

Massively Parallel Algorithms for String Matching with Wildcards

MohammadTaghi Hajiaghayi¹, Hamed Saleh¹, Saeed Seddighin¹,
Xiaorui Sun² [CoRR'19]



Sublinear Space

Have you ever had a dataset so big
that it doesn't fit in the memory?



$o(n)$

Sublinear Space Algorithms!

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RAM model

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$o(n)$

Sublinear Space Algorithms!

Alternative models

Alternative Models of Computation

Massively Parallel Computation

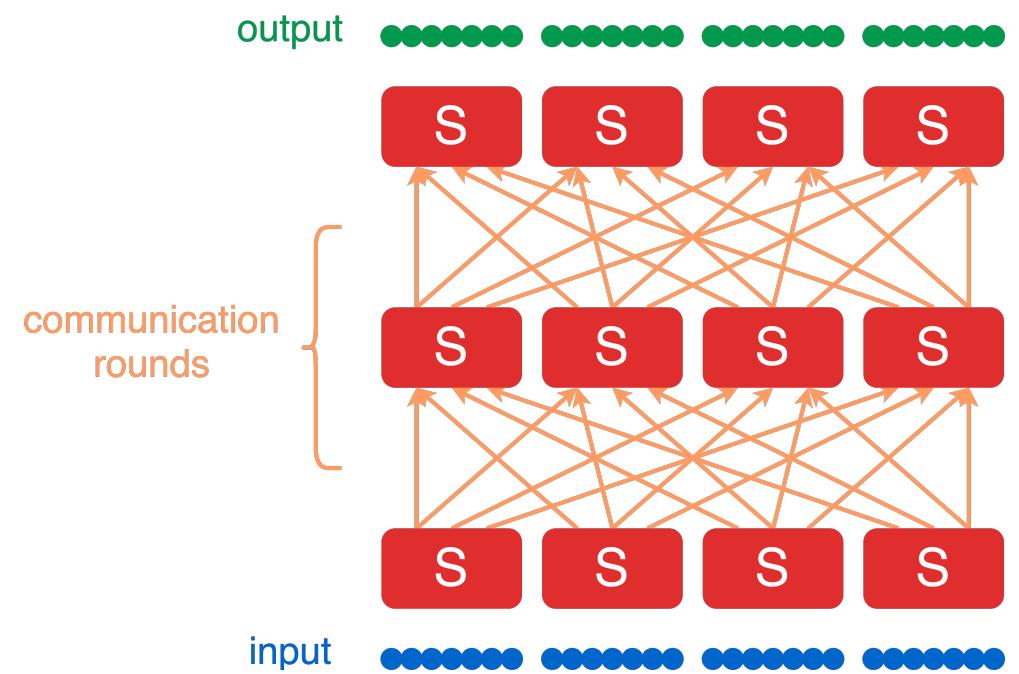
- Modern frameworks for Large-scale parallel/distributed data processing: MapReduce, Hadoop, Spark.
- Key Idea: distribute the workload among several machines.
- The **MPC** model: A theoretical model to abstract out the computational power of these frameworks.

[Karloff et. al 2010]
[Goodrich et. al 2011]
[Beame et. al 2013]
[Andoni et. al 2014]



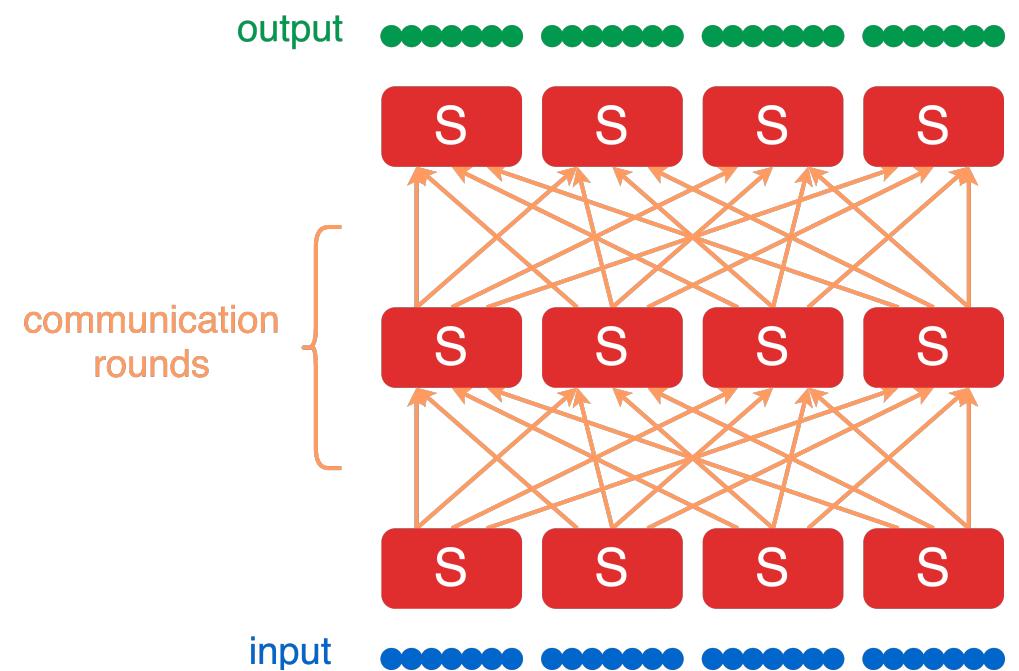
Massively Parallel Computation

- There are M machines each with memory S .
- The input of length N is initially (randomly) distributed among the machines.
 - Sublinear space $S = o(N)$.
 - Usually $N = O(S \cdot M)$.
- The data is processed in several **synchronous rounds**.



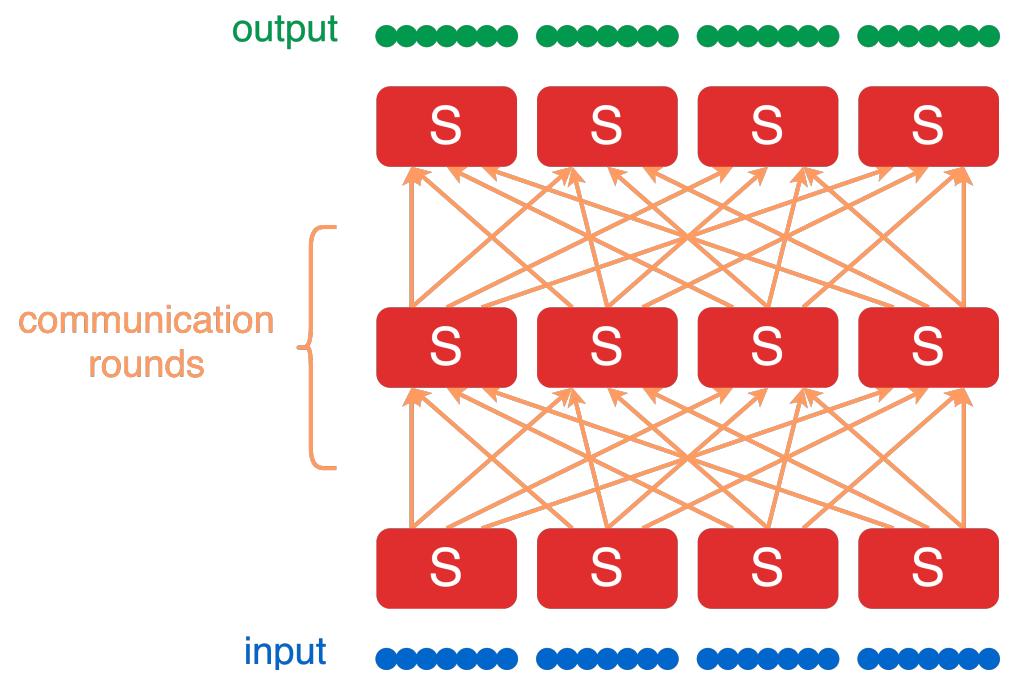
Massively Parallel Computation

- There are M machines each with memory S .
- The data is processed in several **synchronous rounds**. In each round,
 - Machines perform arbitrary computation on their **local** data.
 - Machines **communicate** with each other. Total **incoming/outgoing** messages of each machine is bounded by $O(S)$ words.



Massively Parallel Computation

- There are M machines each with memory S .
- The data is processed in several **synchronous rounds**.
- Main bottleneck: **Communication**. We wish for algorithms with very small number of rounds.
 - Often **sub-logarithmic** rounds.



Related **distributed/parallel** models

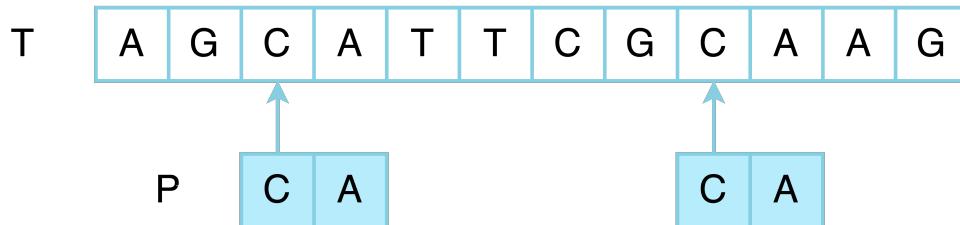
- The **PRAM** model (shared memory)
 - Any PRAM algorithm running in time $t = t(n)$ can be simulated in $O(t)$ MPC rounds.
[Karloff et. al 2010]
- The **LOCAL**, **CONGEST**, and **congested-clique** models
 - Similar techniques can be used for both MPC and these distributed models.
 - Congested-clique is almost equivalent to MPC (in terms of the number of rounds).
[Behnezhad et. al 2018]

String Matching

Massively Parallel Algorithms for String Matching with Wildcards [[arXiv](#)]
Hajiaghayi, **me**, Seddighin, Sun

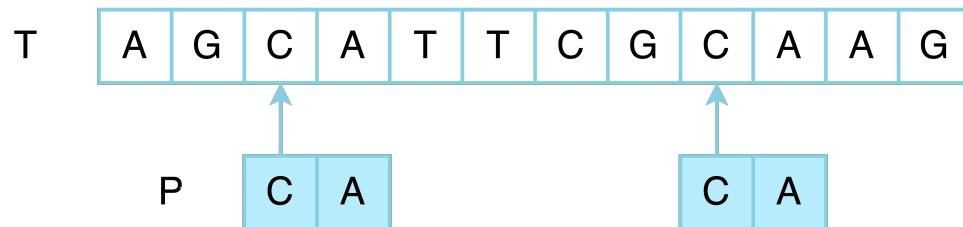
The **String Matching** problem

- An essential problem in bio-informatics and many other areas.
- Given a **text** T and a **pattern** P , we wish to find all substrings of T that match P .



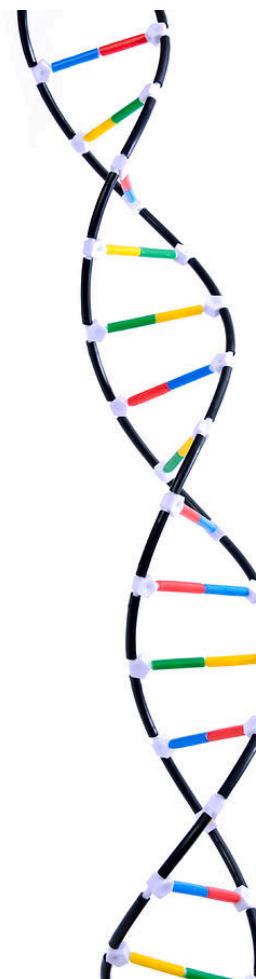
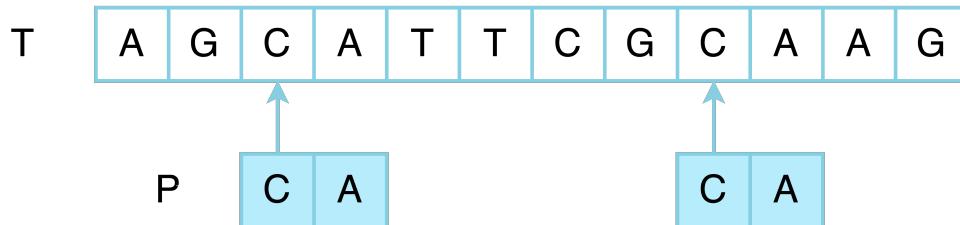
The **String Matching** problem

- Given a **text** T and a **pattern** P , we wish to find all substrings of T that match P .
- In the simplest form both T and P are using the same alphabet Σ , and there is no special character.



The **String Matching** problem

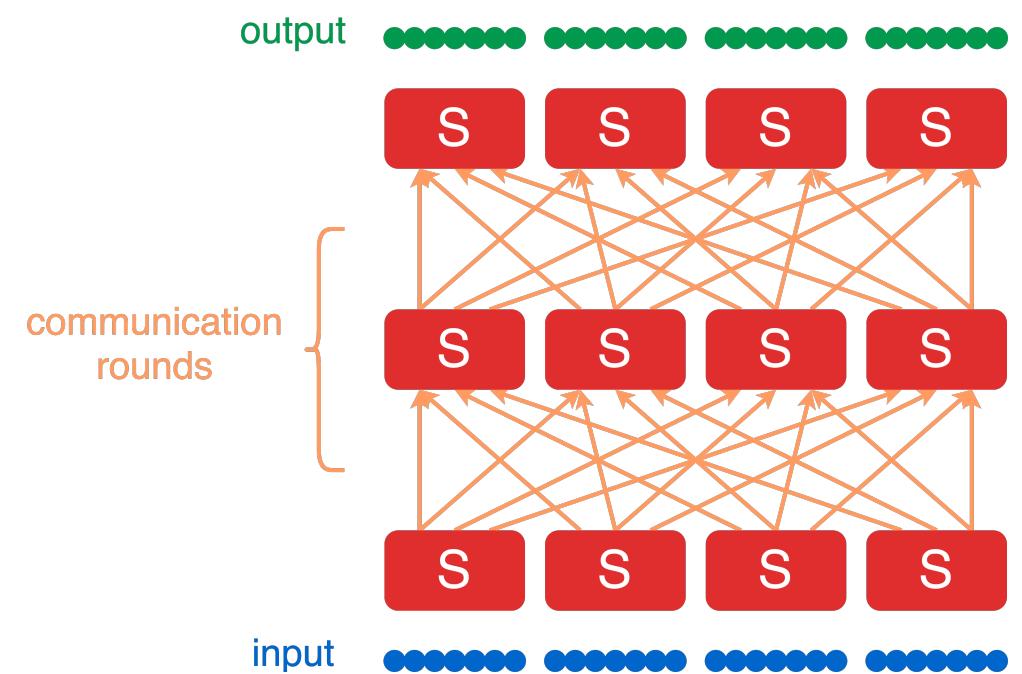
- Given a **text** T and a **pattern** P , we wish to find all substrings of T that match P .
- We also study the case when P can have special characters known as **wildcards**. In particular $\{?, +, *\}$.



String Matching in MPC

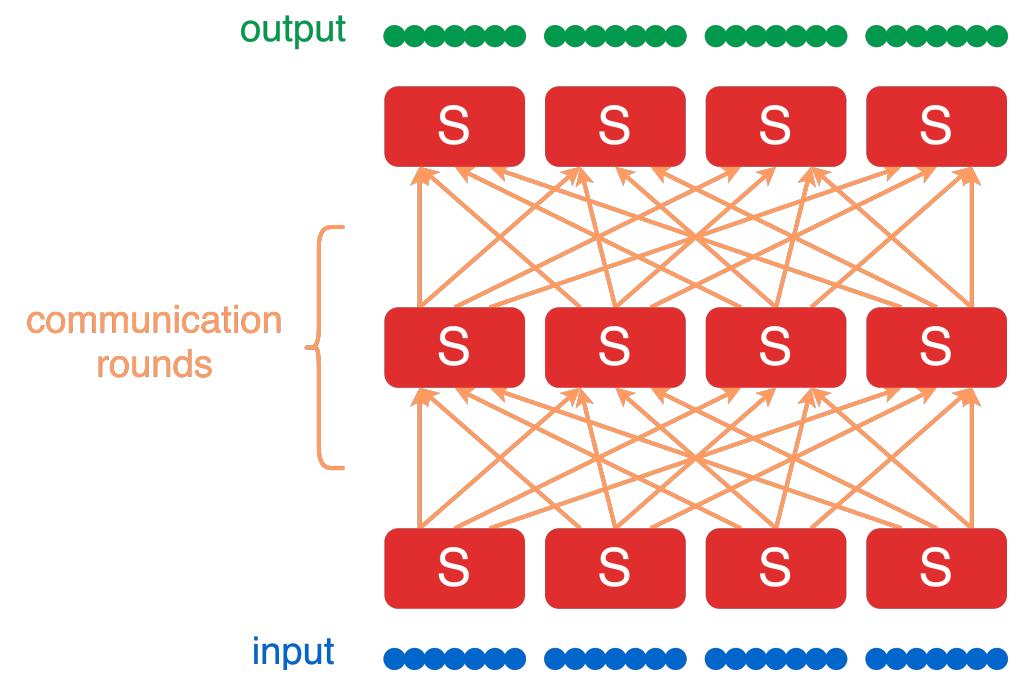
- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq \Sigma^m$ are given.
- There is a **constant**-round MPC algorithm, with $S = O(n^{1-x})$ and $M = O(n^x)$ for any $x < 0.5$, which finds all occurrences of P in T .

[HSSS 2019]



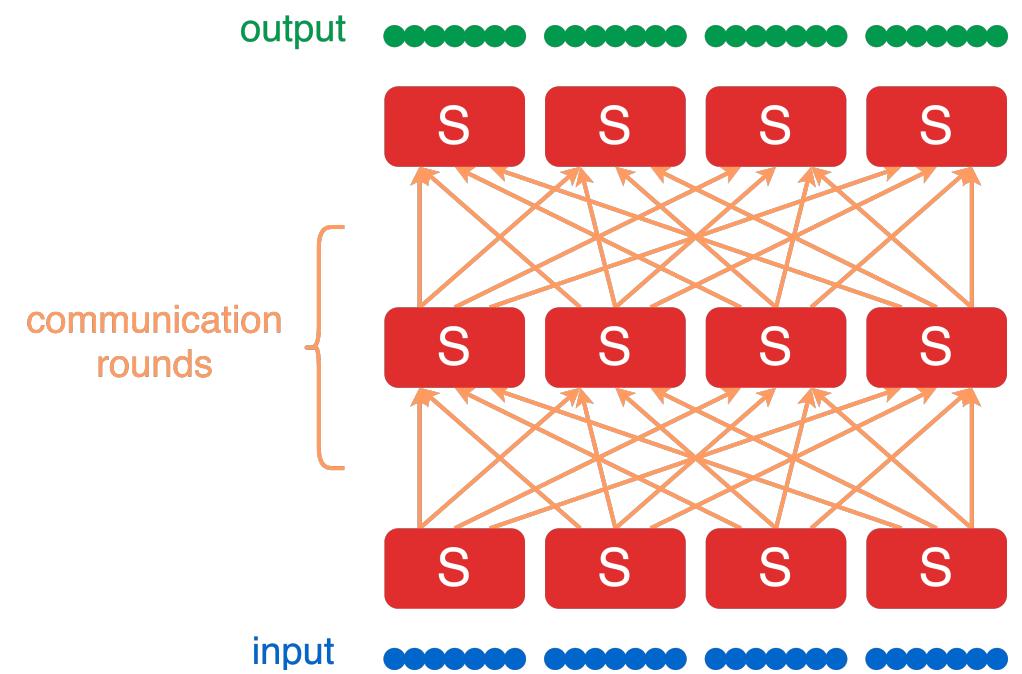
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[HSSS 2019]
- Easy when $m = O(n^{1-x})$: **Double**-covering.



String Matching in MPC

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[HSSS 2019]
- Also easy for general m : Partial **hashing**.

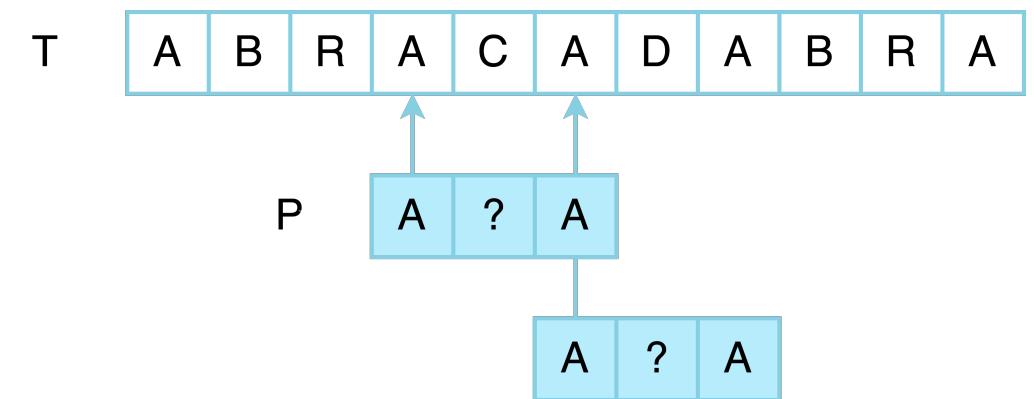


String Matching

with ‘?’ wildcard

String Matching with ‘?’ wildcard

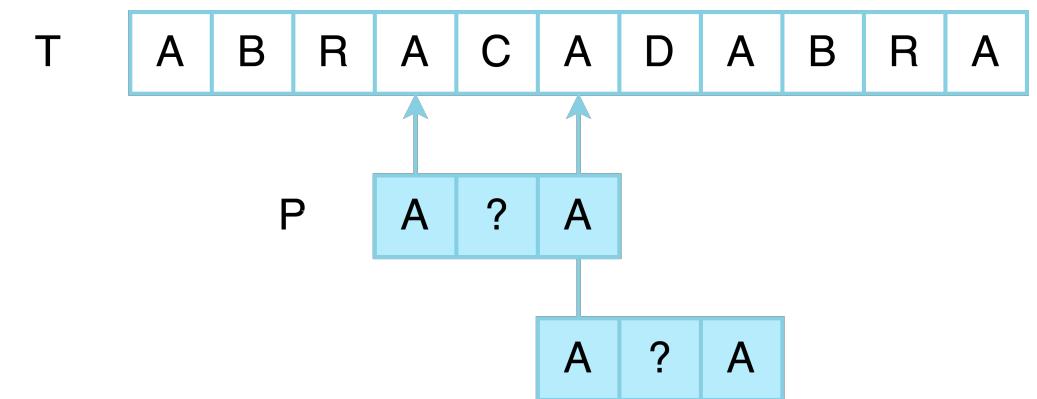
- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{?\})^m$ are given.
- The special character ‘?’ can be replaced with any arbitrary character.



String Matching with ‘?’ wildcard

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[HSSS 2019]

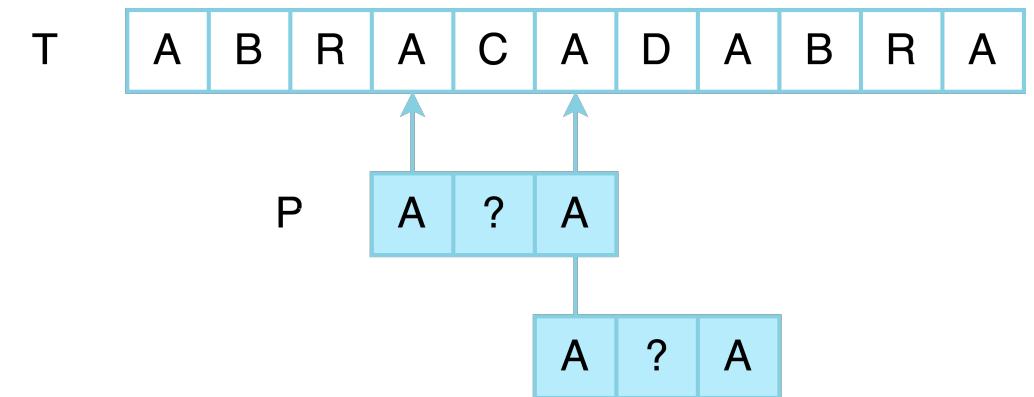


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[HSSS 2019]

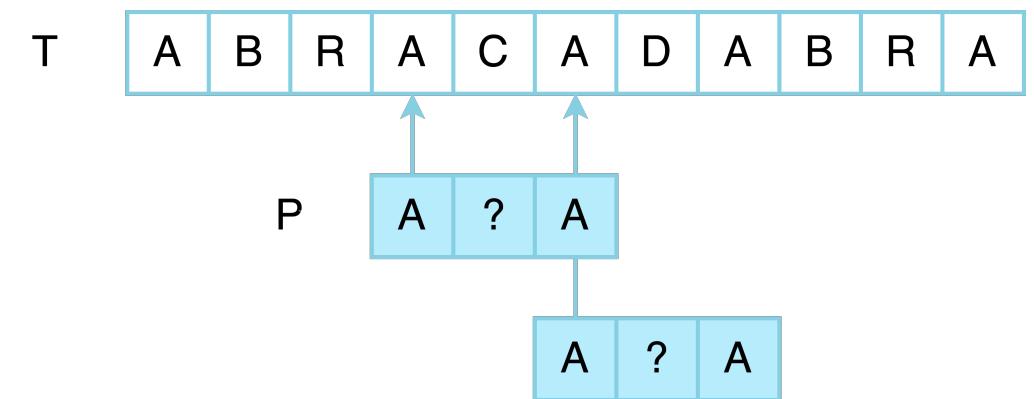
- This can be improved to any constant $x < 1$ at the cost of $O\left(\frac{1}{1-x}\right)$ rounds.



‘?’ wildcard and convolution

- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{?\})^m$ are given.
- This variant of the string matching problem can be reduced to the convolution of two arrays.

[Fischer et. al 1974]



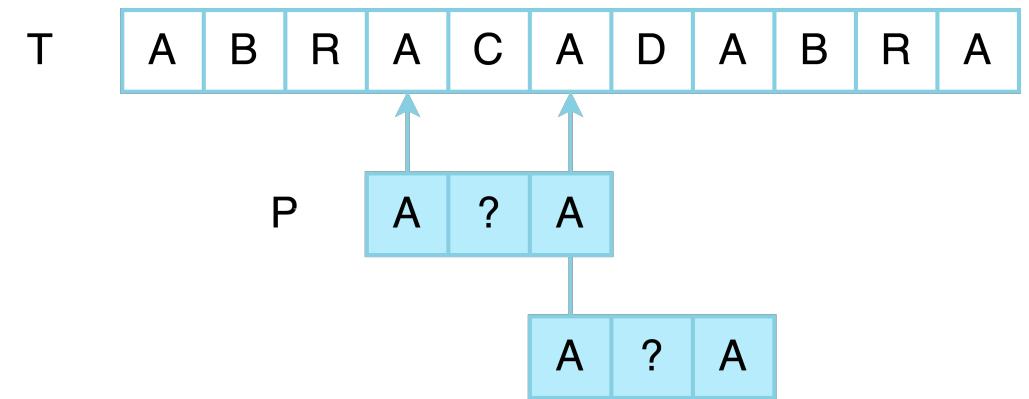
$$T^\dagger = \langle \text{mp}_{T_1}, \text{mp}_{T_1}^{-1}, \text{mp}_{T_2}, \text{mp}_{T_2}^{-1}, \dots, \text{mp}_{T_n}, \text{mp}_{T_n}^{-1} \rangle$$
$$P^\dagger = \langle \text{mp}_{P_1}, \text{mp}_{P_1}^{-1}, \text{mp}_{P_2}, \text{mp}_{P_2}^{-1}, \dots, \text{mp}_{P_n}, \text{mp}_{P_n}^{-1} \rangle$$

$$\text{mp}\{?\} = \text{mp}\{?\}^{-1} = 0, \text{mp}\{C\} = 3, \text{mp}\{C\}^{-1} = 1/3$$

‘?’ wildcard and convolution

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- This variant of the string matching problem can be reduced to the convolution of two arrays.

[Fischer et. al 1974]



$$T^\dagger = \langle 1, \frac{1}{1}, 2, \frac{1}{2}, 18, \frac{1}{18}, 1, \frac{1}{1}, 3, \frac{1}{3}, 1, \frac{1}{1}, 4, \frac{1}{4}, 1, \frac{1}{1}, 2, \frac{1}{2}, 18, \frac{1}{18}, 1, \frac{1}{1} \rangle$$

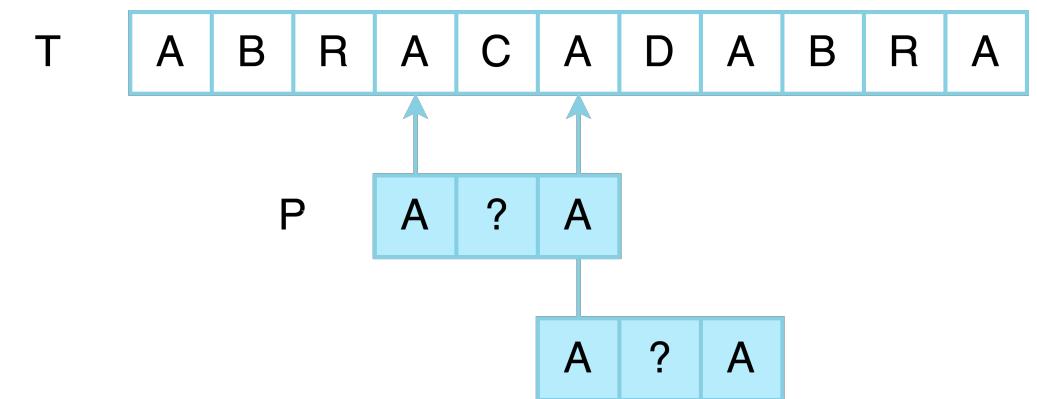
$$P^\dagger = \langle 1, \frac{1}{1}, 0, 0, 1, \frac{1}{1} \rangle$$

‘?’ wildcard and convolution

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$$T^\dagger = \langle 1, \frac{1}{1}, 2, \frac{1}{2}, 18, \frac{1}{18}, 1, \frac{1}{1}, 3, \frac{1}{3}, 1, \frac{1}{1}, 4, \frac{1}{4}, 1, \frac{1}{1}, 2, \frac{1}{2}, 18, \frac{1}{18}, 1, \frac{1}{1} \rangle$$

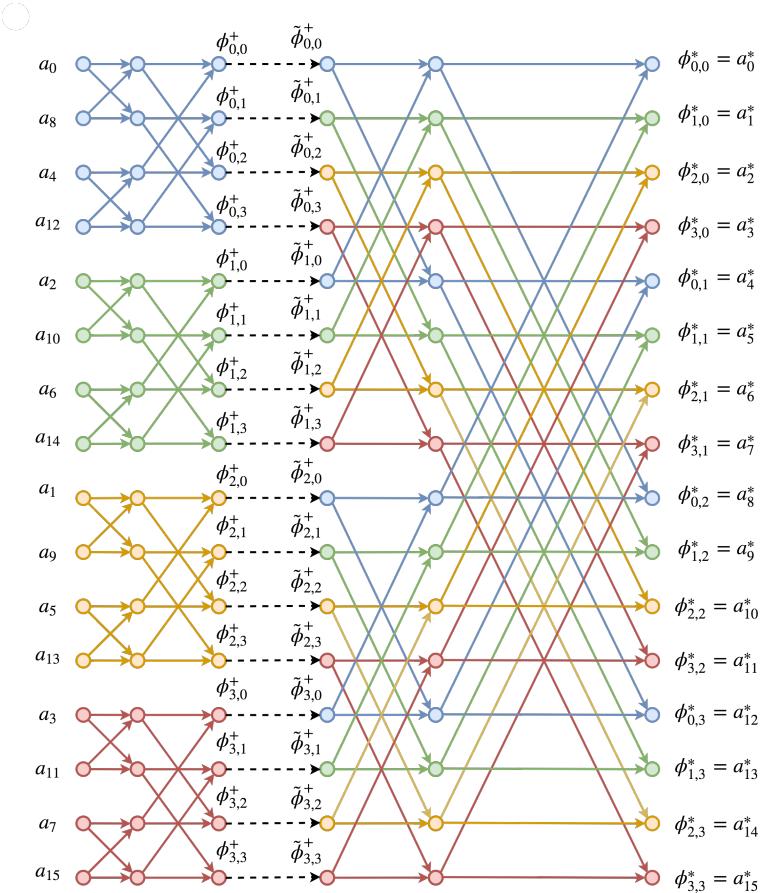
$$P^\dagger = \langle 1, \frac{1}{1}, 0, 0, 1, \frac{1}{1} \rangle$$



$$C = T^\dagger * \text{rev}(P^\dagger)$$

FFT in constant rounds

- Performing a **bit-reversal** operation makes the divide and conquer pattern clean.
- It is easy to decompose the **precedence** graph into the **Butterfly** graphs of different sizes.
- Cooley-Tukey with radix $R = n^{1-x}$.
- Further implications such as the **knapsack** problem.

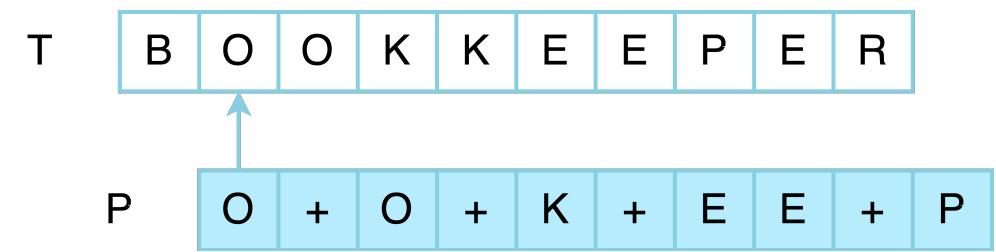


String Matching

with ‘+’ wildcard

String Matching with ‘+’ wildcard

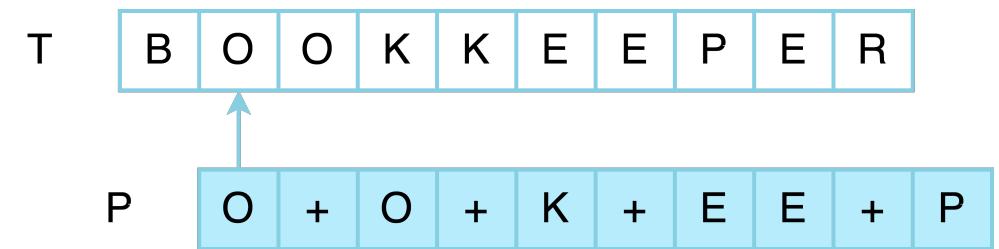
- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{+\})^m$ are given.
- The special character ‘+’ means that the preceding character can be repeated arbitrary times.



String Matching with ‘+’ wildcard

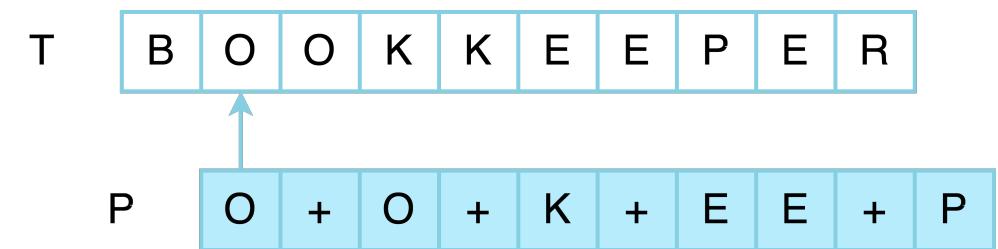
- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{+\})^m$ are given.
- There is a **constant**-round MPC algorithm, with $S = O(n^{1-x})$ and $M = O(n^x)$ for any $x < 0.5$, which finds all occurrences of P in T .

[HSSS 2019]



Run Length Encoding

- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup ' + ')^m$ are given.
- Perform **Run Length Encoding** on both strings.



$$T^\circ = \langle \langle b, 1 \rangle, \langle o, 2 \rangle, \langle k, 2 \rangle, \langle e, 2 \rangle, \langle p, 1 \rangle, \langle e, 1 \rangle, \langle r, 1 \rangle \rangle$$

$$P^\circ = \langle \langle o, 2+ \rangle, \langle k, 1+ \rangle, \langle e, 2+ \rangle, \langle p, 1 \rangle \rangle$$

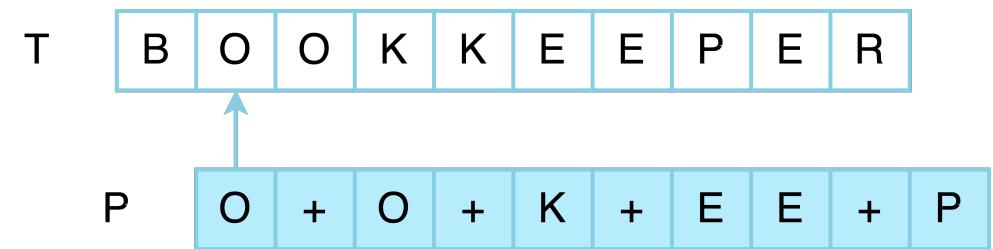
- Reduces to **Greater-than** matching.

Run Length Encoding

- Reduces to **Greater-than** matching.
- A special case of **Subset** matching.
- The **Subset** matching problem can be solved in $O(n \log^2 m)$ by a careful reduction to sparse convolution.

[Cole et. al 2002]

- It's possible to implement it in $O(1)$ MPC rounds.

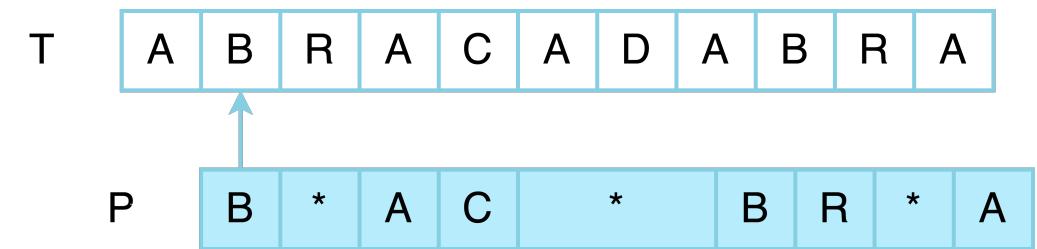


String Matching

with '*' wildcard

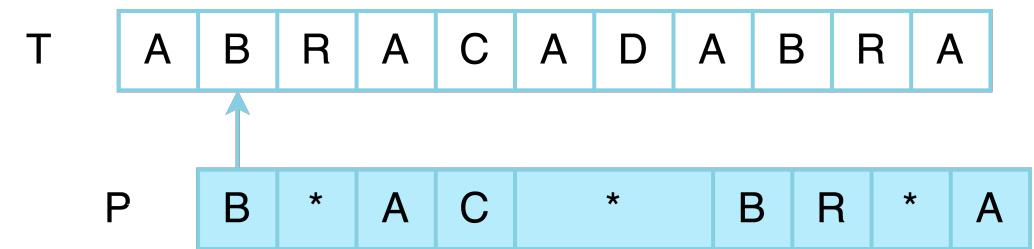
String Matching with ‘*’ wildcard

- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{*\})^m$ are given.
- The special character ‘*’ can be replaced with any arbitrary string.



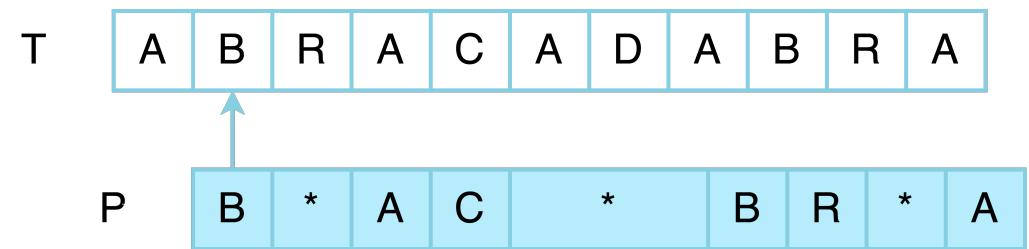
String Matching with ‘*’ wildcard

- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{*\})^m$ are given.
- The special character ‘*’ can be replaced with any arbitrary string.
- Unlike ‘?’ and ‘+’, we have no positive result for this wildcard even in $O(\log n)$ rounds...
- There is a conjecture that we can’t solve graph **connectivity** in $o(\log n)$ rounds.



String Matching with ‘*’ wildcard

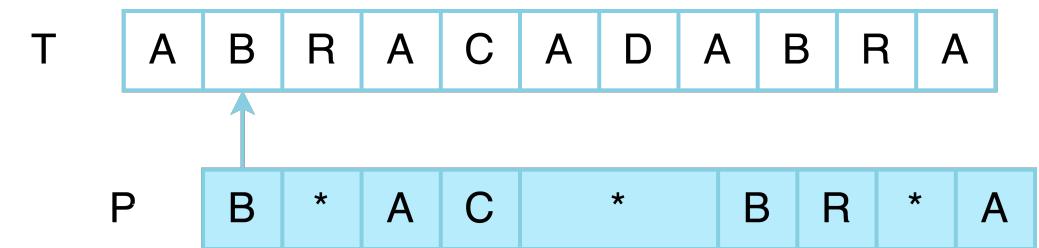
- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{*\})^m$ are given.
- The special character ‘*’ can be replaced with any arbitrary string.
- Unlike ‘?’ and ‘+’, we have no positive result for this wildcard even in $O(\log n)$ rounds...
- But we can solve it in special cases.



‘*’ wildcard in **small** patterns

- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{*\})^m$ are given such that $\mathbf{m} = O(n^{1-x})$.
- There is a $O(\log n)$ -round MPC algorithm, with $\mathbf{S} = O(n^{1-x})$ and $\mathbf{M} = O(n^x)$ for any $x < 0.5$, which finds all occurrences of P in T .

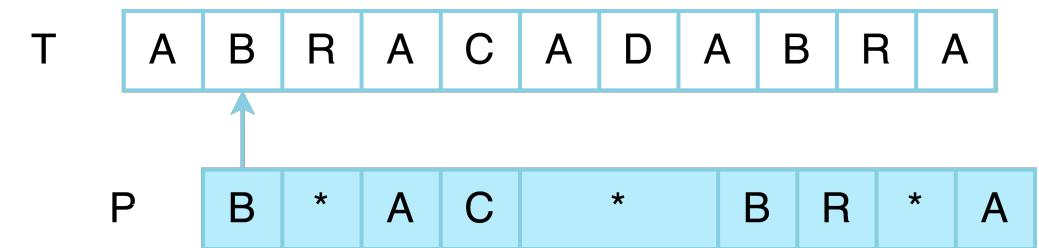
[HSSS 2019]



‘*’ wildcard in no **common prefix** case

- A **text** $T \subseteq \Sigma^n$ and a **pattern** $P \subseteq (\Sigma \cup \{*\})^m$ are given such that no two **sub-patterns** share a common **prefix**.
- There is a $O(\log n)$ -round MPC algorithm, with $S = O(n^{1-x})$ and $M = O(n^x)$ for any $x < 0.5$, which finds all occurrences of P in T .

[HSSS 2019]



Thanks for watching!

Any questions?