Absorbing patterns in BST-like expression-trees

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Joint work with Florent Koechlin

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Introduction



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Automated testing, benchmark testing

• Correctness and performance of algorithms

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Randomly generated input

- Realistic distribution
- Simple implementation, possibility of theoretical analysis.

BST-like trees

Target: produce random tree with given *number of nodes n*.

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```
function RandomFormula(n):
if n = 1 then
      p := random symbol in AP \cup \{\top, \bot\};
      return p;
else if n = 2 then
      op := random unary operator in \{\neg, \mathbf{X}, \Box, \Diamond\};
      f := \mathsf{RandomFormula}(1);
      return op f;
else
      op := random operator in \{\neg, \mathbf{X}, \Box, \Diamond, \land, \lor, \rightarrow, \leftrightarrow, \mathbf{U}, \mathbf{R}\};
      if op in \{\neg, \mathbf{X}, \Box, \Diamond\} then
             f := \mathsf{RandomFormula}(n-1);
             return op f:
      else
             x := uniform integer in [1, n-2];
            f_1 := \text{RandomFormula}(x);

f_2 := \text{RandomFormula}(n - x - 1);

return (f_1 \text{ op } f_2);
```

Figure: Code used in tool 1btt (from TCS) to draw an LTL formula.

BST-like trees: distribution over unary-binary trees



BST-like tree distribution is not uniform.

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BST-like tree distribution is not uniform.

• Binary nodes \approx balanced $\frac{n}{2} - \frac{n}{2}$. not for uniform trees

 $\mathbb{E}_n[\min(|T_L|, |T_R|)] \sim c_0 \sqrt{n} \,.$

Expected height of different order

 $\Theta(\log n)$ vs $\Theta(\sqrt{n})$.



Uniform and BST-like distributions

The uniform distribution:

- naturally maximizes entropy.
- can be sampled efficiently (Boltzmann,Recursive,Devroye's constrainted GW).
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Let us see what happens with uniform expressions first...

Universal result for uniform tree model:

Theorem (Koechlin, Nicaud, R, '20)

Expected size of reduction of uniform tree bounded, as size $\rightarrow \infty$.

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Wide variety of examples:



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Wide variety of examples:



For regular expressions on two letters, constant bound ≈ 77.8 .

In our work we

- draw an random BST-like tree expression of size n.
- ▶ study expected size of reduced expressions as $n \to \infty$.
- answer the question: do BST-like distributions present the same flaw as the uniform one?

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Experimental expected size (10 000 samples)¹ on regular expressions $(+, \bullet, \star)$ on two letters a, b:



 ${}^{1}\text{Left to right } (p_{\star},p_{\bullet},p_{+}): \ (\frac{1}{3},\frac{1}{3},\frac{1}{3})\text{, } (\frac{5}{29},\frac{5}{29},\frac{19}{29})\text{, and } (\frac{1}{10},\frac{1}{10},\frac{8}{10})$

1. Model: BST-like trees and absorbing patterns

2. Main Theorem and outline of the proof

3. Conclusions

Unary-binary trees:

leaves (constants), unary, and binary operators.

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- Case binary operator: pick $k \in \{1, ..., n-2\}$ uniformly. return $p_{\text{BST}(k)} \sim p_{\text{BST}(n-k-1)}^{op}$

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• If
$$n = 2$$
: pick op. of arity $1...$

Definition (Simplification, absorbing pattern)

Consider a family of unary-binary tree expressions, consider

- ▶ an "operation" \circledast with arity a = 2,
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We simplify by applying bottom-up the rule:

$$\bigwedge^{\circledast} _{C_1} \stackrel{\sim}{\underset{C_2}{\longrightarrow}} \mathcal{P} \text{, whenever } C_i = \mathcal{P} \text{ for some } i \in \{1, 2\}.$$

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 \Rightarrow We are interested in the *size* (*number of nodes*) of the trees after simplification.

Denote by $\sigma(T)$ the simplification of T.

Theorem. Consider a family of expression trees defined from unary and binary operators with an absorbing pattern \mathcal{P} for an operator \circledast of arity 2.

Take the simplification consisting in inductively changing a \circledast -node by \mathcal{P} whenever one of its children simplifies to \mathcal{P} .

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Then the expected size of the simplification of a random BST-like tree has an asymptotic behaviour given by the following cases, depending on the probability p_{\circledast} of the absorbing operator:

$$\begin{array}{c|c} \Theta(\frac{n}{(\log n)^{\gamma}}) & \Theta(\log n) \\ \Theta(n) & & \Theta(n^{\theta}) & & \Theta(1) \\ 0 & \text{almost no reduction} & \frac{1}{2} & \underset{\text{reduction}}{\overset{\text{significant}}{\overset{3-p_1}{4}} & \underset{\text{case}}{\overset{\text{degenerate}}{\overset{1}{1}} 1 \end{array}$$

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Then the expected size of the simplification of a random BST-like tree has an asymptotic behaviour given by the following cases, depending on the probability p_{\circledast} of the absorbing operator:



▶ Probability p_{\circledast} of \circledast , and p_{I} of picking unary operator.

▶ Regimes from no reduction $\Theta(n)$ to complete reduction $\Theta(1)$

The main regimes experimentally



The main regimes experimentally



Experimental plots (10 000 samples) for regular expressions on two letters a, b: $\mathcal{P} = (a + b)^*$ absorbing for union $\circledast = +$



Scheme for the proof

We employ Analytic Combinatorics to study the expectation,

Ordinary generating function

$$E(z) := \sum_{n=0}^{\infty} e_n z^n \,,$$

encodes sequence $e_n := \mathbb{E}_n[|\sigma(T)|].$

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► Symbolic Step. We find a formal equation describing E(z). Here this will be an ordinary differential equation

 $E'(z) = B(z) + C(z) \cdot E(z) \,.$

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► Analytic Step. We look at E(z) over the complex z ∈ C. A Transfer Theorem links the behaviour of E(z) at its dominant singularity to asymptotics of e_n ⇒ Study singularities

$$E(z) \sim_{z \to 1} \lambda (1-z)^{-\alpha} \Longrightarrow e_n \sim \lambda n^{\alpha-1} / \Gamma(\alpha)$$

Symbolic step

We consider a fundamental sequence

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First order differential equation for the generating function

$$E'(z) = F(z, A(z)) + \frac{1}{1 - p_{\mathsf{I}} z} \left(\frac{2}{z} - p_{\mathsf{I}} + 2(1 - p_{\mathsf{I}}) \frac{z}{1 - z} - 2p_{\circledast}A(z)\right) \cdot E(z) ,$$

where $A(z) = \sum_{n} \gamma_n z^n$.

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where $A(z) = \sum_{n} \gamma_n z^n$.

Proof. Recurrence for e_n involving γ_n ,

$$e_{n+1} = 1 + (s-1)\gamma_{n+1}\mathbf{1}_{n+1\neq s} + p_{\mathbf{I}}e_n + \frac{2p_{\mathbf{II}}}{n-1}\sum_{j=1}^{n-1}e_j + \frac{2p_{\circledast}}{n-1}\sum_{j=1}^{n-1}(e_j - s\gamma_j)(1 - \gamma_{n-j}),$$

here $p_{II} := 1 - p_I - p_{\circledast}$ and $s = |\mathcal{P}|$.

Analytic step: the singularity z = 1

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To apply the Transfer Theorem and complete the proof:

- we show that A(z) and E(z) are analytic over $\Omega = \mathbb{C} \setminus [1, \infty)$,
- we require precise asymptotics for A(z) at z = 1.

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Solution of ODE gives asymptotics

$$\begin{split} E(z) &\sim \frac{c}{(1-z)^2} \left(2 + \int_0^z F(w,A(w))I(w)dw \right) (I(z))^{-1} \,, \quad z \to 1 \,, \end{split}$$
 where
$$I(z) := \exp\Big(2p_{\circledast} \int_0^z \frac{A(w)}{1-p_{\mathrm{I}}w}dw \Big). \end{split}$$

Fully reducible trees: probabilities

We study the generating function $A(z) = \sum \gamma_n z^n$

Proposition

Generating function satisfies Riccati differential equation

$$A'(z) = (s-2)\gamma_s z^{s-1} + \left(\frac{2}{z} + 2p_{\circledast}\frac{z}{1-z}\right)A(z) - p_{\circledast} \cdot (A(z))^2,$$

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Proof.

The probabilities $\gamma_n = \Pr_n \left\{ \sigma(T) = \mathcal{P} \right\}$ satisfy, for $n \ge |\mathcal{P}|$,

$$\gamma_{n+1} = \frac{p_{\circledast}}{n-1} \sum_{k=1}^{n-1} (\gamma_k + \gamma_{n-k} - \gamma_k \gamma_{n-k}).$$

Analytic step: linearization of Riccati

Considering v(z) such that $p_{\circledast}A(z) = v'(z)/v(z)$, Riccati equation becomes linear

$$v''(z) = p_{\circledast} \cdot (s-2)\gamma_s z^{s-1} v(z) + \left(\frac{2}{z} + 2p_{\circledast} \frac{z}{1-z}\right) v'(z) \,.$$

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For linear ODEs:

- domain of analyticity well-understood.
- Frobenius method characterizes behaviour at singularities.

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Proposition

The generating function A(z) satisfies, $z \rightarrow 1$

where $\gamma_{\infty} := (2p_{\circledast} - 1)/p_{\circledast}$ and D > 0 is a constant.

Probability of full reduction

Theorem

The probability γ_n of being fully reducible tends to the constant $\gamma_{\infty} := (2p_{\circledast} - 1)/p_{\circledast}$ for $p_{\circledast} > \frac{1}{2}$, and to zero for $p_{\circledast} \le \frac{1}{2}$.

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 Absorbing pattern model is general ⇒ consider interactions between operators?

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- Absorbing pattern model is general ⇒ consider interactions between operators?
- 3. Take a concrete case: LTL formulas.

Thank you!