



Finding Diverse Strings and Longest Common Subsequences in a Graph



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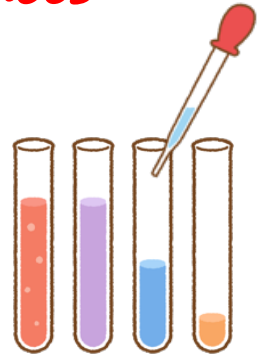
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Pisa, Italy



■ A Classic problem: Longest Common Subsequence (m-LCS)

- The problem of finding *one of the longest (non-contiguous) subsequences common to all M input strings (LCS)*.
- One of the most fundamental problems in computer science and bioinformatics.
- It has been studied for over 50 years in theory and applications.



Longest Common Subsequence (LCS)

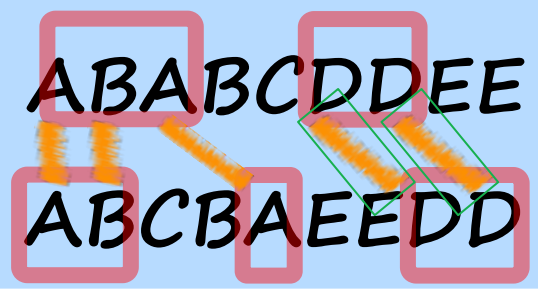


A **longest common subsequence (LCS)** of a set S of input strings is a (non-contiguous) subsequence common to all of m inputs strings.

The set S of $m=2$ input strings

$X = \text{ABABCDDEE}$

$Y = \text{ABCBAEEDD}$



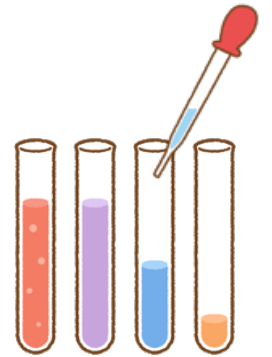
- ϵ, A, B, C, D, E Common subsequences
- AA, AB, AC, AD, AE, BA, ..., CD, CE, DD, EE,
- ABA, ABB, ABC, ABD, ..., CEE,
- ABAD, ABAE, ABBD, . . . , BCEE,

- ABADD, ABAEE, ABBDD,**
ABBEE, ABCDD, ABCEE

Six LCSs of S with length $r=5$



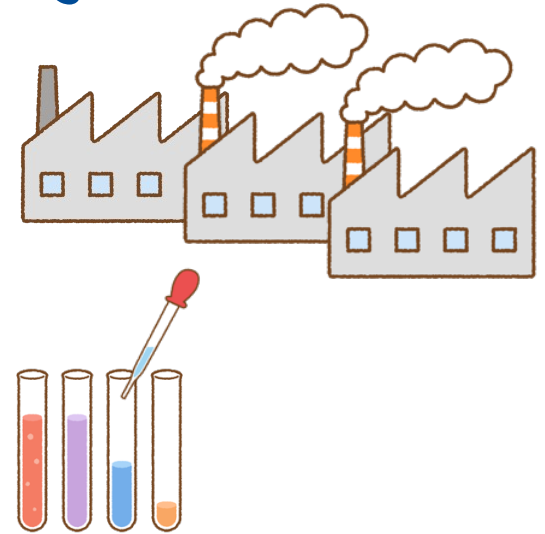
- **A Classic problem: Longest Common Subsequence (m-LCS)**
 - The problem of finding **one of the longest (non-contiguous) subsequences common to all M input strings (LCS)**.
 - One of the most fundamental problems in computer science and bioinformatics.
 - It has been studied for over 50 years in theory and applications.
- **Computational complexity of m-LCS:**
 - **Polynomial-time solvable** if M is a constant (Irving & Fraser, CPM'92), while it is **NP-hard** if M is an input.
 - **$W[t]$ -hard** if M is a parameter, and **$W[2]$ -hard** if L is a parameter (Bodlaender, Downey, Fellows, & Wareham, TCS, 1995).
 - **FPT** by other parameterization (Bulteau, Jones, Niedermeier+, CPM'22)
- **Our goal:** We introduce **the diversity maximization problem for LCSs**, and study **its computational complexity** (approximability & parameterized complexity)





Motivations: Finding Multiple Diverse Solutions

- In combinatorial optimization, much effort has been done for finding a single best solution.
 - Examples: Drug discovery, route planning in delivery networks, factory automation, etc.
- However, there has been growing interest in finding **multiple diverse solutions** in optimization problems
- Reasons:
 - The **specification** may not be perfect
 - There can be **too many optimal solutions** (algorithm-dependent)
 - **Human experts** may want to intervene ("Human-in-the-Loop")
- Diversity maximization problem attracts much attention





To find diverse solutions

■ There has been a variety of methods studied in the past:

- Random generation - Generate solutions randomly.
- Enumeration - Generate solutions exhaustively.
- Top-K search - Generate in decreasing order of objectives

■ However, any of these methods are not satisfactory from the view point of (i) the size of a solution set and (ii) the explicit guarantee of the diversity

■ Our goal: We study the computational complexity of diversity maximization problem for LCSs



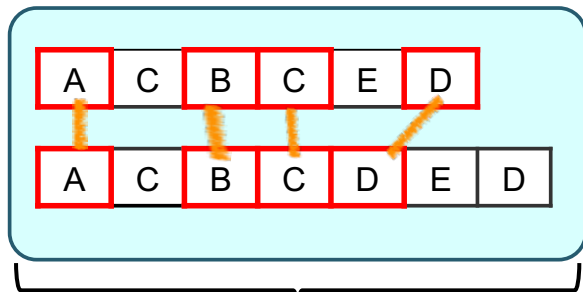
Our problem: Diversity Maximization

Given a set S of m input strings, find a subset X of K longest common subsequence among all of N solutions of the m -LCS problem that maximizes a specified diversity measure $D_d(X)$.

The universe Sol of all solutions of m -LCS problem (all LCSs of S)

Output: a “diverse” subset X of K longest common subsequence

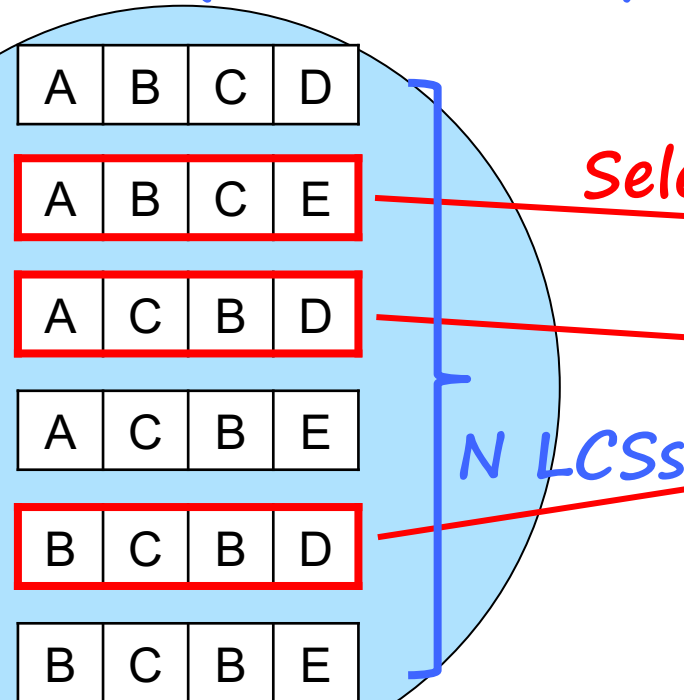
Input: A set S of m input strings (constant m)



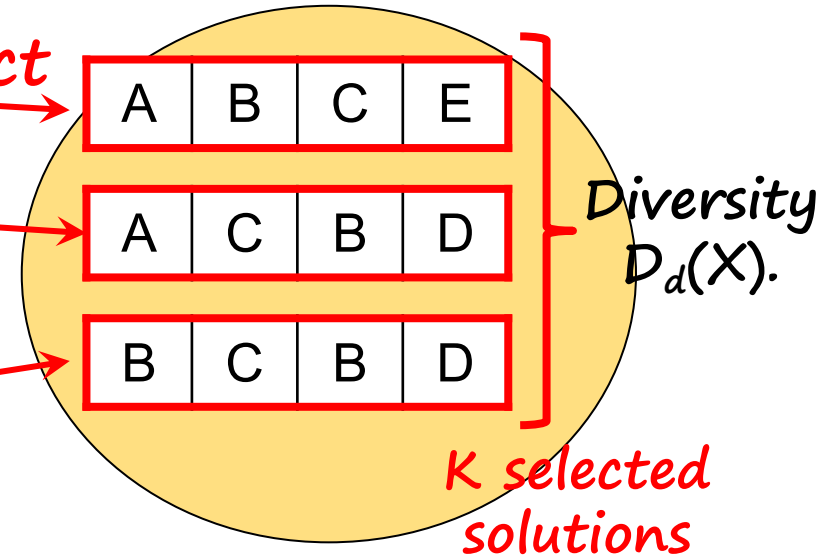
Minimum length L

Defines

Exponentially many LCSs



Select





Def. The *Hamming distance* $d_{HD}(X, Y)$ between two strings X and Y of the same length n is the total number of positions at which the strings disagree (differ).

$$d_{HD}(X, Y) = \sum_{i=1}^n \mathbb{1}\{X[i] \neq Y[i]\}$$

	1	2	3	4	5
X	A	B	A	D	D
Y	A	B	C	D	E

$$d_{HD}(X, Y) = 2$$



Def. Diversity measures for K -solution sets

$$X = \{x_1, \dots, x_K\} \subseteq \text{Sol.}$$

- **Sum-Diversity** $D_d^{\text{sum}}(X)$: the sum of the pairwise distance over all pairs in X .

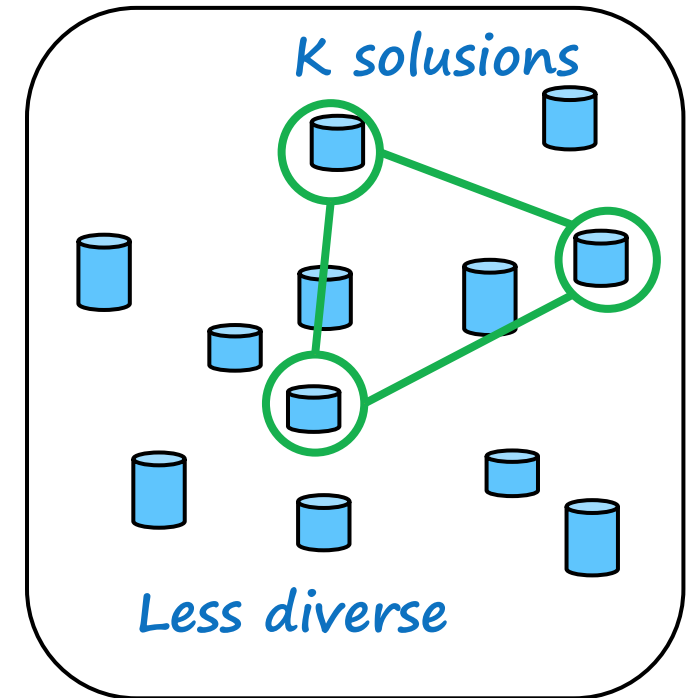
$$D_d^{\text{sum}}(X) := \sum_{i < j} d(x_i, x_j),$$

- **Min-Diversity** $D_d^{\text{min}}(X)$: the minimum of the pairwise distance over all pairs in X

$$D_d^{\text{min}}(X) := \sum_{i < j} d(x_i, x_j)$$

A set Sol of optimal solutions

$$K = 3$$





Sum-Diversity Maximization Problem (Max-Sum Diverse(LCS))

Given: A set of solutions Sol , distance function $d: Sol^2 \rightarrow \mathbb{R}_+$ integers $K \geq 1, \Delta \geq 0$

Task: Find a subset $X = \{x_1, \dots, x_K\} \subseteq Sol$ of K solutions such that

(1) $|X| \leq K$, and (2) $D_d^{\text{sum}}(X) \geq \Delta$ ($D_d^{\text{min}}(X) \geq \Delta$).

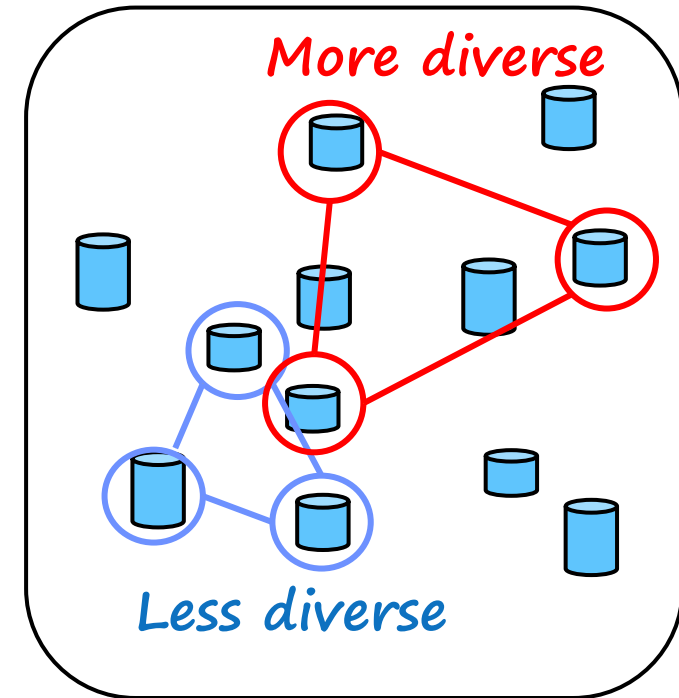
Similarly, we can define *the Max-Min variant* (Max-Min(LCS))

$$D_d^{\text{sum}}(X) := \sum_{i < j} d(x_i, x_j) \quad \text{Sum-Diversity}$$

$$D_d^{\text{min}}(X) := \sum_{i < j} d(x_i, x_j) \quad \text{Min-Diversity}$$

A set Sol of optimal solutions

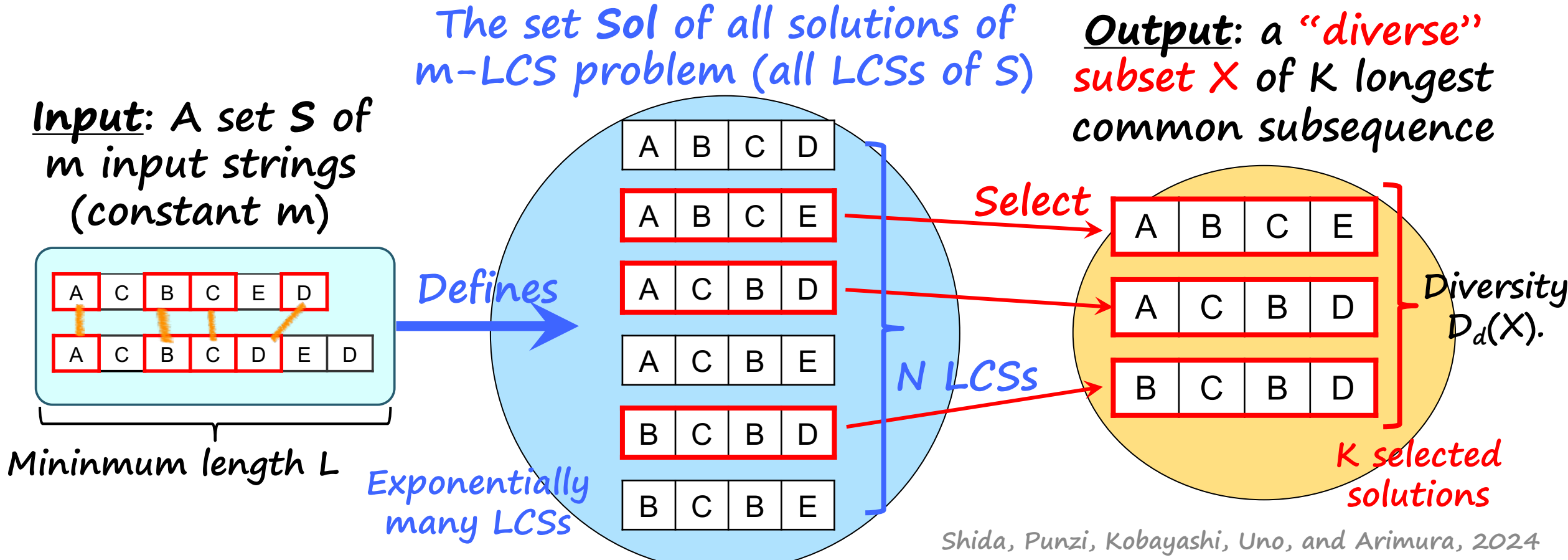
$K = 3$





Our problem: Diversity Maximization

Given a set S of m input strings, find a subset X of K longest common subsequence among all of N solutions of the m -LCS problem that maximizes the specified diversity measure $D_d(X)$.





In 1970s, early days: K-dispersion problem

- Selecting a diverse set of K points among N points in a metric space
- $O(N^K t_{\text{dist}})$ time : try all K combinations among N

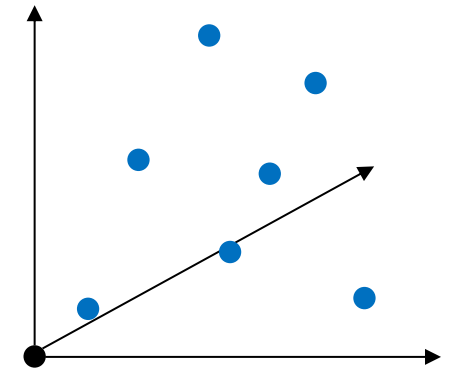
In 2020s: “Diverse-X program”

- proposed by Michael R. Fellows (Dagstuhl seminar, 2019)

Let's study the complexity of finding diverse solutions in various discrete optimization problems X !

- Examples of X = MSTs, Matchings, Shortest paths, etc.
 - Typically, there are exponentially many solutions.
- NP-hard in most case. Sometimes, FPT or Approximable

Our goal: We study the computational complexity of diversity maximization problem in strings, especially LCSs



Michael R. Fellows
U. Bergen, Norway

Notes: All results below except PTAS hold for the Min-Sum(LCS) & MinSum(StringSet)
Assumption: the number m of input strings is always constant, throughout.

When K is bounded

Polynomial time
computable

Proof: Dynamic
programming

When K is input

NP-hard, but

admits PTAS (Min-Sum only)

Approximable within any constant error

Proof: Local search (Cevallos+ 2019)

+ DP + Negative type metric

Parameterized
by K and r
FPT

Proof: Color coding

Parameterized
by K
W[1]-hard

Proof: FPT-reduction
from p -CLIQUE

K : #solutions to select, r : max. length of LCS



Summary: Complexity of Max. Diverse LCSs

CPM2024, Fukuoka



Notes: All results below except PTAS hold for the Min-Sum(LCS) & MinSum(StringSet)
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Main result of Sec.3 (upperbound)

Assumption: the number of input strings $m \geq 2$ is a constant, throughout

- **Corollary**: When K is a constant, **Max-Sum(LCS)** is solvable in polynomial time.
- The same result holds for **Max-Min**.

Our approach: First show a more general result (Thm 6), solve it using dynamic programming on a DAG, and then apply it to the Corollary.

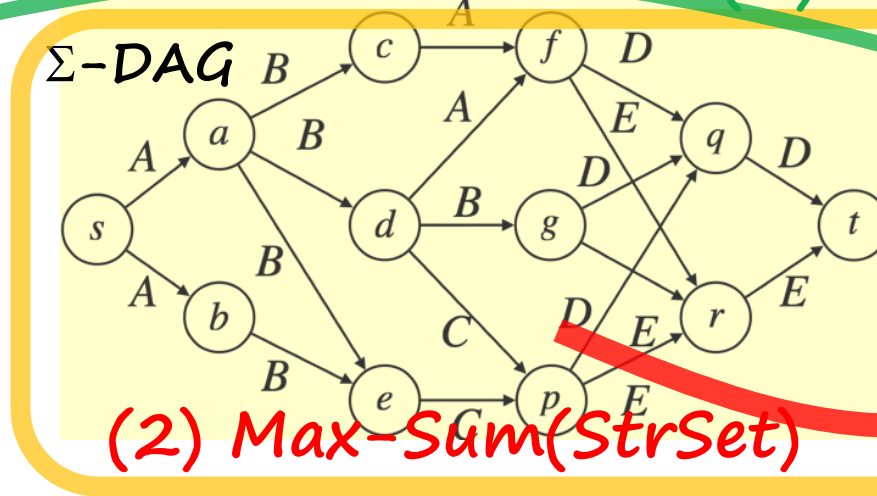
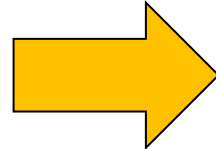
- **Theorem 6 (Max-Sum(StrSet))**:
When K is a constant, for any set L of strings of equal length, and given a Σ -labeled acyclic directed graph (Σ -DAG) representing it, **Max-Sum(StrSet)** can be solved in polynomial time.

Original problem to consider:

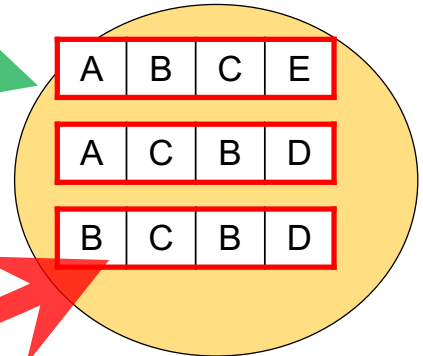
(1) Max-Sum(LCS): the Max-Sum diverse LCSs problem

Input strings ($m=2$)

$X_1 = ABABCDDEE$
 $X_2 = ABCBAEEDD$



(1) Max-Sum(LCS) A diverse subset X of K LCSs



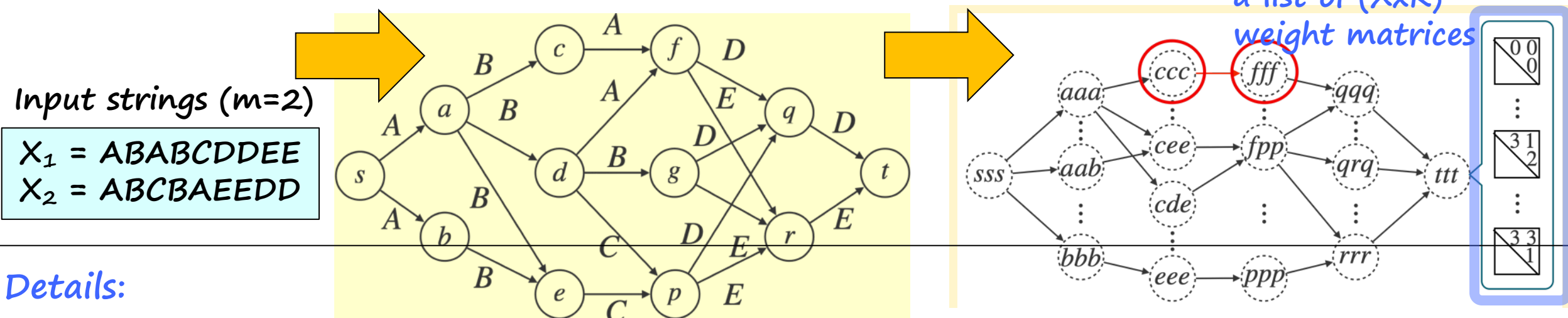
Select

- **Observation (folklore):** The set of all (exponentially many) LCSs can be stored in a polynomial-sized DAG G (called a Σ -DAG).

Hence, we consider a more general problem instead: (2) Max-Sum(StrSet): the Max-Sum diverse String Set problem for Σ -DAGs of equi-length strings



Polynomial-time Algorithm: Performs dynamic programming over all K -tuples of vertices of G to collect the set of $(K \times K)$ -weight matrices for all combinations of K paths that maximize the diversity measure.



- Details:**
- A combination of K paths in G reachable to each K -tuple of vertices uniquely determines a set of K prefixes of solution strings, and thus, the associated $(K \times K)$ -weight matrix $M = (W_{ij})$,
 - where the weight W_{ij} is defined by the pairwise Hamming distances $d(P_i, P_j)$ between string labels of two paths P_i and P_j .
 - To each K -tuple of vertices, maintain the list of all possible weight matrices

Observation: There are at most $(\Delta+1)^{K \times K}$ (polynomially many) distinct weight matrices for constant K and Δ .



Notes: All results below except PTAS hold for the Min-Sum(LCS) & MinSum(StringSet)
Assumption: the number m of input strings is always constant, throughout.

When K is bounded

When K is input

NP-hard, but

admits PTAS (Min-Sum only)

Approximable within any constant error

Proof: Local search (Cevallos+ 2019)

+ DP + Negative type metric



- **Theorem 3 (Hardness):** When K is an input, the Max-Sum maximum diversity LCSs problem is
 - NP-hard. (Even if # of input strings is a constant $m=2$).
 - W[1]-hard if K is a parameter.
- The same result holds for the Max-Min variant.

Approach: We show the NP-hardness of a more basic "string set diversity problem" and reduce it to the original problem.

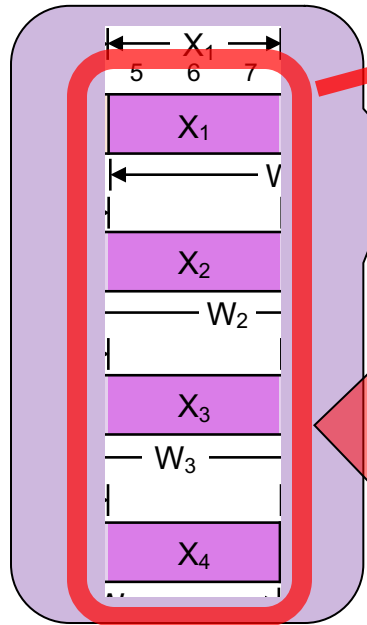
- Lemma 1: When K is an input, Max-Sum(StringSet) is NP-hard, and W[1]-hard if K is a parameter. (Proof: reduction from p -clique)

The next lemma is a key.

- Lemma 2: Max-Sum(StringSet) is FPT-reducible to Max-Sum(LCS)

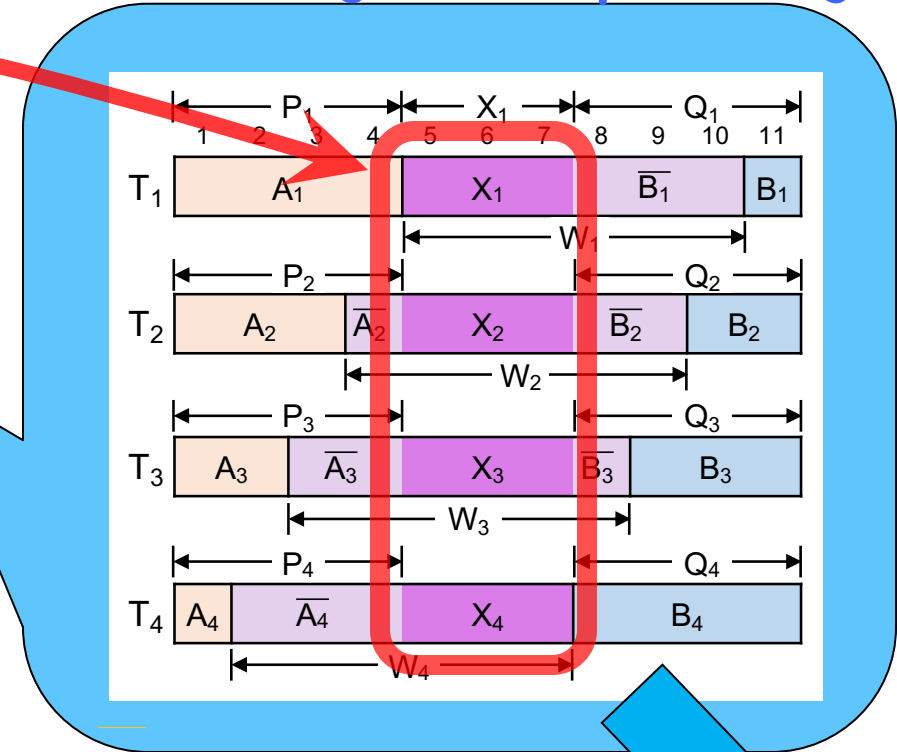
Proof of Lemma 2

(3) A solution of the 2-LCS problem:
Recovered original N strings with paddings

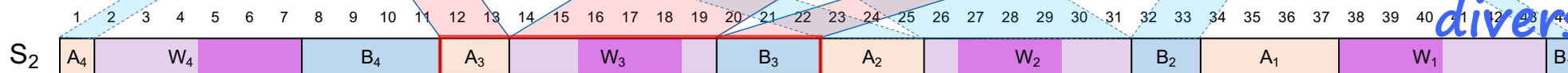
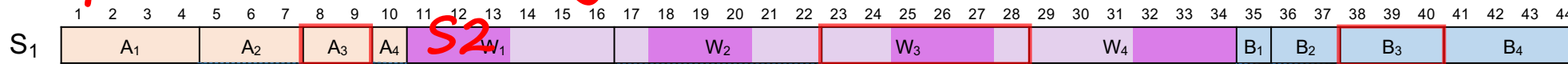


(1) Input to the string set diversity problem: A set of N strings

FPT-reduction computation



(2) Input to the LCS problem: Two strings S_1 and S_2



(4) Solution to the Max-Sum LCS diversity problem



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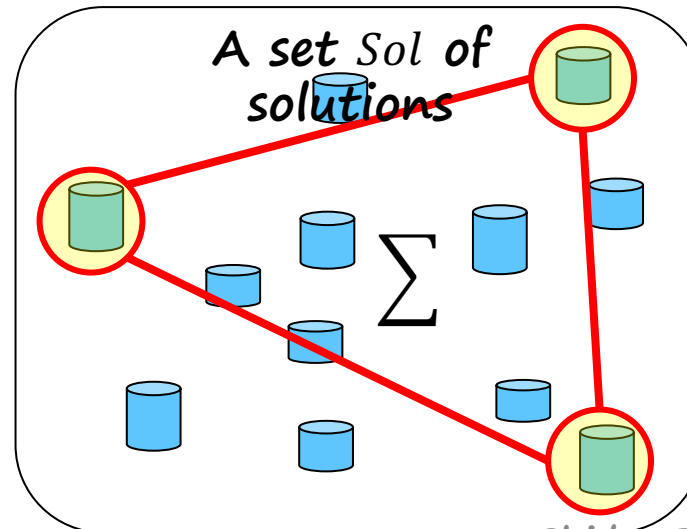
Result 3: When #solutions K is unbounded.

Approximability

- **Theorem 13:** When K is an input, **the Max-Sum diversity problem for LCSs has a PTAS.** (i.e., for any constant $\epsilon > 0$, it is $(1 - \epsilon)$ -approximable in polynomial time).
- **Note:** The same result **does not hold for the Max-Min variant.**

Sum-Diversity:

$$D_d^{\text{sum}}(X) := \sum_{i < j} d(x_i, x_j),$$



**Select
 $K = 3$
solutions**



Theorem 13 follows from Theorem 10, Lemma 11, and Lemma 12 below.

- Theorem 10 (Cevallos, Eisenbrand, Zenklusen et al., 2019):
When K is an input, **the Max-Sum version of the maximum diversity problem for a point set has a PTAS** if the condition (1) and (2) hold:
 - The distance function satisfies the **"negative type inequality"** (NEG).
 - The **"farthest point problem"** is polynomial time solvable.

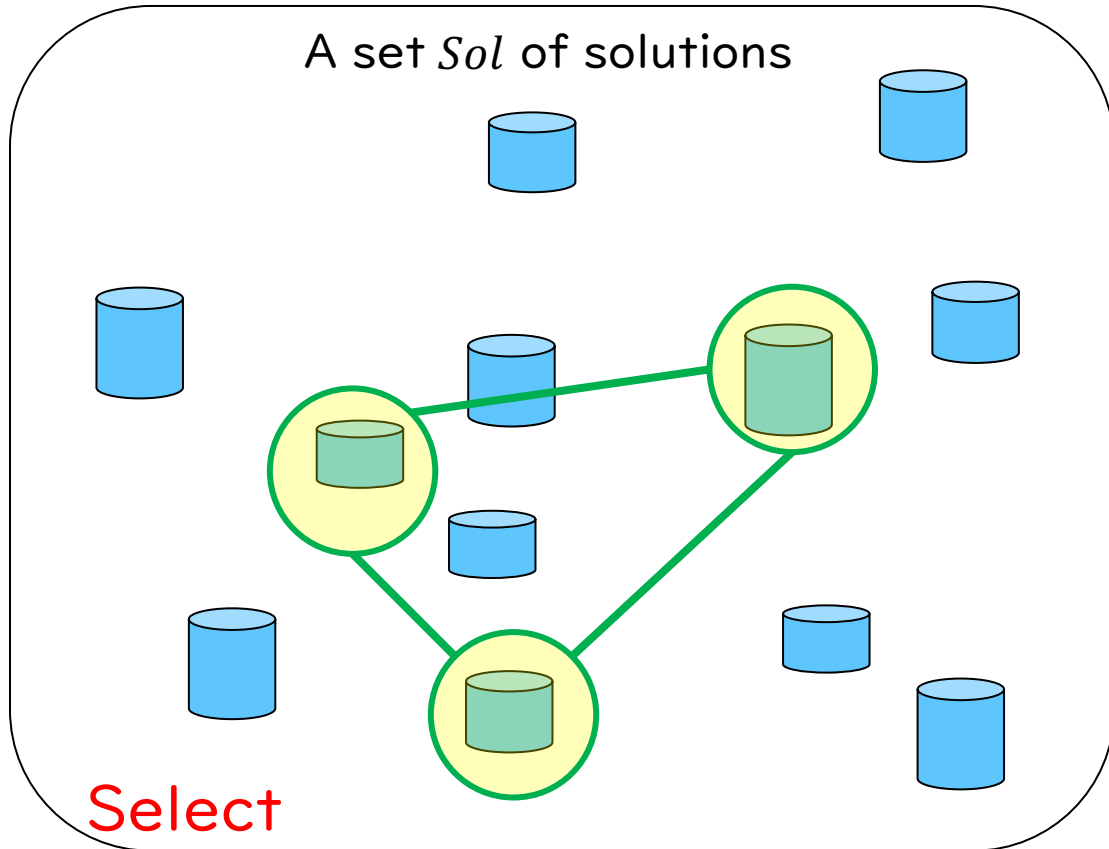
Our case:

- A point = an LCS
- A distance function = Hamming distance

- **Lemma 11:** The Hamming distance between strings is a quasi-metric that satisfies the "negative type inequality".

(Proof: Hamming distance on strings over Σ with length L can be embedded into L_1 -metric over bitvectors of length $|\Sigma| \cdot L$. Since (i) L_1 -metric satisfies NEG, and (ii) NEG is closed under positive linear combinations (i.e., a cone), Lemma follows.)

- **Lemma 12:** The farthest point problem for LCS can be solved in polynomial time.
(Proof: In the DP-algorithm of Thm.6, fix $K-1$ strings, and search for the rest one.)

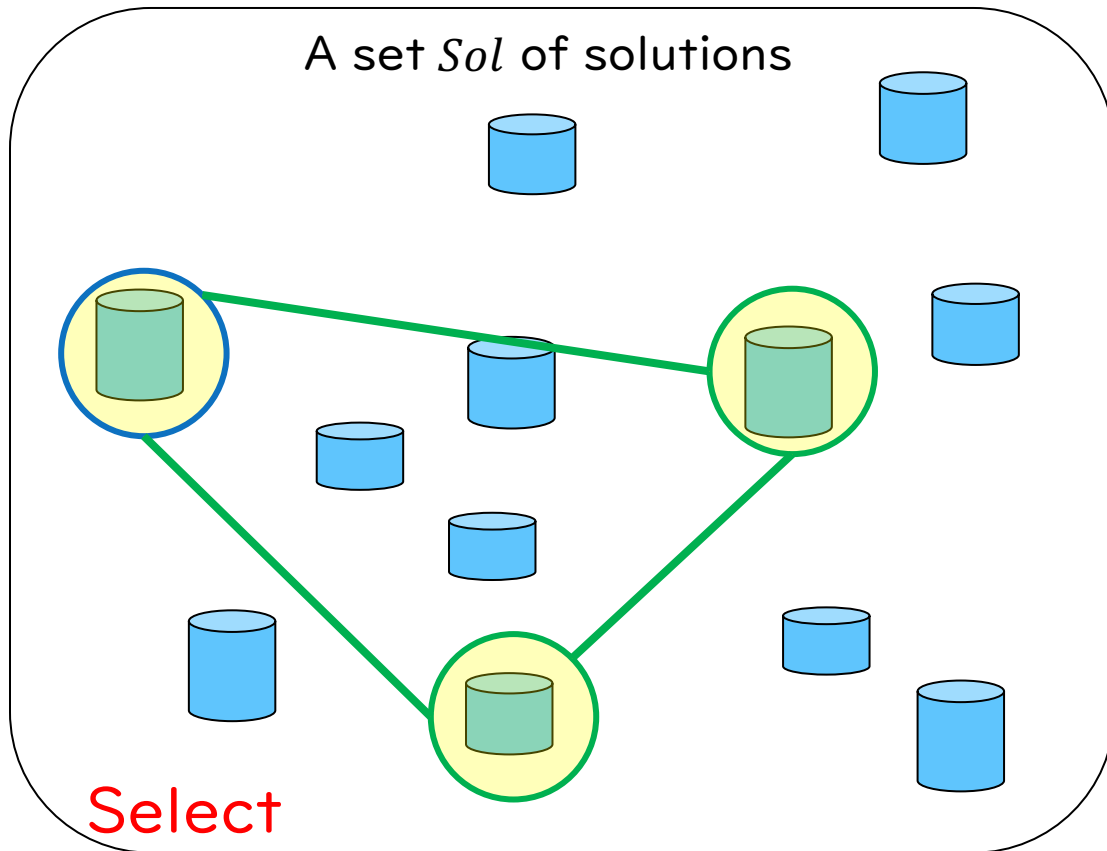


Select
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- Use the local search method below

Local search:

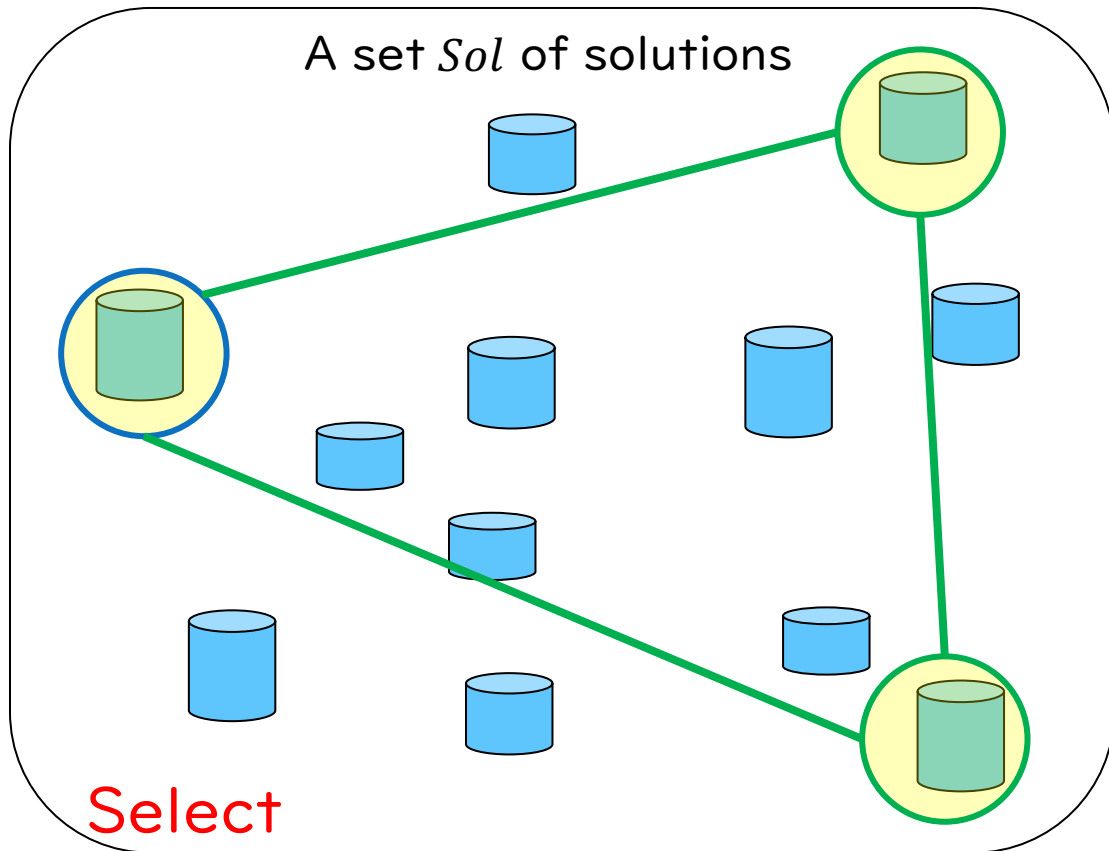
- Initially, select any solution set $S = \{X_1, \dots, X_K\}$.
- Repeat the following process $O(K^2 \log K)$ times as long as the updated diversity $D(S - \{X\} \cup \{Y\})$ increases:
 - Select a solution X from S
 - Replace X with a new solution Y
- Return the solution set S .



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Local search:

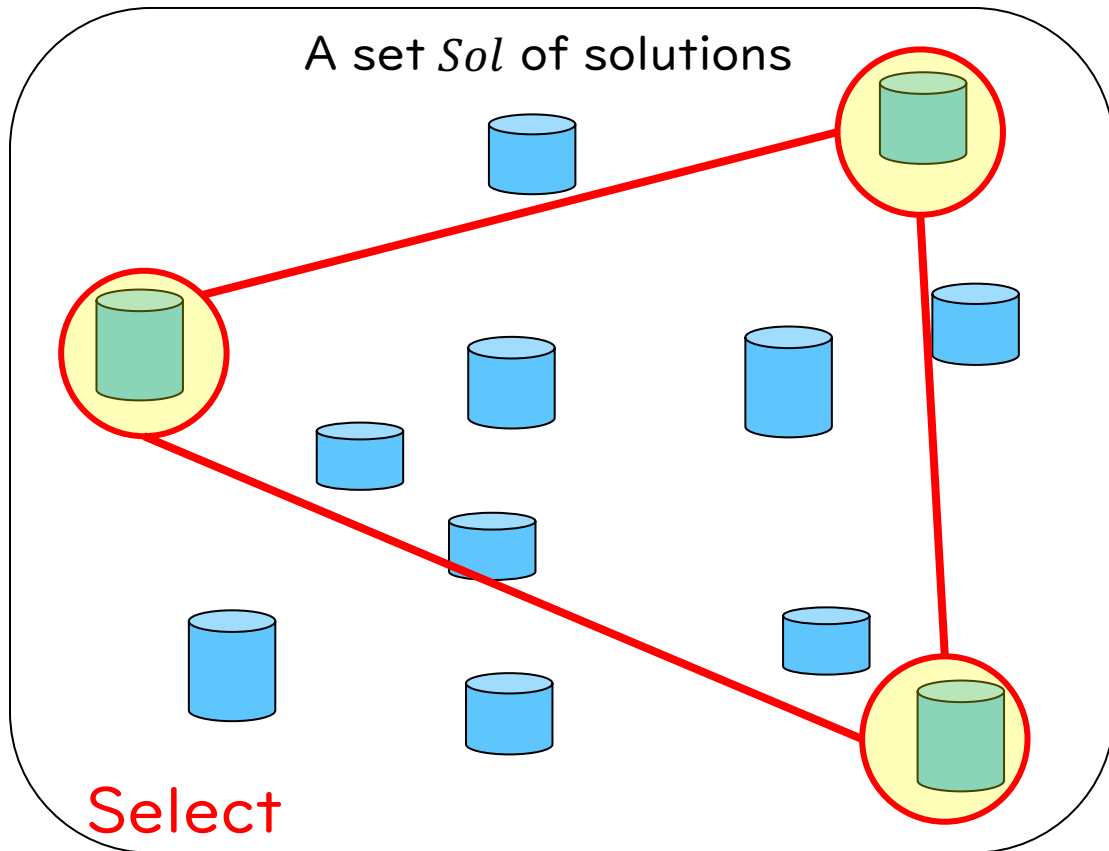
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Result 3: When #solutions K is unbounded. Approximability

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- **Note:** The same result does not hold for the Max-Min variant.



Summary: Complexity of Max. Diverse LCSs

Notes: All results below except PTAS hold for the Min-Sum(LCS) & MinSum(StringSet)
Assumption: the number m of input strings is always constant, throughout.

When K is bounded

When K is input

Parameterized
by K and r
FPT

Proof: Color coding

(This part will be
skipped due to time
constraint)

K : #solutions to select, r : max. length of LCS

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Parameterized
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W[1]-hard

Proof: FPT-reduction
from p -CLIQUE

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Problem	Type	<i>bounded K</i>		<i>unbounded K</i>	
		$K: \text{const}$	$K: \text{param}$	$K: \text{input}$	
MAX-SUM DIVERSE STRING & LCS	Exact	Poly-Time (Theorem 3.2)	W[1]-hard on Σ -DAG (Theorem 6.2)) <i>Intractable</i>	NP-hard on Σ -DAG if $r \geq 3:\text{const}$ (Theorem 6.1)	
	Approx. or FPT	—	FPT if $r: \text{param}$ (Theorem 5.2)	PTAS <i>Approximable</i> (Theorem 4.2)	
MAX-MIN DIVERSE STRING & LCS	Exact	Poly-Time (Theorem 3.1)	W[1]-hard on Σ -DAG (Theorem 6.2)	NP-hard on Σ -DAG if $r \geq 3:\text{const}$ (Theorem 6.1)	
	Approx. or FPT	—	FPT if $r: \text{param}$ (Theorem 5.1)	Open	



■ Diversity Maximization problem for LCS and a Σ -DAG

- w/ Sum- and Min-diversities (for the first time)

■ Investigated the complexity in various settings cases:

- Bounded $K \Rightarrow$ Tractable (PTIME)
- Unbounded $K \Rightarrow$ Intractable (NP-hard); Approximable (PTAS) for Max-Sum
- Parameterized by K only \Rightarrow Param. Intractable (W[1]-hard)



■ Not surprising results. However, we required to establish:

- Negative-typeness of String Hamming distance in $O(L\sigma)$ dimension (seems new)
- Color-coding technique for automata (NFAs or Σ -DAGs)

K : #solutions,
 r : max. length
of input strings

Future work

- Approximability of Max-Min(LCS) in the case of unbounded K
- Max-Sum(MCS)/Max-Min(MCS) for **MCS (maximal common subsequences) under Edit Distance**. (Recently, it was independently shown: m -MCSs have a DAG representation of poly-size with $m=2$ [Punzi, Grossi, Uno, ISAAC'23], [Hirota & Sakai, arXiv'23], **while it is open for any $m \geq 3$.**)



After the start up meeting, Sapporo,
Nov. 2024



Yuto Shida¹



Giulia Punzi^{2,3}



Yasuaki Kobayashi¹



Takeaki Uno²



Hiroki Arimura¹

Thank you!
And Question?



1) Hokkaido
University,
Sapporo, Japan

2) National Inst.
Informatics,
Tokyo, Japan

3) University of Pisa,
Pisa, Italy

Slide pdf will be found at "Code section" of this paper's arXiv site: <https://arxiv.org/abs/2405.00131>