Near-Optimal-Time Quantum Algorithms for Approximate Pattern Matching

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• The edit distance $ED(X, Y)$ measures the minimum number of insertions, deletions, and substitutions of characters to transform X into Y .

Alignment Communication

X a b a a b a b a b a b b b a Y a b a b a b b b a b b a a ED(X, Y) = 3

Text T , $|T| = n$ a b a a b a b a b a b b a b a b a b b a b b a a

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- PM with mismatches: Compute $\mathrm{Occ}_{k}^{H}(P, T) := \{x \mid \mathrm{HD}(T[x \dots x + m), P) \leq k\}.$
- PM with edits: Compute $\mathsf{Occ}^E_k(P, T) := \{x \mid \exists y \; \mathsf{ED}(T[x \mathinner{.\,.} y), P) \leq k\}.$

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- Time complexity $T(n)$: also counts the number of elementary gates

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- \bullet To compute $\mathsf{Occ}^H_k(P, T) \neq \emptyset$ (resp. $\mathsf{Occ}^H_k(P, T) \neq \emptyset$) incurs a $\mathcal{O}(\sqrt{n/m})$ overhead.

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- Time complexity is optimal for $k \leq n^{1/3}$ for mismatches and for $k \leq n^{1/6}$ for edits.
- \bullet Offer advantage over classical algorithms for $k\leq n^{1/4}$ for mismatches and for $k\leq n^{1/7}$ for edits.

Problem: $O(k)$ are too many candidate positions!

Workaround: Communication Complexity

[CKP19]: A subset $S \subseteq \mathrm{Occ}^H_k(P, T)$ of size $|S| = \mathcal{O}(\log n)$ suffices to encode $\mathrm{Occ}^H_k(P, T)$.

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Theorem [CKP19]

There exists a subset $\mathcal{S} \subseteq \mathsf{Occ}^H_k(P,T)$ of size $|\mathcal{S}| = \mathcal{O}(\log n)$ such that the mismatch information for all $x \in S$, defined as

 $MI(x) := \{(i, P[i], T[x + i]) | i \in [0..m) \text{ and } P[i] \neq T[x + i]\},$

provides enough information to construct two strings $P^{\#}$ and $\mathcal{T}^{\#}$ satisfying

$$
\mathsf{Occ}_{k}^{H}(P, T) = \mathsf{Occ}_{k}^{H}(P^{\#}, T^{\#}).
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Definition

 $P^\#,T^\#$ are strings of the same length of P,T , and $T^\#[i]=\sigma$, $P^\#[j]=\sigma'$, and $P^\#[j]=T^\#[i]$ if and only if such equalities can be inferred from $\mathcal{E}(S)$.

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Theorem [CKP19]

For all *x* ∈ [0..*n* − *m*] we have HD($P^{\#}$, $T^{\#}[x..x + m)$) ≥ HD(P , $T[x..x + m)$). Moreover, if x divides $gcd(S)$, then equality holds.

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By choosing S s.t. $\gcd(S) = \gcd(\mathsf{Occ}^H_k(P, T))$ we obtain $P^{\#}, T^{\#}$ s.t. $\mathsf{Occ}^H_k(P, T) = \mathsf{Occ}^H_k(P^{\#}, T^{\#})$. 9 / 14

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Given S and $\text{MI}(\mathsf{x})$ for all $\mathsf{x} \in \mathcal{S}$, we can construct a grammar-like representation $P^\#$ and $T^\#$ of size $\mathcal{O}(k)$ in time $\mathcal{O}(k^2)$. The grammar supports $\mathcal{O}(1)$ time PILLAR operations.

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Theorem [KNW25]

Given N equality equations of the form $X[i..j] = X[i'..j']$ on a length-n string X, we can construct in time $\widetilde{\mathcal{O}}(N^2)$ a grammar-like representation of size $\widetilde{\mathcal{O}}(N)$ of a strings Y which:

- 1. satisfies all N equations, and
- 2. $Y[i] = Y[i]$ only when dictated by the equations.

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- [KNW24]: To encode Occ ${}^H_E(P, T)$ it suffices to consider a set S of size $|S| = \mathcal{O}(\log n)$ of k-edit occurrences $+$ we can translate the encoded information to string equations.
- **[KNW24]**: Adapt to the quantum setting the classical algorithm from **[GKKS22]** for the GAP EDIT DISTANCE problem, i.e., distinguish between $ED(X, Y)$ "small" and $ED(X, Y)$ "large".

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We give a $\widetilde{\mathcal{O}}(\sqrt{n})$ -time algorithm, adapting the quantum algorithm for bounded-error inputs from [HMDW03].

Open Questions

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Thank you!