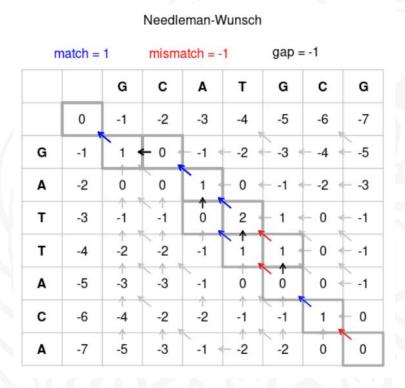
GLOBAL SEQUENCE ALIGNMENT VIA A PARALLEL-PREFIX BASED NEEDLEMAN WUNSCH ALGORITHM Max Farrington





#### What is global sequence alignment?

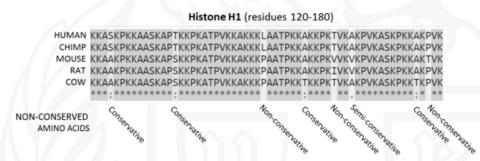
- Global sequence alignment is a bioinformatics technique for aligning two [or more] protein sequences with respects to the whole sequence.
- Every alignment is evaluated by maintaining a scoring matrix.
- Positive and negative scores are granted based on matches or mismatches.
  - Based on the use case, you can change the scoring scheme (ex: +1 match, -1 insertion/deletion, -1 mismatch)
- The best alignment is then found by backtracing from the bottom right



https://en.wikipedia.org/wiki/Needleman%E2%80%93Wunsch\_algorithm

#### What is it actually used for?

- When comparing the sequences of two subjects that share a common ancestor, you can view the mismatches, insertions, and deletions as mutations from that ancestor.
- You can then derive the importance of specific subsequences by how they are preserved in descendants of that ancestor.
- Millions of subsequences have also been tagged/identified for specific behavior.
  - You can find similarities between untagged/tagged sequences to find known genes in a sequence.



https://en.wikipedia.org/wiki/Sequence\_alignment

## Why is it a good parallel programming problem?

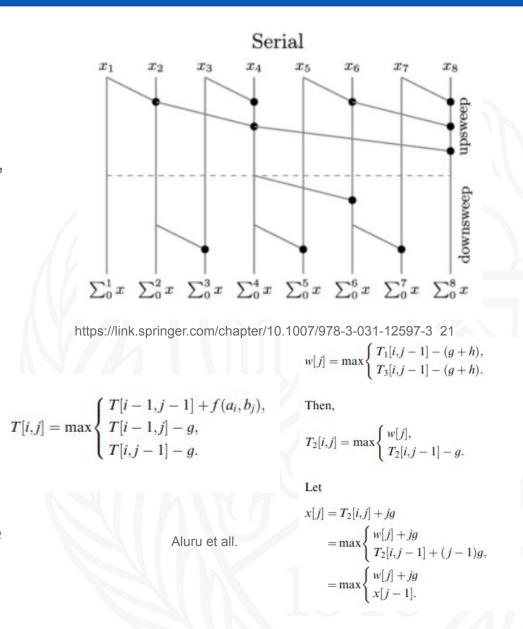
- To compute the running score, there are minimal data dependencies, allowing for computation to be done in parallel either row/column-wise, or along the anti-diagonal.
- These methods have tradeoffs in terms of efficiency and space complexity.
  - For anti-diagonal solutions, you only need to store the current and previous anti-diagonal, which changes in size as you fill the matrix.
  - For row/column wise solutions you need to store the current and previous row, but the size stays fixed.

	0	С	Α	G	С	С	U	С	G	С	U	U	Α	G
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Α	0	0	5	0	0	0	0	0	0	0	?			
A	0	0	5	2	0	0	0	0	0	?				
U	0	0	0	2	0	0	5	0	?					
G	0	0	0	5	0	0	0	?						
С	0	5	0	0	10,	51	?							
С	0	5	2	0	5	**								
A	0	0	10	1	?	3								
U	0	0	1	?										
U	0	0	?											
G	0	?												
С	0													
С	0													
G	0													
G	0													

https://www.researchgate.net/figure/Anti-diagonal-method-an d-dependency-of-the-cells\_fig11\_222408669

# How the algorithm relates to parallel prefix

- Parallel prefix takes a binary associative operator (such as +, -,
  - \*, MAX(), etc.) and an array of n elements, and for each element, computes and stores a running total based on the chosen operator.
- In the case of the Needleman Wunsch algorithm, we are effectively keeping a running total, but the value in each spot depends on the max value of its neighbors that have already been computed.
- Needleman Wunsch also allows for negative values in the matrix, so the work can very easily be split into chunks with the preceding values communicated via parallel prefix.



#### Parallel example

Gap penalty: -2, mismatch: -1, match 1 A = AACTGGAA B = CATG

$$w[j] = \max \begin{cases} T_1[i,j-1] - (g+h), \\ T_3[i,j-1] - (g+h). \end{cases} \quad T_2[i,j] = x[j] - jg.$$

$$\begin{split} x[j] &= T_2[i,j] + jg \\ &= \max \begin{cases} w[j] + jg \\ T_2[i,j-1] + (j-1)g, \\ &= \max \begin{cases} w[j] + jg \\ x[j-1]. \end{cases} \end{split}$$

w[1] = Max(T[0,0] + Match(B[0],A[0]),T[0,1] - 2)= -1

 $x[j] = max(-1 + 1(2), -\infty) = 1$ 

T[1,1] = 1 - 1(2) = -1

w[2] = Max(T[0,1] + Match(B[0],A[1]),T[0,2] - 2) = -3

x[j] = max(-3 + 2(2), 1) = 1

T[1,2] = 1 - 2(2) = -3

w[5] = Max(T[0,4] + Match(B[0],A[4]),T[0,5] - 2)= -9

 $x[j] = max(-9 + 5(2), -\infty) = 1$ 

T[1,5] = 1 - 5(2) = -9

#### Parallel example

Gap penalty: -2, mismatch: -1, match 1 A = AACTGGAA B = CATG

 $w[j] = \max \begin{cases} T_1[i, j-1] - (g+h), \\ T_3[i, j-1] - (g+h). \end{cases} \quad T_2[i, j] = x[j] - jg.$  $x[j] = T_2[i,j] + jg$  $= \max \begin{cases} w[j] + jg \\ T_2[i, j-1] + (j-1)g, \end{cases}$  $\begin{cases} w[j] + jg \\ x[j-1]. \end{cases}$ = max

	0	1	2	3	4	5	6	7	8
		А	А	С	Т	G	G	А	A
	0	-2	-4	-6	-8	-10	-12	-14	-16
С	-2	-1 (1)	-3 (1)	-3* (3)	-5* (3)	-9* (1)	-11* (1)	-15* (1)	-15* (1)
А	-4								
Т	-6								
G	-8								

2

Α

-4

-3 (1)

3

-3\* (3)

С

-6

#### Parallel example

Α

-2

1

-1 (1)

0

0

-2

-4

-6

-8

С

А

Т

G

Gap penalty: -2, mismatch: -1, match 1 A = AACTGGAA B = CATG

6

-11\* (1)

G

-12

5

-9\* (1)

G

-10

4

-5\* (3)

Т

-8

7

-15\* (1)

Α

-14

8

Α

-16

-15\* (1)

 $w[j] = \max \begin{cases} T_1[i, j-1] - (g+h), \\ T_3[i, j-1] - (g+h). \end{cases} \quad T_2[i, j] = x[j] - jg.$  $x[j] = T_2[i,j] + jg$  $= \max \begin{cases} w[j] + jg \\ T_2[i, j-1] + (j-1)g, \end{cases}$  $= \max \begin{cases} w[j] + jg \\ x[j-1]. \end{cases}$ 

Prefix time!

Binary associative operator -Max(x)

#### Parallel example

Gap penalty: -2, mismatch: -1, match 1 A = AACTGGAA B = CATG

	0	1	2	3	4	5	6	7	8
		А	А	С	Т	G	G	А	A
	0	-2	-4	-6	-8	-10	-12	-14	-16
С	-2	-1 (1)	-3 (1)	-3 (3)	-5 (3)	-7 (3)	-9 (3)	-11 (3)	-13 (3)
А	-4								
Т	-6								
G	-8								

 $w[j] = \max \begin{cases} T_1[i, j-1] - (g+h), \\ T_3[i, j-1] - (g+h), \end{cases} \quad T_2[i, j] = x[j] - jg.$  $x[j] = T_2[i,j] + jg$  $\begin{cases} w[j] + jg \\ T_2[i, j-1] + (j-1)g, \end{cases}$  $= \max \{$ w[j] + jg= max

Now, recompute x[j] using the value received during the prefix scan, and use for calculating T[i,j]

Additionally, share the last value in your local row with the processor next to you

		А	A	C	Т	G	G	A	Α
	0	-2	-4	-6	-8	-10	-12	-14	-16
С	-2	-1	<b>↓</b> -3	-3	<mark>↓ -</mark> 5	<b>+</b> -7	<mark>↓ -9</mark>	<b>+</b> -11	<b>↓-1</b> 3

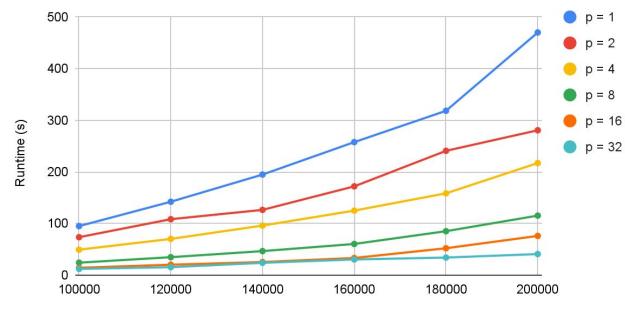
# Benchmarking and scalability



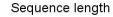


#### Runtime analysis

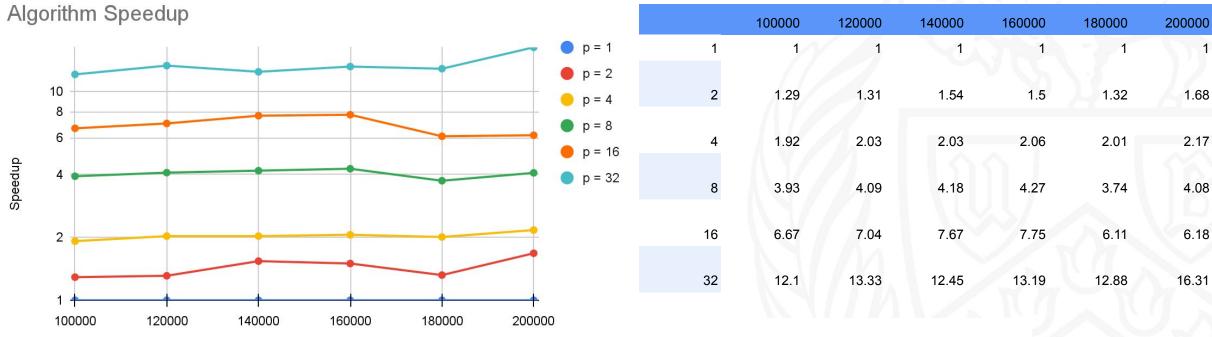
#### Algorithm Runtime Comparison



n/p		100000	120000	140000	160000	180000	200000
	1	94.972	142.176	194.823	257.642	318.329	470.006
	2	73.5367	108.38	126.419	171.986	240.79	280.597
	4	49.3603	70.1906	95.9919	125	158.477	217.044
	8	24.1509	34.7976	46.5727	60.3988	85.1384	115.308
	16	14.2428	20.195	25.385	33.2346	52.0935	75.9925
	32	7.85093	10.6655	15.6509	19.5389	24.7102	28.8214



## Speedup

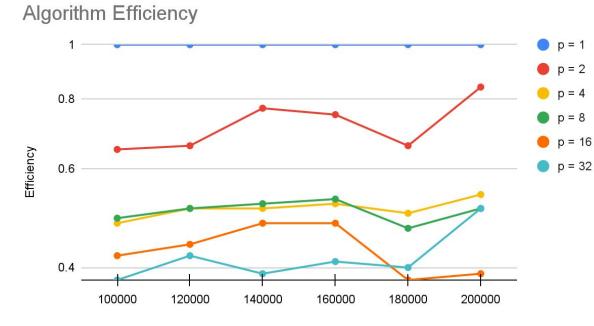


Sequence size

Speedup =  $\frac{T_1}{T_p}$ 

12

## Efficiency



	100000	120000	140000	160000	180000	200000
1	1	1	1	1	1	1
2	0.65	0.66	0.77	0.75	0.66	0.84
4	0.48	0.51	0.51	0.52	0.5	0.54
8	0.49	0.51	0.52	0.53	0.47	0.51
16	0.42	0.44	0.48	0.48	0.38	0.39
32	0.38	0.42	0.39	0.41	0.4	0.51

Sequence size

Efficiency =  $\frac{T_1}{pT_p}$ 

#### Conclusions

- Strong scaling is shown for p=2 (efficiency >= 65%) according to amdahl's law for n = 100000 through n = 200000.
- No weak scaling according to Gustafson's law with fixed ratio of problem size to processors
- Problem size during testing may not have been high enough, as performance showed some promise for higher problem size.
  - Ran into bottleneck with ram, but further testing development is required for implementing a more space efficient solution
- Better utilization of on-chip parallelism would have likely greatly increased the performance of the algorithm with MPI.



#### References

- https://en.wikipedia.org/wiki/Needleman%E2%80%93Wunsch\_al gorithm
- <u>https://en.wikipedia.org/wiki/Sequence\_alignment</u>
- <u>https://www.researchgate.net/figure/Anti-diagonal-method-and-d</u>
  <u>ependency-of-the-cells\_fig11\_222408669</u>
- <u>https://link.springer.com/chapter/10.1007/978-3-031-12597-3\_21</u>
- Srinivas Aluru, Natsuhiko Futamura, Kishan Mehrotra, Parallel biological sequence comparison using prefix computations,
- https://bioboot.github.io/bimm143\_W20/class-material/nw/