

# ppALIGN: A tool to assess the uncertainty in score-based alignments

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# Outline

- 1 Score-based alignment
  - Notion of alignment
  - Finding the best alignment
  - Example
- 2 Probabilistic alignment
  - Alignment biases
  - Gibbs-Boltzman distribution
- 3 Posterior probabilities
  - Examples
  - Partition function calculation
  - ppALIGN in action

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# How to compare two sequences ?

## Problem

How to compare two biological sequences (DNA, proteins)  $X = X_1 \dots X_n$  and  $Y = Y_1 \dots Y_m$  ? Should we:

- compare their respective lengths ?
- compare their compositions (letters, word of size 2, 3, ...)
- look for repetitions ?
- ...

⇒ What about their **proximity** in the evolution process ?

# Mutation and Indel

## Definition

During the evolution of a biological sequence, a letter that changes is called a **mutation** and we called **indel** either the insertion of a letter or the deletion of a letter.

## Example (a DNA sequence)

a	c	c	g	t	t	a	c	a	a	g	a	c	a		
a	c	c	g	t	t	a	c	a	a	g	a	c	a		
a	c	c	g	t	t	a	c	a	a	g	a	c	a		
									•						
a	c	•	g	t	t	a	c	a	t	g	a	c	a		
a	c		g	t	t	a	•	g	c	a	t	g	a	c	a

# What is an alignment ?

## Example (Two DNA sequences)

$$X = \text{a c g t a g c a t g a c a}$$

$$Y = \text{a c c g t a c a a g c a}$$

We denote by  $Z$  a **common ancestral sequence**:

$Z$	a	c	c	g	t	t	a	c	a	a	g	a	c	a	
$X$	a	c		g	t		a	g	c	a	t	g	a	c	a
$Y$	a	c	c	g	t		a		c	a	a	g		c	a

here is the **alignment** we get:

$\tilde{X}$	a	c	-	g	t	-	a	g	c	a	t	g	a	c	a
$\tilde{Y}$	a	c	c	g	t	-	a	-	c	a	a	g	-	c	a

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# Scoring Alignments

## Definition (Score of an Alignment)

Using the **scoring function**  $\sigma : \mathcal{S} \cup \{-\} \times \mathcal{S} \cup \{-\} \rightarrow \mathbb{R}$  we define the **score of an alignment** as the sum of the scoring function over all the columns of the alignment.

**Example** ( $\sigma(\text{match}) = +1$   $\sigma(\text{mismatch}) = -1$   $\sigma(\text{gap}) = -2$ )

the first alignment scores  $8 \times 1 - 0 \times 1 - 9 \times 2 = -10$

$\tilde{X}$	a	c	-	g	t	a	-	-	-	g	c	a	t	g	a	c	a
$\tilde{Y}$	a	c	c	g	t	a	c	a	a	g	c	a	-	-	-	-	-

and the second alignment scores  $10 \times 1 - 1 \times 1 - 2 \times 2 = 5$

$\tilde{X}$	a	c	-	g	t	a	g	c	a	t	g	a	c	a
$\tilde{Y}$	a	c	c	g	t	a	-	c	a	a	g	-	c	a

# How to find the best alignment ?

## Brute force

Scores individually **all possible alignments** and pick up the best one.

⇒ **How many** possible alignments ?

## Proposition

If  $N_{n,m}$  is the number of alignments between two sequences of lengths  $n$  and  $m$  we have the following **recurrence relation**:

$$N(1, m) = 2m + 1 \quad N(n, 1) = 2n + 1$$

$$N(n, m) = N(n, m - 1) + N(n - 1, m) + N(n - 1, m - 1)$$

for all  $n, m \geq 1$ .

# Number of Alignments

## Example

	$m = 1$	$m = 2$	$m = 3$	$m = 4$	$m = 5$	$m = 6$	$m = 7$
$n = 1$	3	5	7	9	11	13	15
$n = 2$	5	13	25	41	61	85	113
$n = 3$	7	25	63	129	231	377	575
$n = 4$	9	41	129	321	681	1289	2241
$n = 5$	11	61	231	681	1683	3653	7183

## Approximation

Idea:  $N(n, m) \sim \rho^{n+m}$  we hence get

$$\rho^2 - 2\rho - 1 = 0 \Rightarrow \rho = 1 + \sqrt{2}$$

this gives us:

$$N(20, 20) \simeq 10^{15} \quad N(100, 100) \simeq 10^{76} \quad N(1000, 1000) \simeq 10^{764}$$

# Dynamic Programming

## Needleman and Wunsch (1970)

We denote by  $B_{i,j}$  the best score of an alignment of  $X_1 \dots X_i$  and  $Y_1 \dots Y_j$  and we get

1:  $B_{0,0} = 0$

2:  $B_{i,0} = \sum_{k=1}^i \sigma(X_k, -)$  and  $B_{0,j} = \sum_{k=1}^j \sigma(-, Y_k)$

3: **for**  $i = 1 \dots n$  **do**

4:   **for**  $j = 1 \dots m$  **do**

5:

$$B_{i,j} = \max \begin{cases} B_{i-1,j-1} + \sigma(X_i, Y_j) \\ B_{i-1,j} + \sigma(X_i, -) \\ B_{i,j-1} + \sigma(-, Y_j) \end{cases}$$

6: return  $B_{n,m}$  and use a **traceback** to find the alignment

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# Example

Example ( $X = \text{gcgacgtgcaag}$   $Y = \text{aggcacgca} +3, -1, -2$ )

	-	a	g	g	c	a	c	g	c	a			
-	0	-2	-4	-6	-8	-10	-12	-14	-16	-18			
g	-2	-1	1	-1	-3	-5	-7	-9	-11	-13			
c	-4	-3	-1	0	2	0	-2	-4	-6	-8			
g	-6	-5	0	2	0	1	-1	1	-1	-3			
a	-8	-3	-2	0	1	3	1	-1	0	2			
c	-10	-5	-4	-2	3	1	6	4	2	0			
g	-12	-7	-2	-1	1	2	4	9	7	5			
t	-14	-9	-4	-3	-1	0	2	7	8	6			
g	-16	-11	-6	-1	-3	-2	0	5	6	7			
c	-18	-13	-8	-3	2	0	1	3	8	6			
a	-20	-15	-10	-5	0	5	3	1	6	11			
a	-22	-17	-12	-7	-2	3	4	2	4	9			
g	-24	-19	-14	-9	-4	1	2	7	5	7			
a	g	g	c	-	a	c	g	-	-	c	a	-	-
-	g	-	c	g	a	c	g	t	g	c	a	a	g

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# Alignment biases

- Problem in score-based alignment: (nearly) optimal alignments **not unique**.
- Reliable regions: common to all high scoring alignments
- Questionable regions: close to gaps and low complexity regions

## Example (protein alignment with blosum62)

SALL <b>ASGGTSSHR</b> WSRT	score = 31
SALL <b>MARKSHRVL</b> WSRT	
SALL <b>ASGGTSSHR</b> - - WSRT	score = 31
SALL <b>MA</b> - - <b>RKSHRVL</b> WSRT	
SALL <b>ASGGTSSHR</b> - - WSRT	score = 28
SALL - - <b>MARKSHRVL</b> WSRT	

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# Gibbs-Boltzman distribution

- Look at weighted space of **all** alignments
- Gibbs-Boltzmann distribution

$$P_T(\mathcal{A}) = \frac{1}{Z_T} \exp \left[ \frac{1}{T} S(\mathcal{A}) \right]$$

temperature:  $T$

partition function:  $Z_T = \sum_{\mathcal{A}} \exp \left[ \frac{1}{T} S(\mathcal{A}) \right]$

- $\lim_{T \rightarrow 0} : P_T(\mathcal{A}) = 1$  if  $\mathcal{A}$  is optimal and 0 otherwise
- $\lim_{T \rightarrow \infty} : P_T(\mathcal{A}) = \text{const} \quad \forall \mathcal{A}$

[S. Miyazawa, Protein Eng. (1995)]

[M. Kschischo and M. Lässig, Pacific Symposium on Biocomputing 5 (2000)]

# Probabilistic alignment

## Example (weighted set of alignments $T=1$ )

$\mathcal{A}$	$P(\mathcal{A})$	$\mathcal{A}$	$P(\mathcal{A})$
A--CGT   III S = 1	32.1 %	A-CG-T   ++   S = -4	0.2 %
ACGCGT		ACGCGT	
A-C-GT   + II S = -2	1.6 %	AC-G-T   +   S = -2	1.6 %
ACGCGT		ACGCGT	
AC--GT   II II S = 1	32.1 %	ACG--T   III   S = 1	32.1 %
ACGCGT		ACGCGT	
A---CGT   +   S = -5	0.8 %	ACG---T   +   S = -5	0.8 %
ACGCG-T		A-CGCGT	

$$P(\mathcal{A}) = e^{S/T} / Z_T$$

$$Z_T = e^1 + e^{-2} + e^1 + e^{-5} + e^{-4} + e^{-2} + e^1 + e^{-5}$$

# Probabilistic alignment

## Example (weighted set of alignments $T=10$ )

$\mathcal{A}$	$P(\mathcal{A})$	$\mathcal{A}$	$P(\mathcal{A})$
A--CGT   III S = 1	16.2 %	A-CG-T   ++   S = -4	9.8 %
ACGCGT A-C-GT   + II S = -2	12.0 %	AC-G-T   +   S = -2	12.0 %
ACGCGT AC--GT     II S = 1	16.2 %	ACG--T         S = 1	16.2 %
ACGCGT A---CGT   +   S = -5	8.9 %	ACG---T   +   S = -5	8.9 %
ACGCG-T		A-CGCGT	

$$P(\mathcal{A}) = e^{S/T} / Z_T$$

$$Z_T = e^{10} + e^{-20} + e^{10} + e^{-50} + e^{-40} + e^{-20} + e^{10} + e^{-50}$$

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# Posterior probabilities

- Alignment accuracy = posterior probabilities:

$$P \begin{bmatrix} \dots & X_j & \dots \\ \dots & Y_j & \dots \end{bmatrix} = \frac{1}{Z_T} \sum_{\mathcal{A}: (X_i, Y_j) \in \mathcal{A}} \exp[S(\mathcal{A})/T]$$

Example ( A pair,  $T=1$  )

$$P \begin{bmatrix} \dots & G_3 & \dots \\ \dots & G_5 & \dots \end{bmatrix} = (e^1 + e^{-2} + e^1)/Z_T \approx 65.9\%$$

A--CGT  
| | |  
ACGCGT

S = 1

A-C-GT  
| + | |  
ACGCGT

S = -2

AC--GT  
| | | |  
ACGCGT

S = 1

# Posterior probabilities

- Alignment accuracy = posterior probabilities:

$$P \begin{bmatrix} \dots & X_i & \dots \\ \dots & Y_j & \dots \end{bmatrix} = \frac{1}{Z_T} \sum_{\mathcal{A}: (X_i, Y_j) \in \mathcal{A}} \exp[S(\mathcal{A})/T]$$

Example ( A gap, T=1 )

$$P \begin{bmatrix} \dots & - & \dots \\ \dots & C_2 & \dots \end{bmatrix} = (e^1 + e^{-2} + e^{-5} + e^{-4})/Z_T \approx 34.0\%$$

A--CGT  
| IIII  
ACGCGT

S = 1

A-C-GT  
| + II  
ACGCGT

S = -2

A---CGT  
| + |  
ACGCG-T

S = -5

A-CG-T  
| ++ |  
ACGCGT

S = -4

# Outline

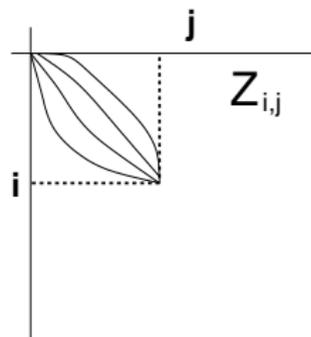
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# Partition function calculation

- $B_{i,j} \rightarrow Z_{i,j}$ ,  $\max \rightarrow \sum$  and  $+$   $\rightarrow \times$
- Recall:  $Z_T = \sum_{\mathcal{A}} e^{S(A)/T}$
- **Forward algorithm**  
 $Z_{i,j}$ : sum over all alignments of  $X_1 \dots X_i$  and  $Y_1 \dots Y_j$
- Here:  $T = 1$

$$Z_{i,j} = \underbrace{Z_{i-1,j-1} e^{\sigma(X_i, Y_j)}}_{\text{match/mismatch}} + \underbrace{Z_{i-1,j} e^{\sigma(X_i, -)} + Z_{i,j-1} e^{\sigma(-, Y_j)}}_{\text{gap}}$$

$$Z_T = Z_{n,m}$$

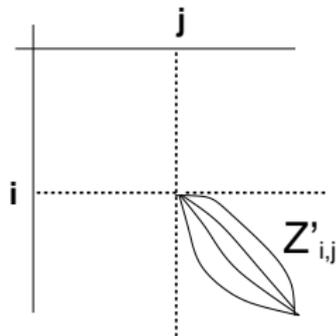


# Partition function calculation

- $B_{i,j} \rightarrow Z_{i,j}$ ,  $\max \rightarrow \sum$  and  $+$   $\rightarrow \times$
- Recall:  $Z_T = \sum_{\mathcal{A}} e^{S(A)/T}$
- **Backward algorithm**  
 $Z'_{i,j}$ : sum over all alignments of  $X_{i+1} \dots X_n$  and  $Y_{j+1} \dots Y_m$
- Here:  $T = 1$

$$Z'_{i,j} = \underbrace{Z'_{i+1,j+1} e^{\sigma(X_{i+1}, Y_{j+1})}}_{\text{match/mismatch}} + \underbrace{Z'_{i+1,j} e^{\sigma(X_{i+1}, -)} + Z'_{i,j+1} e^{\sigma(-, Y_{j+1})}}_{\text{gap}}$$

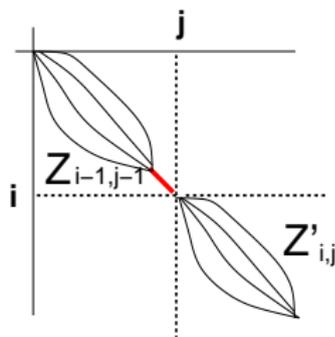
$$Z'_T = Z'_{0,0} \equiv Z_{n,m}$$



# Partition function calculation

- $B_{i,j} \rightarrow Z_{i,j}$ ,  $\max \rightarrow \sum$  and  $+$   $\rightarrow \times$
- Recall:  $Z_T = \sum_{\mathcal{A}} e^{S(\mathcal{A})/T}$
- **Combining forward and backward**
- Here:  $T = 1$

$$P \begin{bmatrix} \dots & X_i & \dots \\ \dots & Y_j & \dots \end{bmatrix} = \frac{Z_{i-1,j-1} e^{\sigma(X_i, Y_j)} Z'_{i,j}}{Z_T}$$



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# Posterior probabilities

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score = 31



score = 31



score = 28

# Our software

## Generalizations / other features

- Affine gap costs:  $g(l) = \alpha + \beta l$  instead of  $g(l) = \gamma l$
- Pair Hidden Markov models.  
Probabilistic description without temperature parameter.
- Other decoding algorithms than optimal alignments,  
**Sampling** from Gibbs-Boltzmann distribution  
 $P_T[\mathcal{A}] \propto e^{S(\mathcal{A})/T}$
- Different alignment models (e.g. **global** or **local**)

## The ppALIGN webserver

- Standalone programs and C++ library (open source)
- Webinterface

<http://www.math-info.univ-paris5.fr/ppblast/>

# Summary / Outlook

## Summary

- Score based alignment common tool
- Alignment biases: many high scoring alignments
- Probabilistic description of alignment space. **Posterior probabilities, alternative alignments.**
- ppALIGN: software to analyse alignments

## Outlook

- Parameter optimization (choice of temperature).
- Distribution of pattern in alignments
- Biological **applications** where correct alignment close to gaps is crucial