

# Reductions from Tree Reconstruction to String Reconstruction

Thomas Jacob Maranzatto

University of Illinois Chicago (Ph.D.)  
University of Maryland College Park

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# Outline

- Some Background/ Definitions
- Combinatorics of the Deletion Channel
- Tree Reconstruction Reductions

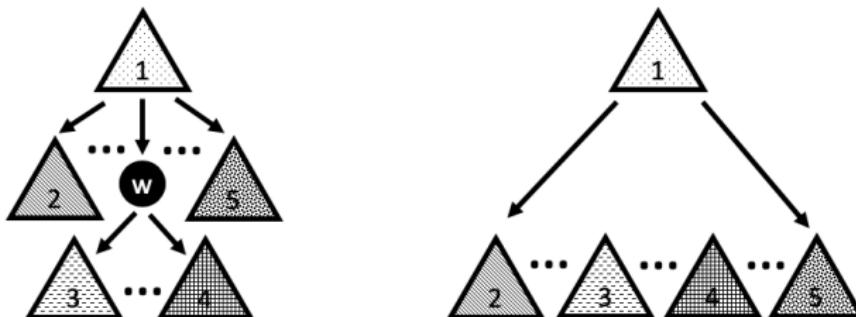
# Deletion Channel Models

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- For any tree  $R$ , in the *tree edit distance* (TED) model when a node  $w$  is removed, the children of  $w$  become the children of  $w$ 's parent. A trace is obtained by removing each node with probability  $q$  (the order of removal does not matter).

# TED Illustration



**Figure:** The generic picture before (left) and after (right) node  $w$  is removed from tree  $R$  in the TED model. The subtrees 3 and 4 are inserted as children of 1 when  $w$  is removed.

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- We relate the sample complexity of the tree reconstruction problem to  $T(n, q, \delta)$

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- Davies, Racz, and Rashtchian (2020) introduced the TED model, and proved upper bounds for spider graphs and complete k-ary trees
- The spider upper bounds generalized some techniques from De et al. (2017), Nazarov and Peres (2017).

# Functions Applied to Traces

## Definition

If  $s \in \{\pm 1\}^n$ , then  $\mathcal{D}_s^k$  is the distribution of traces generated from the deletion channel applied to  $s$  when conditioned to have length  $k$ .

## Theorem

Let  $s \in \{\pm 1\}^n$  be any string. Then for every  $0 \leq k \leq n$  and for any function  $h : \{\pm 1\}^k \rightarrow \mathbb{R}$ ,

$$\mathbb{E}_{\mathcal{D}_s^k}[h] = \frac{1}{\binom{n}{k}} \sum_{I \in \{0,1\}^k} \left( \binom{n-k}{k - \|I\|} \cdot \mathbb{E}_{\mathcal{D}_{s^{k-\|I\|}}^{k-\|I\|}} [h_I^s] \right)$$

# Proof Idea

- Write the expected value in terms of string densities
- Split the sum into parts corresponding to fixed bit values
- Recursively fix bit values until  $h$  is saturated

# Combinatorics of Infinite Deletion Channel

## Theorem

Let  $s \in \{0, 1\}^\infty$ , and suppose the 1's occur at indices  $\mathcal{I} \subset \mathbb{N}$ . Consider the generating function for  $s$ ,  $f(s; x) := \sum_{i \in \mathcal{I}} x^i$ . Then under the deletion channel with rate  $q$ ,

$$\mathbb{P}[j\text{'th bit of the trace is a } 1] = \frac{1}{j!} p^{j+1} \left. \frac{\partial^j f(s; \cdot)}{\partial x^j} \right|_{(1-p)}$$

# Proof

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Observe  $\frac{\partial^j f(t; x)}{\partial x^j} = \sum_{i \in \mathcal{I}} \frac{(i)!}{(i-j)!} x^{i-j}$ .

Also,  $\mathbb{P}[j\text{'th bit of the trace is a } 1] = \sum_{i \in \mathcal{I}} \binom{i}{j} p^{j+1} (1-p)^{i-j}$  by inspecting each 1 in  $s$  and noting the probability it ends up at position  $j$ , and using the fact that no bit  $\mathcal{I} \ni i < j + 1$  contributes anything to the sum, and by convention  $i < j \implies \binom{i}{j} = 0$ . Therefore,

$$\mathbb{P}_{\mathcal{D}_t}[A_j] = \frac{1}{j!} p^{j+1} \sum_{i \in \mathcal{I}} \frac{i!}{(i-j)!} (1-p)^{i-j} = \frac{1}{j!} p^{j+1} \frac{\partial^j f(k; \cdot)}{\partial x^j} (1-p)$$

Note: This argument also applies to  $k$ -mer probabilities with  $s$  replaced by  $s \oplus r$



# TED Lower Bound Reduction

## Definition

In the TED deletion channel, when vertex  $v$  is removed, contract the edge between  $v$  and its parent. Each vertex is removed with probability  $p$ .

## Theorem

*Let  $q \in (0, 1)$  and  $\delta > 0$  be constants. Then,*

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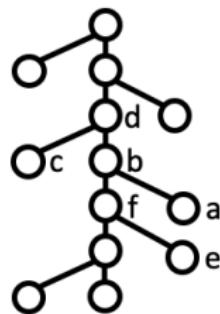
At least  $\Omega(T(n, q^2, \delta))$  TED traces are needed to distinguish arbitrary unlabelled trees with probability at least  $1 - \delta$ .

## Proof Idea for Theorem 3

Consider string  $s = 010110$ . We construct an *unlabelled* tree based on  $s$ :

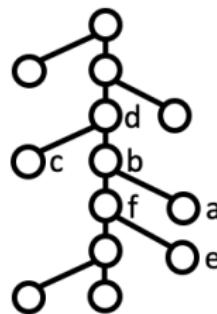
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Deletion of nodes  $a, e$  or  $c$  are OK, have to be careful with nodes  $b, d, f$ .

# Upper Bounds on Tree Families

## Theorem

Let  $q, \delta > 0$ . Given an ordered tree  $R$  with degree at least  $\log_{\frac{1}{q}}(nT(n, q, \delta))$  on  $n$  nodes, we can reconstruct  $R$  using  $T(n, q, \delta)$  traces with probability at least  $1 - \delta$ .

## Theorem

For any ordered tree  $R$  on  $n$  nodes, and  $q, \delta > 0$ , if the leaves of  $R$  have label 0 and internal nodes have label 1, under the TED deletion channel a.a.s. we can reconstruct  $R$  and its labelling using  $T(n, q, \delta)$  traces.

# Proof Idea

- For both theorems, on observing a TED trace from tree  $R$ , do a pre-order traversal, and write down two binary strings corresponding to leaf positions
- Show that the ordering of bits in these strings is preserved with high probability
- Argue that the distribution of these binary strings are equal to the string deletion channel applied to corresponding pre-order traversals of  $R$ .
- Apply an optimal string reconstruction algorithm to recover the leaf positions, and thus  $R$  itself.

# Future Work

- Apply previous two theorems to problems in reconstruction or channel coding
- Find a generic upper bound for tree reconstruction with no bit labels.  
Is  $T(n, q, \delta)$  sufficient in the TED channel?