

# It's a Small Inverse

Kazuhiro Inaba (kinaba@nii.ac.jp)

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## 1 Existence of Linear-Size Inverse

We assume the reader to be very familiar with the theory of tree transducers [Eng77, Bak79, EV85]. The goal of this note is the following theorem.

**Theorem 1.1.** *Let  $f \in D_t\text{MTT}^*$ , i.e.,  $f$  is a tree-to-tree function that can be expressed by a finite composition of total deterministic macro tree transducers. Then,*

$$\exists c \in \mathbb{N}. \forall t \in \text{range}(f). \exists s \in \text{dom}(f). \left( f(s) = t \wedge |s| \leq c|t| \right)$$

where  $|s|$  and  $|t|$  denotes the number of nodes in  $s$  and  $t$ , respectively.

We denote by  $\exists\text{LBI}$  (existentially linearly-bounded-input) the class of translations satisfying the property in the theorem. Hence, the theorem statement can be written as follows:  $D_t\text{MTT}^* \subseteq \exists\text{LBI}$ .

Let us take a look at an example. Consider the following translation  $f$

$$\begin{aligned} f(\mathbf{a}(x_1)) &= f(x_1) \\ f(\mathbf{b}(x_1, x_2)) &= \mathbf{c}(f(x_1)) \\ f(\mathbf{e}) &= \mathbf{d} \end{aligned}$$

and the tree  $t = \mathbf{c}(\mathbf{d})$ . What is the input tree  $s$  that makes  $f$  to output  $t$ ? There are many possibilities, because  $f$  is not injective. Some of them are large (e.g.,  $\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{b}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{a}(\mathbf{e}))))))))))))))$ ), and some are not. In fact, the smallest one  $\mathbf{b}(\mathbf{e}, \mathbf{e})$  is *not so large* compared to the given tree  $t$ . More precisely speaking, the size is  $\leq 2|t|$ . The theorem tells us that this is a universal phenomenon shared among all MTT-definable translations.

### 1.1 The Proof

The problem is broken down to simpler cases by using the following lemma.

**Lemma 1.2** ([Man02], Theorem 12).  $D_t\text{MTT}^* \subseteq D_t\text{LT}^{\text{R}}; \text{LBI}$

$X; Y$  denotes sequential composition:  $X$  followed by  $Y$ .  $D_t\text{LT}^{\text{R}}$  is the class of total deterministic linear top-down tree transducers with regular lookahead, whose property is discussed soon.  $\text{LBI}$  is the class of translations that satisfies:  $\exists c \in \mathbb{N}. \forall s \in \text{dom}(f). |s| \leq c|f(s)|$ . The class  $\text{LBI}$  is a strict subclass of  $\exists\text{LBI}$ .

Let us assume here that  $D_t\text{LT}^{\text{R}} \subseteq \exists\text{LBI}$ , then from Lemma 1.2 we can derive the main theorem, because we can also show the following fact.

**Lemma 1.3.**  $\exists\text{LBI}; \text{LBI} \subseteq \exists\text{LBI}^1$ .

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<sup>1</sup> For those who might wonder: it also holds that  $\text{LBI}; \text{LBI} \subseteq \text{LBI}$ . But,  $\exists\text{LBI}; \exists\text{LBI} \not\subseteq \exists\text{LBI}$  and  $\text{LBI}; \exists\text{LBI} \not\subseteq \exists\text{LBI}$ .

*Proof.* Let  $g_1 \in \exists\text{LBI}$ ,  $g_2 \in \text{LBI}$ , and  $t \in \text{range}(g_1; g_2)$ . Then take arbitrary  $s_0, s_1$  satisfying  $s_1 = g_1(s_0)$  and  $t = g_2(s_1)$  (since  $t \in \text{range}(g_1; g_2)$ , there exists at least one such pair). From the assumption  $g_1 \in \exists\text{LBI}$ , we can make  $s_0$  small. That is, by  $s_1 \in \text{range}(g_1)$ , there exists  $s'_0$  s.t.  $|s'_0| \leq c_{g_1}|s_1|$ . By the LBI-property,  $|s_1| \leq c_{g_2}|t|$ . Hence, we can choose a small input  $s'_0$  such that  $|s'_0| \leq c_{g_1}c_{g_2}|t|$ , as desired.  $\square$

Now, as you expect, the assumption on  $\text{D}_t\text{LT}^{\text{R}}$  does hold.

**Lemma 1.4.**  $\text{D}_t\text{LT}^{\text{R}} \subseteq \exists\text{LBI}$ .

*Proof.* By Theorem 2.6 of [Eng77],  $\text{D}_t\text{LT}^{\text{R}}$  can be represented as a deterministic finite-state bottom-up relabeling followed by  $\text{D}_t\text{LT}$ . Let  $g$  be the relabeling and  $f \in \text{D}_t\text{LT}$ , and  $t$  a tree in the range of  $g; f$ . Let  $s$  be (one of) the minimum input tree such that  $(g; f)(s) = t$ . We will show  $|s|$  is bounded by  $c|t|$  where  $c$  is a constant determined by  $g$  and  $f$  and independent from  $t$  or  $s$ .

For each node  $v$  of  $s$ , we denote by  $f_{\text{Q}}(v)$  the state of  $f$  applied to  $v$  during the computation of  $f(s)$  (since  $f$  is linear transducer, the state, if any, is uniquely determined; if  $f$  never visited  $v$ , let  $f_{\text{Q}}(v) = \perp$ ). Note that if  $f_{\text{Q}}(v) = \perp$  then for all nodes  $v'$  in the subtree rooted at  $v$ , we have  $f_{\text{Q}}(v') = \perp$ . We can show  $|\{v \mid f_{\text{Q}}(v) = \perp\}| \leq r|g|^r|\{v \mid f_{\text{Q}}(v) \neq \perp\}|$  where  $r$  is the maximum rank of the label alphabet, and  $|g|$  is the number of states of the relabeling  $g$ . In other words, unvisited parts are smaller than visited parts (ignoring the constant factor).

The inequation is derived as follows. Let  $v_1, \dots, v_u$  be the set of nodes that they are unvisited ( $f_{\text{Q}}(v_i) = \perp$ ) but all their ancestors are visited. It should be clear that  $u \leq r|\{v \mid f_{\text{Q}}(v) \neq \perp\}|$ ; there can be at most  $|\{v \mid f_{\text{Q}}(v) \neq \perp\}|$  leaves in the visited fragment of  $s$ , and each of them can only have at most  $r$  unvisited children. Furthermore, the number of nodes of each subtree rooted at  $v_i$  is bounded by  $|g|^r$ . Since the subtree of  $v_i$  is unvisited, we can freely substitute the subtree to another one without changing the output  $t$ , as long as the bottom-up relabeling  $g$  reaches the same state at  $v_i$ . Here, for any  $|g|$ -state tree automaton of rank  $r$ , its minimal instance tree is at most  $|g|^r$  (by the pumping lemma of regular languages, the height of minimal instance is bounded by  $|g|$ ). Hence we have the desired inequation.

We now know that the  $\frac{1}{1+r|g|^r}$  fraction of the input tree  $s$  is visited by the translation  $f$ . Then we have  $\frac{|s|}{1+r|g|^r} \leq |t|$ , and the proof is done, isn't it? Unfortunately, no. Even if  $\text{D}_t\text{LT}$  visits an input node, it does not mean that an output node is generated there. In the only one exceptional case, the transducer may *skip* the node without generating any new node. This happens when a rule of the form  $q(\sigma(x_1, \dots, x_m)) = q'(x_i)$  (the form whose rhs is a single state-application) is used at the node. We have to deal with this case.

Let  $V$  be the set of *skipped* nodes of  $s$ , i.e., the nodes  $v$  such that the right-hand side of the rule of  $f$  of for the state-label pair  $f_{\text{Q}}(v)$ ,  $\text{label}(v)$  is a single state-application. Let  $v_1, v_2, \dots, v_u \in V$  be a list of nodes such that  $v_i$  is the parent node of  $v_{i+1}$  for each  $i$ . Then,  $u$  must be less than or equal to  $|f||g|$ . This is again due to the pumping lemma. If the chain is longer than  $|f||g|$ , for some  $i < j$  it would be  $f_{\text{Q}}(v_i) = f_{\text{Q}}(v_j)$  and  $g_{\text{Q}}(v_i) = g_{\text{Q}}(v_j)$  (where  $g_{\text{Q}}$  is the unique state used during the run of  $g$  on  $s$ ), which can be shortened and hence contradicts the minimality of  $s$ .

This upper bound of the length implies that at least 1 out of  $1 + |f||g|$  visited nodes are not skipped and generate some output node. In a summary, we have  $|s| \leq (1 + r|g|^r)(1 + |f||g|)|t|$ .  $\square$

## 1.2 Nondeterministic Version

I believe the same property holds for nondeterministic MTTs. Analogue of Lemma 1.2 also holds for the nondeterministic case (Theorem 5.10 of [Ina09]). Hence, all I have to do is to show  $\text{LT}^{\text{R}} \subseteq \exists\text{LBI}$ . Its proof should also be similar; after we fix one particular run of the transducer that converts  $s$  into  $t$ , the same argument should hold.

## 2 Discussion

Are there any easier proof for the theorem? The theorem may be trivial, just I'm not noticing it...

Lemma 1.4 may have something to do with Theorem 5.2 of [AU71], which (if I understand correctly) says that the growth rate of a single top-down tree transducer must be in the form  $x^c$  or  $c^x$  for some integer  $c$ . In other words, if the growth rate is less than linear, it is constant (no  $\log n$  or  $\sqrt{n}$  translation).

Finally, can we derive something fruitful from the theorem, like linear-time upper bound for some useful problem? Currently I have no idea... :p

## References

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