

# BIOS 477/877 Bioinformatics and Molecular Evolution

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## Today's topics

- Multiple Sequence Alignment
  - Progressive alignment (ClustalW)
  - T-Coffee

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## Multiple alignment as an extension of pairwise alignment

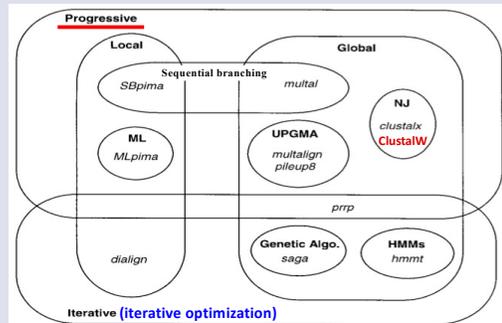
- **Dynamic programming algorithm**
  - Guarantees to find the optimal alignment
  - Optimal alignments are searched based on alignment score
    - How can we score multiple alignment?
      - Sum of pairs score:  $S(A) = \sum_{i,j} S(A_{ij})$
      - No statistical justification
  - Not very efficient for many sequences

We need good/efficient heuristic methods!

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## Multiple alignment methods (heuristic approach)



Thompson et al. (1999)

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## ClustalW

Thompson et al. (1994)

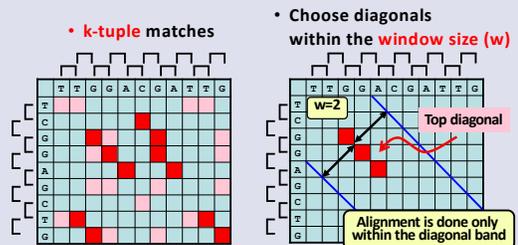
1. Pairwise alignment  
(fast approximation or full dynamic programming)
2. Generate a distance matrix  
(% identities converted to distances)
3. Construct a **guide tree**  
(neighbor-joining phylogenetic tree)
4. Progressive alignment following the guide tree  
(scoring matrix, sequence weight, gap penalties, etc.)

Updated to ClustalΩ and should not be used; to try, see Course Web 'Links' page    BIOS477/877 L15 - 5

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## ClustalW: 1. Pairwise alignment

- Fast approximation using a diagonal band



The default is to use full dynamic programming: better

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### ClustalW: 2 & 3. Guide tree generation

➤ A guide tree is generated from a distance matrix

**Pairwise alignment**

Seq1 xxxxx  
Seq2 yyyyy → S12

Seq1 xxxxx  
Seq3 zzzzz → S13

Seq2 yyyyy  
Seq3 zzzzz → S23

**% identities**

S12 → D12

S13 → D13

S23 → D23

**Distances (100 - % identities)**

-	D12	D13
D12	-	D23
D13	D23	-

Neighbor-joining phylogenetic reconstruction

**Guide tree**

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### ClustalW: steps

Thompson *et al.* (1994)

1. Pairwise alignment  
(fast approximation or full dynamic programming)
2. Generate a distance matrix  
(% identities converted to distances)
3. Construct a **guide tree**  
(neighbor-joining phylogenetic tree)
4. Progressive alignment following the guide tree  
(scoring matrix, sequence weight, gap penalties, etc.)

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### ClustalW: 4. progressive alignment

➤ Alignment is done **following the guide tree**

**Guide tree**

- Closest sequences are aligned first
- Pairwise alignment can be done by using simply the dynamic programming algorithm

Next closest sequence is aligned against the existing alignment → Profile alignment (can be alignment between alignments)

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### ClustalW: profile alignment

➤ Alignment against existing alignment

S1 peeksav

S2 geekaav

S3 egewglv

	P	E	R	S	A	V
G	E	K	A	A	V	
E						
W						
G						
L						
V						

A sequence (S3) is aligned against the alignment (S1, S2)

Existing alignment = profile  
→ Each position of alignment shows the configuration of possible amino acids

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### ClustalW: profile alignment

➤ How cell scores can be calculated

S1 peeksav

S2 geekaav

S3 egewglv

From a scoring matrix

S1	P	
S2	G	
S3	0	-g
E	-g	? -2g

(gap penalty = -g)

S1 vs. S3 = S(P,E)

S2 vs. S3 = S(G,E)

S1/S2 vs. S3 = {S(P,E)+S(G,E)}/2

Simple average from all pairwise scores

Existing alignment = profile  
→ Each position of alignment shows the configuration of possible amino acids

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### ClustalW: profile alignment

➤ Cell scores using simple vs. weighted average

S1 peeksav (W1)

S2 geekaav (W2)

S3 egewglv (W3)

[Simple average (without weighting)]  
Score(P/G,E) = {S(P,E) + S(G,E)} / 2

or

[Weighted average]  
Score(P/G,E) = {W1xW3xS(P,E) + W2xW3xS(G,E)} / 2

S1	P	
S2	G	
S3	0	-g
E	-g	?

Sequence weights from Thompson *et al.* (1994)

If W1=W2=W3=1, the same as simple average

Closely related sequences share information  
→ The sequences with duplicated information should receive smaller weights

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### ClustalW: profile alignment

➤ **Sequence weights:** based on branch lengths

Shorter branch lengths  
→ Fewer changes  
(e.g., S1 and S2 are more similar than S1 and S3)

Distance from the root

Rooted at the midpoint

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### ClustalW: profile alignment

➤ **Sequence weights:** based on branch lengths

$W1 = 0.02 + 0.15/2 + 0.09/3 = 0.125$

Distance from the root

Divided by the number of sequences sharing the branch

Closely related sequences share information  
→ shorter branch lengths

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### ClustalW: profile alignment

➤ **Sequence weights:** based on branch lengths

$W1 = 0.02 + 0.15/2 + 0.09/3 = 0.125$

$W2 = 0.08 + 0.15/2 + 0.09/3 = 0.185$

Distance from the root

Divided by the number of sequences sharing the branch

Closely related sequences share information  
→ shorter branch lengths

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### ClustalW: profile alignment

➤ **Sequence weights:** based on branch lengths

$W1 = 0.02 + 0.15/2 + 0.09/3 = 0.125$

$W2 = 0.08 + 0.15/2 + 0.09/3 = 0.185$

$W3 = 0.38 + 0.09/3 = 0.410$

$W4 = 0.46$

Distance from the root

Divided by the number of sequences sharing the branch

NOTE: Weights are normalized so that the max weight (0.46) becomes 1.0

Closely related sequences share information  
→ The sequences with duplicated information should receive smaller weights

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### ClustalW: profile alignment

➤ How sequence weighting works: example 1

Aligning the 3rd sequence (P) to the first 2 sequences (L, V) previously aligned

[Simple average]  
Score(L,V,P)  
=  $\{S(L,P) + S(V,P)\} / 2$   
=  $0.5 S(L,P) + 0.5 S(V,P)$

[Weighted average]  
Score(L,V,P)  
=  $\{1 \times 1 \times 2 \times S(L,P) + 5 \times 1 \times 2 \times S(V,P)\} / 2$   
=  $66 S(L,P) + 30 S(V,P)$

$S(L,P) \gg S(V,P)$

$S(L,P)$  is weighted more

Closely related sequences share information  
→ The sequences with duplicated information should receive smaller weights

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### ClustalW: profile alignment

➤ How sequence weighting works: example 2

How should 'VG' in S4 be aligned against the alignment of S1/S2/S3?  
(S1, S2, and S3 are previously aligned)

Which is the best alignment, (a) or (b) or (c)?

(a) 123 (b) 123 (c) 123  
S1 VAG VAG VAG  
S2 VGA VGA VGA  
S3 IVG IVG IVG  
S4 VG- V-G -VG

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### ClustalW: profile alignment

➤ How sequence weighting works: example 2

[Simple average] (match=2, mismatch=-1)

Alignment (a) [Score = 1 + gap penalty]  
 1: Score =  $\{S(V,V)+S(V,V)+S(L,V)\}/3 = (2+2-1)/3 = 1$   
 2: Score =  $\{S(A,G)+S(G,G)+S(V,G)\}/3 = (-1+2-1)/3 = 0$   
 3: Score = (gap penalty x 3)/3

Alignment (b) [Score = 2 + gap penalty]  
 1: Score =  $\{S(V,V)+S(V,V)+S(V,V)\}/3 = (2+2-1)/3 = 1$   
 2: Score = (gap penalty x 3)/3  
 3: Score =  $\{S(G,G)+S(A,G)+S(G,G)\}/3 = (2-1+2)/3 = 1$

Alignment (c) [Score = 1 + gap penalty]  
 1: Score = (gap penalty x 3)/3  
 2: Score =  $\{S(A,V)+S(G,V)+S(V,V)\}/3 = (-1-1+2)/3 = 0$   
 3: Score =  $\{S(G,G)+S(A,G)+S(G,G)\}/3 = (2-1+2)/3 = 1$

(a) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 VG-

(b) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 V-G

(c) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 -VG

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### ClustalW: profile alignment

➤ How sequence weighting works: example 2

[Weighted average] (match=2, mismatch=-1)

Alignment (a)  
 1: Score =  $\{1.1x7xS(V,V)+1.1x7xS(V,V)+6.1x7xS(L,V)\}/3 = \{7.7x2+7.7x2+42.7x(-1)\}/3 = -3.97$   
 2: Score =  $\{1.1x7xS(A,G)+1.1x7xS(G,G)+6.1x7xS(V,G)\}/3 = \{7.7x(-1)+7.7x2+42.7x(-1)\}/3 = -11.67$   
 3: Score = (gap penalty x 3)/3

Alignment Score = -15.64 + gap penalty

(a) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 VG-

(b) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 V-G

(c) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 -VG

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### ClustalW: profile alignment

➤ How sequence weighting works: example 2

[Weighted average] (match=2, mismatch=-1)

Alignment (b)  
 1: Score =  $\{1.1x7xS(V,V)+1.1x7xS(V,V)+6.1x7xS(L,V)\}/3 = \{7.7x2+7.7x2+42.7x(-1)\}/3 = -3.97$   
 2: Score = (gap penalty x 3)/3  
 3: Score =  $\{1.1x7xS(G,G)+1.1x7xS(A,G)+6.1x7xS(G,G)\}/3 = \{7.7x2+7.7x(-1)+42.7x(2)\}/3 = 31.03$

Alignment Score = 27.06 + gap penalty

(a) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 VG-

(b) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 V-G

(c) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 -VG

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### ClustalW: profile alignment

➤ How sequence weighting works: example 2

[Weighted average] (match=2, mismatch=-1)

Alignment (c)  
 1: Score = (gap penalty x 3)/3  
 2: Score =  $\{1.1x7xS(A,V)+1.1x7xS(G,V)+6.1x7xS(V,V)\}/3 = \{7.7x(-1)+7.7x(-1)+42.7x(2)\}/3 = 23.33$   
 3: Score =  $\{1.1x7xS(G,G)+1.1x7xS(A,G)+6.1x7xS(G,G)\}/3 = \{7.7x2+7.7x(-1)+42.7x(2)\}/3 = 31.03$

Alignment Score = 54.36 + gap penalty

(a) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 VG-

(b) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 V-G

(c) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 -VG

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### ClustalW: profile alignment

➤ How sequence weighting works: example 2

[Simple average]

(a) Alignment score = 1 + gap penalty  
 (b) Alignment score = 2 + gap penalty  
 (c) Alignment score = 1 + gap penalty

[Weighted average]

(a) Alignment score = -15.64 + gap penalty  
 (b) Alignment score = 27.06 + gap penalty  
 (c) Alignment score = 54.36 + gap penalty

(a) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 VG-

(b) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 V-G

(c) 123  
 S1 VAG  
 S2 VGA  
 S3 IVG  
 S4 -VG

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### ClustalW: parameter selection

➤ How **scoring matrix** is chosen

- Users choose only a scoring matrix series (BLOSUM, PAM, etc.)
- Specific matrix (BLOSUM80, etc.) is determined based on distance between sequences
  - 80 - 100% identity → BLOSUM80
  - 60 - 80% identity → BLOSUM62
  - 30 - 60% identity → BLOSUM45
  - 0 - 30% identity → BLOSUM30

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## ClustalW: parameter selection

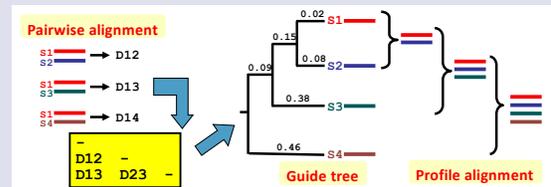
- How **gap penalties** are determined
  - Initial gap penalties: **GOP** (gap opening) and **GEP** (gap extension) are set by the user
  - **Position- or residue-specific** gap penalties
    - Weight (scoring) matrix dependent
    - Similarity level dependent
    - Sequence length dependent
    - Position-specific
      - if gaps already exist
      - Residue-specific (e.g., hydrophilic stretches)

See Thompson *et al.* (1994) for more details

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## ClustalW: summary & limitation



- Progressive alignment
  - **Greedy** (finds local optima, but no guarantee for global optima)
  - **Guide tree** is built only once at the beginning (cannot be fixed)
  - **Errors** in the early alignments (**incorrect gap positions**) cannot be rectified later
  - ClustalW does not rely on one set of alignment parameters
    - **Position- and various feature-specific scoring matrix and gap penalties**
- **Global alignment only** (local similarity may be missed)

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## How to solve progressive-alignment problems

- Incorporate more information to reduce **early errors**
  - **Structural alignment** (e.g., Espresso, PROMALS3D, TM-Coffee, PRALINE, MAFFT-DASH)
  - **Profile/profile-HMM alignment** (e.g., PRALINE, PSI-Coffee, PROMALS3D, ProbCons/CONTRAlign, ClustalΩ, MUSCLE5)
- Avoid the **greedy-algorithm** problem
  - **Iterative refinement** to search the global maxima
    - A good objective function is required (e.g., MUSCLE/MUSCLE5, MAFFT, ProbCons/CONTRAlign)
- **Global (or local) only** alignment problem
  - **Combine both methods** (e.g., T-Coffee)
- More accurate **insertion/deletion placement**
  - **Phylogeny aware gap-placement** (e.g., PRANK, ProPIP, Bali-Phy, SATé)

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## T-Coffee

Notredame, Higgins & Heringa (2000); Taly *et al.* (2011)

- T-Coffee: **Tree-based Consistency Objective Function for alignment Evaluation**
- Progressive alignment
  - Uses a guide tree, fast
- Tries to avoid the **greedy** nature of the progressive algorithm
  - Using **alignment libraries** derived from a mixture of alignment methods (global, local, etc.)

[T-Coffee on the Web]

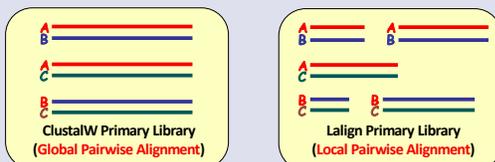
- [T-Coffee home page](#)
- [T-Coffee site from Notredame Lab](#)
- [T-Coffee @ EBI Tools](#) (only T-Coffee, without other associated programs)

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## T-Coffee: primary libraries

- Primary libraries of alignments



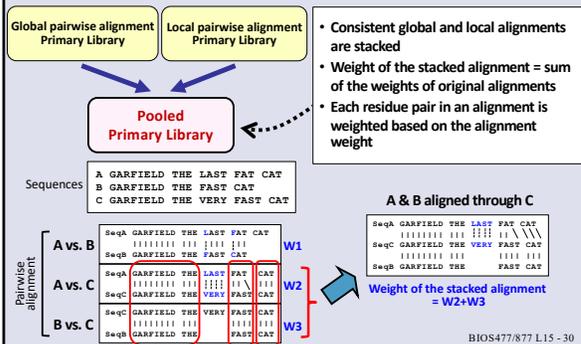
- Global and local, or any combination of pairwise alignment methods
- Each pairwise alignment is given a **weight** based on % identity ignoring gap sites

$\frac{ATTCGG}{ATAGCG} \Rightarrow 3/6 = 50\% \Rightarrow W=50$

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## T-Coffee: pooled primary library

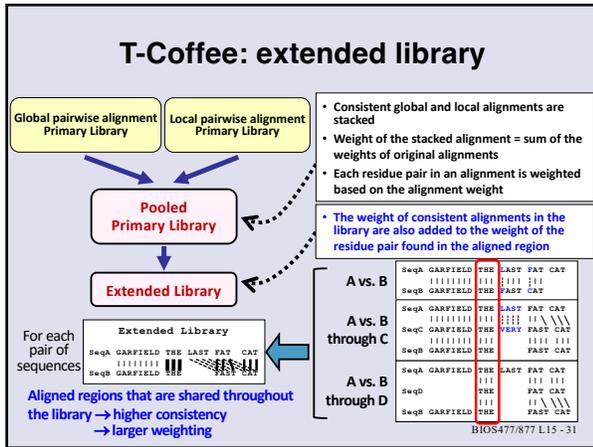


- Consistent global and local alignments are stacked
- Weight of the stacked alignment = sum of the weights of original alignments
- Each residue pair in an alignment is weighted based on the alignment weight

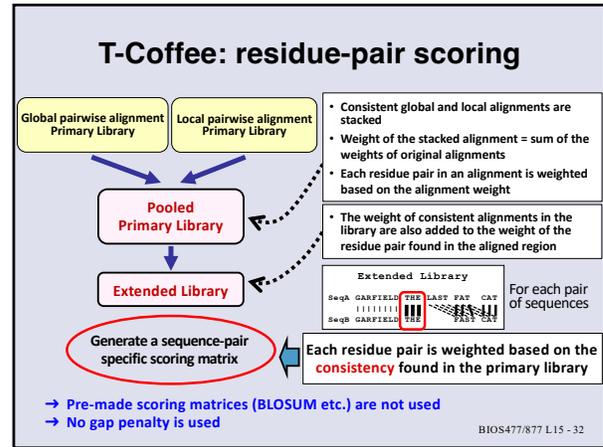
Weight of the stacked alignment =  $W2+W3$

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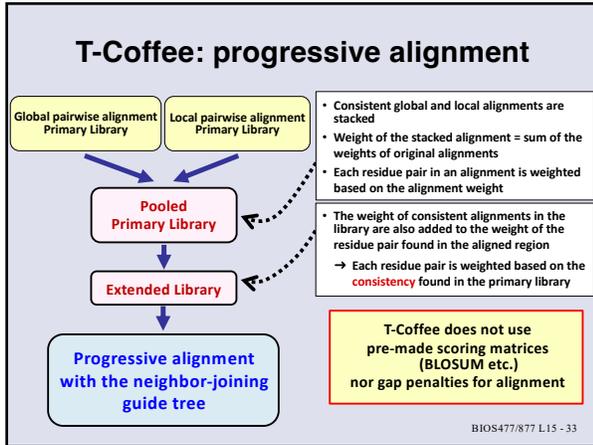
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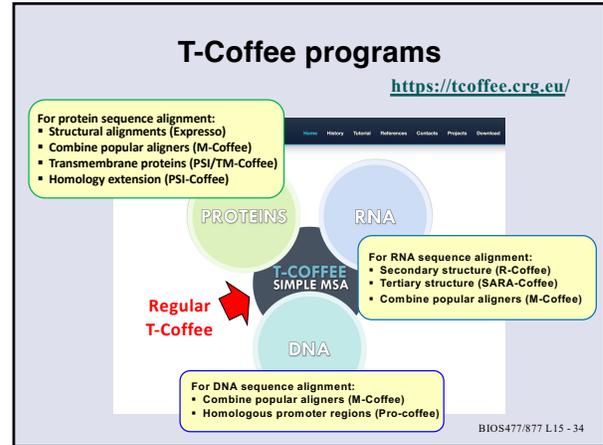
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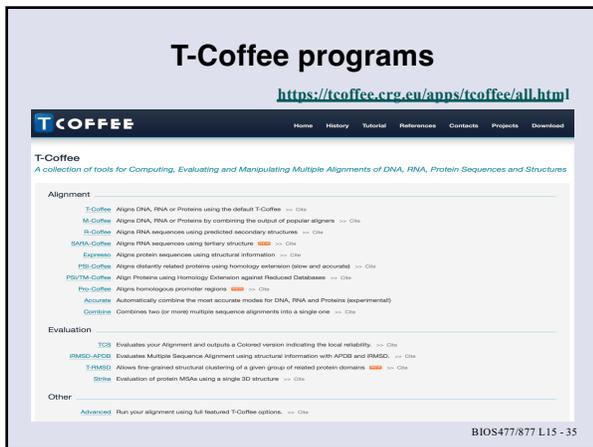
32



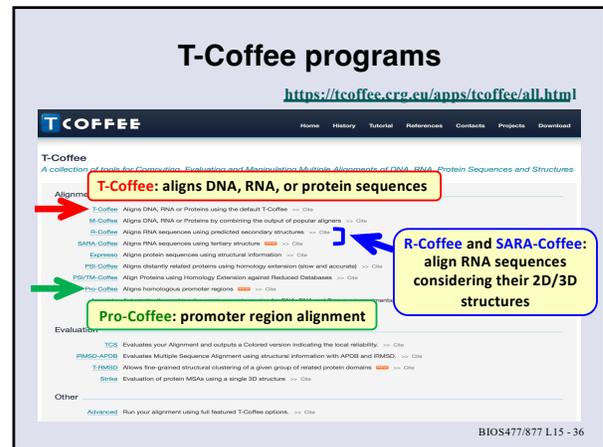
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## T-Coffee programs

<https://tcffee.org/en/apps/tcoffee/all.html>

**Expresso (3D-Coffee):** aligns protein sequences with structural information incorporated as pairwise structural alignments (library)

**PSI-Coffee:** extends each sequence with PSI-BLAST  
 → Structural information can be incorporated  
**PSI/TM-Coffee:** PSI-BLAST against a reduced database (for fast search)  
 → Only TM proteins can be targeted (optional)

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## T-Coffee programs to improve MSA

<https://tcffee.org/en/apps/tcoffee/all.html>

**M-Coffee:** combines multiple alignments  
 → shows consistency between multiple alignments

**Combine:** combines your own multiple alignments

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## T-Coffee programs to evaluate MSA

<https://tcffee.org/en/apps/tcoffee/all.html>

**TCS:** evaluates alignments based on the consistency between pairwise alignment library and multiple alignment

**Structural information based evaluation and clustering**

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## Evaluating multiple alignments using TCS

➤ **TCS (transitive consistency score)** Chang *et al.* (2014)

**T-Coffee**

```

SCORE=395
BAD AVG GOOD
laboA : 48
lyceB : 47
lphc : 45
lyse : 32
llhVA : 32
cons : 39
  
```

**ClustalW2**

```

SCORE=305
BAD AVG GOOD
laboA : 42
lyceB : 41
lphc : 39
lyse : 25
llhVA : 28
cons : 30
  
```

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## Evaluating multiple alignments using TCS

Reference alignment

a T-Coffee

b M-Coffee (kalign+mafft+muscle)

c PSI-Coffee

d Expresso

from Taly *et al.* (2011)

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