String Sanitization: A Combinatorial Approach

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The goal: String sanitization

Conceal sensitive patterns in W while maintaining data utility.

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Given **constraints** and **properties**, determine the **edit operations** to be applied to W so that the properties are satisfied subject to the constraints.

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 - C1 No length-k sensitive pattern occurs in X.

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- **P2** The frequency of length-k non-sensitive patterns is preserved in X.

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 ⇒ No utility loss for tasks based on sequentiality.
- **P2** The frequency of length-k non-sensitive patterns is preserved in X. \Rightarrow No utility loss for tasks based on frequency.

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TFS (Total order, Frequency, Sanitization) problem

Construct the **shortest** string X that satisfies P1, P2, and C1.

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Example. Let $\Sigma = \{a, b\}$, W = aabaaaababbbaab, k = 4, and the set of sensitive patterns be {baaa,aaaa,bbaa}.

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Theorem

The length of X is in $\Theta(k|W|)$. TFS-ALGO solves TFS in the optimal O(k|W|) time.

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PFS-ALGO solves **PFS** in the optimal O(|W| + |Y|) time.

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MCSR is NP-hard via the Multiple-Choice Knapsack (MCK).

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We also develop MCSR-ALGO, an effective heuristic to solve MCSR.

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- Result: Linear time!

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		n	size Σ	patterns	positions $ \mathcal{S} $	length k
OLD	Movement	85,563	100	[30, 240] (60)	[600, 6103]	[3,7] (4)
TRU	Transportation	5,763	100	[30, 120] (10)	[324, 2410]	[2,5] (4)
MSN	Web	4,698,764	17	[30, 120] (60)	[6030, 320480]	[3,8] (4)
DNA	Genomic	4,641,652	4	[25, 500] (100)	[163, 3488]	[5, 15] (13)
SYN	Synthetic	20,000,000	10	[10, 1000] (1000)	[10724, 20171]	[3,6] (6)

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• We compared **TPM** against a greedy baseline, referred to as **BA**.

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OLD	Movement	85,563	100	[30, 240] (60)	[600, 6103]	[3,7] (4)
TRU	Transportation	5,763	100	[30, 120] (10)	[324, 2410]	[2,5] (4)
MSN	Web	4,698,764	17	[30, 120] (60)	[6030, 320480]	[3, 8] (4)
DNA	Genomic	4,641,652	4	[25, 500] (100)	[163, 3488]	[5,15] (13)
SYN	Synthetic	20,000,000	10	[10, 1000] (1000)	[10724, 20171]	[3,6] (6)

- We compared **TPM** against a greedy baseline, referred to as **BA**.
- BA replaces #s greedily from left to right based on letter frequencies.

Experiments: Frequency Distortion

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Experiments: Frequency Distortion $\sum_{U} (Freq_{W}(U) - Freq_{Z}(U))^{2}$, where U is a non-sensitive pattern.

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S denotes the set of occurrences of sensitive patterns.

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 τ -losts are patterns with frequency $> \tau$ in W and $\leq \tau$ in Z.

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 y_y^{\times} on the top of each bar for **BA** denotes $x \tau$ -lost and $y \tau$ -ghost.

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Experiments: Output Size

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Length of X and Y (output of TFS-ALGO and PFS-ALGO, resp.).

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On the top of each pair of bars we plot |X| - |Y|.

Experiments: Speed

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Experiments: Speed



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• Introduced the Combinatorial String Dissemination model which focuses on guaranteeing **privacy-utility trade-offs**.

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Conference version: ECML/PKDD 2019 Full version: arxiv.org/abs/1906.11030