

# Binary Jumbled Pattern Matching on Trees and Tree-Like Structures

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# Regular vs. Jumbled Pattern Matching

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- ▶ (Regular) Pattern Matching:

- ▶ Large text **T** (length **n**): **MISSISIPICITSISIS**

- ▶ Small pattern **P** (length **m**): **SISI**

- ▶ Report whether **P** occurs in **T** (or find all occurrences).

- ▶ Indexing variant:

- ▶ Preprocess **T** to answer query patterns fast.



# Regular vs. Jumbled Pattern Matching

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- ▶ Jumbled Pattern Matching:

- ▶ Large text **T** (length **n**): **MISSISIPICITSISIS**

- ▶ Small pattern **P** (length **m**): **SISI**

- ▶ Report whether some permutation of **P** occurs in **T**.

- ▶ Indexing variant:

- ▶ Same as regular pattern matching.



# Regular vs. Jumbled Pattern Matching

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	Pattern Matching	Indexing
Regular	Several classical (non-trivial) $O(n + m)$ solutions: <ul style="list-style-type: none"><li>• KnuthMorrisPratt</li><li>• BoyerMoore</li><li>• KarpRabin</li><li>• ...</li></ul>	$O(n)$ construction time and space, $\tilde{O}(m)$ query time: <ul style="list-style-type: none"><li>• Suffix Trees</li><li>• Suffix arrays</li><li>• ...</li></ul>
Jumbled	Trivial schoolbook “sliding window” solution gives $O(n+m)$ .	???



# Indices for Binary Jumbled PM

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- ▶  $O(n^2)$  construction time,  $O(n)$  space,  $O(1)$  query time [CicaleseFiciLipták '09].
- ▶  $O(n^2/\lg n)$  construction time,  $O(n)$  space,  $O(1)$  query time [BursciCicaleseFiciLipták '10] and [MoosaRahman '10].
- ▶  $O(n^2/\lg^2 n)$  construction time,  $O(n)$  space,  $O(1)$  query time [MoosaRahman '12].
- ▶ Better bounds only known for approximate indices ...



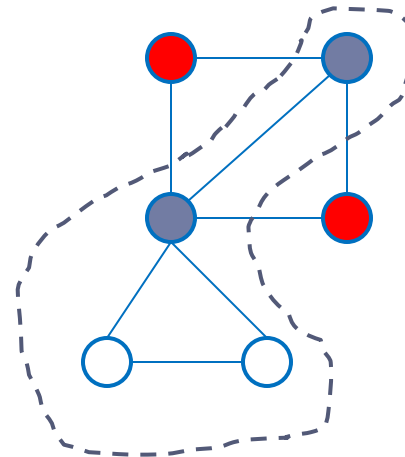
# Jumbled Pattern Matching on Graphs

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- ▶ **Text:** vertex-labeled graph on  $n$  vertices.
- ▶ **Pattern:** multiset of labels of size  $m$ .
- ▶ **Question:** Is there a connected subgraph whose label multiset matches the pattern ?

P = ● ● ○ ○

T =



# Jumbled Pattern Matching on Graphs

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- ▶ Also known as the **Graph Motif** problem.
- ▶ Several work done on this (and variants):
  - ▶ [Lacroix, Fernandes, and Sagot '06]
  - ▶ [Fellows, Fertin, H., Vialette '07]
    - ▶ An  $n^{O(cw)}$  algorithm for graphs with treewidth  $\leq w$  and #labels  $\leq c$ .
  - ▶ [Bruckner, Karp, Shamir, and Sharan '09]
  - ▶ [Dondi, Fertin, Vialette '07+'09+'11]
  - ▶ [Guillemot and Sikora '10]
  - ▶ Several others . . .
- ▶ Trees ??



# Our Results

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- ▶ Index for trees:  $O(n^2/\lg^2 n)$  construction time,  $O(n)$  bits,  $O(1)$  query time.
  - ▶ Matches the performance of the best known index for strings.
- ▶ Index for grammars:  $O(g^{2/3}n^{4/3} / \lg^{4/3} n)$  construction time,  $O(n)$  bits,  $O(1)$  query time.
  - ▶ Time-bound is  $O(n^2/\lg^2 n)$  even when the string is incompressible.
- ▶ Bounded treewidth graphs:  $f(w) \cdot n^{O(c)}$  algorithm.
  - ▶ Beats previous  $n^{O(cw)}$  algorithm.





# $O(n)$ -space index for trees in $O(n^2)$ time

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▶ Some conventions:

- ▶ The tree  $T$  is **rooted** and **complete binary** (for ease of presentation).
- ▶ Binary alphabet = tree nodes are colored either white or black.
- ▶ Query  $(i, j)$  = A connected subgraph on exactly  $i$  nodes with exactly  $j$  black nodes.

**Observation:** *If  $(i, j_1)$  and  $(i, j_2)$  both occur in  $T$ , then  $(i, j)$  also occurs in  $T$  for all  $j_1 \leq j \leq j_2$ .*

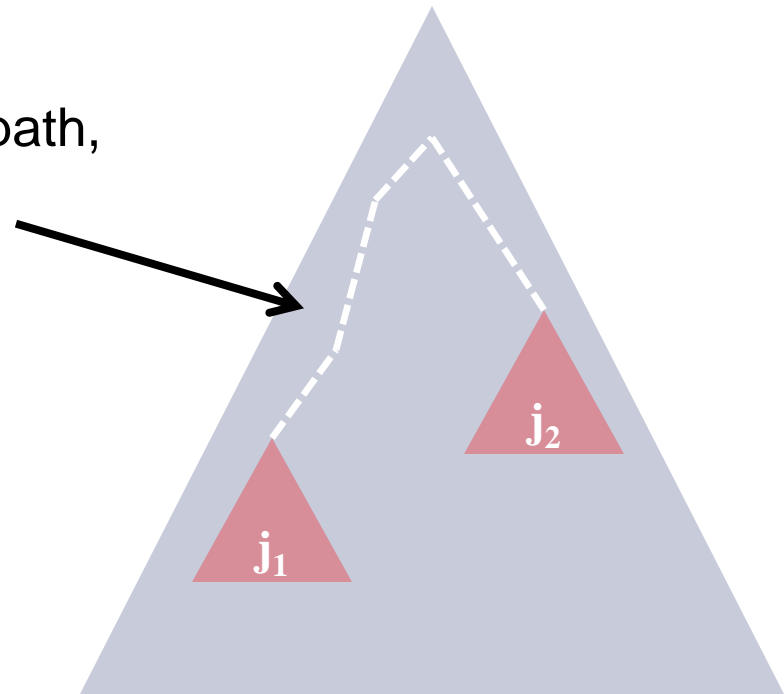


# $O(n)$ -space index for trees in $O(n^2)$ time

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**Observation:** If  $(i, j_1)$  and  $(i, j_2)$  both occur in  $T$ , then  $(i, j)$  also occurs in  $T$  for all  $j_1 \leq j \leq j_2$ .

Each time we move on this path,  
the number of black nodes  
changes by at most 1.



# $O(n)$ -space index for trees in $O(n^2)$ time

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**Observation:** If  $(i, j_1)$  and  $(i, j_2)$  both occur in  $T$ , then  $(i, j)$  also occurs in  $T$  for all  $j_1 \leq j \leq j_2$ .

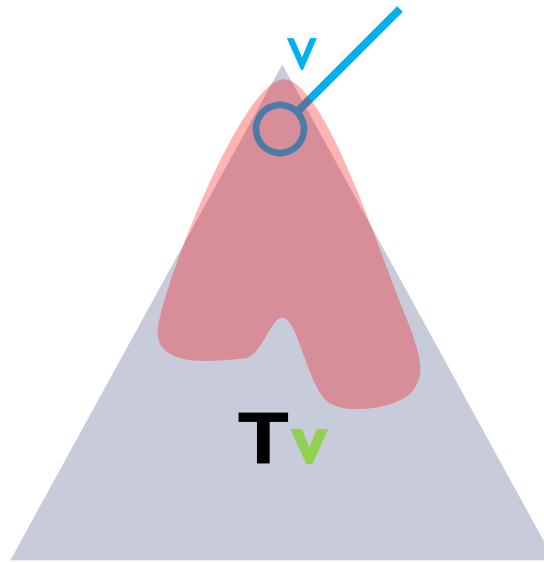
- ▶ For each  $i$ , store only two values:
  - ▶  $i_{\min}$  = minimum  $j$  such that  $(i, j)$  occurs in  $T$ .
  - ▶  $i_{\max}$  = maximum  $j$  such that  $(i, j)$  occurs in  $T$ .
- ▶ On query  $(i, j)$  report yes iff  $i_{\min} \leq j \leq i_{\max}$ .
- ▶  $O(n)$  space.
- ▶ We show how all  $i_{\max}$  values  $O(n^2)$  time.
  - ▶  $i_{\min}$  analogous.



# $O(n)$ -space index for trees in $O(n^2)$ time

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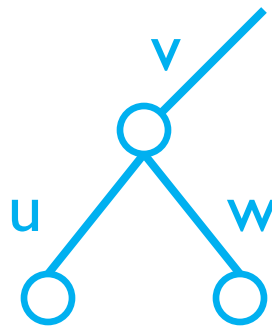
- ▶ For each node  $v$  and each  $i$  define
  - ▶  $Av[i]$  = Maximum number of black nodes in a connected subgraph of size  $i$  in  $T_v$  which includes  $v$ .



# $O(n)$ -space index for trees in $O(n^2)$ time

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- ▶ Simple top-down recursion:
  - ▶  $Av[i] = \max_k Au[k] + Aw[i-1-k] + \text{col}(v)$ .



- ▶ Looks like  $O(n^3)$  time, but its actually  $O(n^2)$  time.



# Succinct $O(n)$ -bits index

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**Observation:** Either  $Av[i] = Av[i-1]$  or  $Av[i] = Av[i-1] + 1$ .

- ▶ Thus, we can store the binary difference vectors instead:
  - ▶  $Bv[i] = Av[i] - Av[i-1]$ .
- ▶ Since  $Av[i] = \sum_{1 \leq k \leq i} Bv[k]$ , we can retrieve  $Av[i]$  from  $Bv$  in  $O(1)$  time using **rank** queries.
  - ▶  $\text{rank}[i] = \#1\text{'s in } Bv[1 \dots i]$ .



# From trees to strings

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$$B_u = 1\ 1\ 0\ 0\ 0\ 0$$



$$\#1 = A_u[4]$$

$$B_w = 0\ 0\ 1\ 1\ 1\ 0$$



$$\#1 = A_w[3]$$

$$S_v = 0\ 0\ 0\ 0\ 1\ 1 \quad \text{col}(v) \quad 0\ 0\ 1\ 1\ 1\ 0$$



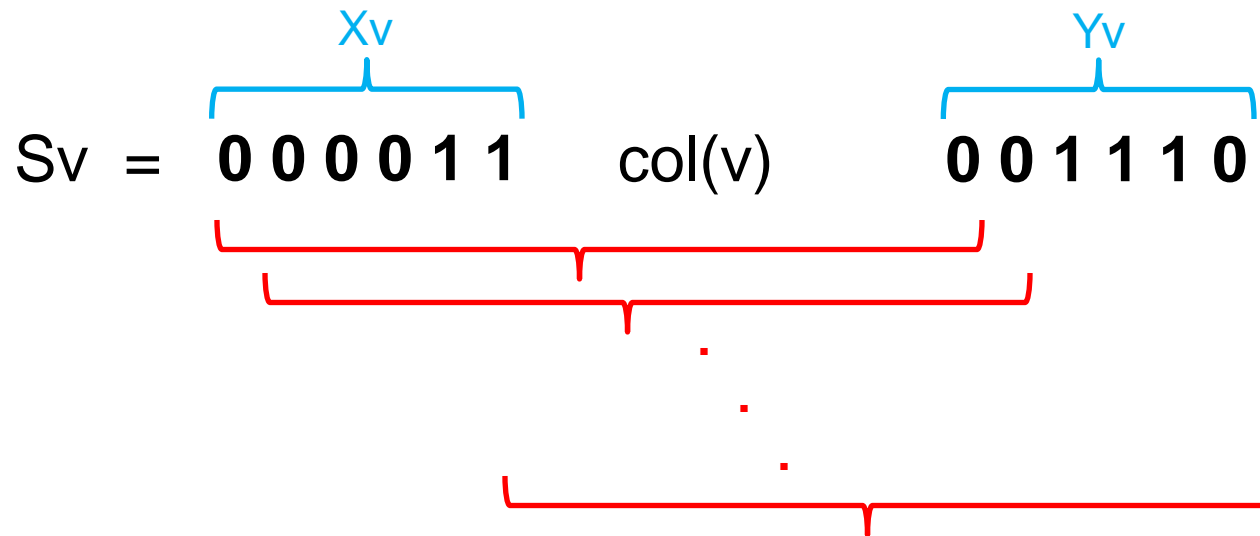
$$\#1 = A_u[4] + A_w[3] + \text{col}(v)$$



# From trees to strings

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- ▶ Recall:  $Av[i] = \max_k Au[k] + Aw[i-1-k] + \text{col}(v)$ .
- ▶ Hence,  $Av[i] = \max$  #1's in a window of size  $i$  in  $Sv$  containing the  $\text{col}(v)$  position.





# Shaving off log-factors

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- ▶ Using the algorithm for strings, we can compute  $A_v$  in  $O(|S_v|^2 / \lg^2 n)$  time.
- ▶ But in order to get  $O(n^2 / \lg^2 n)$  overall we need:
  - ▶  $O(|X_v| \cdot |Y_v| / \lg^2 n)$ , and not ...
  - ▶  $O((|X_v| + |Y_v|)^2 / \lg^2 n)$ .
- ▶ To get  $O(|X_v| \cdot |Y_v| / \lg n)$  is relatively easy.
  - ▶ Use lookup tables of size  $s \approx \lg n$ .
  - ▶ Run sliding windows of sizes which are multiples of  $s$ .
  - ▶ In total,  $O(n / \lg n)$  sliding windows.

as is done in  
MoosaRahaman



# Shaving off log-factors

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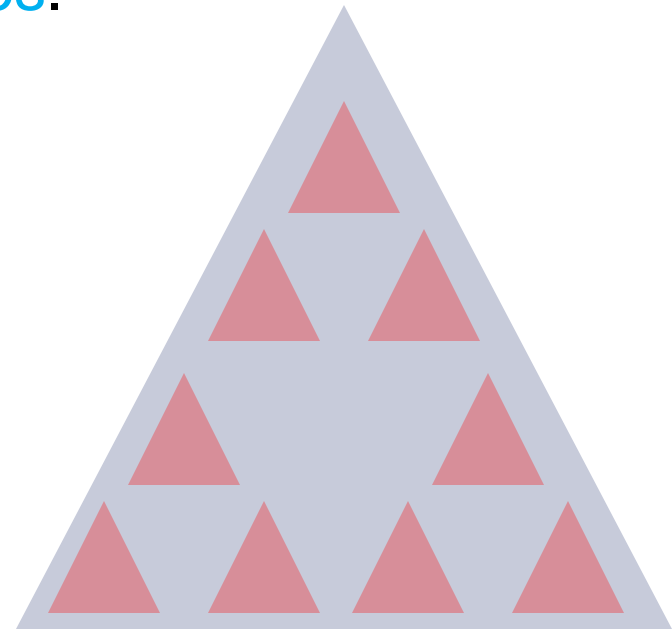
- ▶ To get  $O(|Xv| \cdot |Yv| / \lg^2 n)$  is problematic:
  - ▶ In strings, Moosa and Rahaman use additional lookup tables to slide the window in jumps of size  $s \approx \lg n$ .
  - ▶ Here we can also do this in most cases.
  - ▶ But what happens when, e.g.,  $|Xv| < s$  ?
    - ▶ Since we only consider sliding windows which include the  $\text{col}(v)$  position, we cannot jump !
    - ▶ In this case, we get a running-time of  $O(|Yv| / \lg n) = O(n / \lg n)$ .
  - ▶ **Solution:** Use **micro-macro** decomposition to ensure that the computations above happen only  $O(n / \lg n)$  times.



# Micro-macro decomposition

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- ▶ Decompose the tree into a **macro-tree** of disjoint connected subgraphs, aka **micro-trees**.
  - ▶ Each micro-tree has size  $\leq \lg n$ .
  - ▶ # micro-trees =  $O(n / \lg n)$ .
- 1. Compute (essentially)  $A_v$  arrays for each micro-tree. Requires  $O(\lg^2 n \cdot n / \lg n) = O(n \lg n)$  time.
- 2. In bottom-up fashion, merge micro arrays to macro arrays using string speedups. Requires  $O(n / \lg n \cdot n / \lg n) = O(n^2 / \lg^2 n)$  time.



# Closing Remarks

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- ▶ We can also find a node of an occurrence in  $O(\lg n)$  time, assuming  $(i,j)$  occurs in  $T$ .
- ▶ Our algorithm can be made into a pattern matching algorithm running in  $O(nm/\lg^2 n)$  time, when the pattern is of size  $m$ .
  - ▶ Can this be improved to near-linear (recall the string case) ?
- ▶ Bigger alphabets ?
- ▶ **A reasonable index for strings ?**



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Thank you  
for  
your attention

