

# Finding Optimal Pairs of Cooperative and Competing Patterns with Bounded Distance

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# Pattern Discovery

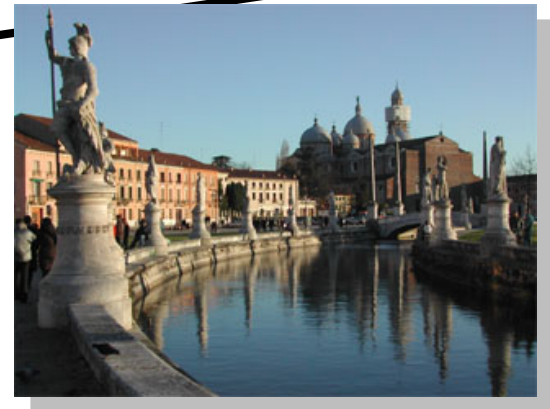
Input : set  $S$  of strings  $s_1, s_2, \dots, s_m$

Output : pattern  $p$  that characterizes input  $S$

- $S$
- ◆ Pattern discovery is a core of Discovery Science.
  - ◆ Jump and dive into the sea during summer vacation.



$p = \text{padova}$



# Pattern Discovery in Bioinformatics

## Machine Discovery System BONSAI

(Shimozono et al. 1994)

- finds **optimal substring** patterns to characterize input sequence set **S**.
- has been extended to finding:

**subsequence** patterns (Hirao et al. 2000)

**episode** patterns (Hirao et al. 2001)

**VLDC** patterns (Inenaga et al. 2002)

**approximate FVLDC** patterns (Takeda et al. 2003)



Refer to the invited talk by Ayumi Shinohara in DS'04

**"String Pattern Discovery"**

# Composite Pattern Discovery





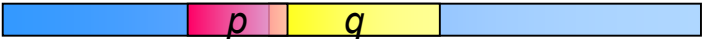
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- More than one sequence element may act in ensemble!
- Sequence elements may not necessarily be **cooperative** - they may be **competitive**.
- Consider finding **Boolean combination** of patterns.

e.g.:  $p \wedge q$  :  $p$  AND  $q$   
 $p \vee q$  :  $p$  OR  $q$   
 $p \wedge \neg q$  :  $p$  AND (NOT  $q$ )

# Examples

$$S = \{s_1, s_2, s_3, s_4, s_5\}$$

		$p$	$q$	$p$ OR $q$	$p$ AND (NOT $q$ )
$s_1$		true	false	true	true
$s_2$		false	true	true	false
$s_3$		false	false	false	false
$s_4$		true	true	true	false
$s_5$		true	true	true	false

  substrings: exact match only (no mismatches)

# Recent Work by Bannai et al.

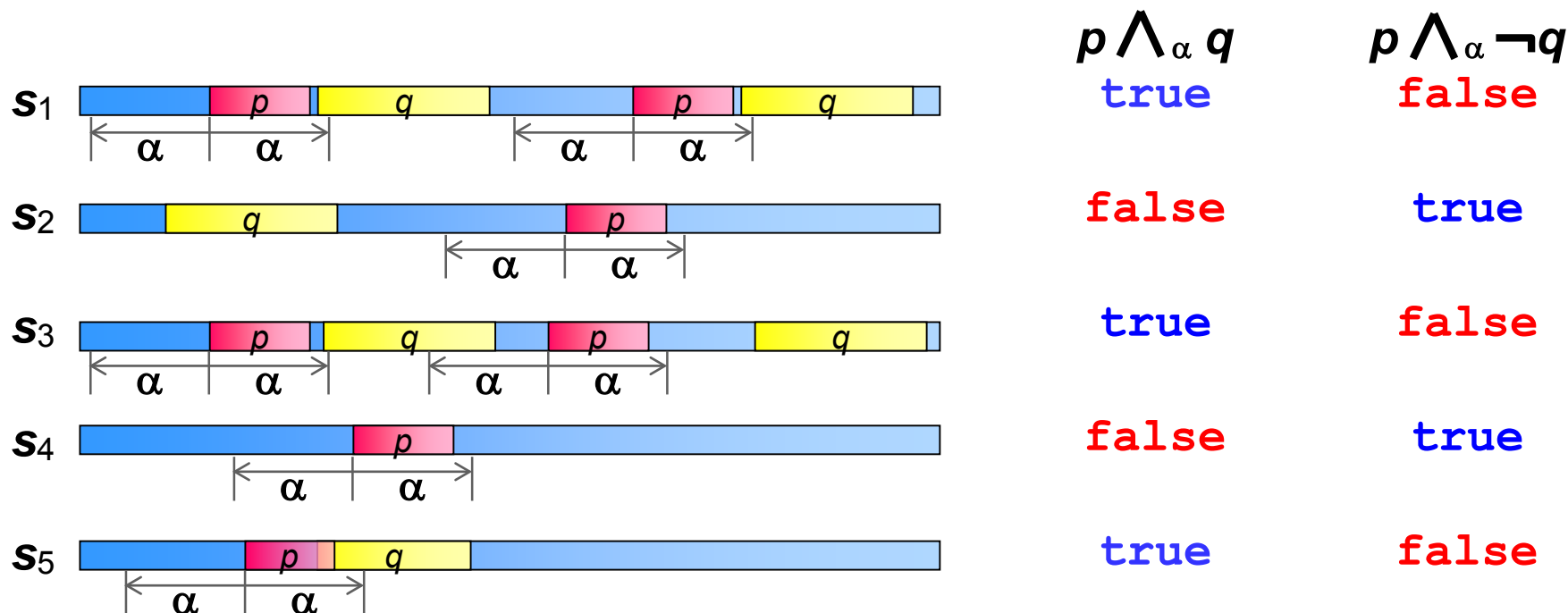
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## "Finding Optimal Pairs of Patterns" (Bannai et al. in WABI'04)

- Efficient algorithm for finding optimal Boolean pattern pair of substring patterns
- $O(N^2)$  time &  $O(N)$  space  
( $N$ : total length of strings in input set  $S$ )
- $O(N^k)$  time &  $O(N)$  space  
for Boolean combination of  $k$  substring patterns

# And This Work...

- ◆ Introduces the notion of  $\alpha$ -distance between patterns in Boolean pair w.r.t.  $\wedge$  (AND).
- ◆ Denoted  $p \wedge_{\alpha} q$  and  $p \wedge_{\alpha} \neg q$ .



# Our Result

Input : set  $S$  of strings and distance  $\alpha$

Output : optimal pattern pair  $p \wedge_{\alpha} q$  and  $p \wedge_{\alpha} \neg q$  w.r.t.  $S$

- $O(N^2)$  time &  $O(N)$  space (not depending on  $\alpha$ )
- $O(N^k)$  time &  $O(N)$  space  
for combination of  $k$  substring patterns
  - Improves the worst case complexity  $O(\alpha^k N^{k+1} \log N)$  due to Arimura et al. (proximity patterns).



# Optimality of Pattern Pair

Pattern pair  $\pi$  is optimal w.r.t.  $\mathbf{S}$



Pattern pair  $\pi$  maximizes  $\text{score}(M(\pi, \mathbf{S}))$

Examples of *score* function:

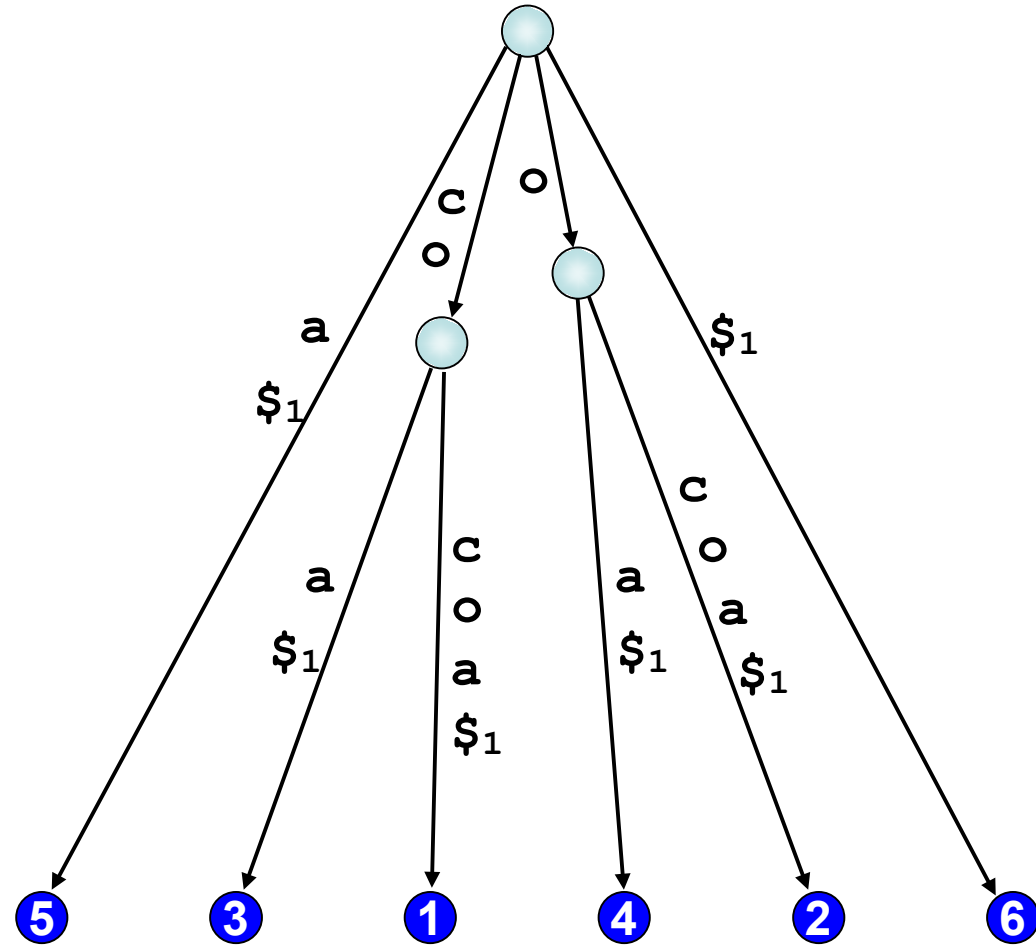
- Gini index
- Chi-square statistic
- Rank-sum test

$M(\pi, \mathbf{S})$  : num. of strings in  $\mathbf{S}$  that  $\pi$  matches.  
Assume *score* can be computed in  $O(1)$  time.

# Suffix Tree

## Suffix Tree of $s$ ( $ST(s)$ )

- Tree structure which represents all suffixes of  $s$ .
- Each leaf is marked with its suffix number.
- $ST(s)$  can be constructed in  $O(n)$  time ( $n = |s|$ ).  
(Weiner 1973, McReight 1976, Ukkonen 1995, etc.)



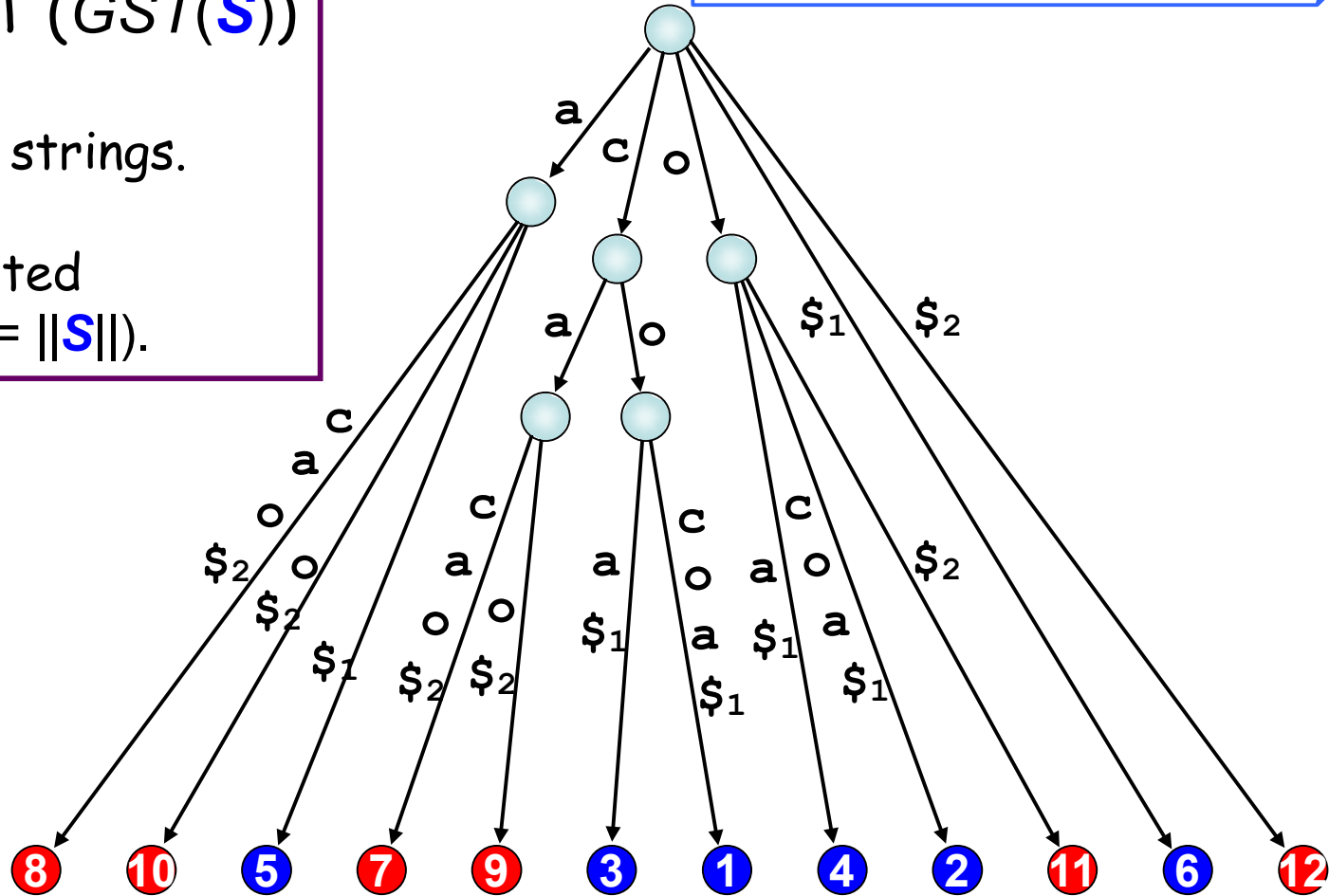
$s = \text{cocoa}\$1$   
123456

# Generalized Suffix Tree

$path(v)$  = concatenation of labels from *root* to node  $v$ .

## Generalized ST (GST( $\mathbf{S}$ ))

- ST for set  $\mathbf{S}$  of strings.
- Can be constructed in  $O(N)$  time ( $N = \|\mathbf{S}\|$ ).



$$\mathbf{S} = \{\text{cocoa}\$1, \text{cacao}\$2\}$$

# Find Best Single Pattern

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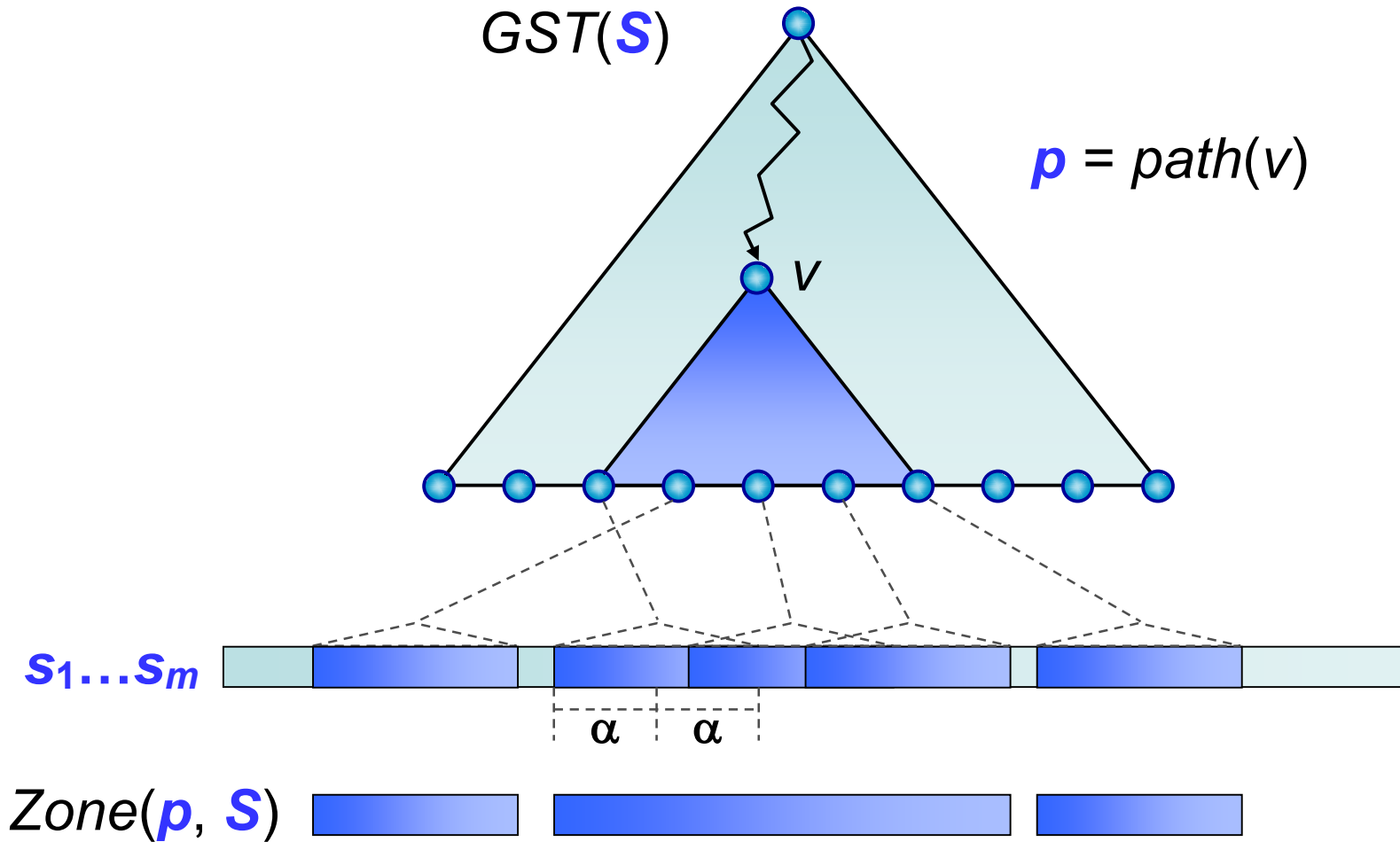
- We can restrict the candidates for pattern  $p$  to those represented by the nodes of  $GST(\mathbf{S})$ .
  - there are only  $O(N)$  nodes
- For all nodes  $v$  in  $GST(\mathbf{S})$ ,  $M(\text{path}(v), \mathbf{S})$  is computable in total  $O(N)$  time.  
(Color Set Size Problem, Hui 1992)
- Output  $\text{path}(v)$  of the best score.
- $O(N)$  time & space for single pattern case.

# Find Best Pair $p \wedge_{\alpha} q$

## Algorithm Sketch:

- For each candidate of first pattern  $p$ 
  1. Compute **Zone**( $p, S$ ) - the region covered by  $\alpha$  from each position of  $p$  in  $S$ .
  2. Build **sparse** suffix tree (SST) on **Zone**( $p, S$ ).
  3. For each node  $u$  in SST, compute score.
- Output pattern pair of the best score.

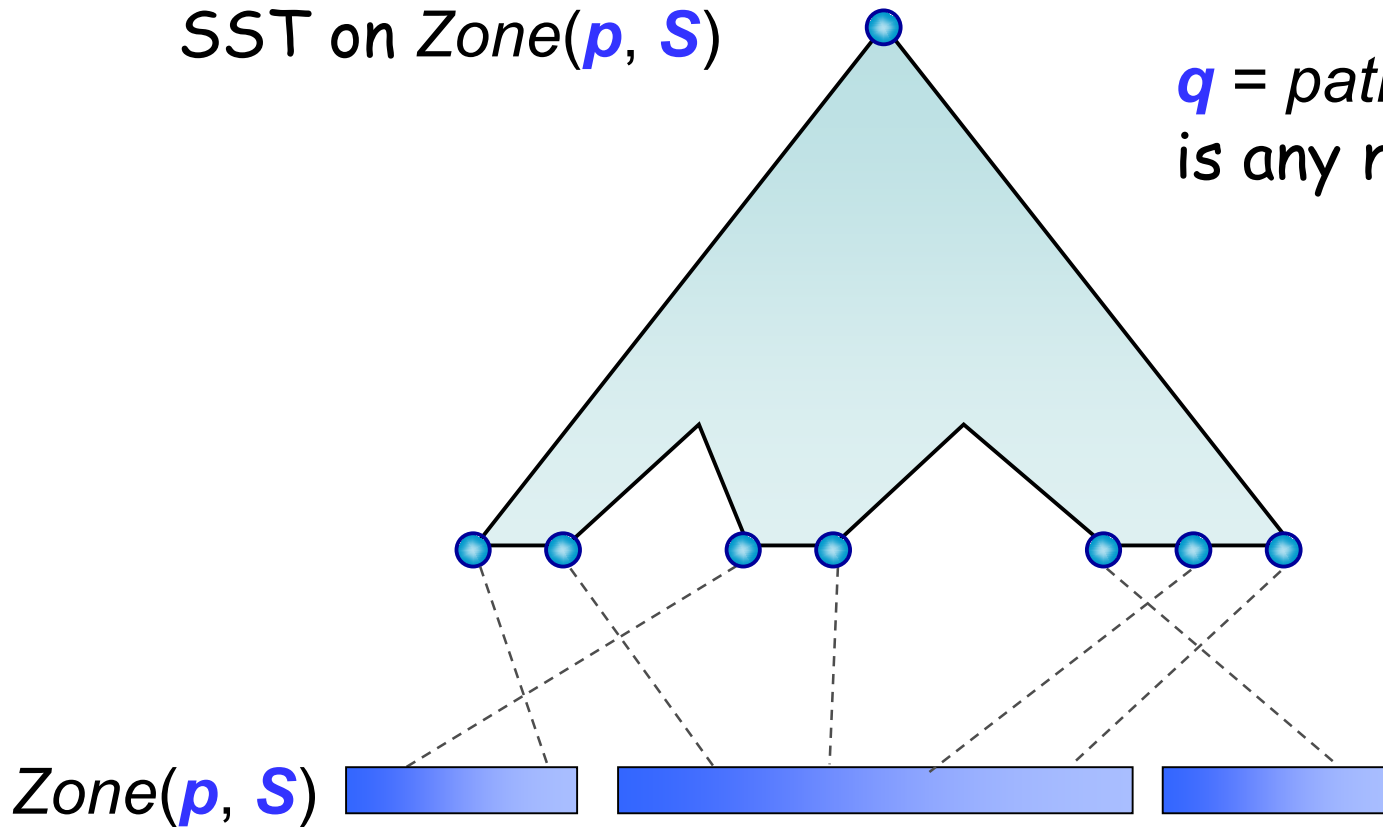
# Zone( $p, S$ )



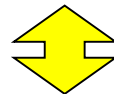
# Build SST on Zone( $p$ , $S$ )

SST on Zone( $p$ ,  $S$ )

$q = \text{path}(u)$  where  $u$  is any node in SST



$q = \text{path}(u)$  for some node  $u$  in SST on Zone( $p$ ,  $S$ )



$q$  is such a pattern that  $p \wedge_{\alpha} q$  matches  $S$

# Analysis

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- We have  $O(N)$  choices for first pattern  $p$ .
  - For each candidate  $p$ 
    - Compute  $\text{Zone}(p, S)$  -  $O(N)$  time
    - Construct SST on  $\text{Zone}(p, S)$  -  $O(N)$  time
- Thus it takes  $O(N^2)$  time in total.
- We need  $O(N)$  space since we use one SST at each stage of the algorithm.



# Find Best Pair $p \wedge_{\alpha} \neg q$

$O(N^2)$  time

$O(N)$  time

$O(N^2)$  time

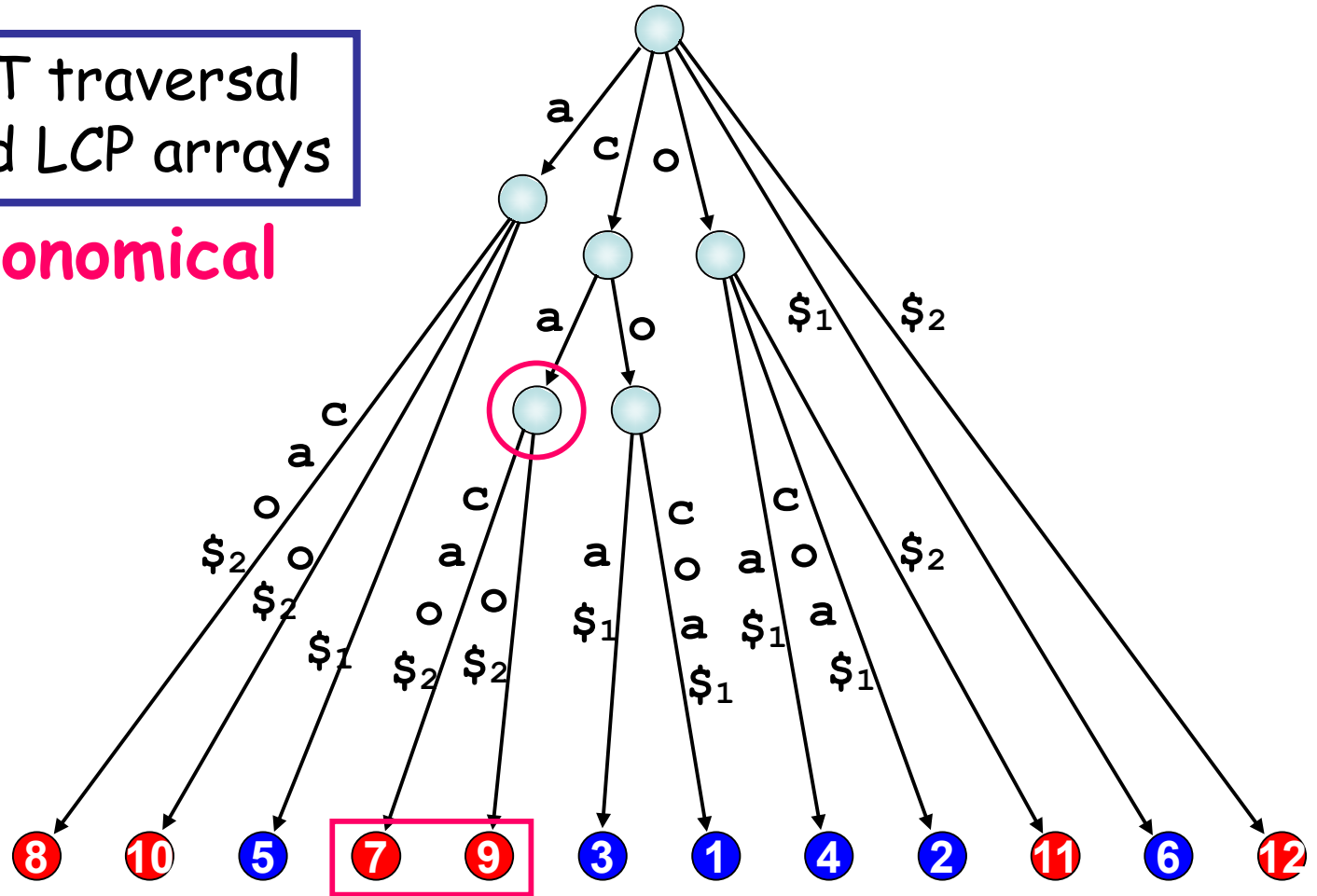
$$M(\bar{\pi}, S) = M(p, S) - M(\pi, S)$$

$$\pi = p \wedge_{\alpha} q, \quad \bar{\pi} = p \wedge_{\alpha} \neg q$$

# Implementation with Suffix and LCP Arrays

Simulate GST traversal  
by suffix and LCP arrays

→ **space-economical**



SA(**S**)

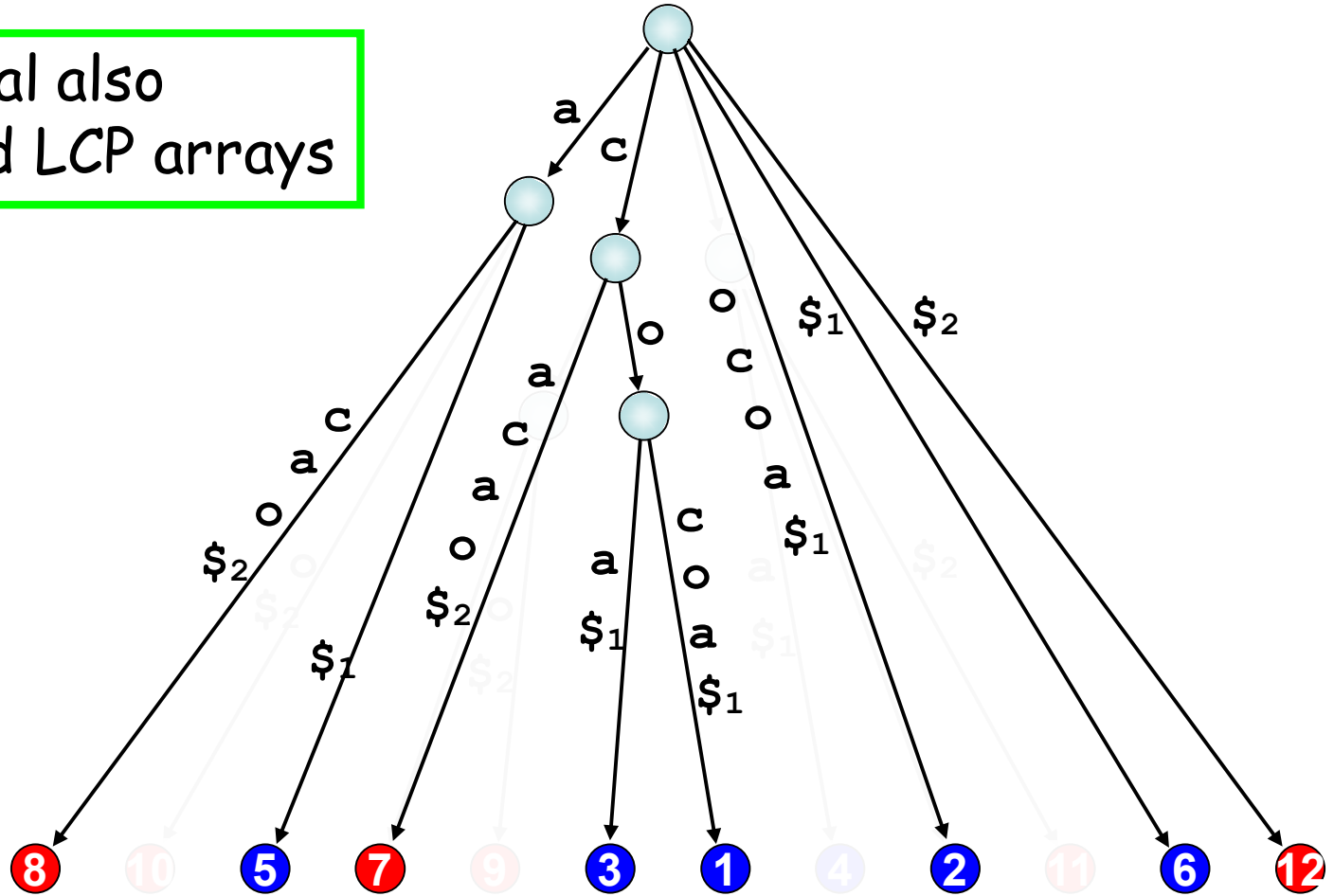
8	10	5	7	9	3	1	4	2	11	6	12
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LCP(**S**)

0	1	1	0	2	1	2	0	1	1	0	0
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# Implementation with Suffix and LCP Arrays

SST traversal also by suffix and LCP arrays



SA( <b>S</b> )	8	-	5	7	-	3	1	-	2	-	6	12
LCP( <b>S</b> )	0	1	1	0	2	1	2	0	1	1	0	0

# Extended to Correlated Patterns

- ◆ Each sequence  $s_i$  in  $S$  may be associated with numeric value  $r_i$ , such as **gene expression level**.
  - ◆ Wanted:  
pattern pair  $\pi$  that matches sequences  $s_i$  with high  $r_i$ ,  
but doesn't match sequences  $s_j$  with low  $r_j$  (or vice versa).
- pattern pair  $\pi$  that maximizes  $\text{score}(M(\pi, S), R(\pi, S))$ .

- $O(N^2)$  time &  $O(N)$  space to find optimal pairs  $p \wedge_{\alpha} q$  and  $p \wedge_{\alpha} \neg q$

# Computational Experiments

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## Yeast mRNA

- 3'UTR predicted processing site sequences (100nt each)

- 379 fast degrading ( $t_{0.5} < 10$  min.)

- 393 slowly degrading ( $t_{0.5} > 50$  min.)

Divided according to mRNA decay rate measurements (Wang et al. 2002, Graber 2003)

- score function: chi-squared test statistic

# Computational Experiments (Contd.)



AUA  $\wedge_{10}$  UGUA

fast  
159/393

slow  
248/379

known binding site of the PUF protein  
which is important in mRNA regulation

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AUA  $\wedge$  UGUA

268/393

190/379

AUA  $\wedge_{10}$  UGUA

231/393

123/379

may influence how efficiently UGUA  
functions, when it is close to UGUA??

# Conclusions

- $O(N^2)$  time  $O(N)$  space algorithm to find optimal pattern pair with bounded distance  $\alpha$
- Efficient implementation with suffix arrays
- Biologically relevant patterns discovered
- It can be extended to more advanced versions of bounded distance:
  - B)  $O(m^2N^2)$  time &  $O(m^2N)$  space
  - C)  $O(N^3)$  time &  $O(N^2)$  space
  - D)  $O(m^2N^3)$  time &  $O(N^2+m^2N)$  space