

From Strings to Seaweeds

Modern Tools for Classical Problems

Philip Wellnitz
National Institute of Informatics
Tokyo

An Example

** ** ** ** **
 * * * * * (TU Kaiserslautern).
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 * * * * * (Simons Institute and UC Berkeley),
 * * * * * (University of Salzburg), * * * * *
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 * * * * * (University of Pennsylvania),
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 (Reichman University, Herzliya, Israel and Birkbeck,
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 SIC, Saarbrücken, Germany), * * * * * (Max
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 * * * * * (Universit  Paris Cit ,
 CNRS, IRIF, F-75013, Paris, France); * * * * *
 * * * * * (Universitat Polit cnica de Catalunya)

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(Alibaba Group); * * * * * (Microsoft Research); * * * * * (Washington University in St. Louis); * * * * * (Columbia University); * * * * * (University of Chicago)

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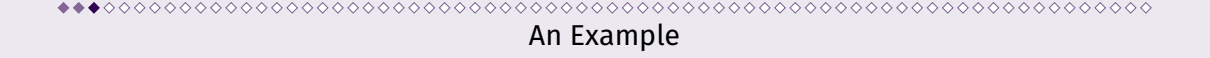
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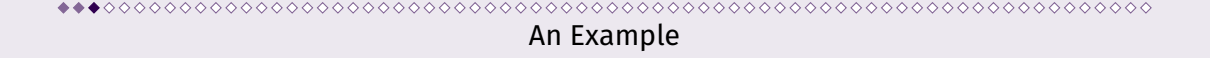
An Example

Task: Find **Saarbrücken** in a text.

An Example

Task: Find **Saarbrücken** in a text.

Or Saarbruecken.



An Example

Task: Find **Saarbrücken** in a text.

Or Saarbruecken. Or Sarrebruck.

An Example

Task: Find **Saarbrücken** in a text.

Or Saarbruecken. Or Sarrebruck. Or Saarbrücken, Saarbrücken, Saarbrücken,

The Approximate String Matching Problem

Approximate String Matching

early 1980's

Given a text T , a pattern string P , and an integer k , identify the (starting positions of) substrings of T that are at **edit distance** of at most k to P .

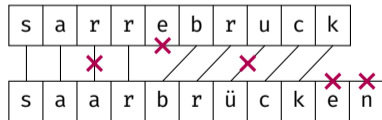
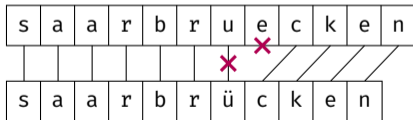
The Approximate String Matching Problem

Approximate String Matching

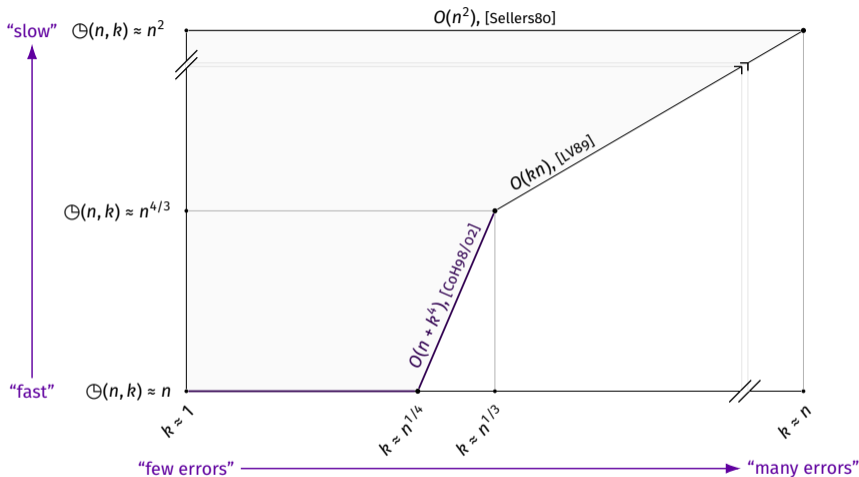
early 1980's

Given a text T , a pattern string P , and an integer k , identify the (starting positions of) substrings of T that are at **edit distance** of at most k to P .

Edit distance: minimum number of insertions, deletions, or substitutions of single characters to transform one string into another string

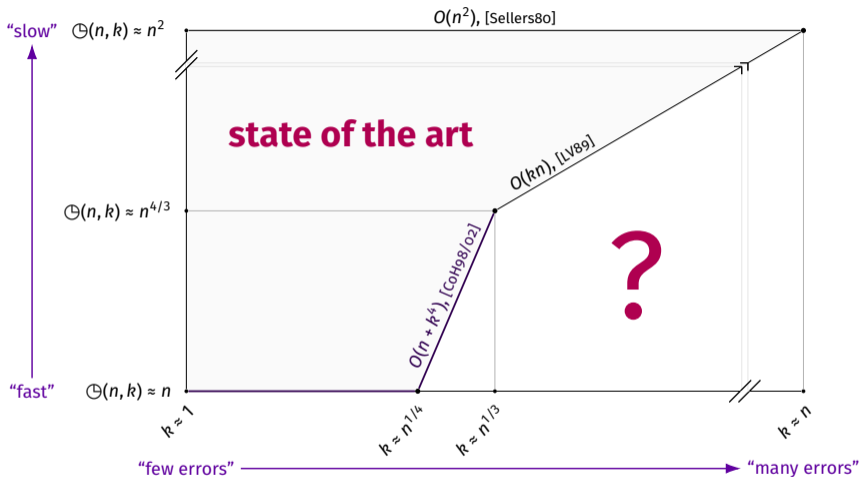


State-of-the-Art Algorithms



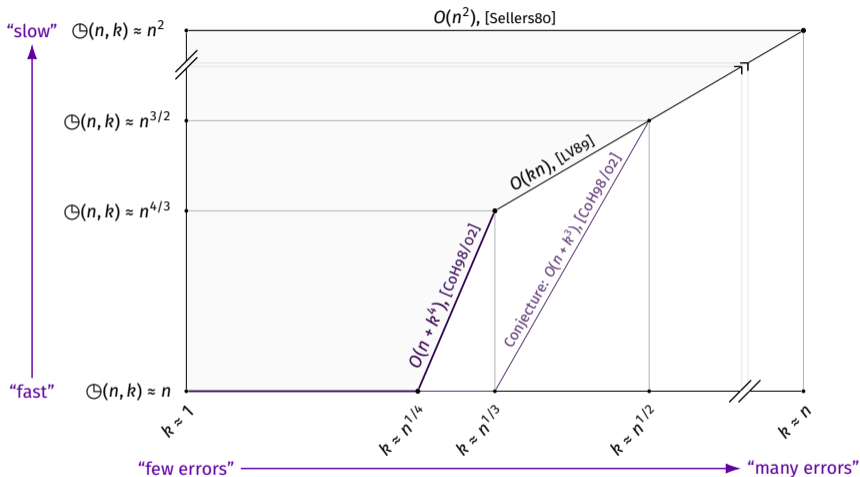
Existing algorithms for the case $n \approx t \approx p$

State-of-the-Art Algorithms



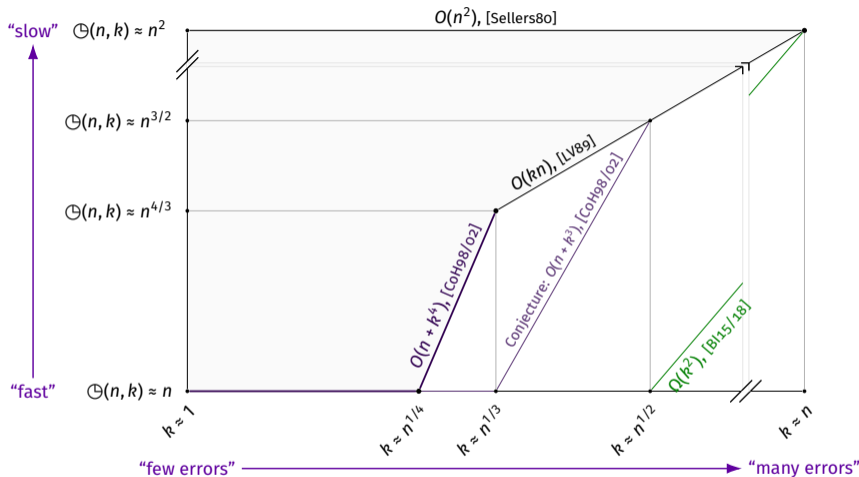
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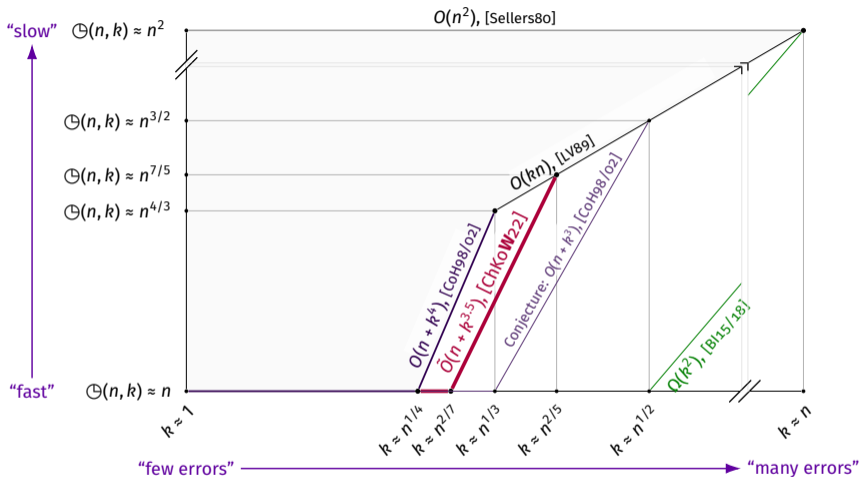
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State-of-the-Art Algorithms



Existing algorithms for the case $n \approx t \approx p$

Intro

- Some Background and Basic Definitions

- (Some) Classical Algorithms for Approximate String Matching

Structural Insights into the Solution Structure

- The Marking Trick and How to Handle Easy Patterns

- Dynamic Puzzle Matching and How to Handle Not So Easy Patterns

On Puzzles and Seaweeds

- Some Technical Definitions

- Detour: On the Name of Seaweeds

- Alignment Graphs and Permutation Matrices of Strings

- Solving Dynamic Puzzle Matching via the Seaweed Method

Outlook: Pattern Matching with Weighted Edits and Open Problems

From Edit Distance to Approximate String Matching

Approximate String Matching

For a text T , a pattern P , and an integer k ,
Identify the (starting positions of) substrings of T that are at **edit distance** of at most k to P .

- ◆ Naive idea: Compute edit distance to pattern starting from any position in the text
 \rightsquigarrow Need to compute edit distance between strings

From Edit Distance to Approximate String Matching

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From Edit Distance to Approximate String Matching

Edit Distance (ED)

For two strings T and P , compute their edit distance
(minimum number of insertions, deletions, and substitutions to transform T into P).

- ◆ Textbook dynamic programming algorithm from around 1970, runs in time $O(tp)$
(writing $t = |T|$, $p = |P|$)

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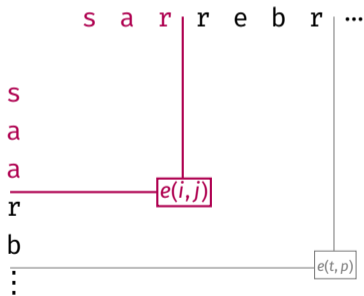
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(writing $t = |T|$, $p = |P|$)
- ◆ Write $e(i, j)$ for the ED of $T[0..i)$ and $P[0..j)$
 $\rightsquigarrow e(t, p)$ is the ED of T and P .

- ◆ Clearly, $e(i, 0) = i$ (delete $T[0..i)$)
and $e(0, j) = j$ (insert $P[0..j)$)
- ◆ Observe

$$e(i, j) = \min \begin{cases} e(i-1, j) + 1 & \text{(delete from } T) \\ e(i, j-1) + 1 & \text{(insert in } T) \\ e(i-1, j-1) + (T[i] \neq P[j]) & \text{(match/subst)} \end{cases}$$



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	s	a	r	r	e	b	r	...
s	0	1	2	3	4	5	6	7
a	1							
a	2							
r	3							
b	4							
⋮	5							

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		s	a	r	r	e	b	r	...
		0	1	2	3	4	5	6	7
s	1	0	1	2	3	4	5	6	
a	2	1	0	1	2	3	4	5	
a	3	2	1	1					
r	4								
b	5								
⋮									

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a	2	1	0	1	2	3	4	5	
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		0	1	2	3	4	5	6	7
s	1	0	1	2	3	4	5	6	
a	2	1	0	1	2	3	4	5	
a	3	2	1	1	2				
r	4								
b	5								
⋮									

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a	2	1	0	1	2	3	4	5
a	3	2	1	1	2	3	4	5
r	4	3	2	1	1	2	3	4
b	5	4	3	2	2	2	2	3
⋮								

From Edit Distance to Approximate String Matching

Approximate String Matching

For a text T , a pattern P , and an integer k ,
Identify the (starting positions of) substrings of T that are at edit distance of at most k to P .

- ◆ Textbook dynamic programming algorithm from ~~~1970~~ 1980, runs in time $O(tp)$ (writing $t = |T|$, $p = |P|$), [Sellers 1980, rediscovered several times]
- ◆ Write $e(i, j)$ for the ED of $T[0..i]$ and $P[0..j]$
after deleting an arbitrary prefix of T
- ◆ Clearly, $e(i, 0) = 0$ (delete $T[0..i]$)
and $e(0, j) = j$ (insert $P[0..j]$)
- ◆ Observe

$$e(i, j) = \min \begin{cases} e(i-1, j) + 1 & \text{(delete from } T) \\ e(i, j-1) + 1 & \text{(insert in } T) \\ e(i-1, j-1) + (T[i] \neq P[j]) & \text{(match/subst)} \end{cases}$$

	s	a	r	r	e	b	r	...
s	0	0	0	0	0	0	0	0
a	1	0	1	1	1	1	1	1
a	2	1	0	1	2	2	2	2
r	3	2	1	1	2	3	3	3
b	4	3	2	1	1	2	3	3
⋮	5	4	3	2	2	2	2	3

From Edit Distance to Approximate String Matching

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	s	a	r	r	e	b	r	...
s	0	0	0	0	0	0	0	0
a	1	0	1	1	1	1	1	1
a	2	1	0	1	2	2	2	2
r	3	2	1	1	2	3	3	3
b	4	3	2	1	1	2	3	3
	5	4	3	2	2	2	2	3

Ending positions of saarb with $k = 2$.

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For a text T , a pattern P , and an integer k ,

Identify the (starting positions of) substrings of T that are at edit distance of at most k to P .

- ◆ Textbook dynamic programming algorithm from ~~~1970~~ 1980, runs in time $O(tp)$ (writing $t = |T|$, $p = |P|$), [Sellers 1980, rediscovered several times]
- ◆ Write $e(i, j)$ for the ED of $T^R[0..i]$ and $P^R[0..j]$
after deleting an arbitrary prefix of T^R
- ◆ Clearly, $e(i, 0) = 0$ (delete $T[0..i]$)
and $e(0, j) = j$ (insert $P[0..j]$)
- ◆ Observe

$$e(i, j) = \min \begin{cases} e(i-1, j) + 1 & \text{(delete from } T) \\ e(i, j-1) + 1 & \text{(insert in } T) \\ e(i-1, j-1) + (T[i] \neq P[j]) & \text{(match/subst)} \end{cases}$$

	r	b	e	r	r	a	s	...
b	0	0	0	0	0	0	0	
r	1	1	0	1	1	1	1	
a	2	1	1	1	1	1	2	
a	3	2	2	2	2	2	1	
s	4	3	3	3	3	3	2	
	5	4	4	4	4	4	3	2

Starting positions of saarb with $k = 2$.

Detour: More on (Historic) Applications

Approximate String Matching

For a text T , a pattern P , and an integer k ,
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(Variations of) Sellers' $O(tp)$ Algorithm were (are?) popular for

- ◆ (Have seen) **Text Retrieval**: Find occurrences of patterns or phrases in a text, accounting for misspellings or conversion errors
- ◆ **Signal Processing**: Find patterns in signals, accounting for transmission errors
- ◆ **Computational Biology**: Find specific patterns in DNA sequences, accounting for mutations or evolutionary alterations


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Using Kangaroo Jumps for Speed-Ups

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⇒ Do not need values $e(i, j) > k$

◆ How do we compute just values $\leq k$?

⇒ Compute values diagonally.

		r	r	a	s	r	a	a	s	r	h	a	s	r	r	a	s	...
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
r	1	0	0	1	1	0	1	1	1	0	1	1	1	0	0	1	1	
a	2	1	1	0	1	1	0	1	2	1	1	1	2	1	1	0	1	
a	3	2	2	1	1	2	1	0	1	2	2	1	2	2	2	1	1	
s	4	3	3	2	1	2	2	1	0	1	2	2	1	2	3	2	1	
r	5	4	3	3	2	1	2	2	1	0	1	2	2	1	2	3	2	
a	6	5	4	3	3	2	1	2	2	1	1	1	2	2	2	2	3	
a	7	6	5	4	4	3	2	1	2	2	2	1	2	3	3	2	3	
s	8	7	6	5	4	4	3	2	1	2	3	2	1	2	3	3	2	

Finding saarsaar with $k \leq 2$ edits.

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a	1	0	0	1	1	0	1	1	1	0	1	1	1	0	0	1	1
a	2	1	1	0	1	1	0	1	2	1	1	1	2	1	1	0	1
s	3	2	2	1	1	2	1	0	1	2	2	1	2	2	2	1	1
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a	5	4	3	3	2	1	2	2	1	0	1	2	2	1	2	3	2
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a	2	1	1	0	1	1	0	1	2	1	1	1	2	1	1	0	1
s	3	2	2	1	1	2	1	0	1	2	2	1	2	2	2	1	1
r	4	3	3	2	1	2	2	1	0	1	2	2	1	2	3	2	1
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	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
r	1	0	0	1	1	0	1	1	1	0	1	1	1	0	0	1	1	
a	2	1	1	0	1	1	0	1	2	1	1	1	2	1	1	0	1	
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r	5	4	3	3	2	1	2	2	1	0	1	2	2	1	2	3	2	
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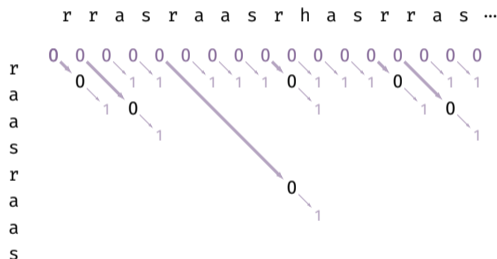
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For a text T , a pattern P , and an integer k ,
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 \rightsquigarrow Compute furthest position on diag d
 reachable with $\ell = 0, \dots, k$ edits.
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 time (Kangaroo Jumps).
 \rightsquigarrow In total: $O((t + p)k)$ time
 [Landau, Vishkin'89].



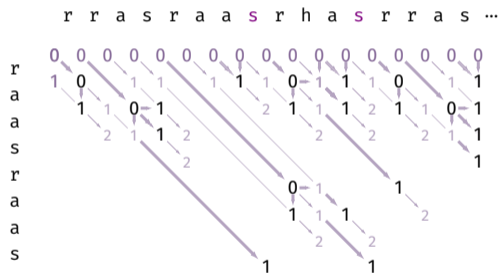
Finding saarsaar with $k \leq 0$ edits.

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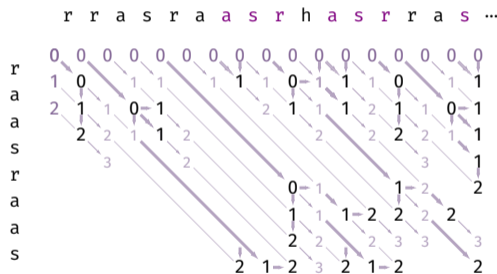
Finding saarsaar with $k \leq 1$ edits.

Using Kangaroo Jumps for Speed-Ups

Approximate String Matching

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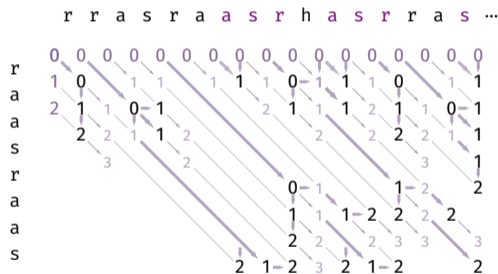
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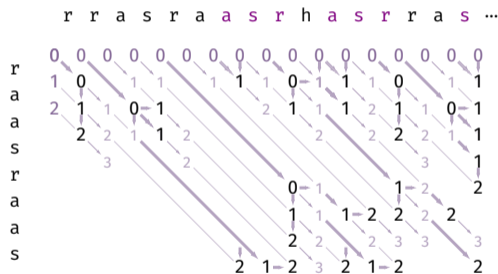
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Finding saarsaar with $k \leq 2$ edits.



Beating Landau and Vishkin's Algorithm

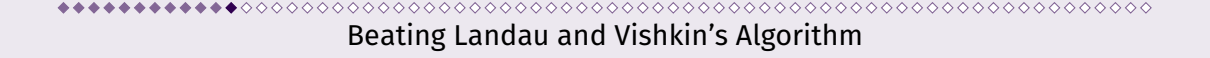
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...



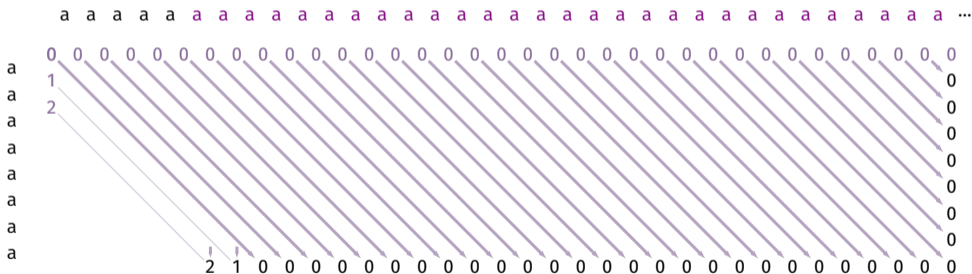


Beating Landau and Vishkin's Algorithm

Idea: Can we **filter** out diagonals that will never lead to an occurrence?

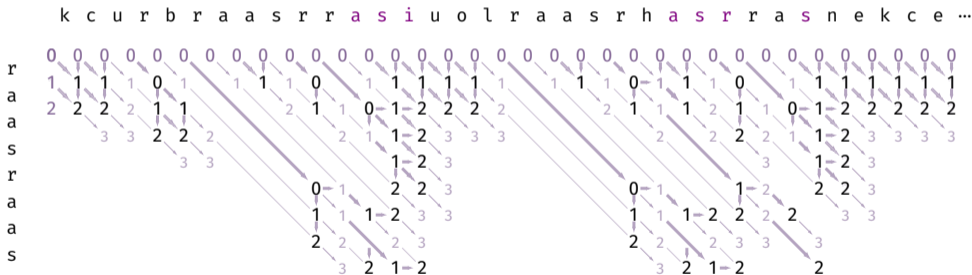


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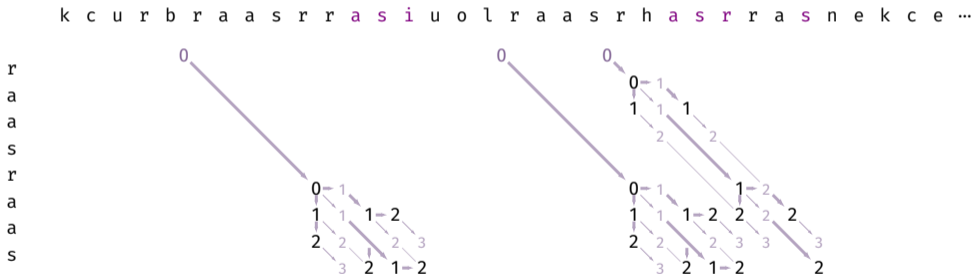
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Sometimes we can; and if we cannot, there might be structure to exploit!



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Unstructured Parts of the Pattern Help!

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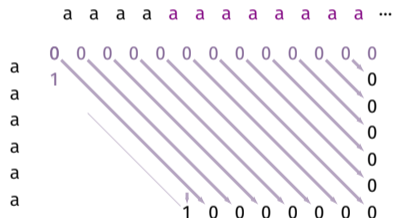
- ◆ Idea: Find parts of P that are **rare** in T
 - ↪ Can filter out candidates for occurrences
- ◆ Seen: Highly repetitive parts R are bad...
 - ↪ R may appear $\approx t$ times in T
- ◆ Non-repetitive parts B are good!
 - ↪ B may appear only $\approx t/b$ times in T
 - ↪ Problem: B may appear approximately in T

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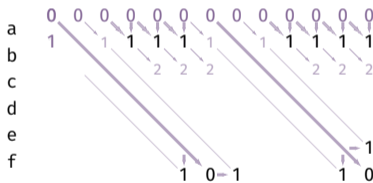
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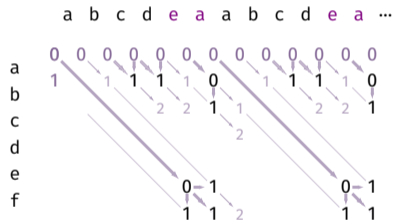


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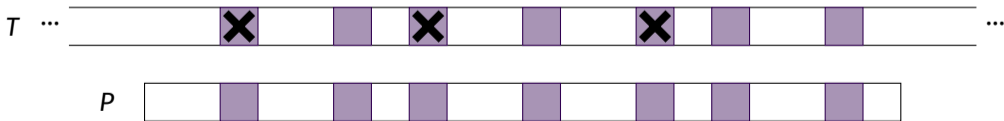


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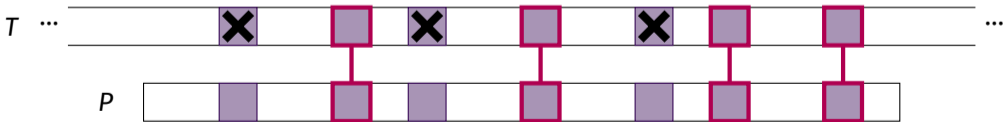


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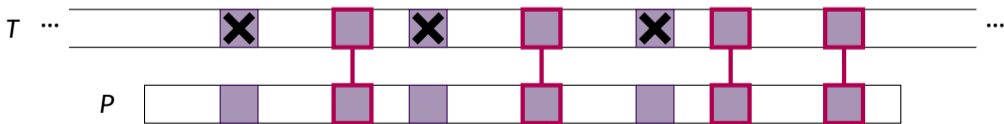


Unstructured Parts of the Pattern Help!

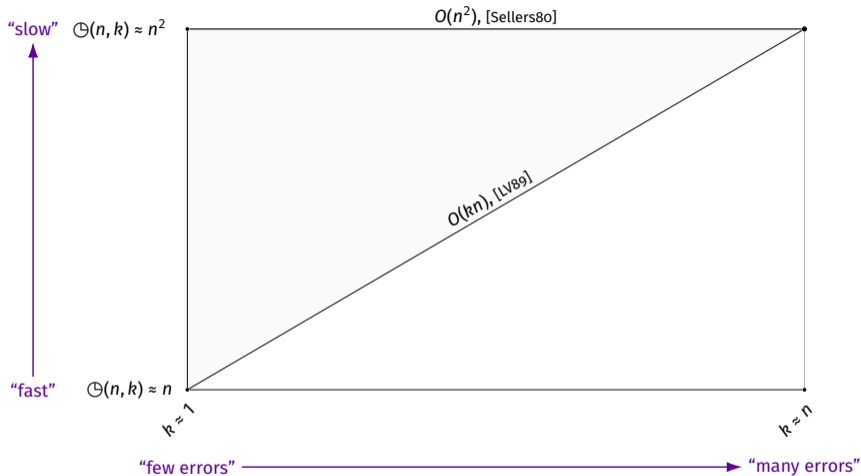
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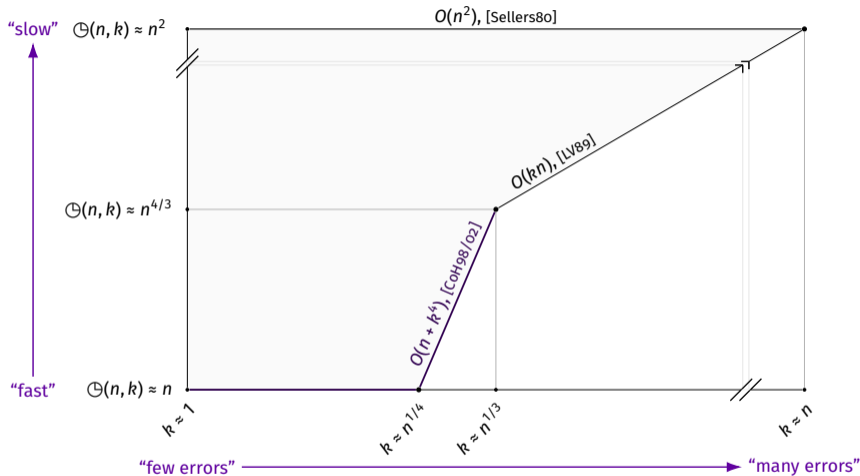


Recap: State-of-the-Art Algorithms



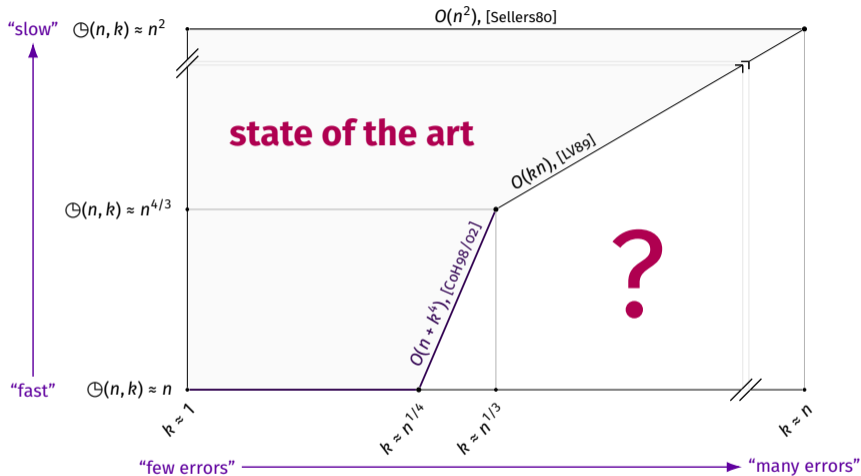
Existing algorithms for the case $n \approx t \approx p$

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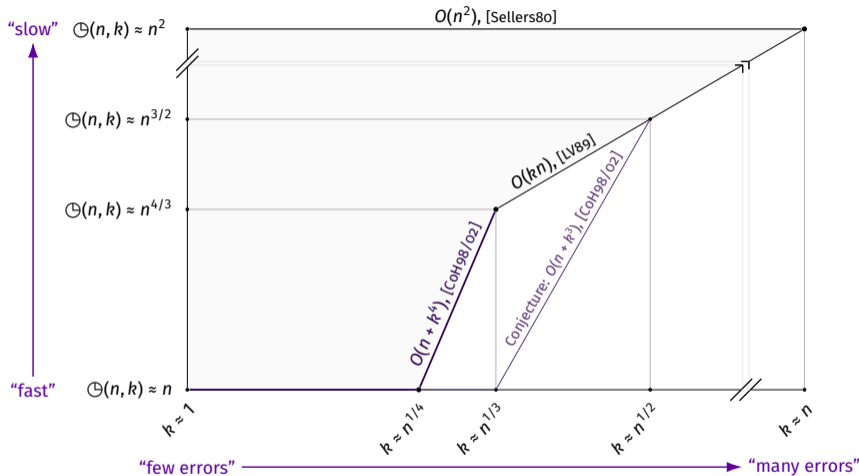
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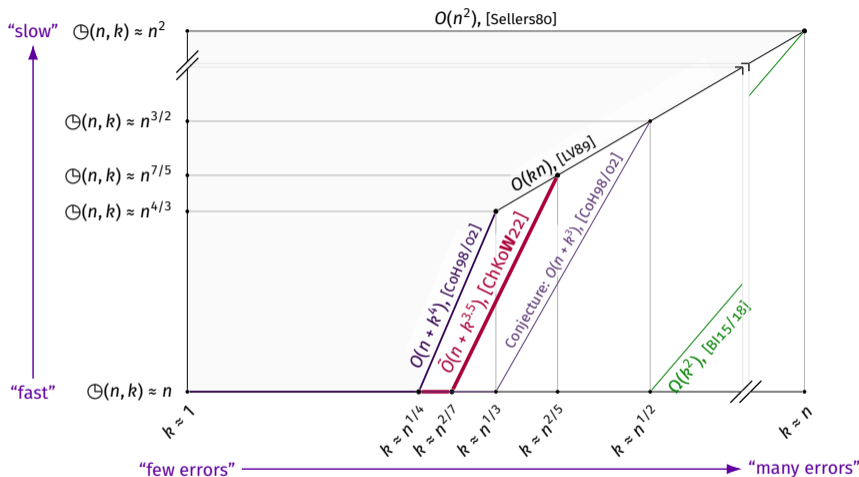
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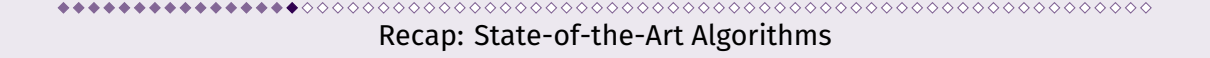


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Why did a new improvement take 24 years?

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[Bl18] (ann. 2014)

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(Unless there is a major breakthrough for the Satisfiability Problem.)

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⇒ Yields lower bound of $O(t + k^2 \cdot t/p)$ for Approximate String Matching.

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 - ↪ $O(tk)$ algorithm [Landau, Vishkin'86]
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Optimal String Matching with Mismatches

[GU18] (ann. 2017)

There is a $\tilde{O}(t + kt/\sqrt{p})$ -time algorithm for String Matching with Mismatches and no significantly faster algorithm exists.

(Unless there is a major breakthrough for combinatorial Boolean Matrix Multiplication.)

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- ◆ Example 2: String Matching with Mismatches (no insertions or deletions of chars)
 - ↪ $O(tk)$ algorithm [Landau, Vishkin'86]
 - ↪ $\tilde{O}(t\sqrt{k})$ algorithm [Amir, Lewenstein, Porat'04]
 - ↪ $\tilde{O}(t + t/p \cdot k^2)$ algorithm [CFPSS'16]

Optimal String Matching with Mismatches

[GU18] (ann. 2017)

There is a $\tilde{O}(t + kt/\sqrt{p})$ -time algorithm for String Matching with Mismatches and no significantly faster algorithm exists.

(Unless there is a major breakthrough for combinatorial Boolean Matrix Multiplication.)

↪ String Matching with Mismatches was “fully” understood only in 2017.

How do we obtain faster algorithms?

New insights into the solution structure of
Approximate String Matching

How do we obtain faster algorithms?

New insights into the solution structure of
Approximate String Matching



Step 0:

What is the solution structure of
Exact String Matching?

The Solution Structure of Exact String Matching

(The) Period of a String

$p > 0$ is a **period** of a string S if $S[i] = S[i + p]$ for all $i = 1, \dots, |S| - p$.

The **period** of S : smallest period of S , write $\text{per}(S)$.

S is **periodic**: $\text{per}(S) \leq |S|/2$



$S = a \ b \ c \ a \ b \ c \ a \ b \ c \ a \ b$

$$\text{per}(S) = 3$$

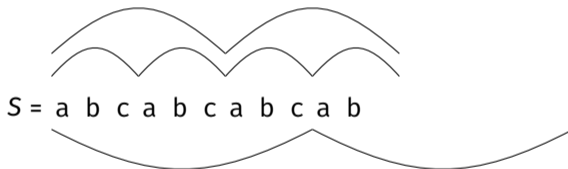
The Solution Structure of Exact String Matching

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$$\text{per}(S) = 3$$

6 and 9 also periods of S

The Solution Structure of Exact String Matching

“Periodicity Lemma”

(Folklore)

Text T , pattern P with $|T| \leq \frac{3}{2} |P|$; one of the following holds

- ◆ P appears ≤ 1 times in T .
- ◆ P (and the relevant part of T) are **periodic** with some period Q .

P



T



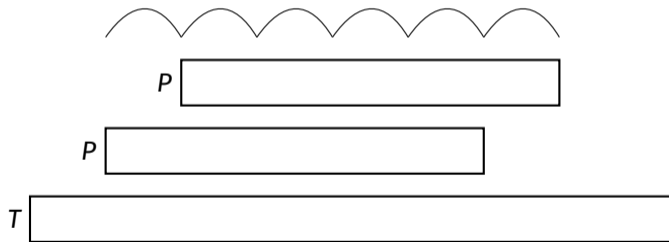
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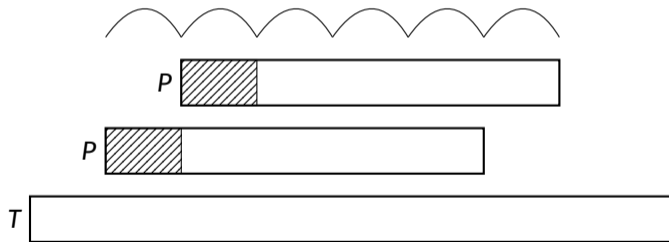
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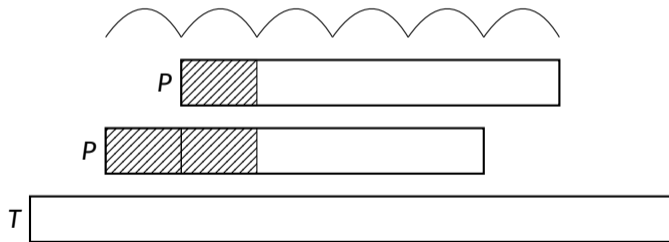
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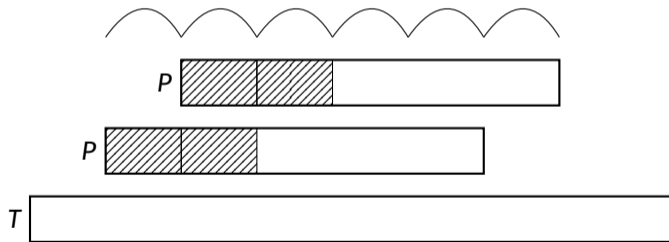
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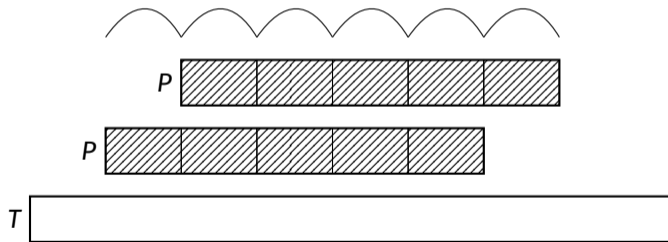
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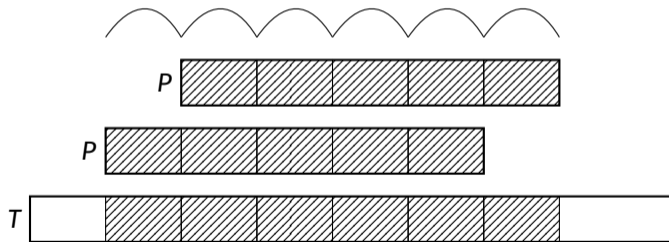
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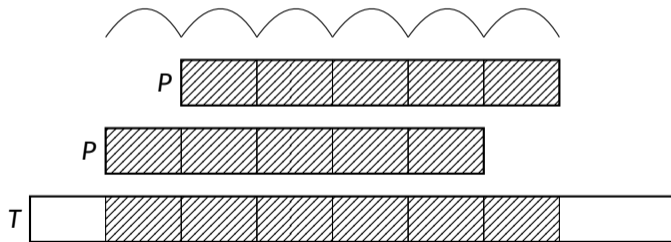
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The Standard Trick: For $|T| \gg |P|$ consider separately $O(n/m)$ fragments of T of length $\leq \frac{3}{2} m$ that overlap by $m - 1$ positions.

Step 1:

What is the solution structure of
String Matching with Mismatches?

Structural Results for Approximate String Matching

Periodicity Lemma

(Folklore)

Text T , pattern P with $t \leq \frac{3}{2}p$; P appears ≤ 1 times in T or P is periodic with some period Q .

Main Result (Mismatches)

[BKüW'19; CKoW'20]

Text T , pattern P with $t \leq \frac{3}{2}p$; threshold k , one of the following holds

P appears $\leq O(k)$ times in T or P is almost periodic with some period Q

T

a	...	a	...	a	c	...	c	...	c
---	-----	---	-----	---	---	-----	---	-----	---

P

a	...	a	c	...	c
---	-----	---	---	-----	---

$a \cdots a c \cdots c$ appears $2k$ times

T

a	...	a	c	a	a	a	...	a	c
---	-----	---	---	---	---	---	-----	---	---

P

a	a	...	a	c	a
---	---	-----	---	---	---

$aa \cdots aca$ has approximate period a
(is at HD $\leq 2k$ from Q^∞)

Structural Results for Approximate String Matching

Periodicity Lemma

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T

a	...	a	...	a	c	...	c	...	c
---	-----	---	-----	---	---	-----	---	-----	---

P

a	...	a	c	...	c
---	-----	---	---	-----	---

$a \cdots a c \cdots c$ appears $2k$ times

T

a	...	a	c	a	a	a	...	a	c
---	-----	---	---	---	---	---	-----	---	---

P

a	a	...	a	c	a
---	---	-----	---	---	---

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Structural Results for Approximate String Matching

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T

a	...	a	...	a	c	...	c	...	c
---	-----	---	-----	---	---	-----	---	-----	---

P

a	...	a	c	...	c
---	-----	---	---	-----	---

$a \cdots a c \cdots c$ appears $2k$ times

T

a	...	a	c	a	a	a	...	a	c
---	-----	---	---	---	---	---	-----	---	---

P

a	a	...	a	c	a
---	---	-----	---	---	---

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Structural Results for Approximate String Matching

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P appears $\leq O(k)$ times in T or P is **almost periodic** with some period Q

T

a	...	a	...	a
---	-----	---	-----	---

c	...	c	...	c
---	-----	---	-----	---

P

a	...	a
---	-----	---

c	...	c
---	-----	---

$a \cdots a c \cdots c$ appears $2k$ times

T

a	...	a	c	a	a	a	...	a	c
---	-----	---	---	---	---	---	-----	---	---

P

a	a	...	a	c	a
---	---	-----	---	---	---

$aa \cdots aca$ has approximate period a
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Step 1.5:

The solution structure of
Approximate String Matching.

Structural Results for Approximate String Matching

Main Result (Mismatches)

[BKüW'19; CKoW'20]

T, P with $t \leq \frac{3}{2}p$; threshold k , P appears $O(k)$ times in T or P is almost periodic with some period Q

Main Result (Edits)

[CKoW'20]

Text T , pattern P with $t \leq \frac{3}{2}p$; threshold k , one of the following holds

P appears $\leq O(k^2)$ times in T or P is almost periodic with some period Q

Structural Results for Approximate String Matching

Main Result (Mismatches)

[BKüW'19; CKoW'20]

T, P with $t \leq \frac{3}{2}p$; threshold k , P appears $O(k)$ times in T or P is almost periodic with some period Q

Main Result (Edits)

[CKoW'20]

Text T , pattern P with $t \leq \frac{3}{2}p$; threshold k , one of the following holds

P appears $\leq O(k^2)$ times in T or P is almost periodic with some period Q

T a a ... a a a ... a a a ... a c a ... a ... c a ... a c a ... a

P a a ... a a a ... a c a ... a ... c a ... a

aa...aaa...aca...a...ca...a appears $\approx k^2$ times

Structural Results for Approximate String Matching

Main Result (Mismatches)

[BKüW'19; CKoW'20]

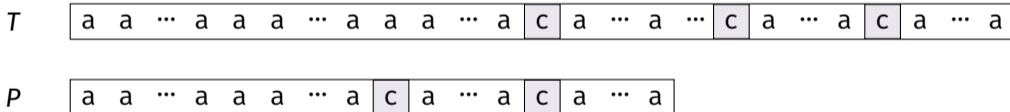
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Main Result (Edits)

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Structural Results for Approximate String Matching

Main Result (Mismatches)

[BKüW'19; CKoW'20]

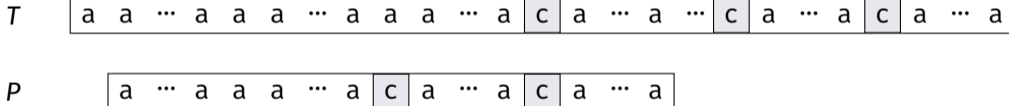
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Structural Results for Approximate String Matching

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T, P with $t \leq \frac{3}{2}p$; threshold k , P appears $O(k^2)$ times in T or P is almost periodic with some period Q

[CKoW'20]

Key Intermediate Result (Analyze)

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
- ◆ P has disjoint **repetitive regions** that cover $\frac{3}{8}P$
- ◆ P is almost periodic.

c*o*o*o*a*o*o*o*a

*a*a*a*a*a*c*c*c*c*

aaacaaaaaaaaaacaaaaaa

Structural Results for Approximate String Matching

[CKoW'20]

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c*****a*****a

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aaaaaccacc**

◆ P is almost periodic.

aaacaaaaaaaaacaaaaaa

Structural Results for Approximate String Matching

[CKoW'20]

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c*~*~*~*~*a*~*~*~*~*a

~~*~*~*~*~*~*~*~*~*~*~*

aaacaaaaaaaaaacaaaaaaaa

Structural Results for Approximate String Matching

[CKoW'20]

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c*~*~*~*~*a*~*~*~*~*a

~~*aaaaa*~*~*ccaccc*~*~*

aaacaaaaaaaaaacaaaaaaaa

Analyze implies Main Result.

Structural Results for Approximate String Matching

[CKoW'20]

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c*****a*****a

*****cacc**

aaacaaaaaaaaacaaaaaa

P



Structural Results for Approximate String Matching

Key Intermediate Result (Analyze)

[CKoW'20]

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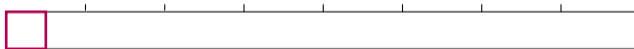
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- ◆ P is almost periodic.

c*****a*****a

aaaaacacc**

aaacaaaaaaaaacaaaaaa

P



- ◆ Process P from left to right, $p/8k$ new characters at a time.

Structural Results for Approximate String Matching

[CKoW'20]

Key Intermediate Result (Analyze)

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
- ◆ P has disjoint **repetitive regions** that cover $\frac{3}{8}P$
- ◆ P is almost periodic.

c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa

P



Breaks $B = \{ \text{red square with cross} \}$

Repetitive Regions $R = \{ \quad \}$

- ◆ If a fragment is a break, add it to the found breaks.

Structural Results for Approximate String Matching

[CKoW'20]

Key Intermediate Result (Analyze)

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
- ◆ P has disjoint **repetitive regions** that cover $\frac{3}{8}P$
- ◆ P is almost periodic.

c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa

P



Breaks $B = \{ \text{*****} \}$

Repetitive Regions $R = \{ \text{*****} \}$

- ◆ Otherwise, find the shortest prefix (longer than $p/8k$) that is a repetitive region.

Structural Results for Approximate String Matching

[CKoW'20]

Key Intermediate Result (Analyze)

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
- ◆ P has disjoint **repetitive regions** that cover $\frac{3}{8}P$
- ◆ P is almost periodic.

c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa



Breaks $B = \{ \text{[grid of asterisks]} \}$

Repetitive Regions $R = \{ \text{[cross-hatch pattern]} \}$

- ◆ Otherwise, find the shortest prefix (longer than $p/8k$) that is a repetitive region.

Structural Results for Approximate String Matching

[CKoW'20]

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Every string P satisfies at least one of

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c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa



Breaks $B = \{ \text{[purple box]}, \text{[green box]}, \text{[purple box]}, \text{[red box]} \}$

Repetitive Regions $R = \{ \text{[cross-hatch box]}, \text{[cross-hatch box]} \}$

- ◆ If we found $2k$ breaks, return the breaks.

Structural Results for Approximate String Matching

Key Intermediate Result (Analyze)

[CKoW'20]

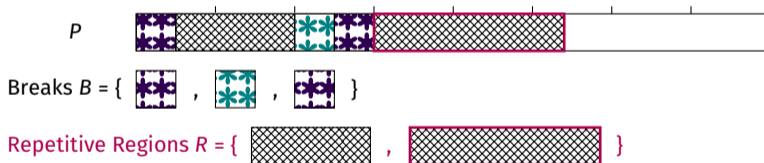
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- ◆ P is almost periodic.

c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa



- ◆ If the total length of the repetitive regions is $> \frac{3}{8} \cdot p$, return the repetitive regions.

Structural Results for Approximate String Matching

[CKoW'20]

Key Intermediate Result (Analyze)

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
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- ◆ P is almost periodic.

c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa



Breaks $B = \{ \text{purple floral}, \text{teal floral}, \text{purple floral} \}$

Repetitive Regions $R = \{ \text{black cross-hatch} \}$

- ◆ If we reach the end of P , try to find a single repetitive region starting from the end.

Structural Results for Approximate String Matching

[CKoW'20]

Key Intermediate Result (Analyze)

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
- ◆ P has disjoint **repetitive regions** that cover $\frac{3}{8}P$
- ◆ P is almost periodic.

c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa

P



Breaks $B = \{ \}$

Repetitive Regions $R = \{ \}$

- ◆ If we reach the end of P , try to find a single repetitive region starting from the end.

Structural Results for Approximate String Matching

[CKoW'20]

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- ◆ P is almost periodic.

c*****a*****a
aaaaaccacc**
aaacaaaaaaaaacaaaaaa



Breaks $B = \{ \}$

Repetitive Regions $R = \{ \text{[cross-hatched bar]} \}$

- ◆ If we found a repetitive region, return it.

Structural Results for Approximate String Matching

[CKoW'20]

Key Intermediate Result (Analyze)

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
- ◆ P has disjoint **repetitive regions** that cover $\frac{3}{8}P$
- ◆ P is almost periodic.

c*****a*****a

*****cacc*****

aaacaaaaaaaaacaaaaaa

P



Breaks $B = \{ \}$

Repetitive Regions $R = \{ \}$

- ◆ If we again don't obtain a repetitive region, P is almost periodic.

Structural Results for Approximate String Matching

Key Intermediate Result (Analyze) ✓

[CKoW'20]

Every string P satisfies at least one of

- ◆ P has $2k$ disjoint, long **breaks**
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c*~*~*~*~*a*~*~*~*~*a

~~*~*~*~*~*~*~*~*~*~*~*~*~*~*

aaacaaaaaaaaaacaaaaaa

How do we turn our insights
into faster algorithms?

How do we turn our insights into faster algorithms?

Need to tackle three cases.

- ◆ P contains $2k$ disjoint breaks;
- ◆ P contains disjoint repetitive regions R_i ;
- ◆ P is almost periodic

New Insights, Same Old Problems

How do we turn our insights into faster algorithms?

Need to tackle ~~three~~ two cases.

- ◆ P contains $2k$ disjoint breaks;
- ◆ ~~P contains disjoint repetitive regions R_i ,~~ \rightsquigarrow Follows from the other two cases.
- ◆ P is almost periodic

The Marking Trick and How to Handle Easy Patterns [Variation of CH'98]

Have: $2k$ disjoint breaks B_1, \dots, B_{2k} in P such that

◆ $|B_i| = \Theta(|P|/k)$ and ◆ $|B_i|$ is not periodic

◆ In any k -edit occ, at least 1 break is matched exactly.

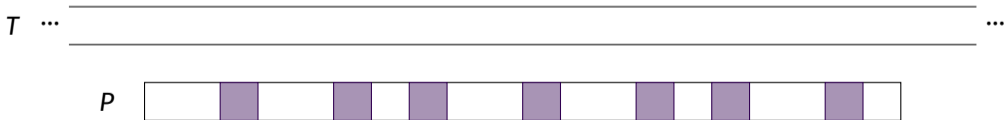


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- ◆ Algorithm idea: For each B_i , find **exact** matches of B_i in T

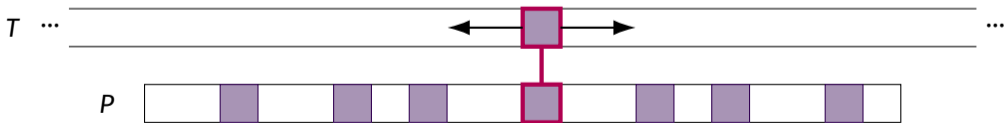


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- ◆ Algorithm idea: For each B_i , find **exact** matches of B_i in T ,
Try to extend each exact match into an occ using [LV'89]

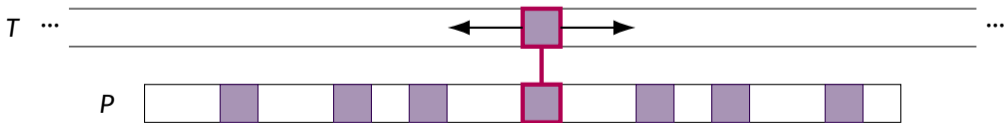


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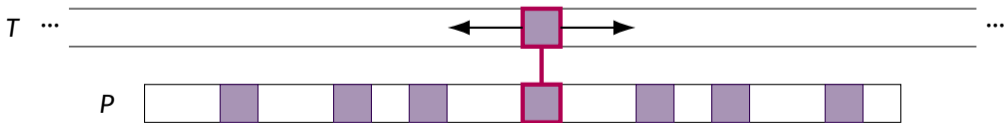


The Marking Trick and How to Handle Easy Patterns [Variation of CH'98]

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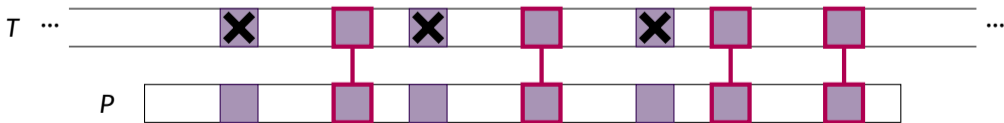
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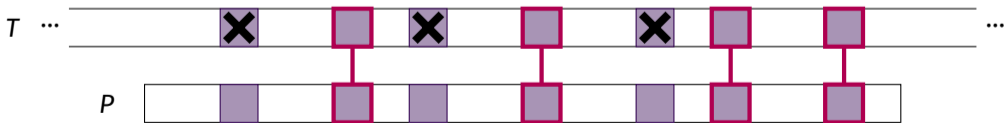
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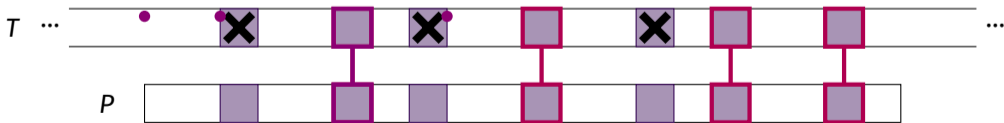
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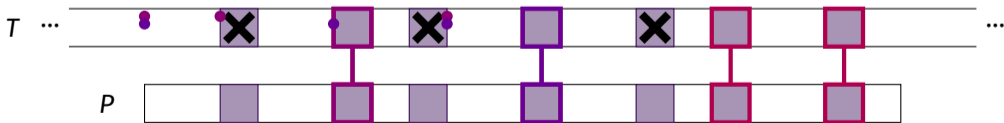
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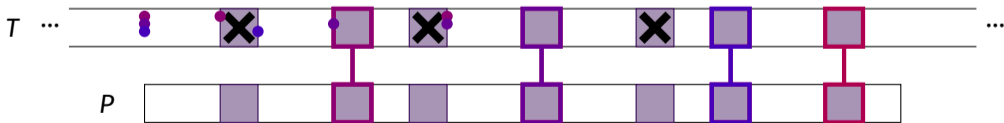


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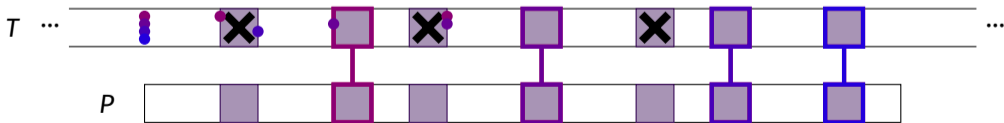
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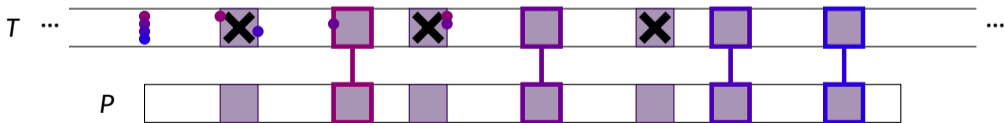
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How do we turn our insights into faster algorithms?

Need to tackle ~~three~~ **one** case.

- ◆ ~~P contains $2k$ disjoint breaks;~~ \rightsquigarrow **Adaption of [Cole,Hariharan'98] yields $O(|T| + |T|/|P| \cdot k^3)$.**
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Reduction to Periodic Patterns

[CKoW'22]

Algorithm for almost periodic case in time $\tilde{O}(|T| + k^a \cdot |T|/|P|)$, for $a \geq 3$
 \implies Algorithm for general case in time $\tilde{O}(|T| + k^a \cdot |T|/|P|)$

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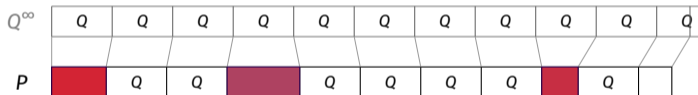
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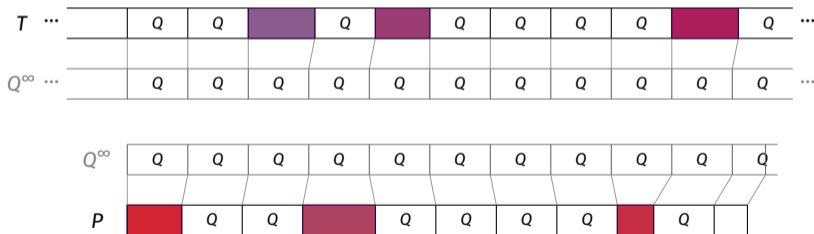
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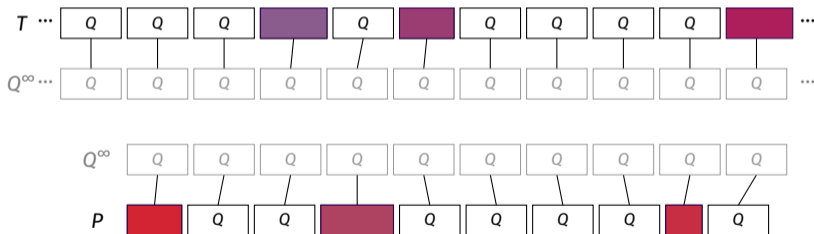
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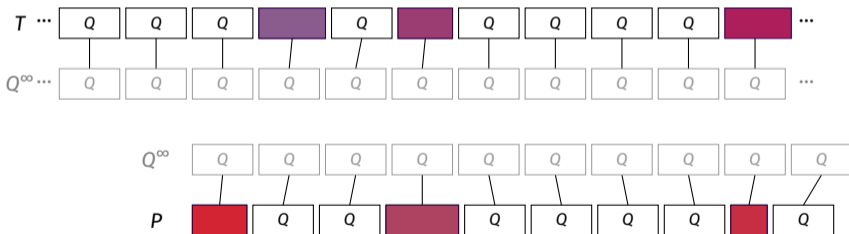
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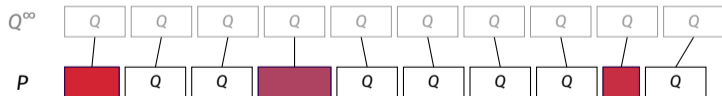
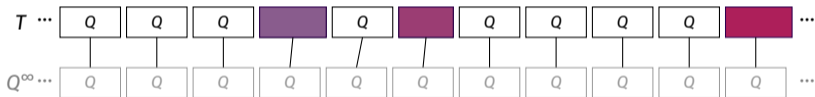
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Main Result (using Seaweeds)

[CKoW'22]

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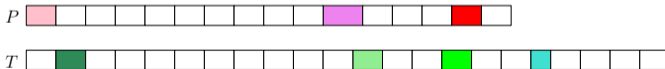
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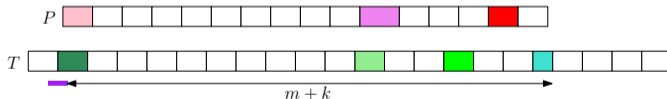
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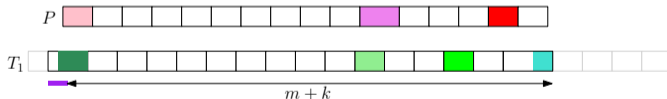
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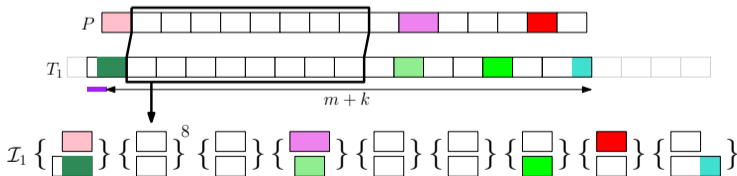
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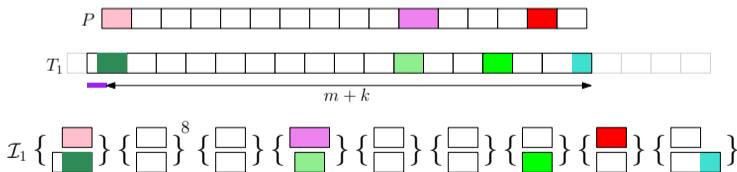
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- ◆ Goal: Iterate over all I_j 's with one DPM instance



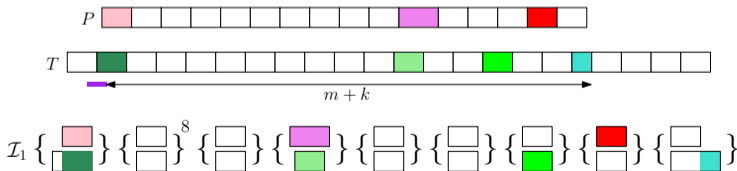
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[CKoW'22]

Input: integer k , family F of strings w/ special string Q , s.t. $\sum_{F \in F} \text{ED}(F, Q) = O(k)$.
Maintain: sequence $I = (U_1, V_1) \dots (U_z, V_z)$ of pairs from F^2
Updates: Insertions and Deletions of pairs in I
Queries: Compute k -edit occ's of $U_1 \dots U_z$ in $V_1 \dots V_z$
 Main Result: After $\tilde{O}(k^3)$ preprocessing, updates and queries in time $\tilde{O}(k)$

- ◆ For simplicity: assume $|Q| \approx \sqrt{|P|}$; ignore handling of initial and final pairs
- ◆ Goal: Iterate over all I_j 's with one DPM instance



Dynamic Puzzle Matching

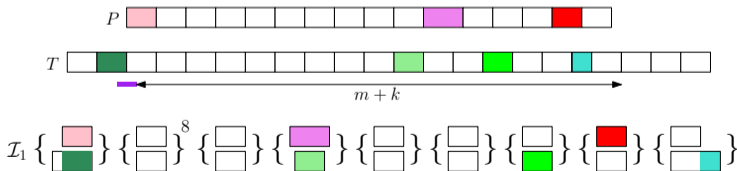
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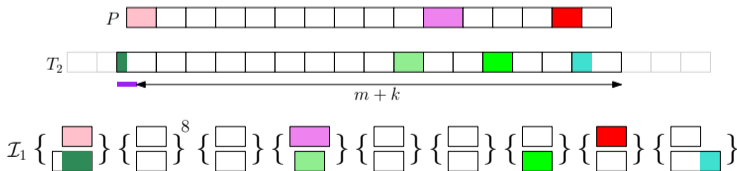
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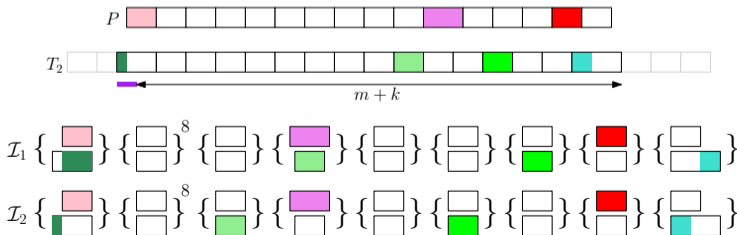
Using Dynamic Puzzle Matching

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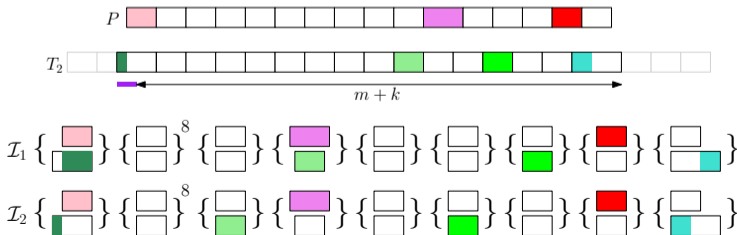
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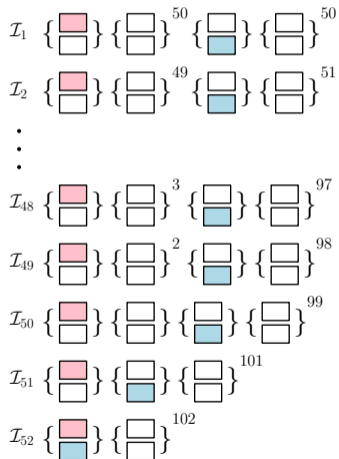
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- ◆ Goal: Iterate over all I_j 's with one DPM instance
 \rightsquigarrow Over $\Theta(\sqrt{|P|})$ shifts of P , need $O(\sqrt{|P|}k)$ DPM-updates $\rightsquigarrow \tilde{O}(k^3 + \sqrt{|P|}k^2)$ time



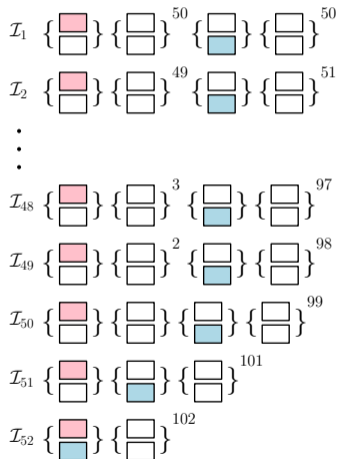
Using Dynamic Puzzle Matching *Better*

Example $k = 2$.



Using Dynamic Puzzle Matching *Better*

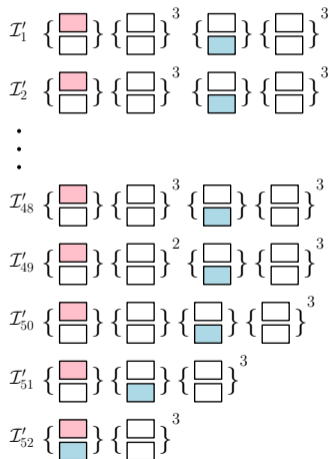
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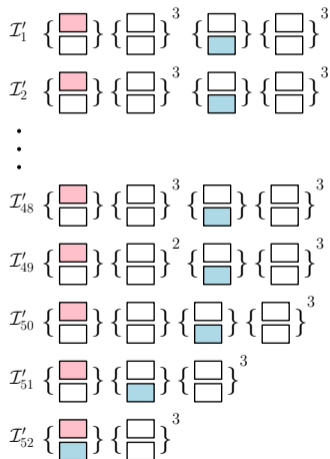
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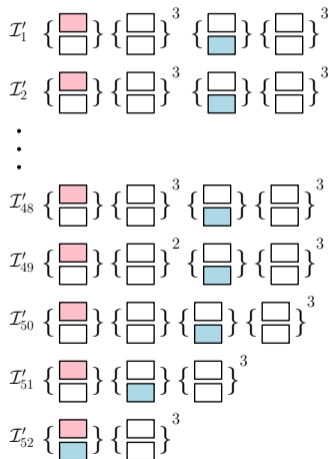
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 - $\rightsquigarrow O(k)$ DPM-updates per pair of special tiles
 - $\rightsquigarrow O(k^2)$ pairs of special tiles $\rightsquigarrow \tilde{O}(k^4)$ time

Using Dynamic Puzzle Matching *Better*

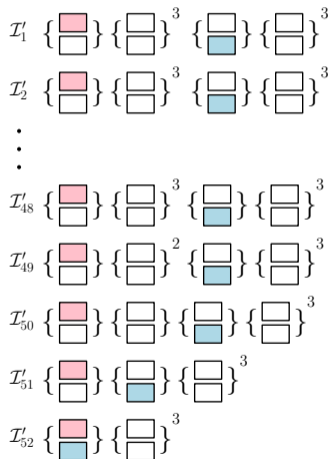
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Using Dynamic Puzzle Matching *Better*

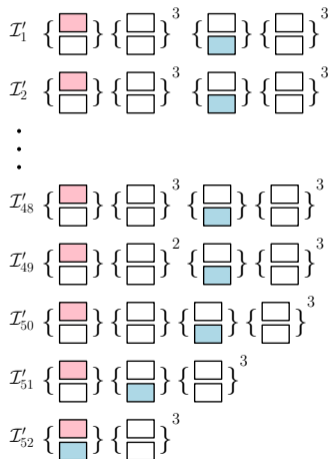
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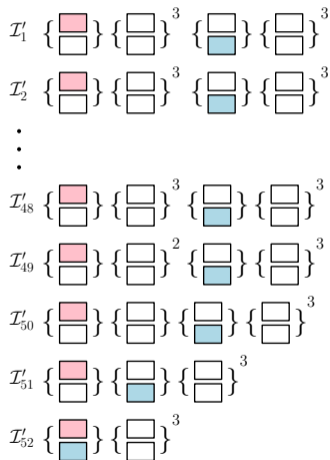
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 \rightsquigarrow Filter out false positives using another marking scheme $\rightsquigarrow \tilde{O}(k^{3.5})$ time in total (**boring**)

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Main Result

[CKoW'22]

Pattern P , text T , threshold k ; can compute starting pos of all k -edit occ's in time $\tilde{O}(|T| + k^{3.5} |T|/|P|)$.

Solving Dynamic Puzzle Matching

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General Idea

- 1 Store edit distance information for each pair $(U_i, V_i) \rightsquigarrow$ suitable permutation matrices allow this in $O(k)$ space
- 2 Show how to compose the information of different pairs in a suitable way \rightsquigarrow “seaweed product”
- 3 Show how to compute said product efficiently

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Content Warning: Some technical computations ahead!

Some Technical Definitions

- ◆ $A \in \{0, 1\}^{n \times n}$ is **permutation matrix**: every row and every column has exactly one 1
 \rightsquigarrow 1-to-1 corresponds to permutation in S_n
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$$A^\Sigma[i, j] := \sum_{i' \geq i} \sum_{j' < j} A[i', j']$$

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$$\begin{matrix} (0, 0) & & (n-1, 0) \\ \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \\ (0, n-1) & & (n-1, n-1) \end{matrix}$$

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For any A , have $(A^\Sigma)^\square = A$; for **simple** A , have $(A^\square)^\Sigma = A$ (first row and last col of A are 0)

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Lemma (Properties of Permutation Matrices)

Perm. matrix A ; $A^\Sigma[i, 0] = A^\Sigma[n - 1, j] = 0$ and $A^\Sigma[0, j] = j$; $A^\Sigma[i, 0] = n - 1 - i$ and $A^\Sigma[i, j] \geq j - i$

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Lemma (Working with Permutation Matrices)

[Tiso7, CP10]

For $n \times n$ permutation matrix A
can store in $O(n)$ space and can access $A^\Sigma[i, j]$ in $O(\log n / \log \log n)$

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Detour: Characterizing Unit-Monge Matrices

Lemma

Any matrix $A \in \mathbb{Z}_{\geq 0}^{(n+1) \times (n+1)}$ with

$$\blacklozenge A[i, 0] = 0$$

$$\blacklozenge A[i, n] = n - i$$

$$\blacklozenge A[n, j] = 0$$

$$\blacklozenge A[0, j] = j$$

$$A \stackrel{!}{=} \begin{pmatrix} 0 & 0 & \dots & 0 & 0 \\ 1 & & & & 0 \\ \dots & & & & \dots \\ n-1 & & & & 0 \\ n & n-1 & \dots & 1 & 0 \end{pmatrix}$$

is a unit Monge matrix.

\blacklozenge Show: each row/column of A^\square sums to 1

\blacklozenge Consider

$$\begin{aligned} \sum_j A^\square[i, j] &= \sum_j A[i+1, j] + A[i, j+1] - A[i+1, j+1] - A[i, j] \\ &= A[i+1, 0] + A[i, n] - A[i+1, n] - A[i, 0] \\ &= 0 + (n-i) - (n-i-1) - 0 \\ &= 1 \end{aligned}$$

$$A \stackrel{!}{=} \left(\begin{array}{ccc|cc|ccc} 0 & & \dots & 0 & & 0 & & \dots & 0 \\ 1 & & & & & & & & 0 \\ \dots & & & & & & & & \dots \\ n-j & & & - & & + & & & 0 \\ n-j+1 & & & + & & - & & & 0 \\ \dots & & & & & & & & \dots \\ n-1 & & & & & & & & 0 \\ n & n-1 & \dots & n-i & n-i-1 & \dots & 1 & & 0 \end{array} \right)$$

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$$A \stackrel{!}{=} \begin{pmatrix} 0 & 0 & \dots & 0 & 0 \\ 1 & & & & 0 \\ \dots & & & & \dots \\ n-1 & & & & 0 \\ n & n-1 & \dots & 1 & 0 \end{pmatrix}$$

is a unit Monge matrix.

\blacklozenge Show: each row/column of A^\square sums to 1

\blacklozenge Consider

$$\begin{aligned} \sum_j A^\square[i, j] &= \sum_j A[i+1, j] + A[i, j+1] - A[i+1, j+1] - A[i, j] \\ &= A[i+1, 0] + A[i, n] - A[i+1, n] - A[i, 0] \\ &= 0 + (n-i) - (n-i-1) - 0 \\ &= 1 \end{aligned}$$

$$A \stackrel{!}{=} \left(\begin{array}{cc|cc|cc} 0 & \dots & 0^- & 0^+ & \dots & 0 \\ 1 & & +^- & -^+ & & 0 \\ \dots & & +^- & -^+ & & \dots \\ n-j & & +^- & -^+ & & 0 \\ n-j+1 & & +^- & -^+ & & 0 \\ \dots & & +^- & -^+ & & \dots \\ n-1 & & +^- & -^+ & & 0 \\ n & \dots & n-i^+ & n-i-1^- & \dots & 0 \end{array} \right)$$

Detour: Characterizing Unit-Monge Matrices

Lemma

Any matrix $A \in \mathbb{Z}_{\geq 0}^{(n+1) \times (n+1)}$ with

$$\blacklozenge A[i, 0] = 0$$

$$\blacklozenge A[i, n] = n - i$$

$$\blacklozenge A[n, j] = 0$$

$$\blacklozenge A[0, j] = j$$

$$A \stackrel{!}{=} \begin{pmatrix} 0 & 0 & \dots & 0 & 0 \\ 1 & & & & 0 \\ \dots & & & & \dots \\ n-1 & & & & 0 \\ n & n-1 & \dots & 1 & 0 \end{pmatrix}$$

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Monge Matrix \times (min,+)-Product

- ◆ A is **simple**: first row and last col are 0; for simple A have $A^{\Sigma \square} = A^{\square \Sigma} = A$
- ◆ A is **Monge** if $A^{\square} \geq 0$; A is **unit-Monge** if A^{\square} is a permutation matrix
- ◆ For matrices A, B , the **(min,+)-product** $A \odot B$ is

$$(A \odot B)[i, k] := \min_j (A[i, j] + B[j, k])$$

Recall: For adjacency matrix A of (weighted) graph G , $A^{\odot \ell}$ stores ℓ -hop dist in G

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Lemma (Unit-Monge Monoid)

[Tis'07,'15]

For Monge A, B , the matrix $A \odot B$ is Monge.

For simple unit-Monge A, B , the matrix $A \odot B$ is simple unit-Monge.

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Proof: Board or exercise.

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- ◆ $A \boxdot B$ is different from normal perm concat: $I^R = \begin{pmatrix} 0 & \dots & 0 & 1 \\ | & 0 & 1 & 0 \\ 0 & 1 & 0 & | \\ 1 & 0 & \dots & 0 \end{pmatrix}$ is a zero

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Monge Matrix \times (min,+)-Product—Examples

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \boxtimes \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} =$$

Monge Matrix \times (min,+)-Product—Examples

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \boxtimes \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = \\
 \begin{matrix} \downarrow \wr (\cdot)^\Sigma \\ \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \odot \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \end{matrix}$$

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$$\begin{array}{ccc}
 \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} & \boxtimes & \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = \\
 \downarrow \wr (\cdot)^\Sigma & & \downarrow \wr (\cdot)^\Sigma \\
 \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} & \odot & \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}
 \end{array}$$

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$\Downarrow (\cdot)^\Sigma$
 $\Downarrow (\cdot)^\Sigma$
 $\Uparrow (\cdot)^\square$

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 $\Downarrow (\cdot)^\Sigma$
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Monge Matrix \times (min,+)-Product—Computation

Naive computation: $O(n^3) \rightsquigarrow$ too slow

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Theorem (Monge Multiplication)

[SMAWK'87]

Can compute (min,+)-product of two $n \times n$ Monge matrices in time $\tilde{O}(n^2)$.

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(Proof skipped.)

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(Proof skipped.)

Theorem (Unit Monge Multiplication)

[Tis'07,'15]

Can compute (min,+)-product of two $n \times n$ simple unit-Monge matrices in time $O(n \log n)$.

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(Proof skipped.)

Theorem (Unit Monge Multiplication)

[Tis'07,'15]

Can compute (min,+)-product of two $n \times n$ simple unit-Monge matrices in time $O(n \log n)$.

Proof idea (details messy, but not too complicated).

- ◆ Divide and Conquer: Split each matrix into two; reduce to just two subproblems
- ◆ For conquer step, use elaborate, but easy to compute function to decide which solution to propagate from subproblem

Monge Matrix \times (min,+)-Product—Examples, revisited

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \boxdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

$\downarrow \downarrow (\cdot)^\Sigma$
 $\downarrow \downarrow (\cdot)^\Sigma$
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$$\begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \odot \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}$$

Monge Matrix \times (min,+)-Product—Examples, revisited

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \boxtimes \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

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$$\begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \odot \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}$$

$$\begin{array}{c} \bullet & & \bullet \\ & \diagdown & / \\ & \diagup & \diagdown \\ \bullet & & \bullet \end{array} \begin{array}{c} \bullet \\ | \\ \bullet \end{array} \boxtimes \begin{array}{c} \bullet \\ | \\ \bullet \end{array} \begin{array}{c} \bullet & & \bullet \\ & \diagdown & / \\ & \diagup & \diagdown \\ \bullet & & \bullet \end{array} =$$

Monge Matrix \times (min,+)-Product—Examples, revisited

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \boxtimes \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

$\Downarrow (\cdot)^\Sigma$
 $\Downarrow (\cdot)^\Sigma$
 $\Uparrow (\cdot)^\square$

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$$\begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \boxtimes \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} = \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array}$$

Monge Matrix \times (min,+)-Product—Examples, revisited

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Diagrammatic representation of the Monge matrix product. The first diagram shows a crossing of two strands (a cup) and a vertical strand. The second diagram shows a vertical strand and a crossing of two strands (a cap). The result is a diagram with two crossings and two vertical strands.

Monge Matrix \times (min,+)-Product—Examples, revisited

$$\begin{array}{ccc}
 \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} & \boxtimes & \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \\
 \downarrow \wr (\cdot)^\Sigma & & \downarrow \wr (\cdot)^\Sigma \quad \quad \quad \uparrow \wr (\cdot)^\square \\
 \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} & \odot & \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 1 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}
 \end{array}$$

Example matches normal multiplication of permutation matrices.

Monge Matrix \times (min,+)-Product—Examples, revisited

$$\begin{array}{ccc}
 \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} & \boxtimes & \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \\
 \downarrow (\cdot)^\Sigma & \downarrow (\cdot)^\Sigma & \uparrow (\cdot)^\square \\
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 \end{array}$$

Monge Matrix \times (min,+)-Product—Examples, revisited

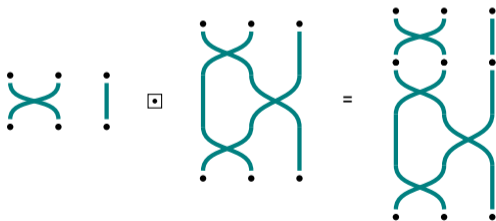
$$\underbrace{\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} \boxtimes \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} = \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\uparrow (\cdot)^\square}$$

$$\begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \odot \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 2 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 2 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}$$

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$$\underbrace{\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} \boxtimes \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} = \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\uparrow (\cdot)^\square}$$

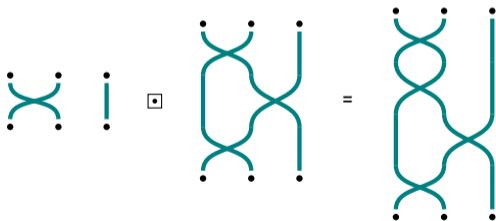
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Monge Matrix \times (min,+)-Product—Examples, revisited

$$\underbrace{\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} \boxtimes \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} = \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\uparrow (\cdot)^\square}$$

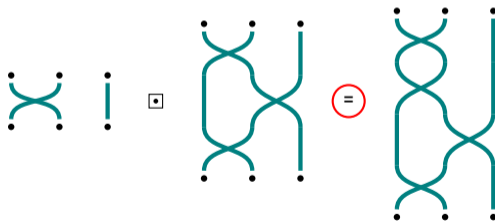
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Monge Matrix \times (min,+)-Product—Examples, revisited

$$\underbrace{\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} \boxtimes \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} = \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\uparrow (\cdot)^\square}$$

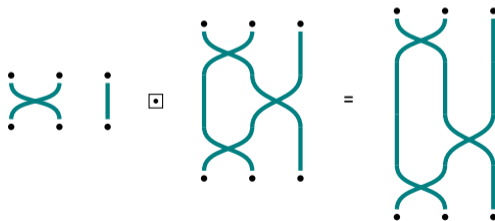
$$\begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \odot \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 2 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 2 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}$$



Monge Matrix \times (min,+)-Product—Examples, revisited

$$\underbrace{\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} \boxtimes \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\downarrow (\cdot)^\Sigma} = \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}}_{\uparrow (\cdot)^\square}$$

$$\begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} \odot \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 2 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 2 & 2 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{pmatrix}$$



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 \end{array}$$

Can view seaweed product in terms of these diagrams, but need some extra simplification rules.

Seaweed Monoid

Can characterize monoid (simple unit-Monge, \square) in terms of seaweed diagrams with their stitching, when using as generators the elementary seaweeds

$$\{ \begin{array}{c} \cdot \\ | \\ \cdot \end{array} \begin{array}{c} \cdot \\ | \\ \cdot \end{array} \dots \begin{array}{c} \cdot \\ | \\ \cdot \end{array}, \begin{array}{c} \cdot \\ \diagdown \quad \diagup \\ \cdot \quad \cdot \end{array} \dots \begin{array}{c} \cdot \\ | \\ \cdot \end{array}, \dots, \begin{array}{c} \cdot \\ | \\ \cdot \end{array} \dots \begin{array}{c} \cdot \\ \diagdown \quad \diagup \\ \cdot \quad \cdot \end{array} \}$$

and the following relations

$$\begin{array}{c} \cdot \\ \diagdown \quad \diagup \\ \cdot \quad \cdot \end{array} \dots \begin{array}{c} \cdot \\ | \\ \cdot \end{array} = \begin{array}{c} \cdot \\ \diagdown \quad \diagup \\ \cdot \quad \cdot \end{array} \dots \begin{array}{c} \cdot \\ | \\ \cdot \end{array} \quad (\text{idempotence for every generator})$$

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A Final Example: Stitching with I

- ◆ Consider following task:

Have permutations $(1, \dots, n) \mapsto (\sigma(1), \dots, \sigma(n))$ and $(1, \dots, n) \mapsto (\varrho(1), \dots, \varrho(n))$

Want permutation $(1, \dots, n, n+1, \dots, 2n) \mapsto (\sigma(1), \dots, \sigma(n), \varrho(1), \dots, \varrho(n))$

\rightsquigarrow Prime application: string concatenation

- ◆ Claim:

$$P_{\sigma\varrho} = \begin{pmatrix} P_\sigma & 0 \\ 0 & P_\varrho \end{pmatrix} = \begin{pmatrix} P_\sigma & 0 \\ 0 & I_{|\varrho|} \end{pmatrix} \boxplus \begin{pmatrix} I_{|\sigma|} & 0 \\ 0 & P_\varrho \end{pmatrix}$$

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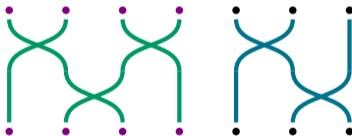
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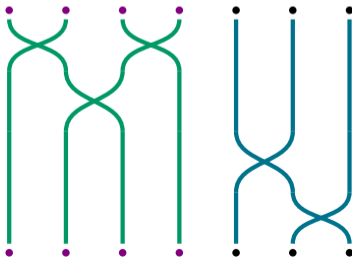
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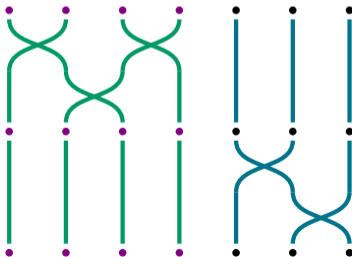
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- ◆ Somewhat weird matrix product (min-plus product of prefix-sum matrices of permutation matrices)
- ◆ Seaweed product can be computed efficiently
 - ↪ Important special case: stitching with I)
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Seaweeds in a Nutshell

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Using Seaweeds for Faster ED Computations

Given P, T , need to compute ED between arbitrary substrings

~> **Goal:** Use seaweeds for speed-ups over standard DP

DP table $e(i, j)$

- ◆ $e(i, j) = \text{ED}(P[0..i], T[0..j])$
- ◆ naive: $O(|P| \cdot |T|)$ computation
- ◆ kangaroo jumps [LV89]
 - ~> time $O(|P| + |T| + k^2)$ for ED
 - ~> time $O((|P| + |T|)k)$ for APM



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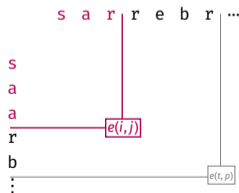
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From DP to Alignment Graphs

Recall classical DP (again):

- ◆ Write $e(i, j)$ for the ED of $T[0..i)$ and $P[0..j)$
 $\rightsquigarrow e(t, p)$ is the ED of T and P .
- ◆ Clearly, $e(i, 0) = i$ (delete $T[0..i)$)
 and $e(0, j) = j$ (insert $P[0..j)$)
- ◆ Observe

$$e(i, j) = \min \begin{cases} e(i-1, j) + 1 & \text{(delete from } T) \\ e(i, j-1) + 1 & \text{(insert in } T) \\ e(i-1, j-1) & \\ \quad + (T[i] \neq P[j]) & \text{(match/subst)} \end{cases}$$



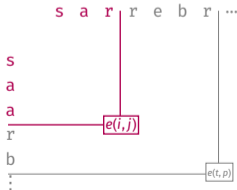
	s	a	r	r	e	b	r	...
	0	1	2	3	4	5	6	7
s	1	0	1	2	3	4	5	6
a	2	1	0	1	2	3	4	5
a	3	2	1	1	2	3	4	5
r	4							
b	5							
⋮								

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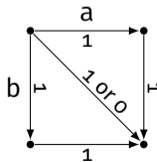
	s	a	r	r	e	b	r	...
s	0	1	2	3	4	5	6	7
a	1	0	1	2	3	4	5	6
a	2	1	0	1	2	3	4	5
r	3	2	1	0	1	2	3	4
b	4							
⋮	5							

Consider different,
but related **alignment graph**:

- ◆ Vertex (i, j) for every $i \in [0..|T|], j \in [0..|P|]$
- ◆ Horizontal edge $(i, j) \rightarrow (i+1, j)$ of weight 1 (deletion from T)
- ◆ Vertical edge $(i, j) \rightarrow (i, j+1)$ of weight 1 (insertion in T)
- ◆ Diagonal edge $(i, j) \rightarrow (i+1, j+1)$ of weight $(T[i] \neq P[j])$ (match/subst)

\rightsquigarrow shortest $(0, 0)$ to $(|T|, |P|)$ path is ED of T and P

\rightsquigarrow recover DP by running Dijkstra from $(0, 0)$



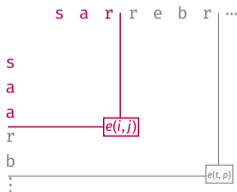
		a	b	c	d	e	f
z							
x							
w							
v							
u							
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	s	a	r	r	e	b	r	...
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a	1	0	1	2	3	4	5	6
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r	3	2	1	0	1	2	3	4
b	4							
⋮	5							

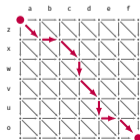
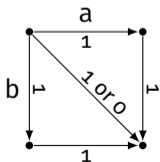
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- ◆ Vertex (i, j) for every $i \in [0..|T|], j \in [0..|P|]$
- ◆ Horizontal edge $(i, j) \rightarrow (i+1, j)$ of weight 1 (deletion from T)
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\rightsquigarrow shortest $(0, 0)$ to $(|T|, |P|)$ path is ED of T and P

\rightsquigarrow recover DP by running Dijkstra from $(0, 0)$

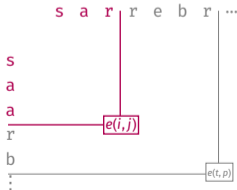


From DP to Alignment Graphs

Recall classical DP (again):

- ◆ Write $e(i, j)$ for the ED of $T[0..i]$ and $P[0..j]$
 $\rightsquigarrow e(t, p)$ is the ED of T and P .
- ◆ Clearly, $e(i, 0) = i$ (delete $T[0..i]$)
 and $e(0, j) = j$ (insert $P[0..j]$)
- ◆ Observe

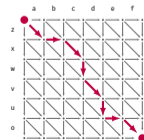
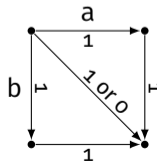
$$e(i, j) = \min \begin{cases} e(i-1, j) + 1 & \text{(delete from } T) \\ e(i, j-1) + 1 & \text{(insert in } T) \\ e(i-1, j-1) + (T[i] \neq P[j]) & \text{(match/subst)} \end{cases}$$



		s	a	r	r	e	b	r	...
s	0	1	2	3	4	5	6	7	
a	1	0	1	2	3	4	5	6	
a	2	1	0	1	2	3	4	5	
r	3	2	1	0	1	2	3	4	
b	4								
⋮	5								

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but related **alignment graph**:

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- \rightsquigarrow shortest $(0, 0)$ to $(|T|, |P|)$ path is ED of T and P
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From DP to Alignment Graphs

Alignment Graph of P and T

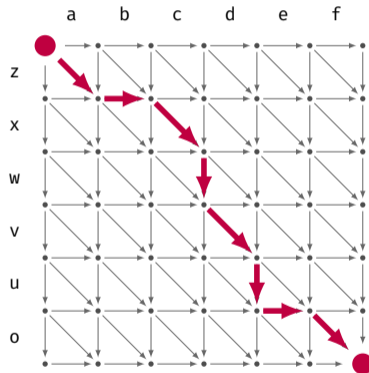
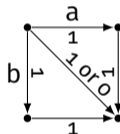
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⇒ shortest $(0, 0)$ to $(|T|, |P|)$ path is ED of T and P

- ◆ shortest (i, j) to (i', j') path is ED of $T[i..i']$ and $P[j..j']$

⇒ For stitching: Need (just) boundary-to-boundary dist's

- ◆ Define $D_{P,T}[a, b] := \text{dist}(\text{in}_a, \text{out}_b)$

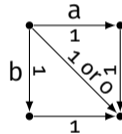


$$\text{dist}((0, 0), (|T|, |P|)) = \text{ED}(T, P)$$

From DP to Alignment Graphs

Alignment Graph of P and T

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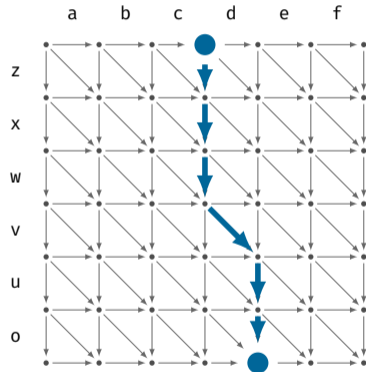


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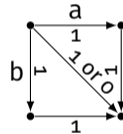


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From DP to Alignment Graphs

Alignment Graph of P and T

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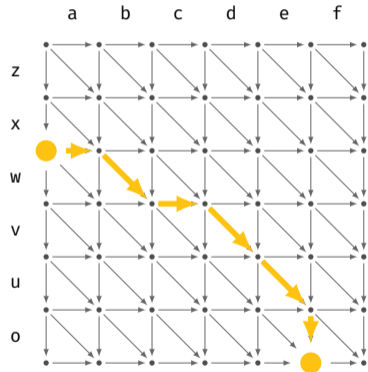


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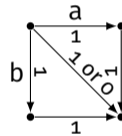


$$\text{dist}((i, j), (i', j')) = \text{ED}(T[i..i'], P[j..j'])$$

From DP to Alignment Graphs

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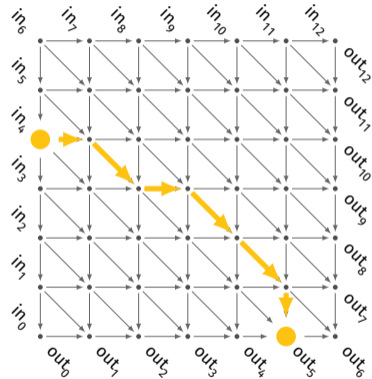


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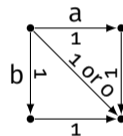


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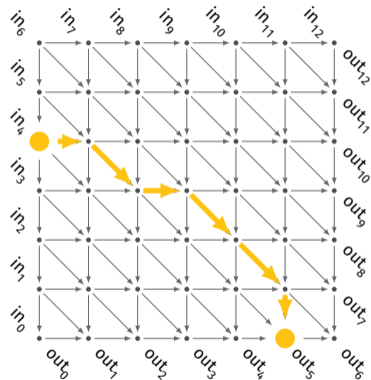


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From DP to Alignment Graphs

Consider Alignment Graph of P and T

◆ Define $D_{P,T}[a, b] := \text{dist}(\text{in}_a, \text{out}_b)$

↪ Hope: $D_{P,T}^\square$ is permutation matrix

◆ Problem: Some $D_{P,T}[a, b]$ are ∞

↪ use undirected edges

◆ Suffices to show (recall earlier slide):

◆ $D_{P,T}$ is non-negative and integer

◆ Border satisfies ($n := |P| + |T|$)

$$D_{P,T} = \begin{pmatrix} 0 & 0 & \dots & 0 & 0 \\ 1 & & & & \\ \vdots & & & & \\ n-1 & & & & \\ n & n-1 & \dots & 1 & 0 \end{pmatrix}$$

◆ $D_{P,T}[a, 0] = 0$

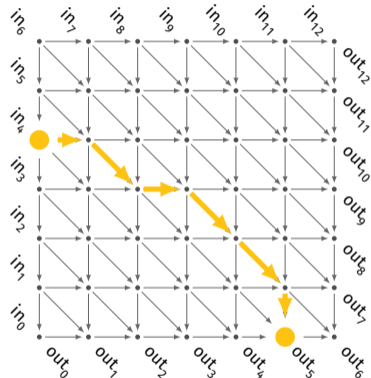
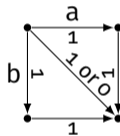
◆ $D_{P,T}[a, n] = n - a$

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Idea: Force boundary values to be correct

↪ $D'_{P,T}[a, b] := (D_{P,T}[a, b] - a + b)/2$



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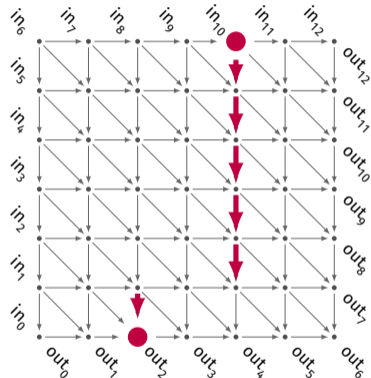
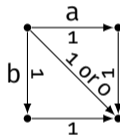
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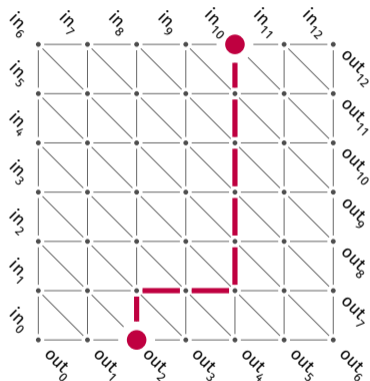
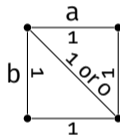
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From DP to Alignment Graphs

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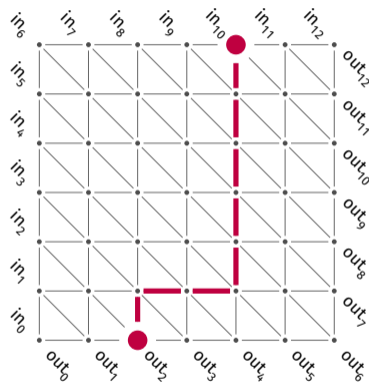
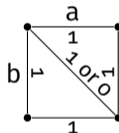
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From DP to Alignment Graphs

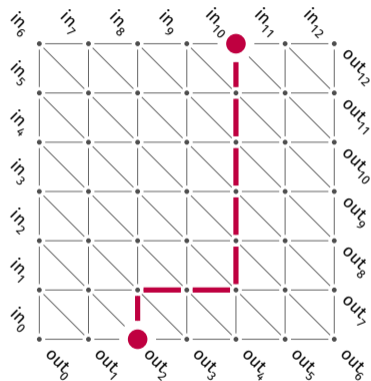
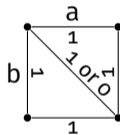
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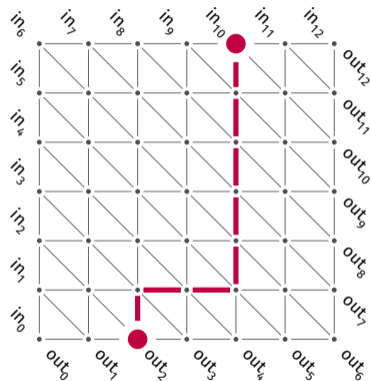
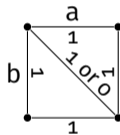
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From DP to Alignment Graphs

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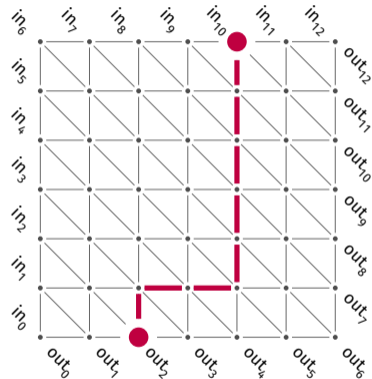
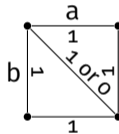
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$\text{dist}(\text{in}_a, \text{out}_b) \stackrel{*}{=} \text{ED}(\mathcal{T}[\max(a - |P|, 0) .. \min(b, |T|)],$

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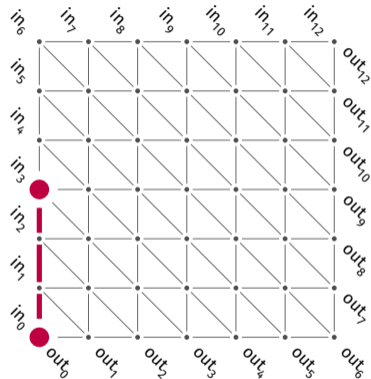
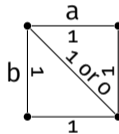
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$$\text{dist}(\text{in}_a, \text{out}_0) \stackrel{*}{=} \text{ED}(T[\max(a - |P|, 0) \dots 0], \\ P[\max(|P| - a, 0) \dots |P|])$$

Idea: Force boundary values to be correct

→ $D'_{P,T}[a, b] := (D_{P,T}[a, b] - a + b) / 2$

From DP to Alignment Graphs

Consider Alignment Graph of P and T

◆ Define $D_{P,T}[a, b] := \text{dist}(\text{in}_a, \text{out}_b)$

◆ **Problem:** Some $D_{P,T}[a, b]$ are ∞

→ use **undirected** edges

→ Hope: $D_{P,T}^\square$ is permutation matrix

◆ Suffices to show (recall earlier slide):

◆ $D_{P,T}$ is non-negative and integer ✓

◆ Border satisfies ($n := |P| + |T|$)

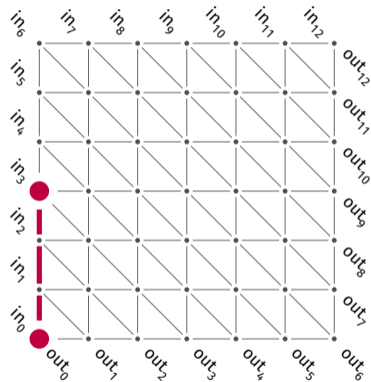
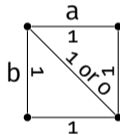
$$D_{P,T} \stackrel{!}{=} \begin{pmatrix} 0 & 0 & \dots & 0 & 0 \\ 1 & & & & 0 \\ \dots & & & & \dots \\ n-1 & & & & 0 \\ n & n-1 & \dots & 1 & 0 \end{pmatrix}$$

◆ $D_{P,T}[a, 0] = 0$

◆ $D_{P,T}[n, b] = 0$

◆ $D_{P,T}[a, n] = n - a$

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Idea: Force boundary values to be correct

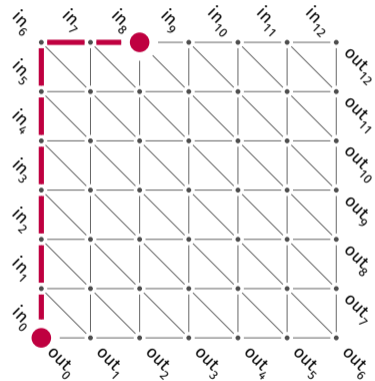
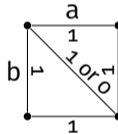
→ $D'_{P,T}[a, b] := (D_{P,T}[a, b] - a + b) / 2$

$$\text{dist}(\text{in}_a, \text{out}_0) = \text{ED}(\epsilon, P[\max(|P| - a, 0) .. |P|]) = a$$

From DP to Alignment Graphs

Consider Alignment Graph of P and T

- ◆ Define $D_{P,T}[a, b] := \text{dist}(\text{in}_a, \text{out}_b)$
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$$\text{dist}(\text{in}_a, \text{out}_0) = a$$

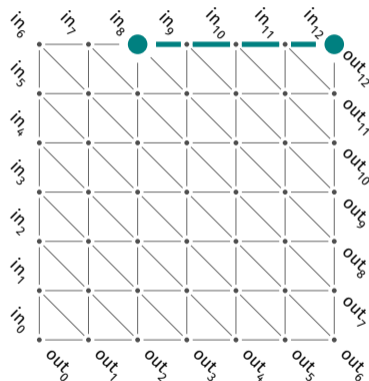
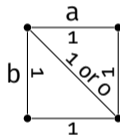
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From DP to Alignment Graphs

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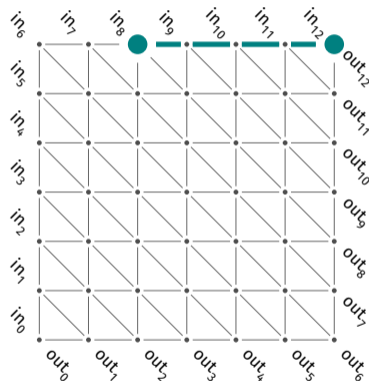
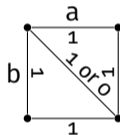
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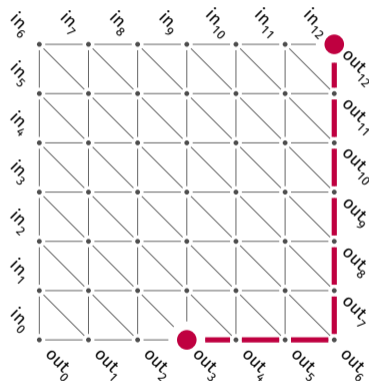
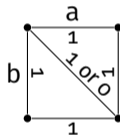


$$\text{dist}(\text{in}_a, \text{out}_n) = n - a$$

From DP to Alignment Graphs

Consider Alignment Graph of P and T

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 - ◆ $D_{P,T}[a, 0] = \ominus a$
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 - ◆ $D_{P,T}[0, b] = b$



$$\text{dist}(\text{in}_n, \text{out}_b) = n - b$$

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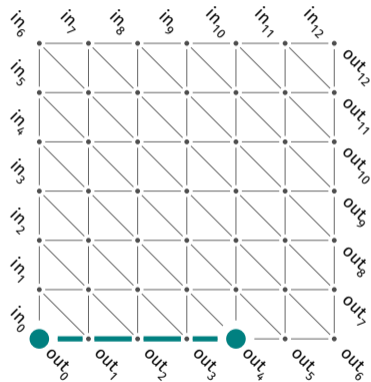
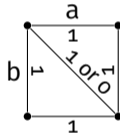
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◆ $D_{P,T}[n, b] = \ominus n - b$ ◆ $D_{P,T}[0, b] = b$

Idea: Force boundary values to be correct

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$\text{dist}(\text{in}_0, \text{out}_b) = b$

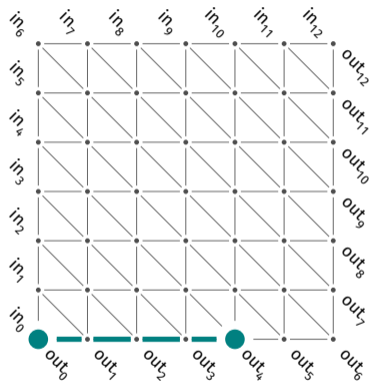
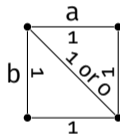
From DP to Alignment Graphs

Consider Alignment Graph of P and T

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$$\text{dist}(\text{in}_a, \text{out}_b) \triangleq \text{ED}(T[\max(a - |P|, 0) .. \min(b, |T|)], P[\max(|P| - a, 0) .. \min(|P| + |T| - b, |P|)])$$

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Have:

$$\text{◆ } D'_{P,T}[a, 0] = (a - a + 0)/2 = 0$$

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◆ Border satisfies ($n := |P| + |T|$)

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Substitutions Are Too Cheap!

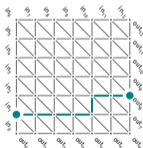
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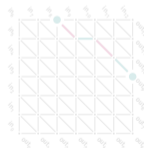
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$$\text{dist}(\text{in}_a, \text{out}_b) = |b - a|$$

good ($|b - a| - a + b$ always divisible by 2)



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red: cost 1 (bad), green: cost 0 (saving of 2, good)

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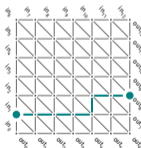
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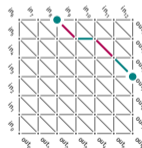
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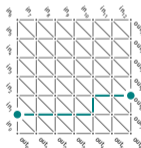
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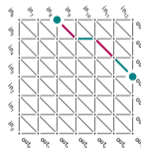
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The Deletion Distance to the Rescue!

- ◆ Define $D_{P,T}[a,b] := \text{dist}(\text{in}_a, \text{out}_b)$ and $D'_{P,T}[a,b] := (D_{P,T}[a,b] - a + b)/2$
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Deletion Distance (DD)

Min number of insertions or deletions of characters to transform T into P .

Embedding DD into ED

Write $S^\$:= S[0]\$S[1]\$ \cdots S[|S|]\$$. Then $\text{DD}(P^\$, T^\$) = 2\text{ED}(P, T)$.

- ◆ In alignment graph: corresponds to removing all diag edges of weight 1 (and adding vertices/edges corresponding to \$)

$$D_{P,T}[a,b] = \text{dist}(\text{in}_a, \text{out}_b) = \text{dist}((x_a, y_a), (x_b, y_b))$$

$$= \begin{cases} |b-a| = |x_a - x_b| + |y_a - y_b| & \text{if } b \leq a - |P| \text{ or } b \geq a + |T| \text{ (out is left or above of in, no diag edges)} \\ \dots & \dots \end{cases}$$

The Deletion Distance to the Rescue!

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- ↪ Hope: $D'_{P,T}[\square]$ is permutation matrix ↪ Problem: $D'_{P,T}[\square][a,b]$ not integer
- ◆ Every character subst. flips parity of $\text{dist}(\text{in}_a, \text{out}_b)$ ↪ “forbid subst’s” / increase cost to 2

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Min number of insertions or deletions of characters to transform T into P .

Embedding DD into ED

Write $S^\$:= S[0]\$S[1]\$ \cdots S[|S|]\$$. Then $\text{DD}(P^\$, T^\$) = 2\text{ED}(P, T)$.

- ◆ In alignment graph: corresponds to removing all diag edges of weight 1 (and adding vertices/edges corresponding to \$)

$$D_{P,T}[a,b] = \text{dist}(\text{in}_a, \text{out}_b) = \text{dist}((x_a, y_a), (x_b, y_b))$$

$$= \begin{cases} |b-a| = |x_a - x_b| + |y_a - y_b| & \text{if } b \leq a - |P| \text{ or } b \geq a + |T| \text{ (out is left or above of in, no diag edges)} \\ \dots & \dots \end{cases}$$

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The Deletion Distance to the Rescue!—Recap

Main Result

Consider (deletion distance) alignment graph of $P^\$$ and $T^\$$;
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Then, $D'_{P^\$, T^\$}$ is a permutation matrix.

- ◆ Can (easily) show: $(\min, +)$ -product of $D'_{P^\$, T^\$}$ matrices (“stitching”) yields deletion distances for concatenated strings
 \rightsquigarrow Can use seaweed product for fast computation
- ◆ Some technicalities (skipped):
 - ◆ stitching requires some overlap in at least one string
 - ◆ want to compute only $\approx k$ diagonals; requires a “restriction” operation
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Dynamic Puzzle Matching

Input: integer k , family F of strings w/ special string Q , s.t. $\sum_{F \in F} \text{ED}(F, Q) = O(k)$.

Maintain: sequence $I = (U_1, V_1) \dots (U_z, V_z)$ of pairs from F^2

Updates: Insertions and Deletions of pairs in I

Queries: Compute k -edit occ's of $U_1 \dots U_z$ in $V_1 \dots V_z$

Main Result: After $\tilde{O}(k^3)$ preprocessing, updates and queries in time $\tilde{O}(k)$

General Idea

- 1 Store edit distance information for each pair (U_i, V_i)
 \rightsquigarrow Seaweed/permutation matrix of (U_i, V_i) allow this in $O(k)$ space
- 2 In preprocessing, build matrix for every possible pair of strings from family F
- 3 Use seaweed product of [Tis'07,'15] to compose the information of different pairs
 $\tilde{O}(k)$ per stitch
- 4 Use BST on top to support update queries

What about the weighted setting?

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10. The weighted case. In the weighted case, deletions of different characters and the various substitutions may have differing costs, but, by way of normalization, all will be required to have cost at least 1.

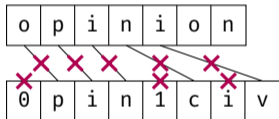
The approximate matches with cost $\leq k$ can be found using essentially the same algorithm; the only change needed is to the Landau–Vishkin algorithm to take into account the differing costs. The details are left to the reader.

—Cole, Hariharan, 2002

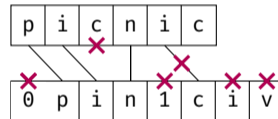
Edit Distance

$ED(X, Y)$

Min number of character insertions, deletions, and substitutions that transform X to Y



$$ED(0PIN1CIV, OPINION) = 5$$



$$ED(0PIN1CIV, PICNIC) = 5$$

Background

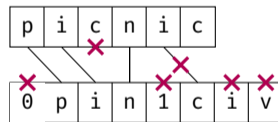
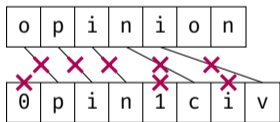
 $ED^w(X, Y)$

Weighted Edit Distance

Min cost of transforming X to Y using character edits, where:

- ◆ inserting y costs $w(\varepsilon, y)$;
- ◆ deleting x costs $w(x, \varepsilon)$;
- ◆ substituting x for y costs $w(x, y)$.

$$w(\emptyset, \emptyset) := 1 \quad w(1, 1) := 1 \quad w(C, \emptyset) := 1 \quad w(*, *) := 2 \quad w(*, \varepsilon) := 1 \quad w(\varepsilon, *) := 10$$



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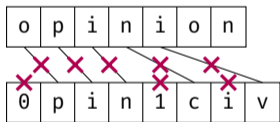
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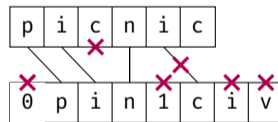
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$$ED^w(0PIN1CIV, OPINION) = 6$$



$$ED^w(0PIN1CIV, PICNIC) \leq 14$$

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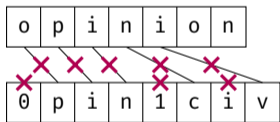
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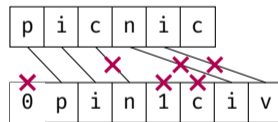
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How fast can we compute the (weighted) edit distance of two strings?

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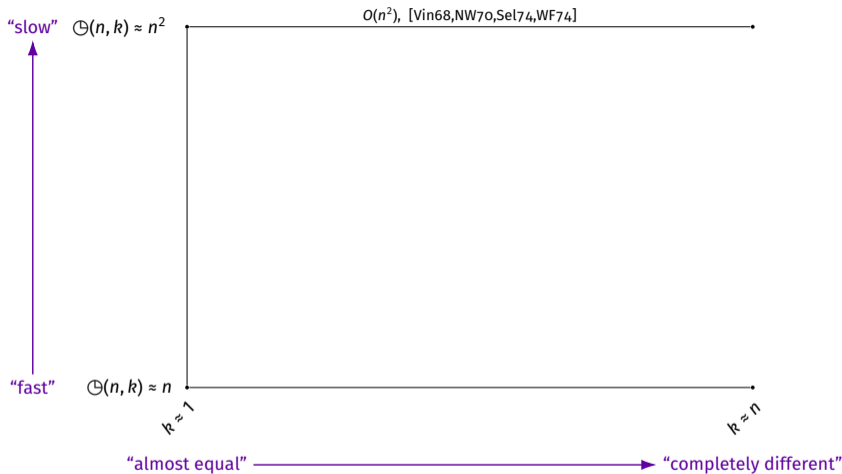
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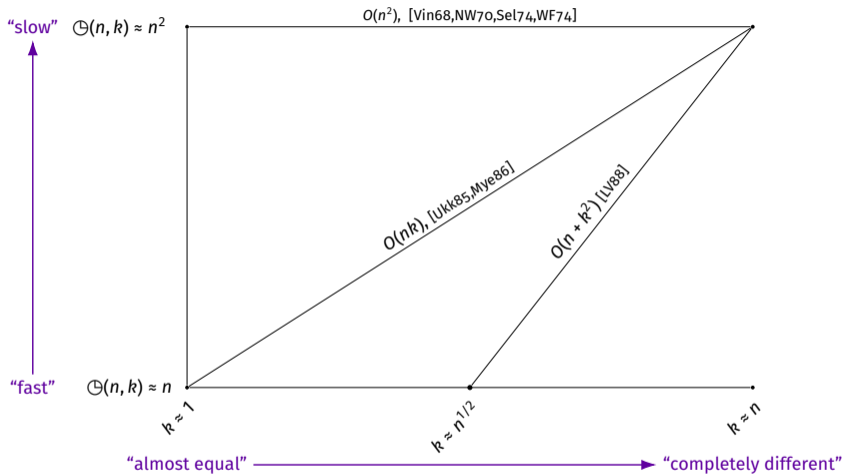
\rightsquigarrow Bounded (Weighted) Edit Distance

Algorithms for (Bounded) Edit Distance



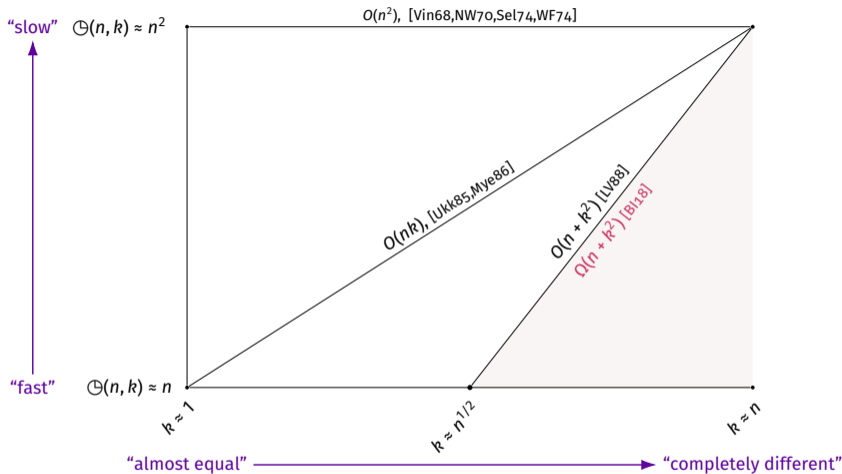
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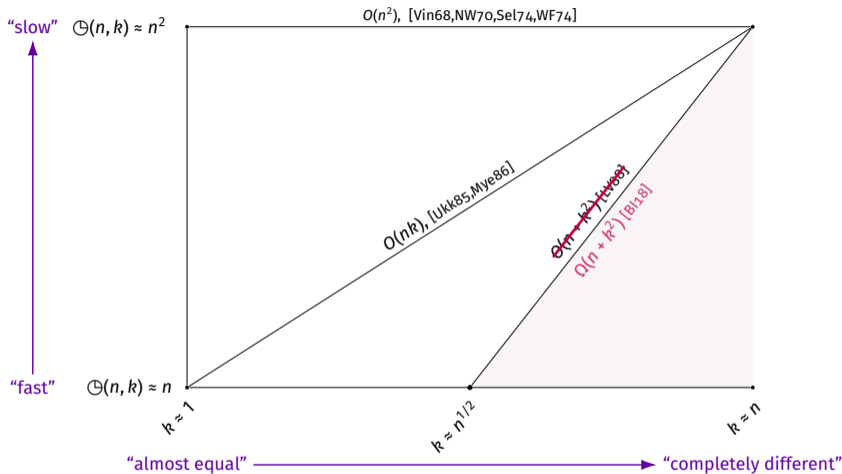
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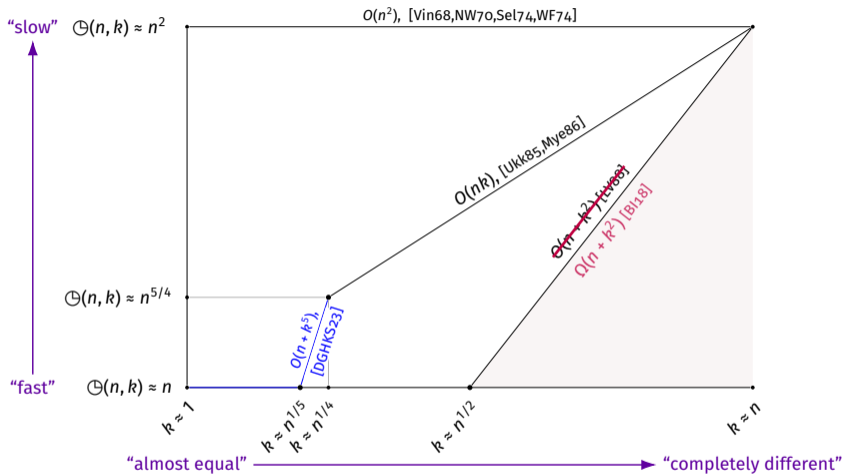
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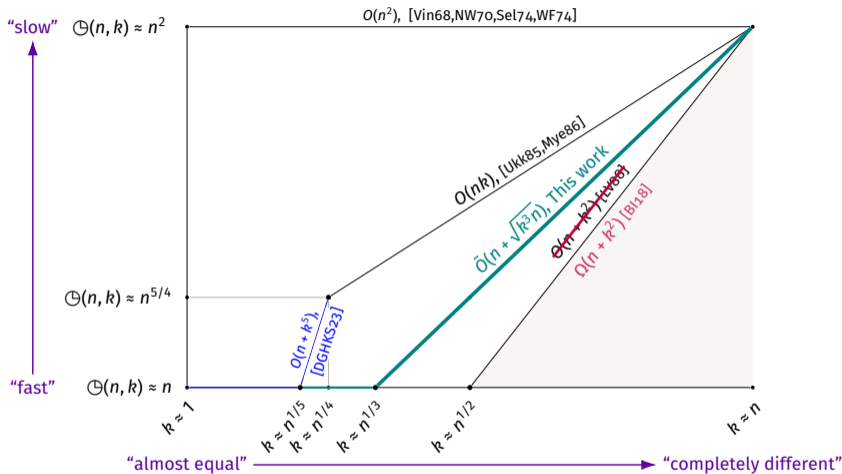
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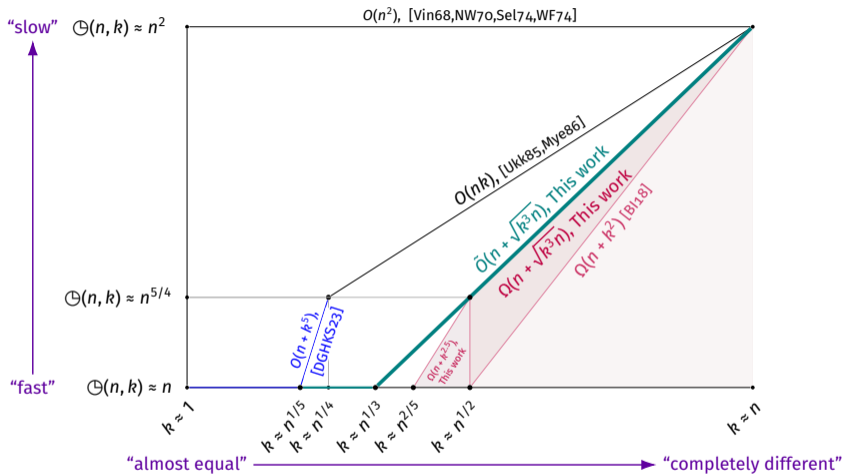
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[CaKoW'23]

Main Theorem 1 (Upper Bound)

Strings X, Y each of length at most n
Oracle access to (normalized) weight function w
Can compute $k = \text{ED}^w(X, Y)$ in time $O(n + \sqrt{nk^3} \log^3 n)$

[CaKoW'23]

Main Theorem 2 (Lower Bound)

Assuming the APSP Hypothesis and for $\sqrt{n} \leq k \leq n$,
Main Theorem 1 is tight (up to $n^{o(1)}$ -factors)

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Theorem (Universal Kernel)

[DGHS23]

Can trim X and Y to length- $O(k^4)$ strings X', Y' with $ED_{\leq k}^w(X, Y) = ED_{\leq k}^w(X', Y')$.

Tool o: Unweighted ED and [DGHKS23]-Kernel

$$ED_{\leq k}^w(X, Y)$$

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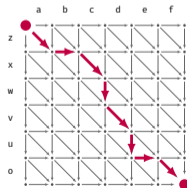
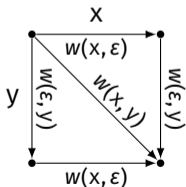
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Compute all b-to-b dist [Kleino5] + stitch together results ((min, +)-product [SMAWK87])

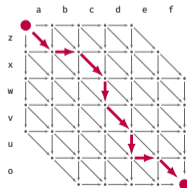
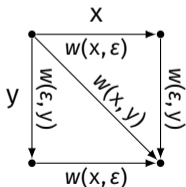
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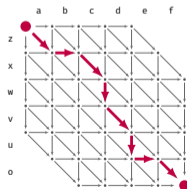
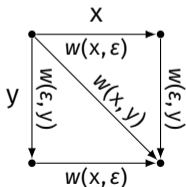
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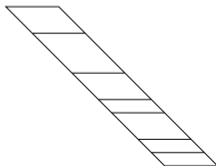
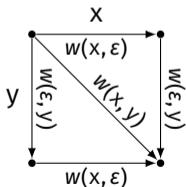
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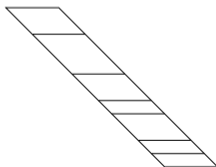
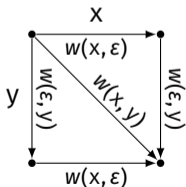
Tool 1: Alignment Graphs and Multiple-Source Shortest Path

- ◆ X and Y consist in $O(k^2)$ pieces of length $O(k^2)$ and with period $O(k)$ each
- ◆ Idea: Use (directed) alignment graph AG of X and Y (seaweeds don't work here)
 $\rightsquigarrow \text{ED}^W(X, Y)$ is distance $(0, 0) \rightsquigarrow (|X|, |Y|)$
- ◆ Idea²: Trim AG to $O(k)$ diagonals $\rightsquigarrow \text{ED}_{\leq k}^W(X, Y)$ is distance $(0, 0) \rightsquigarrow (|X|, |Y|)$
- ◆ Idea³: Split AG according to structure of X and Y
 Compute all b-to-b dist [Kleino5] + stitch together results ((min, +)-product [SMAWK87])



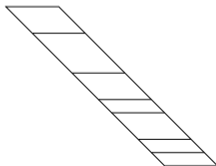
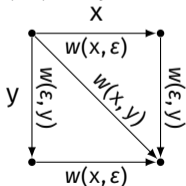
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 for periodic pieces: fast exponentiation;
 $\rightsquigarrow k^2 \cdot \tilde{O}(k^2)$ for periodic pieces + $k^2 \cdot \tilde{O}(k^2)$ for stitching



Tool 2: Divide and Conquer

- ◆ X and Y consist in $O(k^2)$ periodic pieces of length $O(k^2)$ and with period $O(k)$ each
- ◆ Idea: Use AG, trimmed to $O(k)$ diags $\rightsquigarrow \text{ED}_{\leq k}^w(X, Y)$ is distance $(0, 0) \rightsquigarrow (|X|, |Y|)$
- ◆ Idea³: Compute all b-to-b dist for periodic pieces [Kleino5] + fast exponentiation;
stitch together results using min-plus product [SMAWK87]
- ◆ Idea⁴: Use Divide-and-Conquer to reduce number of periodic pieces to $O(k)$
 $\rightsquigarrow k \cdot \tilde{O}(k^2)$ for periodic pieces (and padding) + $k \cdot \tilde{O}(k^2)$ for stitching

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Tool 3: Compressibility instead of Periodicity

- ◆ X and Y consist in $O(k)$ pieces of length $O(k^2)$ each
- ◆ Idea: Use AG, trimmed to $O(k)$ diags $\rightsquigarrow \text{ED}_{\leq k}^w(X, Y)$ is distance $(0, 0) \rightsquigarrow (|X|, |Y|)$
- ◆ Idea³: Compute all b-to-b dist for periodic pieces [Kleino5] + fast exponentiation;
stitch together results using min-plus product [SMAWK87]
- ◆ Idea⁴: Use Divide-and-Conquer to reduce number of periodic pieces to $O(k)$
- ◆ Idea⁵: Use tailor-made compressibility measure instead of periodicity
+ (w)ED algorithms for compressed strings
 $\rightsquigarrow \tilde{O}(n + \sqrt{k^3 n})$ time in total

Main Results

Main Theorem 1 (Upper Bound) ✓

Strings X, Y each of length at most n
Oracle access to (normalized) weight function w
Can compute $k = \text{ED}^w(X, Y)$ in time $O(n + \sqrt{nk^3} \log^3 n)$

Main Theorem 2 (Lower Bound)

Assuming the APSP Hypothesis and for $\sqrt{n} \leq k \leq n$,
Main Theorem 1 is tight (up to $n^{o(1)}$ -factors)

- ◆ Have good algorithm for Bounded Weighted Edit Distance
- ◆ (ongoing work) Use structural result and directed alignment graph for $\tilde{O}(|T| + k^4 |T| / |P|)$ for PM w/ WE
- ◆ Seaweeds don't help,
but fast multiplication of Monge Matrices still does (at a cost of an extra $O(k)$)
- ◆ Still need to solve a lot of extra technical challenges
(actually obtaining an analogue of [LV89] for verifying starting positions; need to generalize DPM; ...)

Main Results of today

- ◆ Structural characterization of strings and PM w/ Edits
- ◆ Main tools required for $\tilde{O}(|T| + k^{3.5}|T|/|P|)$ algo for PM w/ E
- ◆ Some ideas of how to generalize to the weighted setting

Big Open Questions

- ◆ PM w/ Edits in $o(|T| + k^3 \cdot |T|/|P|)$
- ◆ PM w/ Weighted Edits in $o(|T| + k^4 \cdot |T|/|P|)$
- ◆ Better algorithms known for small integer weights [GK24⁺], might give better PM algos in that setting

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