

## The Smith-Waterman algorithm

a multithreaded implementation

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- 1. The algorithm
- 2. The implementation
- 3. Performance analysis
- 4. Conclusions



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## The problem



Genome and protein alignment are fundamental bio-informatics techniques used to compare sequences of DNA, RNA, or proteins to identify regions of similarity.

These alignments reveal evolutionary relationships, functional similarities, or conserved regions across species or within different genes and proteins of the same organism.

## The challenge



Genomes can consist of billions of base pairs (the human genome has over 3 billion base pairs!).

Aligning such large datasets requires significant computational resources.



Figure: Human genome printout



Figure: Enter Caption

## How does the algorithm work?



The Smith-Waterman algorithm is a dynamic programming algorithm used specifically for local sequence alignment, that is to identify the most similar subsequences between two sequences.



## How does the algorithm work?



A scoring matrix H is created and filled as follows. The first row and column are initialized to zero.

$$H(i,j) = \max egin{cases} 0 \ H(i-1,j-1) + s(i,j) & (\mathsf{match/mismatch}) \ H(i-1,j) + g & (\mathsf{deletion}) \ H(i,j-1) + g & (\mathsf{insertion}) \end{cases}$$

- H(i,j) is the score at position (i,j),
- s(i,j) is the score for a match or mismatch,
- g is the penalty for a gap. It can be a function of the gap size.

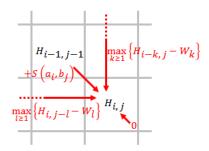
## Algorithm data dependencies



Highlight the dependence of H(i, j) from:

- the previous element along the diagonal H(i-1, j-1),
- the previous element along the column H(i, j-1),
- the previous element along the row H(i-1,j).

Furthermore, sequences can be insanely huge!





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## Implementation challenges



#### This is a great and fun challenge!

- How do we decompose the problem to handle the dependency chain?
- How do we handle synchronization, shared memory access and cache efficiency?
- How do we balance the load?

#### Our solutions



We have experimented 3 different approaches using OpenMP:

- 1. Using locks
- 2. Using atomic counters
- 3. Using tasks

## Locks



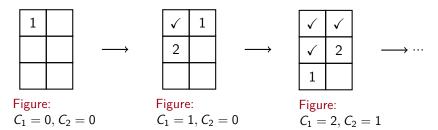
1			<b>√</b>	1		<b>√</b>	<b>√</b>	
		$\longrightarrow$	2		$\longrightarrow$	<b>√</b>	2	$\longrightarrow \cdots$
						1		*

Divide H in blocks and assign a lock to each. All threads set their locks.

When a thread wants to work on a block, it sets the lock corresponding to the previous block along the column. When it's done, it destroys such lock, and unsets its own.

#### Atomic counters





Each thread holds a counter of how many blocks it has processed. Thread i+1 waits until thread i has processed at least one more block.

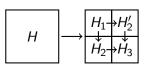
When threads jump the condition is a little different.

### **Tasks**



#### This is a recursive Divide & Conquer approach:

```
int solve(int i0, int i1, int i0, int i1,
      char*A, char*B, scores t*scores.int*
     H, int*Mj, int*Mi){
   int ans0.ans1.ans2.ans3:
   if(/*block is big enough*/){
       int im = (i0+i1)/2, jm = (j0+j1)/2;
      ans0 = solve(i0, im, j0, jm, A, B,
           scores, H. Mi. Mi):
#pragma omp task
      ans1 = solve(im, i1, j0, jm, A, B,
           scores, H. Mi. Mi):
#pragma omp task
      ans2 = solve(i0, im, jm, j1, A, B,
           scores, H. Mi. Mi):
#pragma omp taskwait
      ans3 = solve(im, i1, jm, j1, A, B,
           scores, H, Mj, Mi);
      return max(ans0,ans1,ans2,ans3);
   }else ; // solve normally...
```





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## Test setup



We generated test data starting from the genome of an organism found online. We then copied it and modified it. We then compared the original and the modified versions.

Hardware characteristics:

• CPU: Intel i7-1165G7 (8) @ 2.8GHz

RAM: 8 GB

• OS: Arch Linux (I use arch btw)

#### Performances



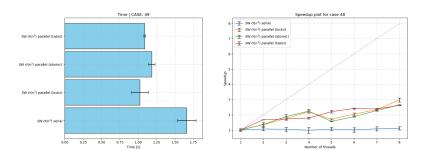


Figure: Running times and speedup plots for each of the algorithms.

## Insights



Algorithm	1 thread	2 threads	4 threads	8 threads
Serial				
Locks				
Tasks				

Table: Average running times for a sample input of size X and Y, for different numbers of maximum number of threads.

It is interesting to see approximately the same results for each of the approaches. The only notable difference is that the one using locks has a larger variation of the duration between runs.



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## Moving forward



A number of different approaches are possible:

- Using a similar approach as the ones used, MPI could be well suited to solve this problem: the communication between process is much smaller than the computation done by each of them.
- 2. **GPUs** could also be very effective, as the computation can proceed row by row and column by column.

# Thank you!

# Backup slide



Some additional content