

# Multiple Sequence Alignment Based on Profile Alignment of Intermediate Sequences

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# Content

- Biology Motivation
- Computation Problem
- Algorithm
- Performance

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- **Biology Motivation**
- Computation Problem
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# Biology Motivation

- Multiple Sequence Alignment:
  - Assess sequence conservation of protein domains, tertiary and secondary structures and even individual amino acids or nucleotides.
  - Evolutionary relationships or sequence conservation among homologous.
  - Simultaneously compare several sequences.

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# Computation Problem

- Methods:
  - Pairwise alignments
  - Prograssive alignment construction
  - Iterative methods
  - Hidden Markov models
- Problems:
  - Accuracy
  - Computational complexity

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# Algorithm-Introduction

- Incorporate additional hits into the input sequences
  - Hits that are not intermediate will introduce noise
  - Use carefully defined intermediate sequences
- Align profiles instead of the sequences
  - Construct a profile for each sequence
  - Align the profiles by modifying the pair-HMM
  - Obtain a secondary structure prediction

# Algorithm

- Finding intermediate sequences
- Choosing intermediate sequences
- Constructing sequence profiles
- Alignment via modified pair-HMM

# Finding Intermediate Sequence

- Definitions of Intermediate Sequence
  - Between two input sequences:

**Definition 1.** Given two sequences  $s_1$  and  $s_2$ , and a distance score  $d(s_1, s_2)$  between them, a sequence  $r$  is intermediate between  $s_1$  and  $s_2$  if  $d(r, s_1) < d(s_1, s_2)$  and  $d(r, s_2) < d(s_1, s_2)$ .

- Between multiple sequences:

**Definition 2.** Given  $n$  input sequences  $s_1, \dots, s_n$ , and  $m$  hits  $r_1, \dots, r_m$  from database search of these sequences, find all hits  $r_k$  that are intermediate between some pair of input sequences  $s_i$  and  $s_j$ .

# Finding Intermediate Sequence

- No need to compute pairwise distances between the potentially very large number of hits.
- The number of pairwise distance score computations:  $O(mn+n^2)$
- The number of score comparisons is  $O(mn^2)$ .

# Choosing Intermediate Sequences

- The number of intermediate sequences can be very large
- Use a subset of intermediate sequences
- Similar sequences are likely to contain redundant information
- Choose a small subset of intermediate sequences using a greedy strategy
- Goal: identify a combined set of sequences as divergent as possible

# Choosing Intermediate Sequences

- Definition

**Definition 3.** Given  $n$  input sequences  $s_1, \dots, s_n$ ,  $m$  intermediate sequences  $r_1, \dots, r_m$ , add  $k$  intermediate sequences from among  $r_1, \dots, r_m$ , denoted by  $s_{n+1}, \dots, s_{n+k}$ , so that the minimum distance between sequences in the combined set  $s_1, \dots, s_{n+k}$  is the largest possible when distances between the input sequences  $s_1, \dots, s_n$  are ignored.

# Choosing Intermediate Sequences

- Greedy algorithm

Input:  $n$  input sequences  $s_1, \dots, s_n$ ,  $m$  intermediate sequences  $r_1, \dots, r_m$ ,  
distance score  $d(r, s)$  between two sequences  $r$  and  $s$ .

Output:  $k$  intermediate sequences  $s_{n+1}, \dots, s_{n+k}$  added to  $s_1, \dots, s_n$ .

$R \leftarrow \{r_1, \dots, r_m\};$

for each  $r_i$  in  $R$  do {  $d_i \leftarrow \min_{1 \leq j \leq n} d(r_i, s_j);$  }

for  $j \leftarrow 1$  to  $k$  do {

$s_{n+j} \leftarrow r_i$  with the maximum  $d_i$ ; remove  $r_i$  from  $R$ ;

for each  $r_i$  in  $R$  do {  $d_i \leftarrow \min(d_i, d(r_i, s_{n+j}));$  } }

# Choosing Intermediate Sequences

- Iteratively add the farthest intermediate sequence.
- Does not guarantee optimum divergence, but still reasonable.
- The number of pairwise score computations is  $O(m(n+k))$ .

# Constructing Sequence Profiles

- Assign each intermediate sequence  $r_i$  ( $i=1..m$ ) to the most similar sequence  $s_j$  ( $j=1..n+k$ ).
- Use star alignment for each sequence  $s_j$  and the intermediate sequence assigned to it.
- The relative frequency of each residue of  $s_j$  is used to construct a profile as a probability distribution.

# Constructing Sequence Profiles

- If the number of very closely related sequences assigned to  $s_j$  is very large, it will have over-contribution.
- Solution: before choosing intermediate sequences, remove sequences from the original set so that none of the remaining sequences are very similar to each other.

# Modified Pair-HMM

- Original model:

- |             |          |        |        |
|-------------|----------|--------|--------|
| X           | x        | x      | -      |
| Y           | y        | -      | y      |
| Probability | $e(x,y)$ | $e(x)$ | $e(y)$ |

- $\delta$ : the gap opening probability
- $\varepsilon$ : the gap extension probability

# Modified Pair-HMM

- Add the probability distribution of residues at each position:

- $p_1(x, i)$ : residue  $x$  at position  $i$  in  $X$ .

- $p_2(y, j)$ : residue  $y$  at position  $j$  in  $Y$ .

- New emission probability of state M:

$$e'(i, j) = \sum_x \sum_y p_1(x, i) p_2(y, j) e(x, y)$$

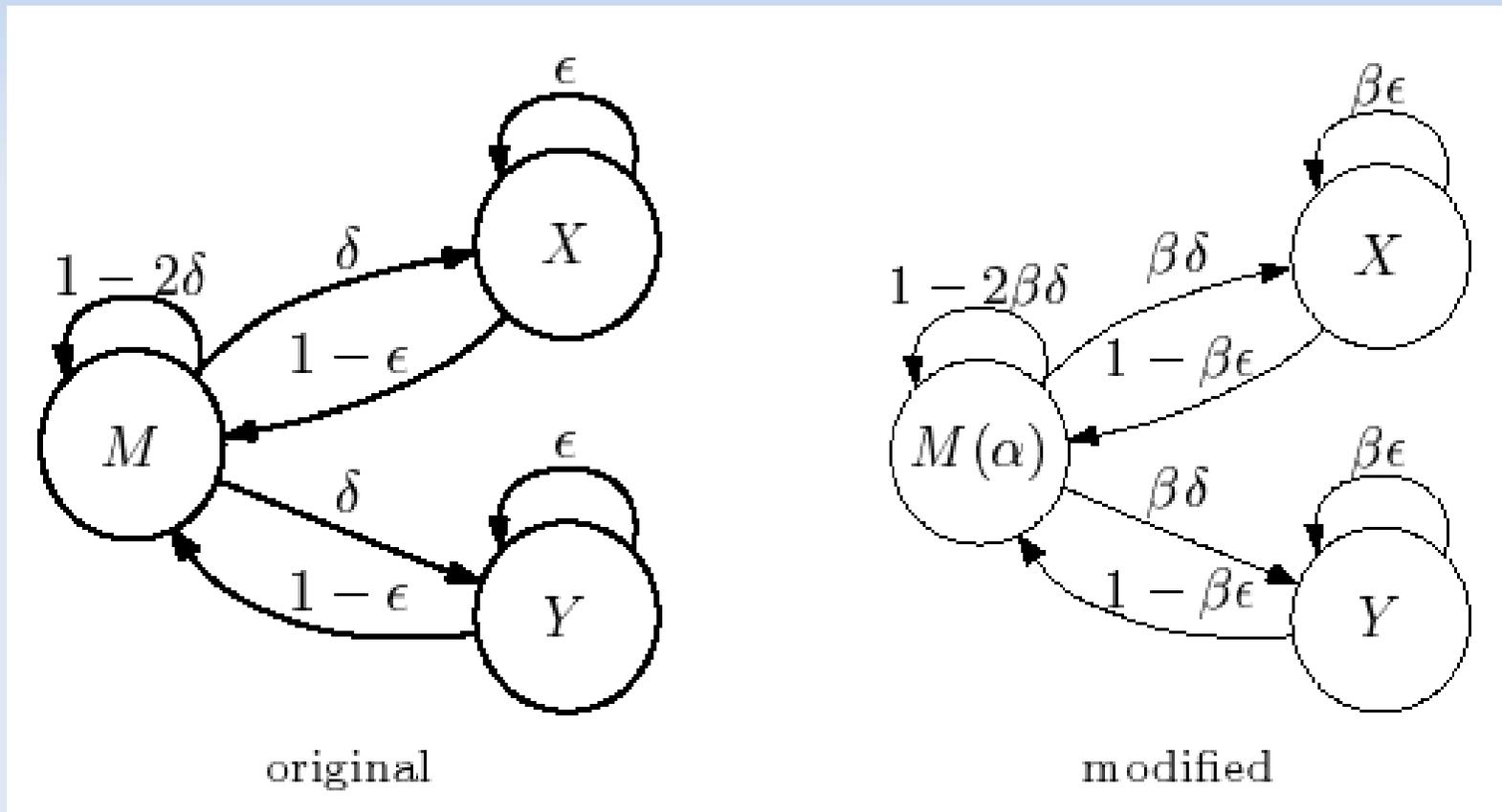
$$e'(i) = \sum_x p_1(x, i) e(x) \quad e'(j) = \sum_y p_2(y, j) e(y)$$

# Modified Pair-HMM

- Secondary structure predictions:
  - In state  $M$ , introduce an additional parameter  $\alpha$
  - Subdivide the emission probability  $e'(i,j)$  into two cases to obtain the state  $M(\alpha)$  with emission probability  $\alpha e'(i,j)$  if  $(x,y)$  at position  $i$  in  $X$  and  $j$  in  $Y$  have the same secondary structure type.
  - $(1-\alpha)e'(i,j)$  otherwise.
- Decrease in emission will allow more gaps:
  - Use  $\beta$  to compensate for the change

# Modified Pair-HMM

- Secondary structure prediction



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# Performance

- Benchmark Sets:
  - BALiBASE 3.0
  - HOMSTRAD
  - PREFAB
  - SABmark
- Compare with:
  - MAFFT 5.8
  - ProbCons 1.10
  - SPEM

# Performance

	SPS				CS			
	MAFFT	ProbCons	SPEM	ISPAAlign	MAFFT	ProbCons	SPEM	ISPAAlign
1V1 {38}	64.8	64.5	73.1	<b>76.0</b>	44.6	40.4	51.6	<b>56.9</b>
1V2 {42}	92.8	93.4	92.1	<b>93.5</b>	83.9	85.6	82.6	<b>85.8</b>
1 (V1-V2) {80}	79.5	79.7	83.1	<b>85.2</b>	65.2	64.2	67.9	<b>72.1</b>
(vs MAFFT)			(4e-5)	(5e-8)			(0.01)	(2e-7)
(vs ProbCons)			(7e-4)	(2e-6)			(0.01)	(2e-5)
(vs SPEM)				(0.002)				(9e-5)
2 {37}	91.8	89.7	88.0	<b>91.9</b>	46.0	40.8	47.1	<b>53.8</b>
3 {29}	81.4	78.8	82.8	<b>83.5</b>	56.8	54.3	51.4	<b>59.9</b>
4 {36}	89.2	86.8	87.5	<b>90.3</b>	<b>67.9</b>	60.9	55.4	63.3
5 {14}	88.2	87.5	87.0	<b>90.3</b>	57.6	59.4	55.9	<b>63.9</b>
All (1-5) {196}	84.5	83.3	85.0	<b>87.5</b>	60.3	57.3	58.3	<b>64.6</b>
(vs MAFFT)			(0.005)	(2e-11)			(—)	(2e-10)
(vs ProbCons)			(5e-4)	(2e-13)			(—)	(4e-10)
(vs SPEM)				(3e-7)				(5e-11)

# Performance

	SPS			CS			SPEM	ISPAIalign	ISPAIalign
	ProbCons	SPEM	ISPAIalign	ProbCons	SPEM	ISPAIalign	(vs ProbCons)	(vs ProbCons)	(vs SPEM)
0-20% {156}	49.7	67.2	<b>68.5</b>	43.1	61.0	<b>62.7</b>	(4e-23)	(5e-24)	(4e-5)
20-40% {459}	80.5	85.6	<b>86.8</b>	74.7	80.4	<b>81.9</b>	(2e-29)	(2e-53)	(7e-7)
40-70% {348}	94.8	94.9	<b>95.5</b>	92.2	92.3	<b>93.2</b>	(0.03)	(2e-9)	(0.003)
70-100% {69}	<b>99.1</b>	98.5	99.0	<b>99.1</b>	98.4	98.9	(0.007*)	(—)	(—)
All {1032}	81.9	86.8	<b>87.8</b>	77.4	82.7	<b>84.0</b>	(2e-46)	(8e-87)	(1e-12)

	MAFFT <sup>2</sup>		MAFFT <sup>50</sup>		SP <sup>2</sup>	ISPAIalign <sup>2</sup>	ISPAIalign <sup>2</sup>	ISPAIalign <sup>2</sup>	
	ProbCons <sup>2</sup>	MAFFT <sup>50</sup>	ProbCons <sup>50</sup>	SP <sup>2</sup>	ISPAIalign <sup>2</sup>	(vs MAFFT <sup>50</sup> )	(vs MAFFT <sup>50</sup> )	(vs SP <sup>2</sup> )	
0-20% {887}	36.2	38.9	56.7	55.6	64.6	<b>64.8</b>	(3e-36)	(5e-46)	(0.03)
20-40% {588}	81.0	82.8	87.1	87.2	89.7	<b>90.1</b>	(2e-16)	(6e-28)	(0.01)
40-70% {112}	96.2	96.4	96.0	95.4	95.3	<b>97.6</b>	(0.02*)	(—)	(—)
70-100% {95}	97.9	97.8	<b>98.0</b>	97.3	97.2	<b>98.0</b>	(6e-4*)	(—)	(0.005)
All {1682}	59.4	61.4	72.3	71.7	77.3	<b>77.7</b>	(1e-46)	(7e-69)	(2e-4)

# Performance

	$f_D$			$f_M$		
	ProbCons	SPEM	ISPAalign	ProbCons	SPEM	ISPAalign
Twilight {205} (vs ProbCons) (vs SPEM)	29.3	44.2 (2e-26)	<b>46.1</b> (6e-29) (0.01)	21.0	30.8 (1e-27)	<b>32.0</b> (3e-29) (0.005)
Superfamily {422} (vs ProbCons) (vs SPEM)	57.1	68.3 (4e-49)	<b>69.0</b> (1e-51) (0.02)	43.6	50.9 (1e-48)	<b>51.6</b> (1e-51) (7e-4)

	HOMSTRAD CS					PREFAB Q				
	ProbCons	Method1	Method2	Method3	Method4	ProbCons	Method1	Method2	Method3	Method4
0-20% (vs previous)	43.1	59.1 (3e-22)	59.2 (—)	59.4 (0.04)	<b>62.7</b> (6e-8)	38.9	58.2 (2e-103)	58.6 (—)	61.3 (6e-12)	<b>64.8</b> (7e-29)
20-40% (vs previous)	74.7	79.1 (2e-24)	79.6 (0.003)	81.4 (7e-14)	<b>81.9</b> (0.005)	82.8	88.7 (9e-45)	89.0 (—)	89.7 (2e-4)	<b>90.1</b> (0.004)
40-70% (vs previous)	92.2	92.1 (—)	92.5 (8e-4)	93.1 (0.001)	<b>93.2</b> (—)	96.4	94.4 (—)	96.6 (0.002)	<b>97.8</b> (—)	97.6 (0.008*)
70-100% (vs previous)	99.1	98.2 (6e-4*)	99.1 (1e-4)	<b>99.2</b> (—)	98.9 (0.003*)	97.8	97.0 (0.04*)	96.9 (0.02)	<b>98.1</b> (—)	98.0 (—)
All (vs previous)	77.4	81.7 (5e-38)	82.2 (1e-6)	83.2 (1e-14)	<b>84.0</b> (1e-6)	61.4	73.5 (7e-146)	73.9 (—)	75.7 (2e-15)	<b>77.7</b> (4e-28)

# Future Work

- Adding intermediate sequence
  - Rather than a fixed number, the number to add depends on the number of the input.
  - Or until the minimum distances fall below a threshold.
- Retain the pair-HMM using a set of confirmed secondary structures.
- Use other profile method
- Use 3D structures if possible

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Thank you!

Questions or Comment?