

A THE DETAILS OF PROOFS

A.1 Proof of Proposition 1

Proof. According to Proposition 18 in [25], if Σ is a finite alphabet of n alphabet symbols, the number of pairwise non-equivalent SOREs over Σ is $s(n)$ with $n!2^{3n-r \log n} \leq s(n) \leq n!2^{7n}$, where r is a constant. This implies there is a finite number of non-equivalent SOREs.

For k -OREs, every symbol in Σ occurs at most k times, we treat the same symbol in a k -ORE as distinct. Then, let $\Sigma(k)$ for k -OREs be a finite alphabet of nk alphabet symbols, the number of pairwise non-equivalent k -OREs over $\Sigma(k)$ is $s(nk)$ with $(nk)!2^{3nk-r \log(nk)} \leq s(nk) \leq (nk)!2^{7nk}$. This implies there is a finite number of non-equivalent k -OREs over $\Sigma(k)$. According to the definition 1, k -REs ($k \geq n$) is a subclass of k -OREs. the number of non-equivalent k -REs over $\Sigma(k)$ is also finite.

Let \mathcal{D} denote the class of k -REs. Assume that there is a language $L \subseteq \Sigma_k^*$ such that no expression $\alpha \in \mathcal{D}$ is a descriptive generalization of L (w.r.t. the class \mathcal{D}). If an expression $\alpha_1 \in \mathcal{D} : \mathcal{L}(\alpha_1) \supseteq L$, then there is an expression $\alpha_2 \in \mathcal{D} : \mathcal{L}(\alpha_1) \supset \mathcal{L}(\alpha_2) \supseteq L$. There are infinite expressions $\alpha_1, \alpha_2, \dots, \alpha_i, \dots \in \mathcal{D}$ such that $\mathcal{L}(\alpha_1) \supset \mathcal{L}(\alpha_2) \supset \dots \supset \mathcal{L}(\alpha_i) \supset \dots \supseteq L$. This contradicts the fact that there is only a finite number of non-equivalent k -REs over $\Sigma(k)$. Hence, for every language L , there exists a k -RE r that is a descriptive generalization of L (w.r.t. the class of k -REs).

A.2 Proof of Theorem 1

Proof. In algorithm 1, a GA \mathcal{A} is obtained by deleting nodes and the corresponding edges in the given directed graph G . The rule R_1 working on G corresponds the operations in lines 13~14. The rules R_2 and R_3 work on G in line 21 and line 11, respectively. This implies that the finally obtained GA \mathcal{A} is reducible. SCCs are searched in G in a recursive way in line 4, and the characters about stable and transverse are added into \mathcal{A} for each SCC in line 7. Actually, the algorithm *Trans2GA* is a variant of algorithm *Soa2Sore* [25]. Then, according to Theorem 27 in [25], for any given directed graph G , G can be correctly transformed into an expression. In this paper, each unit able to be transformed into an expression has been processed by *Trans*(G, \mathcal{A}) until $|G.V| = 2$. The finally obtained GA \mathcal{A} has the glushkov characters: HM, SS, ST and RD. According to Theorem 5.1 in [19], a directed graph is a glushkov graph if and only if the directed graph satisfies characteristics of glushkov graph (glushkov characters) [19]. Hence, \mathcal{A} is a glushkov graph.

A.3 Proof of Theorem 2

Proof. For any given finite sample S and the directed graph G obtained by using *PSO* algorithm, there does not exist another directed graph G' such that $\mathcal{L}(G) \supset \mathcal{L}(G') \supseteq S$; otherwise, G is not the optimal result of *PSO* algorithm.

The directed graph G obtained by using *PSO* algorithm is then transformed into a GA \mathcal{A} . Let f denote the function used to transform G into \mathcal{A} (i.e., $f(G) = \mathcal{A}$). \mathcal{A} must satisfy glushkov characters and any character is unambiguous. f is unique, there does not exist another function f' such that $f'(G) = \mathcal{A}$. It implies that there does not exist another GA \mathcal{A}' such that G can be transformed into \mathcal{A}' .

For any given finite sample S , $\mathcal{L}(\mathcal{A}) \supseteq \mathcal{L}(G) \supseteq S$. There does not exist another GA \mathcal{A}' such that $\mathcal{L}(\mathcal{A}) \supset \mathcal{L}(\mathcal{A}') \supseteq S$. The constructed GA \mathcal{A} is a descriptive generalization of S .

A.4 Proof of Theorem 3

Proof. According to Theorem 2, for any given finite sample S , the obtained GA \mathcal{A} is a descriptive generalization of S . \mathcal{A} is a precise representation for S . In algorithm *LearnRE*, the algorithm *Soa2Sore* is used to transform GA \mathcal{A} into an expression. For any given finite sample S' , since *Soa2Sore* can transform a directed graph G which is a descriptive generalization of S' into an expression that is also a descriptive generalization of S' (Theorem 27 in [25]), *Soa2Sore* can also transform GA \mathcal{A} into a k -RE that is also a descriptive generalization of S (w.r.t. the class of k -REs). Hence, the k -RE r_k returned by *LearnRE* is a descriptive generalization of S (w.r.t. the class of k -REs).

A.5 Proof of Corollary 1

Proof. For any k -RE r_k , we can construct an equivalent GA A_k , i.e., $\mathcal{L}(A_k) = \mathcal{L}(r_k)$. The GA A_k is also a directed graph, we can obtain a finite sample S by traversing A_k . For a string $s \in S$, s is a path from A_k . Then, there exist a finite sample S that can be obtained from A_k such that A_k is a descriptive generalization of S . When S is as input of algorithm *LearnRE*, the obtained GA \mathcal{A} is also a descriptive generalization of S according to Theorem 2. There is $\mathcal{L}(\mathcal{A}) = \mathcal{L}(A_k) = \mathcal{L}(r_k)$. \mathcal{A} equivalent to r_k is then transformed into an expression r in *LearnRE*, r is also a descriptive generalization of S , it implies that $\mathcal{L}(r) = \mathcal{L}(\mathcal{A}) \supseteq S$. Hence, there exists a expression r can be generated by *LearnRE* such that $\mathcal{L}(r_k) = \mathcal{L}(r)$. r_k can be successfully derived by *LearnRE*.