

The Communication Complexity of Pattern Matching with Edits Revisited

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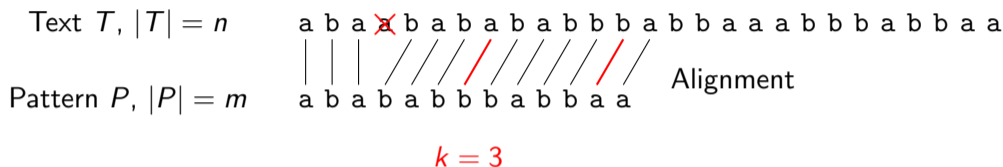
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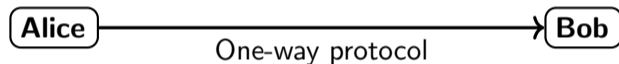
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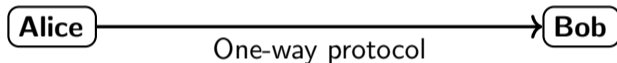


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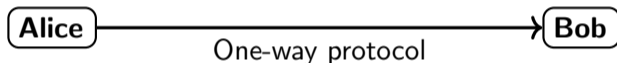
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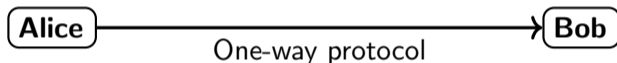
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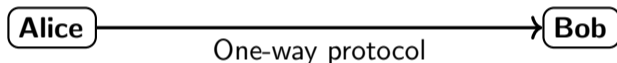
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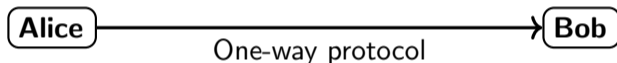
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Communication Complexity = “minimum bits to send to Bob”

Example for Exact Pattern Matching

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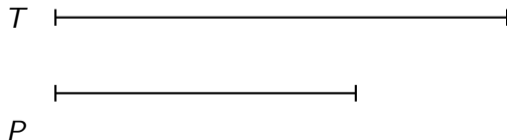
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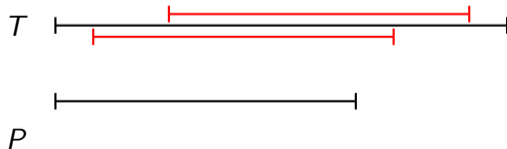
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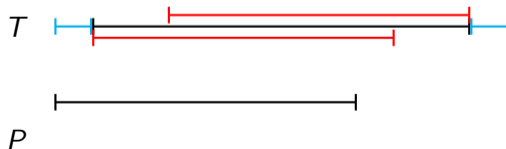
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- The last results are at the basis of our quantum algorithms [KNW24], [KNW25].

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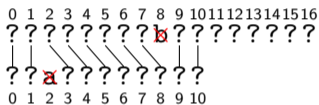
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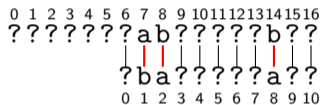
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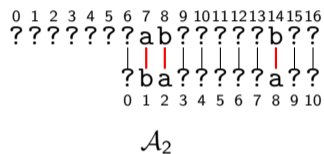
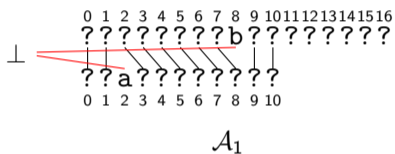


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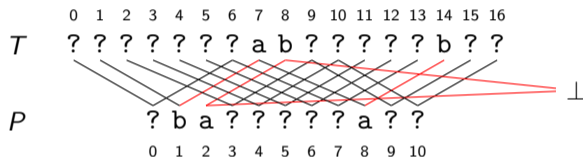
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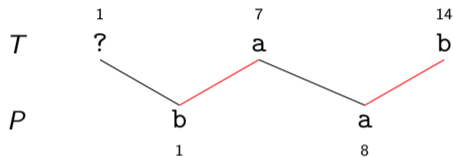
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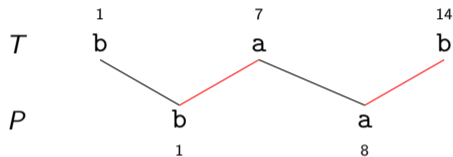
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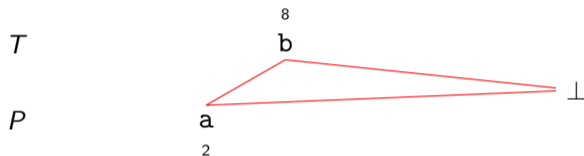
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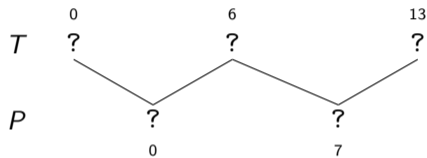
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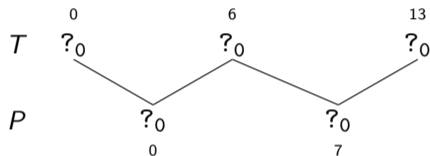
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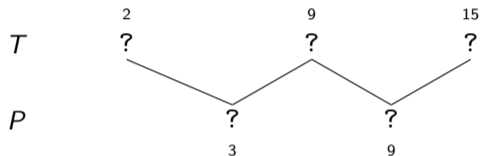
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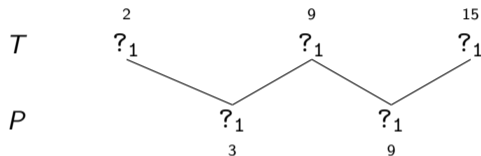
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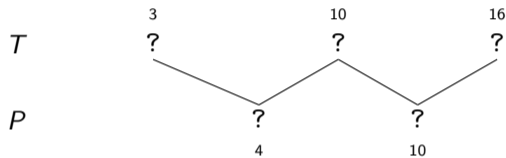
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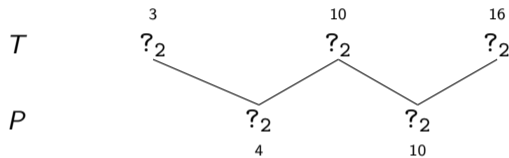
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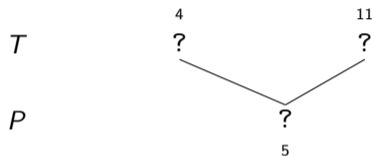
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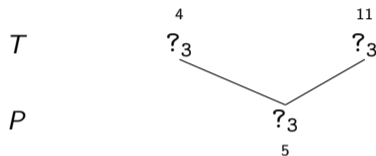
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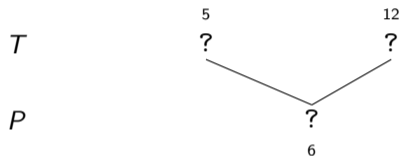
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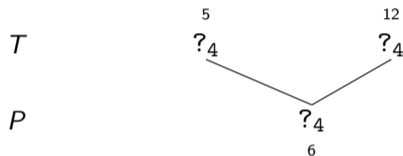
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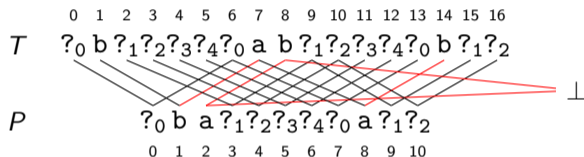
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T_S ?₀ ?₁?₂?₃?₄?₀ ?₁?₂?₃?₄?₀ ?₁?₂

P_S ?₀ ?₁?₂?₃?₄?₀ ?₁?₂

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$$T_S = \text{ ?}_0\text{?}_1\text{?}_2\text{?}_3\text{?}_4\text{?}_0\text{?}_1\text{?}_2\text{?}_3\text{?}_4\text{?}_0\text{?}_1\text{?}_2$$

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$T_S =$ 0 1 2 4 0 1 2 3 4 0 1 2

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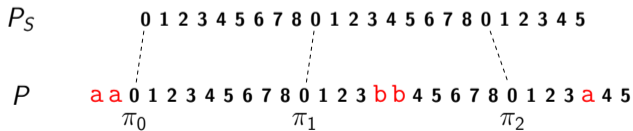
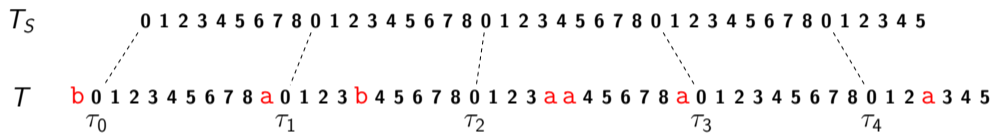
T **b** 0 1 2 3 4 5 6 7 8 **a** 0 1 2 3 **b** 4 5 6 7 8 0 1 2 3 **a a** 4 5 6 7 8 **a** 0 1 2 3 4 5 6 7 8 0 1 2 **a** 3 4 5

P_S 0 1 2 3 4 5 6 7 8 0 1 2 3 4 5 6 7 8 0 1 2 3 4 5

P **a a** 0 1 2 3 4 5 6 7 8 0 1 2 3 **b b** 4 5 6 7 8 0 1 2 3 **a** 4 5

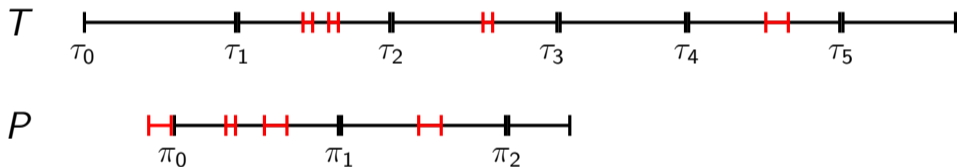
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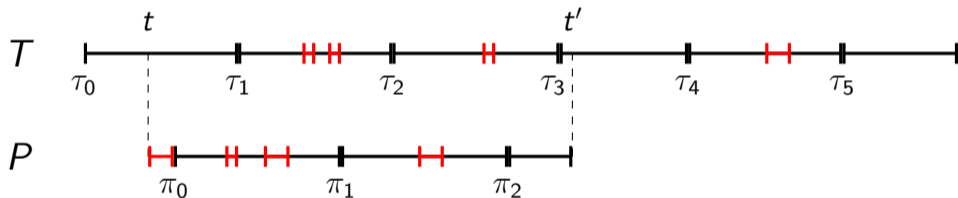
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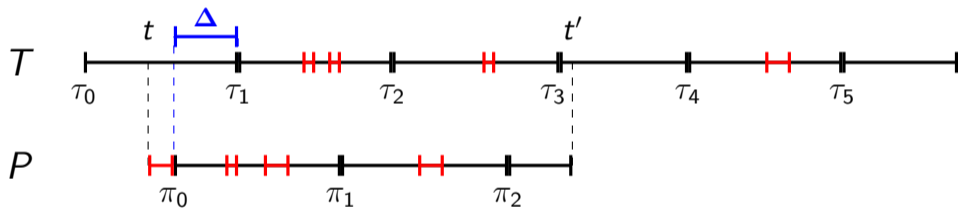
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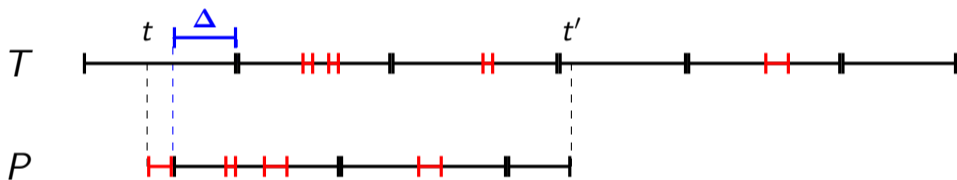
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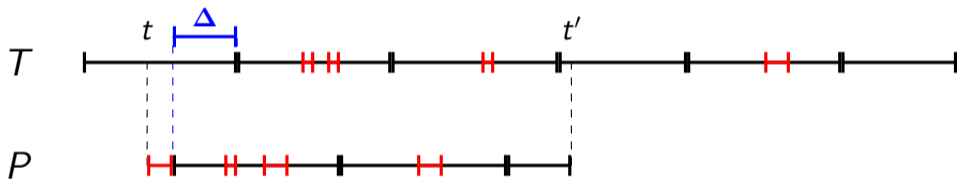


Case $\Delta \gg k$ (large)

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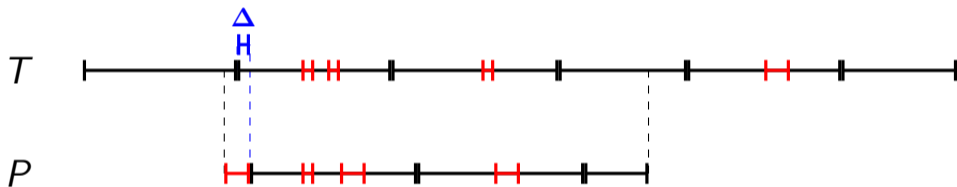
Case $\Delta \gg k$ (large) \implies if \mathcal{X} is added to S , the # of black components at least halves.

N.B. We can assume no such alignments exist. Communication cost = $\mathcal{O}(\underbrace{|S|}_{\leq \log m} \cdot k \cdot \log(m|\Sigma|))$

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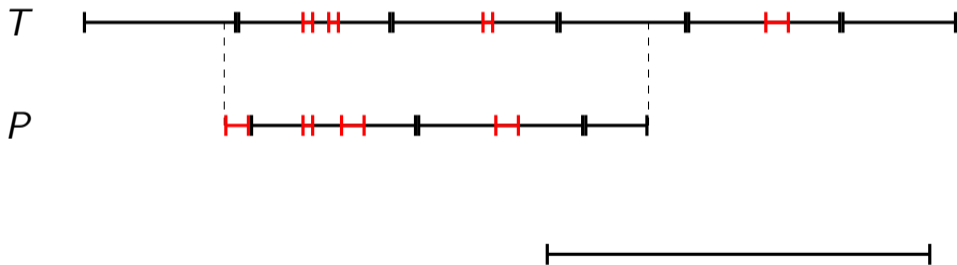


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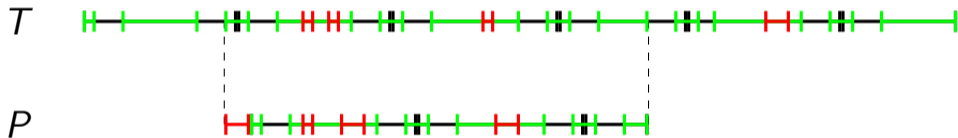


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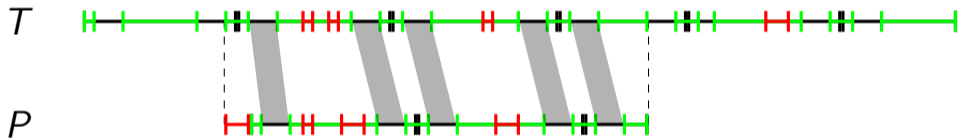


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Case $\Delta \lesssim k$ (small) \implies an optimal alignment $P \rightsquigarrow T[t..t']$ matches characters in the same uncovered black components.

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Challenge 3: For each alignment we send $\Omega(\log m)$ bits, this is a problem when S grows to size $\omega(k)$.
When this happens we: P and T are periodic in a stronger sense (in the sense of [CKW20]), and that in such case we can construct S of size 3.

Thanks!