# Linguagem Chapel

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#### **Timeline**

- Introdução
- Chapel
  - Instalação e configuração
  - Tipos de dados
  - Semelhanças com outras linguagens
  - Data Parallelism
  - Task Paralleism
  - Multi-locale Paralleism
  - The Model and the Problem
  - The Integrated Approach
- 3 Bad News: Hardness Results
  - Hardness of PP-Partitioning of Haplotype Matrices
  - Hardness of PP-Partitioning of Genotype Matrices
- Good News: Tractability Results
  - Perfect Path Phylogenies
  - Tractability of PPP-Partitioning of Genotype Matrices

Quantas linhas são necessários para a sua implementação?

#### **OpenMP**

```
1 #pragma omp shared( A , B , C , chunk ){
2     #pragma omp for schedule( static , chunk )
3     for( i = 0 ; i < N ; i++)
4     for( j = 0 ; j < N ; j++)
5     C[ i ][ j ] = A[ i ][ j ] + B[ i ][ j ] ;
6 }</pre>
```

#### **CUDA**

#### **OpenMPI**

```
1 MPI Comm size (MPI COMM WORLD , &npes ) ;
2 MPI Comm rank (MPI COMM WORLD , &myrank );
3 if ( mvrank == ROOT )
     for ( target = 0 ; target < npes ; i++)
5
         MPI Send( A+N*target , N , MPI INT , target , target ,
             MPI COMM WORLD ) ;
         MPI Send( B+N*target , N , MPI INT , target , target ,
6
             MPI COMM WORLD ) :
8 MPI Recv ( myA , N , MPI INT , ROOT , myrank , MPI COMM WORLD , &
       status ) :
9 MPI Recv( myB , N , MPI INT , ROOT , myrank , MPI COMM WORLD , &
       status ) :
10 \text{ for}(i = 0 ; i < N ; i++) \text{ myC}[i] = \text{myA}[i] + \text{myB}[i];
11 MPI Send( \mathsf{myC} , \mathsf{N} , \mathsf{MPI} \mathsf{INT} , \mathsf{ROOT} , \mathsf{0} , \mathsf{MPI} \mathsf{COMM} \mathsf{WORLD} ) ;
12 if ( myrank === ROOT )
for ( sender = 0 ; sender < npes ; sender++)
14
         MPI Recv( C+N*sender , N , MPI INT , sender , 0 ,
             MPI COMM WORLD , &status ) ;
```

#### Linhas

- OpenMP: 6
- CUDA: 9
- OpenMPI: 14

Então, quantas linhas são necessários com Chapel?

# Our formalization of haplotyping.

#### **Inputs**

- A genotype matrix G.
- The rows of the matrix are individuals / taxa.
- The columns of the matrix are SNP sites / characters.
- The problem is directed: one haplotype is known.
- The input is biallelic: there are only two homozygous states (0 and 1) and one heterozygous state (2).

#### **Outputs**

- A haplotype matrix H.
- Pairs of rows in H explain the rows of G.
- The haplotypes in H form a perfect phylogeny.

## We can do perfect phylogeny haplotyping efficiently, but . . .

- Data may be missing.
  - This makes the problem NP-complete . . .
  - ... even for very restricted cases.

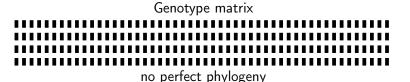
#### Solutions:

- Additional assumption like the rich data hypothesis.
- 2 No perfect phylogeny is possible.
  - This can be caused by chromosomal crossing-over effects.
  - This can be caused by incorrect data.
  - This can be caused by multiple mutations at the same sites.

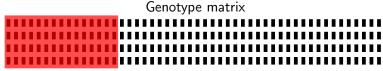
#### Solutions:

- Look for phylogenetic networks.
- Correct data.
- Find blocks where a perfect phylogeny is possible.

- Partition the site set into overlapping contiguous blocks.
- Compute a perfect phylogeny for each block and combine them.
- Use dynamic programming for finding the partition.

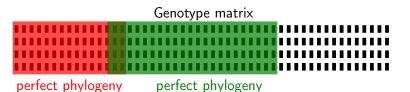


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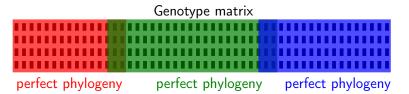
perfect phylogeny

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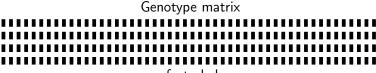
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- Partition the site set into overlapping contiguous blocks.
- Compute a perfect phylogeny for each block and combine them.
- Use dynamic programming for finding the partition.



## Objective of the integrated approach.

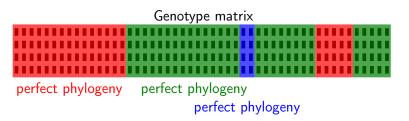
- Partition the site set into noncontiguous blocks.
- Compute a perfect phylogeny for each block and combine them.
- Ompute partition while computing perfect phylogenies.



no perfect phylogeny

## Objective of the integrated approach.

- Partition the site set into noncontiguous blocks.
- Compute a perfect phylogeny for each block and combine them.
- Compute partition while computing perfect phylogenies.



## The formal computational problem.

We are interested in the computational complexity of the function  $\chi_{PP}$ :

- It gets genotype matrices as input.
- It maps them to a number k.
- This number is minimal such that the sites can be covered by k sets, each admitting a perfect phylogeny.
   (We call this a pp-partition.)

## Finding pp-partitions of haplotype matrices.

#### We start with a special case:

- The inputs *M* are already haplotype matrices.
- The inputs *M* do not allow a perfect phylogeny.
- What is  $\chi_{PP}(M)$ ?

#### Example

0 1 0 0

M: 1 0 0 0 0 1 0 1 1 1 0 0 0 0 1 0 1 0 1 0 No perfect phylogeny is possible.

## Finding pp-partitions of haplotype matrices.

#### We start with a special case:

- The inputs M are already haplotype matrices.
- The inputs M do not allow a perfect phylogeny.
- What is  $\chi_{PP}(M)$ ?

#### Example

	0	0	0	1
	0	1	0	0
	1	0	0	0
	0	1	0	0
М:	1	0	0	0
	0	1	0	1
	1	1	0	0
	0	0	1	0
	1	0	1	0

Perfect phylogeny Perfect phylogeny  $\chi_{PP}(M) = 2$ .

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## Bad news about pp-partitions of haplotype matrices.

#### **Theorem**

Finding optimal pp-partition of haplotype matrices is equivalent to finding optimal graph colorings.

#### Proof sketch for first direction.

- $\bigcirc$  Let G be a graph.
- ② Build a matrix with a column for each vertex of G.
- **3** For each edge of G add four rows inducing the submatrix  $\begin{pmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}$ .
- The submatrix enforces that the columns lie in different perfect phylogenies.

## Implications for pp-partitions of haplotype matrices.

## **Corollary**

If  $\chi_{PP}(M) = 2$  for a haplotype matrix M, we can find an optimal pp-partition in polynomial time.

## **Corollary**

Computing  $\chi_{PP}$  for haplotype matrices is

- NP-hard.
- not fixed-parameter tractable, unless P = NP,
- very hard to approximate.

## Finding pp-partitions of genotype matrices.

### Now comes the general case:

- The inputs M are genotype matrices.
- The inputs M do not allow a perfect phylogeny.
- What is  $\chi_{PP}(M)$ ?

### Example

No perfect phylogeny is possible.

## Finding pp-partitions of genotype matrices.

#### Now comes the general case:

- The inputs M are genotype matrices.
- The inputs M do not allow a perfect phylogeny.
- What is  $\chi_{PP}(M)$ ?

#### Example

Perfect phylogeny Perfect phylogeny  $\chi_{PP}(M) = 2$ .

## Bad news about pp-partitions of haplotype matrices.

#### **Theorem**

Finding optimal pp-partition of genotype matrices is at least as hard as finding optimal colorings of 3-uniform hypergraphs.

#### Proof sketch.

- $\bullet$  Let G be a 3-uniform hypergraph.
- 2 Build a matrix with a column for each vertex of G.
- **3** For each hyperedge of G add four rows inducing the submatrix  $\begin{pmatrix} 2 & 2 & 2 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$ .
- The submatrix enforces that the three columns do not all lie in the same perfect phylogeny.

## Implications for pp-partitions of genotype matrices.

## **Corollary**

Even if we know  $\chi_{PP}(M) = 2$  for a genotype matrix M, finding a pp-partition of any fixed size is still

- NP-hard.
- not fixed-parameter tractable, unless P = NP,
- very hard to approximate.

#### Automatic optimal pp-partitioning is hopeless, but...

- The hardness results are worst-case results for highly artificial inputs.
- Real biological data might have special properties that make the problem tractable.
- One such property is that perfect phylogenies are often perfect path phylogenies:
  - In HapMap data, in 70% of the blocks where a perfect phylogeny is possible a perfect path phylogeny is also possible.

## Example of a perfect path phylogeny.

#### **Genotype matrix**

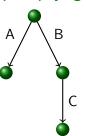
	А	Ь	C
	2	2	2
G:	0	2	0
	2	0	0
	0	2	2

## Haplotype matrix

	Α	В	C
	1	0	0
	0	1	1
	0	0	0
:	0	1	0
	0	0	0
	1	0	0
	0	0	0
	0	1	1

Н

#### Perfect path phylogeny



### The modified formal computational problem.

We are interested in the computational complexity of the function  $\chi_{PPP}$ :

- It gets genotype matrices as input.
- It maps them to a number k.
- This number is minimal such that the sites can be covered by k sets, each admitting a perfect path phylogeny.
   (We call this a ppp-partition.)

## Good news about ppp-partitions of genotype matrices.

#### **Theorem**

Optimal ppp-partitions of genotype matrices can be computed in polynomial time.

## **Algorithm**

- Build the following partial order:
  - Can one column be above the other in a phylogeny?
  - Can the columns be the two children of the root of a perfect path phylogeny?
- 2 Cover the partial order with as few compatible chain pairs as possible. For this, a maximal matching in a special graph needs to be computed.

▶ The algorithm in action

#### **Summary**

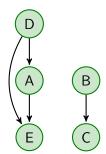
- Finding optimal pp-partitions is intractable.
- It is even intractable to find a pp-partition when just two noncontiguous blocks are known to suffice.
- For perfect path phylogenies, optimal partitions can be computed in polynomial time.

Computation of the partial order.

#### Genotype matrix

	, ,		_		_
	2	2	2	2	2
G:	0	1	2	1	0
	1	0	0	1	2
	0	2	2	0	0

#### Partial order



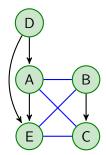
Partial order: →

Computation of the partial order.

### Genotype matrix

	٠,		_		_
	2	2	2	2	2
G:	0	1	2	1	0
	1	0	0	1	2
	0	2	2	0	0

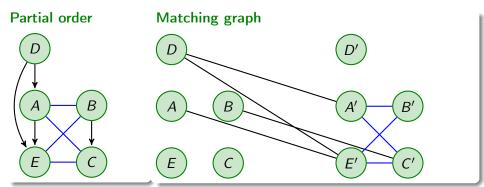
### Partial order



Partial order: →

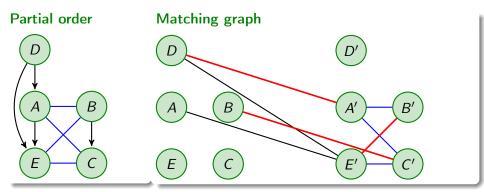
Compatible as children of root: —

The matching in the special graph.





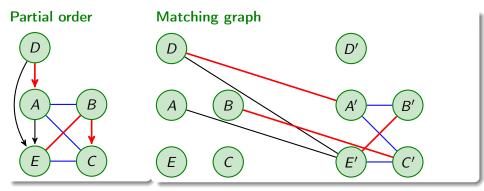
The matching in the special graph.



A maximal matching in the matching graph



The matching in the special graph.



A maximal matching in the matching graph induces perfect path phylogenies.



## Simple slide with three points shown all at once

- Point 1
- Point 2
- Point 3

## Simple slide with three points shown in succession

• Point 1 (Click "Next Page" to see Point 2)

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## Slide with two columns: items and a graphic

First item

Insert graphic here

## Slide with two columns: items and a graphic

- First item
- Second item

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