# Improving Multiple Sequence Alignments with Constraint Programming and Local Search

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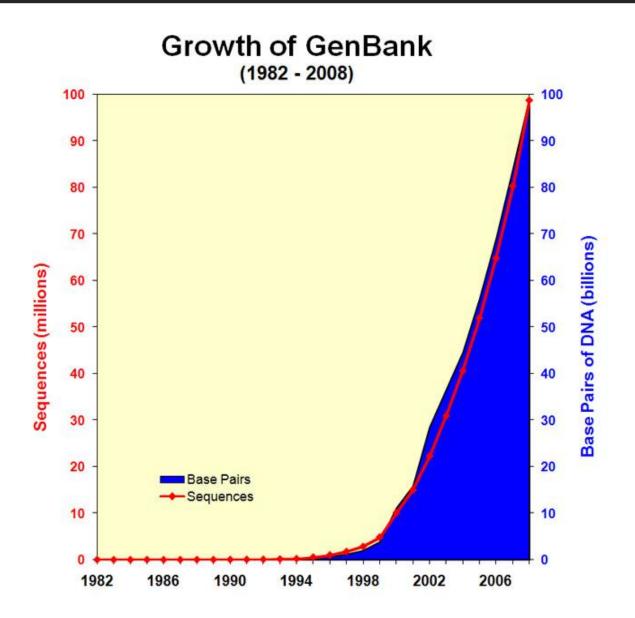






# **OUTLINE**

- 1. Introduction
  - i. Our proposal
- 2. Method
  - i. CP Model
  - ii. Search
  - iii. Local Search
  - iv. Experiments
- 3. Results and Discussion
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	QFGLFSPEEIRASSVALIRTPYPETLENGVPKESGLVCAGHFGHIELVK	S1
Protein sequences	QFGLFSPEEIKRMSVVHVEYPETMDEQRQRPRTKGLECPGHFGHIELAT	s2
	ELGVLDPEIIKKISVCEIVPNVDIYKDGRFPREGGLYCPGHFGHIELAK	s3
Pairwise alignment	QFGLFSPEEIRASSVALIRTPYPETLENGVPKESGLVCAGHFGHIELVK QFGLFSPEEIKRMSVVHVEYPETMDEQRQRPRTKGLECPGHFGHIELAT	s1 s2
Pairwise alignment	QFGLFSPEEIKRMSVVHVE-YPETMDEQRQRPRTKGLECPGHFGHIELAT ELGVLDPEIIKKISVCEIVPNVDIYKDGR-FPREGGLYCPGHFGHIELAK	S2 S3

\$1 QFGLFSPEEIRASSVALIRTPYPETLENGVPKESGLVCAGHFGHIELVK
\$2 QFGLFSPEEIKRMSVVHVEYPETMDEQRQRPRTKGLECPGHFGHIELAT
\$3 ELGVLDPEIIKKISVCEIVPNVDIYKDGRFPREGGLYCPGHFGHIELAK

Protein sequences



OFGLFSPEEIRASSVALIR--YPETLENG--VPKESGLVCAGHFGHIELVK S1 **S2** OFGLFSPEEIKRMSVVHVE--YPETMDEORORPRTKGLECPGHFGHIELAT S3ELGVLDPEIIKKISVCEIV--NVDIYKDG--FPREGGLYCPGHFGHIELAK **S4** OFGVLSPDELKRMSVTEGGIKYPETTE--GGRPKLGGLECPGHFGHIELAK **S5** OFGVLSPDELKRMSVTEGGIKYPETTE--GGRPKLGGLECPGHFGHIELAK **S6** OFGILSPDEIRRMSVTEGGVOFAETME--GGRPKLGGLECPGHFGHIDLAK **S7** OFGILGPEEIKRMSVAH--VEFPEVYE--NGKPKLGGLDCPGHFGHLELAK **S8** QFGILSPEEIRSMSVAK--IEFPETMDESGQRPRVGGLDCPGHFGHIELAK S9 QFGILSPDEIRQMSVIH----VEHSETTEKGKPKVGGLECPGHFGYLELAK S10 -----ECPGHFGHIELAK **S11** ----ECPGHFGFIELAK **S12** QFEIFKERQIKSYAVCLVEHAKSYANA----ADQSGEAECPGHFGYIELAE **S13** QFEVFKEAQIKAYAKCIIEHAKSYEHG----QPVRGGIECPGHFGYVELAE

Multiple
Sequence
Alignment
(MSA)



### **Motif and Domain Identification**

#### **Evolutionary**

**Structural** 

**Functional** 



**Phylogenetics** 



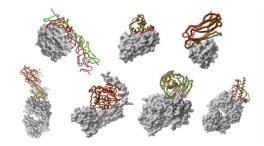


**Physiological Studies** 





**Protein Structure and Interaction** 



# column

QFGLFSPEEIRASSVALIR--YPETLENG--VPKESGLVCAGHF GHIELVK

OFGLFSPEEIKRMSVVHVE--YPETMDEQRORPRTKGLECPGHFGHIELAT

ELGVLDPEIIKKISVCEIV--NVDIYKDG--FPREGGLYCPGHFGHIELAK

QFGVLSPDELKRMSVTEGGIKYPETTE--GGRPKLGGLECPGHFGHIELAK

QFGVLSPDELKRMSVTEGGIKYPETTE--GGRPKLGGLECPGHFGHIELAK

QFGILSPDEIRRMSVTEGGVQFAETME--GGRPKLGGLECPGHFGHIDLAK

QFGILSPEEIKRMSVAH--VEFPEVYE--NGKPKLGGLDCPGHFGHLELAK

QFGILSPEEIRSMSVAK--IEFPETMDESGQRPRVGGLDCPGHFGHIELAK

QFGILSPDEIRQMSVIH---VEHSETTEKGKPKVGGLECPGHFGYLELAK

------ECPGHFGHIELAK

QFEIFKERQIKSYAVCLVEHAKSYANA---ADQSGEAECPGHFGYIELAE

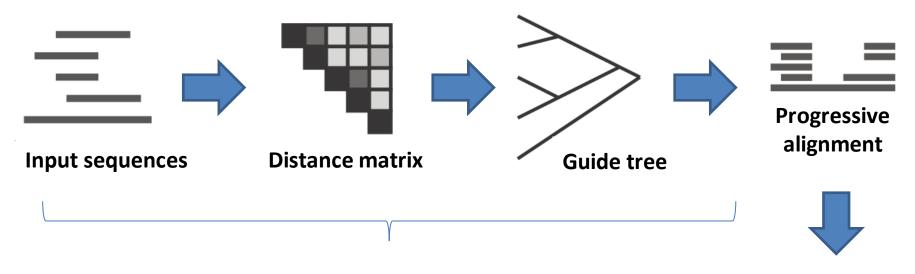
QFEVFKEAQIKAYAKCIIEHAKSYEHG----QPVRGGIECPGHFGYVELAE

MSA =

consecutive gaps

row

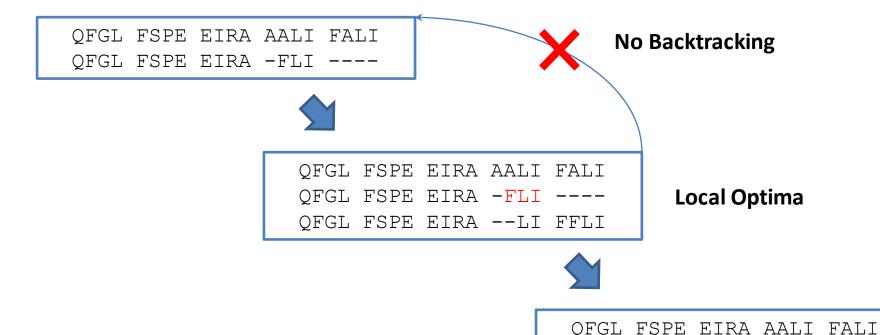
### MSA algorithms workflow:



- Reduced search space
- Limited scoring functions

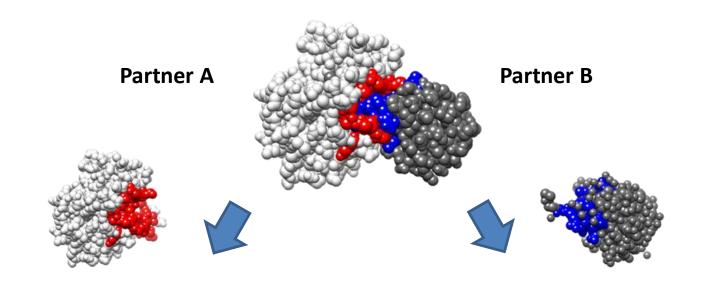


### Progressive vs Iterative Methods:



7

QFGL FSPE EIRA -FLI ----QFGL FSPE EIRA --LI FFLI --GL F--E EIRA -ALI FFL-



DPSLDRPFISEGTTLKDLIYDMTT
----EPRPDSGRDWSVELQEL-----KPGPDLGRDWSVELQEL-----EPESDSGRDWSAELPEL-----EPEPGSGGDCSEELPEL-PEPEQEPEPDSGGDCSAELPEL-RKOGLHSMNMMEAACSEPSLDL--

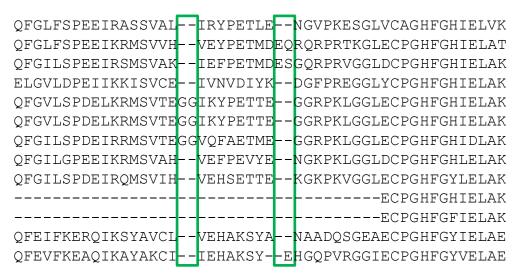
VEGMIKLALSTASGLAHLHMEI
WGSSLRMALSLAQGLAFLHEER
WGSSLRMALSLAQGLAFLHEER
WGSSLRMALSLAQGLAFLHEER
WGSSLSMALSLAEGLAFLHERR
--SSMSMALSLAQGLAFLHERR
--SSCRLAHSITRGLAYLHTRR

```
QFGLFSPEEIRASSVA IR--YPETLENG--V PKESGLVCAGHFGHIELVK
QFGLFSPEEIKRMSVV IVE--YPETMDEQRQR PRTKGLECPGHFGHIELAT
ELGVLDPEIIKKISVC IV--NVDIYKDG--F PREGGLYCPGHFGHIELAK
QFGVLSPDELKRMSVT EGGIKYPETTE--GGR PKLGGLECPGHFGHIELAK
QFGVLSPDEIRRMSVT EGGVQFAETME--GGR PKLGGLECPGHFGHIELAK
QFGILSPDEIRRMSVT EGGVQFAETME--GGR PKLGGLECPGHFGHIDLAK
QFGILSPEEIKRMSVA I--VEFPEVYE--NGK PKLGGLDCPGHFGHLELAK
QFGILSPEEIRSMSVA K--IEFPETMDESGQR PRVGGLDCPGHFGHIELAK
QFGILSPDEIRQMSVI I---VEHSETTEKGK PKVGGLECPGHFGYLELAK
QFGILSPDEIRQMSVI I---VEHSETTEKGK PKVGGLECPGHFGYLELAK
QFGILSPDEIRQMSVI I---VEHSETTEKGK PKVGGLECPGHFGYLELAK
QFEIFKERQIKSYAVC LVEHAKSYANA---A DQSGEAECPGHFGYIELAE
QFEVFKEAQIKAYAKC LIEHAKSYEHG----Q PVRGGIECPGHFGYVELAE
```

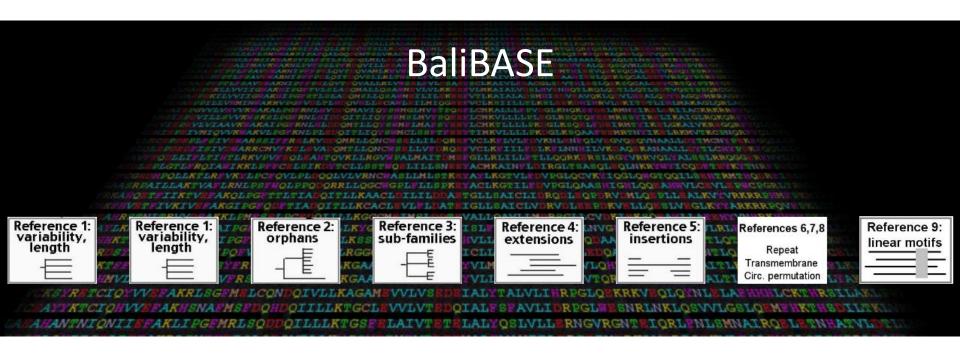
# Manual fixing of misalignments:



#### **Evolutionary sense: Only two insertions**



**Evaluation and assessment of MSA:** 



BAliBASE provides high quality alignments, manually refined and based on 3D structural superpositions

# **OUR APPROACH**

MSA computed with established programs



QFGLFSPEEIRASSVA LIR--YPETLENG--V PKESGLVCAGHFGHIELVK
QFGLFSPEEIKRMSVV HVE--YPETMDEQRQR PRTKGLECPGHFGHIELAT
ELGVLDPEIIKKISVC EIV--NVDIYKDG--F PREGGLYCPGHFGHIELAK
QFGVLSPDELKRMSVT EGGIKYPETTE--GGR PKLGGLECPGHFGHIELAK
QFGVLSPDEIRRMSVT EGGIKYPETTE--GGR PKLGGLECPGHFGHIELAK
QFGILSPDEIRRMSVT EGGVQFAETME--GGR PKLGGLECPGHFGHIELAK
QFGILSPEEIKRMSVA H--VEFPEVYE--NGK PKLGGLDCPGHFGHIELAK
QFGILSPEEIRSMSVA K--IEFPETMDESGQR PRVGGLDCPGHFGHIELAK
QFGILSPDEIRQMSVI H---VEHSETTEKGK PKVGGLECPGHFGYLELAK
QFGILSPDEIRQMSVI H---VEHSETTEKGK PKVGGLECPGHFGYLELAK
QFGILSPDEIRQMSVI H---VEHSETTEKGK PKVGGLECPGHFGYLELAK
QFEIFKERQIKSYAVC LVEHAKSYANA---A DQSGEAECPGHFGYIELAE
QFEVFKEAQIKAYAKC LIEHAKSYEHG----O PVRGGIECPGHFGYVELAE

#### Less conserved region



Repair the MSA using a CP and Local Search approach

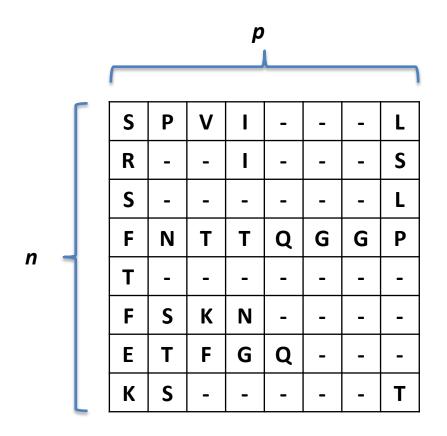


**Constraints and Scoring functions** 

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# **METHOD**



 $a_{i,j}$  -residue at sequence  $\emph{i}$  and position  $\emph{j}$ 

$$\mathbf{s}_i = \langle a_{i,1}, \dots, a_{i,p} 
angle$$
 - sequence  $_i$ 

$$\sigma_{A}\left(a_{1,j},a_{2,j}
ight)$$
 - scoring function

 $g_i$  - number of gaps in  $\mathbf{S}_i$ 

$$\gamma\left(\mathbf{s}_{1},\mathbf{s}_{2}
ight)$$
 - gap penalty

# **METHOD**

Score for the alignment of two sequences,  $S_1$  and  $S_2$ :

$$\sigma_S\left(\mathbf{s}_1, \mathbf{s}_2\right) = \sum_{i=1}^p \sigma_A\left(a_{1,i}, a_{2,i}\right) - \gamma\left(\mathbf{s}_1, \mathbf{s}_2\right)$$

**Substitution Matrix** 

The Score for the multiple alignment is derived from pairwise scores:

$$\sigma = \sum_{i=1}^{n} \sum_{j=i+1}^{n} \sigma_{S}\left(\mathbf{s}_{i}, \mathbf{s}_{j}\right)$$

### **METHOD**

#### **GONNET substitution matrix:**

```
\mathbf{E}
                                               H
                                                    R
                                                        K M I
11.5 0.1 -0.5 -3.1 0.5 -2.0 -1.8 -3.2 -3.0 -2.4 -1.3 -2.2 -2.8 -0.9 -1.1 -1.5 0.0 -0.8 -0.5 -1.0 C
     2.2 1.5 0.4 1.1 0.4 0.9 0.5 0.2 0.2 -0.2 -0.2 0.1 -1.4 -1.8 -2.1 -1.0 -2.8 -1.9 -3.3 S
         2.5 0.1 0.6 -1.1 0.5 0.0 -0.1 0.0 -0.3 -0.2 0.1 -0.6 -0.6 -1.3 0.0 -2.2 -1.9 -3.5 T
              7.6 0.3 -1.6 -0.9 -0.7 -0.5 -0.2 -1.1 -0.9 -0.6 -2.4 -2.6 -2.3 -1.8 -3.8 -3.1 -5.0 P
                   2.4 0.5 -0.3 -0.3 0.0 -0.2 -0.8 -0.6 -0.4 -0.7 -0.8 -1.2 0.1 -2.3 -2.2 -3.6 A
                        6.6 0.4 0.1 -0.8 -1.0 -1.4 -1.0 -1.1 -3.5 -4.5 -4.4 -3.3 -5.2 -4.0 -4.0 G
                            3.8 2.2 0.9 0.7 1.2 0.3 0.8 -2.2 -2.8 -3.0 -2.2 -3.1 -1.4 -3.6 N
                                 4.7 2.7 0.9 0.4 -0.3 0.5 -3.0 -3.8 -4.0 -2.9 -4.5 -2.8 -5.2 D
                                      3.6 1.7 0.4 0.4 1.2 -2.0 -2.7 -2.8 -1.9 -3.9 -2.7 -4.3 E
                                          2.7 1.2 1.5 1.5 -1.0 -1.9 -1.6 -1.5 -2.6 -1.7 -2.7 0
                                               6.0 0.6 0.6 -1.3 -2.2 -1.9 -2.0 -0.1 2.2 -0.8 H
                                                    4.7 2.7 -1.7 -2.4 -2.2 -2.0 -3.2 -1.8 -1.6 R
                                                        3.2 -1.4 -2.1 -2.1 -1.7 -3.3 -2.1 -3.5 K
                                                             4.3 2.5 2.8 1.6 1.6 -0.2 -1.0 M
                                                                 4.0 2.8 3.1 1.0 -0.7 -1.8 I
                                                                      4.0 1.8 2.0 0.0 -0.7 L
                                                                           3.4 0.1 -1.1 -2.6 V
                                                                               7.0 5.1 3.6 F
                                                                                    7.8 4.1 Y
```

# CP MODEL

#### Variables:

 $x_{i,j} \in X$  - for each cell in the matrix with domain

$$D(x_{i,j}) = \{'-', 'A', 'C', 'D', \dots, 'Y', 'W'\}$$

 $x_{i,k}^G \in X_i^G$  - models the position of the k'th gap in Si;  $\ 1 \cdot k \cdot g_i$ 

 $c_i$  - specify the number of consecutive gaps

#### **Constraint:**

VALIDSEQUENCE(
$$\mathbf{s}_i$$
)=INTABLE ( $\langle x_{i,1}, \dots, x_{i,p}, c_i \rangle, T_i$ )

each sequence is obtained by changing the positions of the gaps

each table is created so that each row is obtained by placing the gaps in distinct positions

$$T_i = C_{g_i}^p$$

# **SEARCH**

#### Use of a greedy variable and value heuristics:

$$var(X) = \arg\max_{x_{i,j}} \max_{v_1, v_2 \in D(x_{i,j})} q(x_{i,j}, v_1) - q(x_{i,j}, v_2)$$

$$val(x_{i,j}) = \arg\max_{v \in D(x_{i,j})} q(x_{i,j}, v)$$

$$q\left(x_{i,j},v\right)=q^{A}\left(x_{i,j},v\right)+nq^{G}\left(x_{i,j},v\right)\quad\text{- function which estimates the cost of assigning value $\nu$ to variable }\;x_{i,j}$$

# **SEARCH**

$$q\left(x_{i,j},v\right)=q^{A}\left(x_{i,j},v\right)+nq^{G}\left(x_{i,j},v\right) \quad \text{- function which estimates the cost of assigning value $v$ to variable $x_{i,j}$}$$

#### (A) Cost based on the set of residues already assigned:

$$q^{A}\left(x_{i,j},v\right) = \sum_{k=1}^{n} \begin{cases} \sigma_{A}\left(x_{k,j},v\right) & \in |D\left(x_{k,j}\right)| = 1\\ 0 & \in \text{ otherwise} \end{cases}$$

#### (B) Cost based on the number of consecutive gaps:

$$nq^G\left(x_{i,j},v\right) = \begin{cases} -10 & \Leftarrow \text{ if it creates a new gap} \\ 10 & \Leftarrow \text{ if it does not creates a new gap} \\ 0 & \Leftarrow \text{ not known} \end{cases}$$

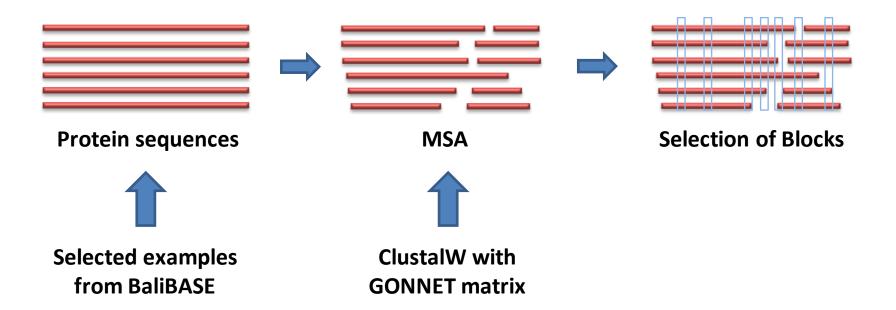
# LOCAL SEARCH

The same model was tested using a greedy hill-climbing heuristic optimization:

Local Maximum

- Constrained local search (COMET)
- The same objective function
- Randomization of start point

Parameter 2



#### **Example of a Block:**

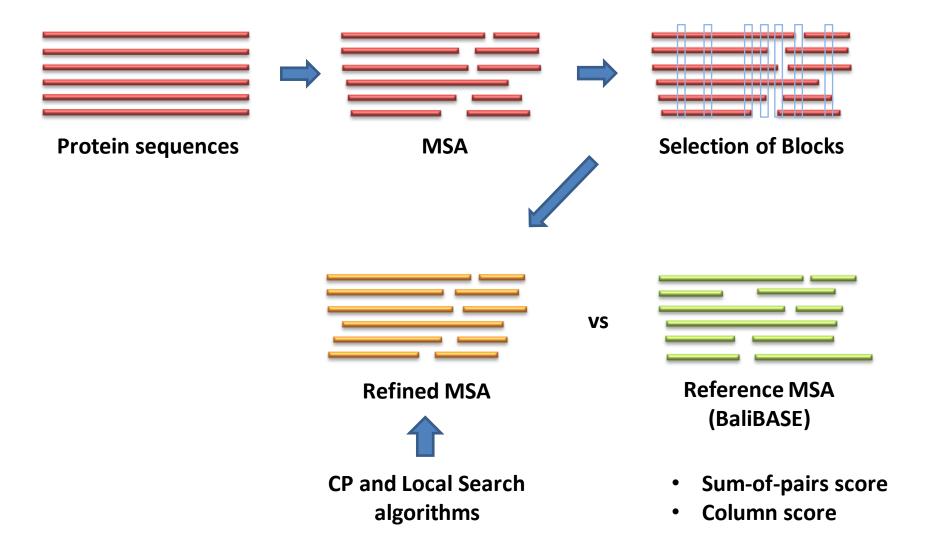
```
QNLDFAVALPAINVAAFSKN-----TTRKLELAVQNMSQFEKGAYTGEISAQ
DNLDFAIAPSFTSLALISTS-----KID-KLKVAAQNLSQFDSGAFTGEISAK
AETEALVCVPATLLSHAAE----IL-RT--PVHAGGEDCHTKESGAYTGCISAE
ARVDALICPPATLLYVATA----LC-DS--PLMIGAQDCHQKQSGAHTGEVSAE
RLFEALICVPATLLSRAFD----IL-GE--NILLGGQNCHFDDYGPYTGDISAF
PQIITGIIPPFTLLSACQQAV----SDS--PIFLGAQTTHEADSGAFTGEISAP
PKIITGIIPPFTLLSSCQQII----KNT--PIRLGAQTLHEVDSGAFTGEISAP
LSCTIGIASPFTSLRAIHEMI----NTTG-FLWLGAQNVHPELSGAFTGEISLP
KAVLGIAPVHVHLTEVNKVLP-----N-NLLLLAQDANFIASGSYTGTVSYT
VNADYSVGVPSIYLNQAKEI-----LKG-IKVIAQDAHFKNEGAYTGNISWS
RVLIGLAAPTVYLLQLHN---AMQIVLN-NRILTCAQDVSRFPDNGAYTGEVSAE
EKNTVIIAPPTIYLERVCKNIS-----NMNIFLGSQNVDINLNGAFTGETSIL
KNNIIIIAPPTVFLERVYKDIN------INIHLAAQNIDVNLTGAFTGENSAL
```

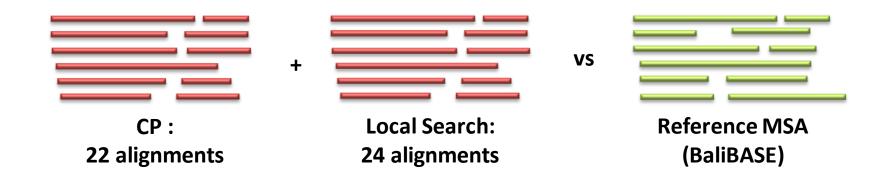
#### **CP and Local Search algorithms:**

SKNTTRKL
STSKID-KL
AEIL-RTPV
TALC-DSPL
FDIL-GENI
QQAVSDSPI
QQIIKNTPI
HEMINTTG-FL
KATbN-NT
KEILKGI
NAMQIVLN-NRIL
CKNISNMNI
AKDININI



```
SKN----TTRKL
STS----KID-KI
AE-T---L-RT--PV
TAI----ICDS--PI
FDTT----T-GE--NT
QQAV----SDS--PI
OOII----KNT--PI
HEMI----NTTGFL
KVT.P----N-NT.
KE-T----TKG--T
NA--MOIVL-N-NRIL
CKNI----SNMNI
YKDT----NTNT
```





#### Sum-of pairs score (SP):

The percentage of correctly aligned pairs of residues in the test alignment, relative to the reference alignment.

#### Column score (CS):

The percentage of correctly aligned columns, which tests the ability of the programs to align all of the sequences correctly at any given position.

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# RESULTS AND DISCUSSION

#### CP:

- Average improvement per column above the average score attributed to a match
- Average number of columns was 54 (in 22 MSA)
- 77% of MSAs (17 out of 22) were improved
- Average improvement of 11% in the SP score
- Several minutes per alignment

#### **Local Search:**

- Insignificant average improvement
- 58% of MSAs (14 out of 24) were improved
- Few seconds per alignment

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# CONCLUSIONS

# A framework for improving MSA obtained with classic algorithms

- Allows the use of different scoring functions
- Too much room for optimizations, both on the heuristics and on the scoring functions
- Possible inclusion of structural information

# **ACKNOWLEDGEMENTS**



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Marco Correia, PhD

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