

## Multiple Sequence Alignment Sum-of-Pairs and Clustal-W

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## This Lecture

- Multiple Sequence Alignment
- The problem
- Theoretical approach: Sum-of-Pairs scores
- Practical approach: Clustal-W


## Multiple Sequence Alignment

- We now align multiple ( $k>2$ ) sequences
- Note: Also BLAST aligns only two sequences
- Why?
- Imagine $k$ sequences of the promoter region of genes, all regulated by the same transcription factor $f$. Which subsequence within the $k$ sequences is recognized by $f$ ?
- Imagine $k$ sequences of proteins that bind to DNA. Which subsequence of the $k$ sequences code for the part of the proteins that performs the binding?
- General
- We want to know the common part(s) in $k$ sequences
- "common" does not mean identical
- This part can be anywhere within the sequences


## Definition

- Definition
- A multiple sequence alignment (MSA) of $k$ Strings $s_{j} 1 \leq i \leq k$, is a table of $k$ rows and I columns (I $\geq \max \left(/ s_{i} /\right)$ ), such that
- Row i contains the sequence of $s_{j}$ with an arbitrary number of blanks being inserted at arbitrary positions
- Every symbol of every $s_{i}$ stands in exactly one column
- No column contains only blanks

$$
\begin{aligned}
& \text { AACGTGATTGAC } \\
& \text { TCGAGTGCTTTACAGT } \\
& \text { GCCGTGCTAGTCG } \\
& \text { TTCAGTGGACGTGGTA } \\
& \text { GGTGCAGACC }
\end{aligned}
$$



## Good MSA

- We are searching for good (optimal) MSAs
- Defining „optimal" here is not as simple as in the $\mathrm{k}=2$ case
- Intuition
- All sequences had a common ancestor and evolved by evolution
- We want to assume as few evolutionary events as possible
- Thus, we want few columns ( ~ few INSDELs)
- Thus, we want homogeneous columns ( $\sim$ few replacements)



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## What Should we Count?

- For two sequences
- We scored each column using a scoring matrix
- Find the alignment such that the total score is maximal
- But - how do we score a column with $5 * T, 3^{*} A, 1^{*}$ ?
- We would need an exponentially large scoring matrix
- Alternative: Sum-of-Pairs Score
- We score an entire MSA
- We score the alignment of each pair of sequences in the usual way
- We aggregate over all pairs to score the MSA
- We need a clever algorithm to find the MSA with the best score


## Formally

- Definition
- Let $M$ be a MSA for the set $S$ of $k$ sequences $S=\left\{s_{1}, \ldots, s_{k}\right\}$
- The alignment of $s_{i}$ with $s_{j}$ induced by $M$ is generated as follows
- Remove from M all rows except i and j
- Remove all columns that contain only blanks
- The sum-of-pairs score (sop) of $M$ is the sum of all pair-wise induced alignment scores
- The optimal MSA for $S$ wrt. to sop is the MSA with the lowest sopscore over all possible MSA for S


## Example



- Given a MSA over k sequences of length I - how complex is it to compute its sop-score?
- How do we find the best MSA?


## Analogy

|  |  | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | w | r | i | t | e | r | s |
| 0 |  | 0. | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| 1 | v | 1 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| 2 | i | 2 | 2 | 2 | 2 | 3 | 4 | 5 | 7 |
| 3 | n | 3 | 3 | 3 | 3. | 3 | 4 | 5 | 6 |
| 4 | t | 4 | 4 | 4 | 4 | 3 | 4 | 5 | 6 |
| 5 | n | 5 | 5 | 5 | 5 | 4. | 4 | 5 | 6 |
| 6 | e | 6 | 6 | 6 | 6 | 5 | 4 | 5 | 6 |
| 7 | $r$ | 7 | 7 | 6 | 7 | 6 | 5 | 4 | 5 |

- Think of the k=2 case
- Every alignment is a path through the matrix
- The three possible directions (down, right, down-right) conform to the three possible constellations in a column (XX, X_, _X)
- With growing paths, we align growing prefixes of both sequences


## Analogy

- Assume k=3
- Think of a 3-dimensional cube with the three sequences giving the values in each dimension
- Now, we have paths aligning growing prefixes of three sequences
- Every column has seven possible constellations (XXX, XX_, X_X, _XX, X__, _X_, __X)



## Dynamic Programming in three Dimensions

- We compute the best possible alignment $d(i, j, k)$ for every triple of prefixes (lengths $\mathrm{i}, \mathrm{j}, \mathrm{k}$ ) using the following formula

$$
\left.\begin{array}{l}
d(i, j, k)=\min \left\{\begin{array}{lll}
d(i-1, j-1, k-1) & +c_{i j}+c_{i k}+c_{j k} & \text { Three (mis)matches } \\
d(i-1, j-1, k) & +c_{i j}+2 & \text { One (mis)match, two ins } \\
d(i-1, j, k-1) & +c_{i k}+2 \\
d(i, j-1, k-1) & +c_{j k}+2 \\
d(i-1, j, k) & +2 \\
d(i, j-1, k) & +2 \\
d(i, j, k-1) & +2
\end{array}\right. \\
\text { Let } c_{i j}=0, \text { if } s_{1}(i)=s_{2}(j), \text { else } 1 \\
\text { Let } c_{i k}=0, \text { if } s_{1}(i)=s_{3}(k), \text { else } 1
\end{array}\right]
$$

## Concrete Examples

$$
d(i, j-1, k)
$$

- Best sop-score for $\mathrm{d}(\mathrm{i}, \mathrm{j}-1, \mathrm{k})$ is known
- We want to compute d(i,j,k)
- This requires to align one symbol with two blanks (blank/blank does not count)
- $d(i, j, k)=d(i, j-1, k)+2$

$$
d(i-1, j, k-1)
$$



- Best sop-score for d(i-1, j,k-1) is known
- We want to compute d(i,j,k)
- This requires aligning a blank with $\mathrm{s}_{1}[\mathrm{i}-1]$ and with $\mathrm{s}_{3}[\mathrm{k}-1]$ and to align $\mathrm{s}_{1}[\mathrm{i}-1]$ and $\mathrm{s}_{3}[\mathrm{k}-1]$
- $d(i, j, k)=d(i-1, j, k-1)+2+c_{i k}$


## Initialization

- Of course, we have $d(0,0,0)=0$
- Aligning in one dimension: $\mathrm{d}(\mathrm{i}, 0,0)=2 * i$
- Same for $d(0, j, 0), d(0,0, k)$
- Aligning in two dimensions: $\mathrm{d}(\mathrm{i}, \mathrm{j}, \mathrm{O})=\ldots$
- Let $d_{a, b}(i, j)$ be the alignment score for $S_{a}[1 . . i]$ with $S_{b}[1 . . j]$
$-\mathrm{d}(\mathrm{i}, \mathrm{j}, 0)=\mathrm{d}_{1,2}(\mathrm{i}, \mathrm{j})+(\mathrm{i}+\mathrm{j})$
$-d(i, 0, k)=d_{1,3}(i, k)+(i+k)$
$-d(0, j, k)=d_{2,3}(j, k)+(j+k)$


## Algorithm

```
initialize matrix d;
for i := 1 to |S |
    for j := 1 to |S |
        for k := 1 to |S |
            if (S (i) = S (j)) then coij := 0; else cocij := 1;
            if (S (i) = S (k)) then cocik := 0; else conik := 1;
```



```
            d
            d
            d
            d
            d5 := d[i - 1,j,k) + 2;
            d
            d7 := d[i,j,k - 1) + 2;
            d[i,j,k] := min(d
    end for;
    end for;
end for;
```


## Bad News: Complexity

- For 3 sequences of length n
- There are $n^{3}$ cells in the cube
- For each cell (top-left-front corner), we need to look at 7 corners
- Together: O(7*n³) operations
- For $k$ sequences of length $n$
- There are $\mathrm{n}^{\mathrm{k}}$ cell corners in the cube
- For each corner, we need to look at $2^{k}-1$ other corners
- Together: O(2* * $\left.\mathrm{n}^{\mathrm{k}}\right)$ operations


## Bad News: Biological Meaningfulness

- Let's take one step back
- What happened during evolution?



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## Different Scoring Function

- If we knew the phylogenetic tree of the $k$ sequences
- Align every parent with all its children
- Aggregate all alignment scores
- This gives the "real" number of evolutionary operations
- But: Finding the true phylogenetic tree requires a MSA
- Not covered in this lecture
- Use a heuristic: ClustalW
- Thompson, J. D., Higgins, D. G. and Gibson, T. J. (1994). "CLUSTAL W: improving the sensitivity of progressive multiple sequence alignment through sequence weighting, position-specific gap penalties and weight matrix choice." Nucleic Acids Res 22(22): 4673-80.


## Clustal-W

- Main idea
- Compute a "good enough" phylogeny - the guide tree
- Use the guide tree to iteratively align small MSA to larger MSA
- Starting from single sequences
- To escape the curse of dimensionality, compute MSA iteratively
- "Progressive" MSA
- Does not necessarily find the best solution
- Needs a fast method to align two MSAs
- Works quite well in practice
- Many other, newer (better) proposals
- DAlign, T-Coffee, HMMT, PRRT, MULTALIGN, ...


## Step 1: Compute the Guide Tree

- Compute all pair-wise alignments and store in similarity matrix M
- $M[i, j]=\operatorname{sim}\left(s_{i}, s_{j}\right)$
- Compute the guide tree using hierarchical clustering
- Choose the smallest M[i,j]
- Let $s_{i}$ and $s_{j}$ form a new (next) branch of the tree
- Compute the distance from the ancestor of $s_{i}$ and $s_{j}$ to all other sequences as the average of the distances to $\mathrm{s}_{\mathrm{i}}$ and $\mathrm{s}_{\mathrm{j}}$
- Set M' = M
- Delete rows and columns i and j
- Add a new column and row (ij)
- For all $\mathrm{k} \neq \mathrm{ij}: \mathrm{M}^{〔}[\mathrm{ij}, \mathrm{k}]=(\mathrm{M}[\mathrm{i}, \mathrm{k}]+\mathrm{M}[\mathrm{j}, \mathrm{k}]) / 2$
- Iterate until M' has only one column / row


## Sketch



## Example

|  | A | B | C | D | E |
| :--- | :--- | :--- | :--- | :--- | :--- |
| A |  | 17 | 59 | 59 | 77 |
| B |  |  | 37 | 61 | 53 |
| C |  |  |  | 13 | 41 |
| D |  |  |  |  | 21 |



|  | A | B | E | CD |
| :--- | :--- | :--- | :--- | :--- |
| A |  | 17 | 77 | 59 |
| B |  |  | 53 | 49 |
| E |  |  |  | 31 |



|  | $E$ | $C D$ | $A B$ |
| :--- | :--- | ---: | ---: |
| $\mathbf{E}$ |  | 31 | 65 |
| $\mathbf{C D}$ |  |  | 54 |



## Example



C PADKTNVKAAWGKVGAHAGEYGA
D AADKTNVKAAWSKVGGHAGEYGA

A PEEKSAVTALWGKVNVDEYGG
B GEEKAAVLALWDKVNEEEYGG

C PADKTNVKAAWG_KVGAHAGEYGA
D AADKTNVKAAWS_KVGGHAGEYGA
E AA__TNVKTAWSSKVGGHAPA A

A PEEKSAV_TALWG_KVN__VDEYGG
B GEEKAAV_LALWD_KVN__EEEYGG
C PADKTNVKAA_WG_KVGAHAGEYGA
D AADKTNVKAA_WS_KVGGHAGEYGA
E AA__TNVKTA_WSSKVGGHAPA__A

## Once a gap, always a gap

## Step 2: Progressive MSA

- Pair-wise alignment of MSAs in the order of the guide tree
- Aligning a MSA $M_{1}$ with a MSA M2
- Use the usual (global) alignment algorithm
- To score a column, compute the average score over all pairs of symbols in these columns
- Example

| A | ...P... |  |
| :---: | :---: | :---: |
| B | ...G... | Score of this column |
| C | ...P... | ( $2 * S(P, A)+s(P, Y)+$ |
|  |  | 2*s(G,A)+s(G,Y)+ |
| D | ...A... | 2*s(P,A)+s(P,Y) ) / 9 |
| E | ...A... |  |
| F | ...Y... |  |

## Issues

- There is a lot to say about whether hierarchical clustering actually computes the "correct" tree
- Clustal-W actually uses a different, more accurate phylogenetic algorithm called "neighbor-joining"
- Clustal-W is fast ( $\mathrm{O}\left(\mathrm{k}^{2 *} \mathrm{n}^{2}+\mathrm{k}^{2 *} \log (\mathrm{k})\right.$ )
- For k sequences; plus cost for computing pairwise alignments
- Idea behind progressive alignment
- Find strong signals (highly conserved blocks) first
- Outliers are added last
- Increases the chances that conserved blocks survive
- Several improvements to this scheme are known


## Problems with progressive MSA



## Further Reading

- Merkl \& Waack, chapter 13
- Böckenhauer \& Bongartz, chapter 5.3

