$  in  $\mathbf{Q}_1; \dots; \mathbf{Q}_{i-1}; \mathbf{Q}_{i+1}; \dots; \mathbf{Q}_l$ );  $\mu(\text{foreach } \mathbf{X} [\mathbf{Q}]) := \mu(\mathbf{Q})$ , otherwise.

To focus on the critical case where  $\mathbf{P}$  is a loop `foreach`  $\mathbf{X} [\mathbf{Q}]$ , assume that  $\mu(\mathbf{Q})$  is already determined. Suppose that  $\mathbf{Q}$  is a sequence  $\mathbf{Q}_1; \dots; \mathbf{Q}_l$ , in which case  $\mu(\mathbf{Q})$  is  $\max(\mu(\mathbf{Q}_1), \dots, \mu(\mathbf{Q}_l))$ . Then a blow up in running time can only occur if  $\mathbf{Q}$  has a top circle, that is,  $\mathbf{Q}$  has a circle with respect to a control variable  $\mathbf{Y}$  of some component  $\mathbf{Q}_i$  of maximal  $\mu$ -measure  $\mu(\mathbf{Q})$ . In this case,  $\mu(\mathbf{P})$  is defined as  $\mu(\mathbf{Q})+1$ . In all other cases,  $\mu(\mathbf{P})$  is defined as  $\mu(\mathbf{Q})$ . Given that all primitive instructions receive  $\mu$ -measure 0, one easily verifies for the examples above that  $\mu(\mathbf{P}_1)=\mu(\mathbf{P}_3)=0$ , whereas  $\mu(\mathbf{P}_2)=1$ .

The core of [10] is the following bounding theorem.

*Lemma 3.1:* Every function  $f$  computed by a stack program of  $\mu$ -measure  $n$  has length bound  $b \in \mathcal{E}^{n+2}$  satisfying  $|f(\vec{w})| \leq b(|\vec{w}|)$ , for all  $\vec{w}$ . In particular, if  $\mathbf{P}$  computes a function  $f$ , and  $\mu(\mathbf{P}) = 0$ , then  $f$  has a polynomial length bound, that is, there exists a polynomial  $p$  satisfying  $|f(\vec{w})| \leq p(|\vec{w}|)$ .

Let  $\mathcal{L}_\mu^n$  be the class of all functions which can be computed by a stack program of  $\mu$ -measure  $n \geq 0$ , and let  $\mathcal{G}^n$  be the class of all functions which can be computed by a Turing machine in time  $b(|\vec{w}|)$ , for some  $b \in \mathcal{E}^n$ . As a consequence of the bounding lemma, the following result holds.

*Theorem 3.1:* For  $n \geq 0$ :  $\mathcal{L}_\mu^n = \mathcal{G}^{n+2}$ .

#### IV. THE $\sigma$ -MEASURE AND A NEW BOUNDING THEOREM

In the rest of the paper, we denote with `imp`( $\mathbf{Y}$ ) an imperative `pop`( $\mathbf{Y}$ ), `push`( $\mathbf{a}, \mathbf{Y}$ ), or `nil`( $\mathbf{Y}$ ); we denote with `mod`( $\bar{\mathbf{X}}$ ) a *modifier*, that is a sequence of imperatives operating on the variables occurring in  $\bar{\mathbf{X}} = \mathbf{X}_1, \dots, \mathbf{X}_n$ . We introduce a modified definition of *circle*, which better matches our new measure.

*Definition 4.1:* Let  $\mathbf{Q}$  be a sequence in the form  $\mathbf{Q}_1; \dots; \mathbf{Q}_l$ . There is a *circle* in  $\mathbf{Q}$  if there exists a sequence of variables  $\mathbf{Z}_1, \mathbf{Z}_2, \dots, \mathbf{Z}_l$ , and a permutation  $\pi$  of  $\{1, \dots, l\}$  such that  $\mathbf{Z}_1 \xrightarrow{\alpha_{\pi(1)}} \mathbf{Z}_2 \xrightarrow{\alpha_{\pi(2)}} \dots \mathbf{Z}_l \xrightarrow{\alpha_{\pi(l)}} \mathbf{Z}_1$ . The subprograms  $\mathbf{Q}_1, \dots, \mathbf{Q}_l$  and the variables  $\mathbf{Z}_1, \dots, \mathbf{Z}_l$  are *involved* in the circle.

For sake of simplicity, we will consider  $\pi(i) = i$ , that is the case  $\mathbf{Z}_1 \xrightarrow{\alpha_1} \mathbf{Z}_2 \xrightarrow{\alpha_2} \dots \mathbf{Z}_l \xrightarrow{\alpha_l} \mathbf{Z}_1$ ; proofs and definitions holds in the general case too.

*Definition 4.2:* Let  $\mathbf{P}$  be a stack program and let  $\mathbf{Y}$  be a given variable. The  $\sigma$ -measure of  $\mathbf{P}$  with respect to  $\mathbf{Y}$  (denoted with  $\sigma_\mathbf{Y}(\mathbf{P})$ ) is defined as follows, inductively (with  $sg(z) = 1$  if  $z \geq 1$ ,  $sg(z) = 0$  otherwise):

- 1)  $\sigma_\mathbf{Y}(\text{mod}(\bar{\mathbf{X}})) := sg(\sum \hat{\sigma}_\mathbf{Y}(\text{imp}(\mathbf{Y})))$ , for each  $\text{imp}(\mathbf{Y}) \in \text{mod}(\bar{\mathbf{X}})$ , where  $\hat{\sigma}_\mathbf{Y}(\text{push}(\mathbf{a}, \mathbf{Y})) := 1$ ;

$$\hat{\sigma}_\mathbf{Y}(\text{pop}(\mathbf{Y})) := -1;$$

$$\hat{\sigma}_\mathbf{Y}(\text{nil}(\mathbf{Y})) := -\infty;$$

$$\hat{\sigma}_\mathbf{Y}(\text{imp}(\mathbf{X})) := 0, \text{ with } \mathbf{Y} \neq \mathbf{X};$$

$$2) \sigma_\mathbf{Y}(\text{if top } \mathbf{Z} \equiv \mathbf{a}[\mathbf{P}]) := \sigma_\mathbf{Y}(\mathbf{P});$$

$$3) \sigma_\mathbf{Y}(\mathbf{P}_1; \mathbf{P}_2) := \max(\sigma_\mathbf{Y}(\mathbf{P}_1), \sigma_\mathbf{Y}(\mathbf{P}_2)), \text{ with } \mathbf{P}_1; \mathbf{P}_2 \text{ not a modifier};$$

$$4) \sigma_\mathbf{Y}(\text{foreach } \mathbf{X} [\mathbf{Q}]) := \sigma_\mathbf{Y}(\mathbf{Q}) + 1, \text{ if there exists a circle in } \mathbf{Q}, \text{ and a subprogram } \mathbf{Q}_i \text{ s.t.}$$

$$(a) \mathbf{Y} \text{ and } \mathbf{Q}_i \text{ are involved in the circle};$$

$$(b) \sigma_\mathbf{Y}(\mathbf{Q}) = \sigma_\mathbf{Y}(\mathbf{Q}_i);$$

$$(c) \text{ the circle is increasing};$$

$$\sigma_\mathbf{Y}(\text{foreach } \mathbf{X} [\mathbf{Q}]) := \sigma_\mathbf{Y}(\mathbf{Q}), \text{ otherwise,}$$

where a circle is *not increasing* if, denoted with  $\mathbf{Q}_1, \mathbf{Q}_2, \dots, \mathbf{Q}_l$  and with  $\mathbf{Z}_1, \mathbf{Z}_2, \dots, \mathbf{Z}_l$  the sequences of subprograms and, respectively, of variables involved in the circle, we have that  $\sigma_{\mathbf{Z}_i}(\mathbf{Q}_j) = 0$ , for each  $i := 1 \dots l$  and  $j := 1 \dots l$ . If the previous condition doesn't hold, we say that the circle is *increasing*.

Note that the  $\sigma_\mathbf{Y}$ -measure of a modifier (see (1) in the previous definition) is equal to 1 only when, in absence of `nil`'s, the overall number of `push`'s over  $\mathbf{Y}$  is greater than the number of `pop`'s over the same variable, that is, only when a growth in the length of  $\mathbf{Y}$  is produced. Moreover, note that the "otherwise" case in (4) can be split in three different cases. First, there are no circles in which  $\mathbf{Y}$  is involved. Second,  $\mathbf{Y}$  is involved, together with a subprogram  $\mathbf{Q}_i$ , in a circle in  $\mathbf{Q}$ , but it happens that  $\sigma_\mathbf{Y}(\mathbf{Q}_i)$  is lower than  $\sigma_\mathbf{Y}(\mathbf{Q})$  (this means that there is a blow-up in the complexity of  $\mathbf{Y}$  in  $\sigma_\mathbf{Y}(\mathbf{Q}_i)$ , but this growth is still bounded by the complexity of  $\mathbf{Y}$  in a different subprogram of  $\mathbf{Q}$ ). Third,  $\mathbf{Y}$  is involved in some circles in  $\mathbf{Q}$ , but each of them is not increasing (that is, according to the previous definition, each variable  $\mathbf{Z}_i$  involved in each circle doesn't produce a growth in the complexity of the subprograms  $\mathbf{Q}_j$  involved in the same circle). This implies that the space used during the execution of the external loop `foreach`  $\mathbf{X} [\mathbf{Q}]$  is basically the same used by  $\mathbf{Q}$  (this is not a surprising fact: one can freely iterate a not increasing program without leading an harmful growth). In all three cases the  $\sigma$ -measure must remain unchanged: it is increased only when an increasing top circle occurs and when at least one of the variables involved in that circle causes a growth in the space complexity of the related subprogram, simultaneously (that is, if there exists a  $p$  such that  $\sigma_{\mathbf{Z}_p}(\mathbf{Q}_p) > 0$ ).

In the following definition, we extend the measure to the whole set of variables occurring in a stack program.

*Definition 4.3:* Let  $\mathbf{P}$  be a stack program. The  $\sigma$ -measure of  $\mathbf{P}$  is  $\sigma(\mathbf{P}) := \tilde{\sigma}(\mathbf{P}) - 1$ , where  $\tilde{\sigma}$  is the usual cut-off subtraction, and

$$1) \tilde{\sigma}(\text{mod}(\bar{\mathbf{X}})) := 0$$

$$2) \tilde{\sigma}(\text{if top } \mathbf{Z} \equiv \mathbf{a} [\mathbf{Q}]) := \max(\sigma_\mathbf{Y}(\text{if top } \mathbf{Z} \equiv \mathbf{a} [\mathbf{Q}])), \text{ for all } \mathbf{Y} \text{ occurring in } \mathbf{P};$$

$$3) \tilde{\sigma}(\mathbf{P}_1; \mathbf{P}_2) := \max(\sigma_\mathbf{Y}(\mathbf{P}_1; \mathbf{P}_2)), \text{ for all } \mathbf{Y} \text{ occurring in } \mathbf{P}, \text{ with } \mathbf{P}_1; \mathbf{P}_2 \text{ not a modifier};$$

- 4)  $\tilde{\sigma}(\text{foreach } X [Q]) := \max(\sigma_V(\text{foreach } X [Q]))$ , for all  $Y$  occurring in  $P$ .

Note that  $\sigma(P) \leq \mu(P)$ , for each stack program  $P$ . Note also that we are using the previously defined  $\tilde{\sigma}_V$  to detect all the *increasing* modifiers, for a given variable  $Y$  (this is done setting  $\tilde{\sigma}_V$  equal to 1); but, once detected, we don't have to consider them in the evaluation of the  $\sigma$ -measure. This is the reason of the "-1" part in the previous definition.

In the rest of the paper we introduce a reduction procedure between stack programs, denoted with  $\rightsquigarrow$ , and we prove a new bounding theorem.

*Definition 4.4:*  $P$  and  $Q$  are *space equivalent* if  $\{\bar{X} = \bar{w}\}P\{|\bar{X}| = m\}$  implies that  $\{\bar{X} = \bar{w}\}Q\{|\bar{X}| = O(m)\}$ . This is denoted with  $P \approx_s Q$ .

*Definition 4.5:* Let  $A$  be a stack program such that  $\mu(A) = n + 1$ , and  $\sigma(A) = m$ , with  $m < n + 1$ ; the program  $\rightsquigarrow A$  is obtained as follows:

- 1) if  $A$  is `foreach X [R]`, with  $\mu(R) = \sigma(R) = n$ , and denoted with  $C_1, \dots, C_l$  the top circles in  $R$ , and with  $A_{i1}, \dots, A_{ip}$  the variables involved in  $C_i$ , for each  $i$ , we have that  $\rightsquigarrow A$  is the result of changing each `imp(Aij)` into `nop(Aij)` (a *no-operation* imperative);
- 2) if  $A$  is `foreach X [R]`, with  $\mu(R) > \sigma(R)$ , we have that  $\rightsquigarrow A$  is equal to `foreach X [rightsquigarrow R]`;
- 3) if  $A$  is  $A_1; A_2$  and  $\max(\mu(A_1), \mu(A_2)) = \mu(A_1)$ , we have that  $\rightsquigarrow A$  is equal to  $\rightsquigarrow A_1; A_2$ ; symmetrically, if  $\max(\mu(A_1), \mu(A_2)) = \mu(A_2)$ , we have that  $\rightsquigarrow A$  is equal to  $A_1; \rightsquigarrow A_2$ ; if  $\mu(A_1) = \mu(A_2)$ , we have that  $\rightsquigarrow A$  is equal to  $\rightsquigarrow A_1; \rightsquigarrow A_2$ ;
- 4) if  $A$  is `if top(X)≡a [R]`, we have that  $\rightsquigarrow A$  is equal to `if top(X)≡a [rightsquigarrow R]`.

*Lemma 4.1:* Given a stack program  $P$ , with  $\mu(P) = n + 1$  and  $\sigma(P) = n$ , there exists a stack program  $\rightsquigarrow P$  such that  $\mu(\rightsquigarrow P) = n$ ,  $\sigma(\rightsquigarrow P) = n$ , and  $P \approx_s \rightsquigarrow P$ .

*Proof.* (by induction on  $n$ ). Base. Let  $\mu(P) = 1$  and  $\sigma(P) = 0$ . In the main case,  $P$  is in the form `foreach X [Q]`, with a not-increasing circle occurring in  $Q$ . Applying  $\rightsquigarrow$  to  $P$ , we obtain a program  $P'$  whose  $\sigma$ -measure is still 0, and whose  $\mu$ -measure is reduced to 0, because  $\rightsquigarrow$  has broken off the circle in  $P$  that leads  $\mu$  from 0 to 1 (i.e., in  $P'$ , there are no more `push`'s on the variables involved in the circle). Note that  $P$  can decrease the length of the stacks involved in the circle, while  $P'$  does not perform any operation in the same circle. Thus,  $P'$  can increase its variables only by a linear factor; indeed, if  $\{\bar{X} = \bar{w}\}P\{|\bar{X}| = m\}$  we have that  $\{\bar{X} = \bar{w}\}P'\{|\bar{X}| = cm\}$ , where  $c$  is a constant depending on the structure of  $P$ : thus,  $P \approx_s P'$ .

Step. Let  $\mu(P) = n + 2$  and  $\sigma(P) = n + 1$ . Let  $P$  be in the form `foreach X [Q]`, and let  $C$  be a top circle occurring in  $Q$ , with  $\mu(Q) = n + 1$ ; we have two cases: (1)  $\sigma(Q) = n + 1$ , or (2)  $\sigma(Q) = n$ .

(1) In this case  $C$  is a not-increasing circle, because it has been detected by  $\mu$ , but not by  $\sigma$ . Applying  $\rightsquigarrow$  to  $P$ , we

obtain a program  $P'$  such that  $\sigma(P') = n + 1$ ,  $\mu(P') = n + 1$ , and  $P \approx_s P'$ .

(2) In this case  $C$  is an increasing circle, detected by  $\mu$  and  $\sigma$ . We have that (by the inductive hypothesis) there exists a program  $Q'$  such that  $\mu(Q') = n$ ,  $\sigma(Q') = n$ , and  $Q \approx_s Q'$ . Starting from  $P$ , we build a new program  $P' = \text{foreach } X [Q']$ . We have that  $\mu(P') = \mu(Q') + 1 = n + 1$ ,  $\sigma(P') = \sigma(Q') + 1 = n + 1$ , and  $P \approx_s P'$  as expected.

The cases  $P_1; P_2; \dots; P_k$  and `if top(X)≡a [P]` can be proved in a similar way.

*Theorem 4.1:* Every function  $f$  computed by a stack program  $P$  such that  $\mu(P) = n$  and  $\sigma(P) = m$ , with  $n > m$ , has a length bound  $b \in \mathcal{E}^{m+2}$  satisfying  $|f(\bar{w})| \leq b(|\bar{w}|)$ .

*Proof.* Let  $k$  be  $\mu(P) - \sigma(P)$ . Then by  $k$  applications of Lemma 4.1, we obtain a sequence  $P =: P_0, P_1, \dots, P_k$  of stack programs such that, for all  $i < k$ ,

$$\mu(P_{i+1}) = \mu(P) - i, \sigma(P_i) = \sigma(P_{i+1}), \text{ and } P_i \approx_s P_{i+1}.$$

By Kristiansen and Niggel's bounding theorem,  $P_k$  has a length bound in  $\mathcal{E}^{\sigma(P)+2}$ , and so does  $P$ , by transitivity of  $\approx_s$ .

Let  $\mathcal{L}_\sigma^n$  be the class of all functions that can be computed by a stack program of  $\sigma$ -measure  $n \geq 0$ , and let  $\mathcal{G}^n$  be the class of all functions which can be computed by a Turing machine in time  $b(|\bar{w}|)$ , for some  $b \in \mathcal{E}^n$ . As a consequence of Theorem 4.1, and similarly to what has been recalled in Section III, the following result holds.

*Theorem 4.2:* For  $n \geq 0$ :  $\mathcal{L}_\sigma^n = \mathcal{G}^{n+2}$ .

## V. CONCLUSIONS AND FUTURE WORK

We have defined a syntactical measure  $\sigma$  that considers how the iteration of imperative stack programs affects the complexity of the programs themselves. In particular, this measure only counts those loops in which programs with a size-increasing effect (w.r.t. the final length of the result) are iterated. Each time such a loop is built over other loops, the  $\sigma$ -measure is increased by 1. Other measures detect potentially harmful loops, but are not able to distinguish between size-increasing loops and the non-size-increasing one's. It is undecidable to know if a function computed by a given stack program lies in a given complexity class, but our measure represents an improvement when compared to previously defined measures. We can assign a function computed by a stack program of  $\sigma$ -measure  $n$  to the  $n+2$ -th Grzegorzczuk class, and this class is lower in the hierarchy, when compared to the class obtained via the  $\mu$ -measure.

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