## Haplotype Assembly

Lesson of Bioinformatics Milano-Bicocca, 6/5/2015

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### Overview

- Problem: introduction and motivation
- Formulations of the problem
- Approaches: FPT and approximation
- Proposal of Thesis

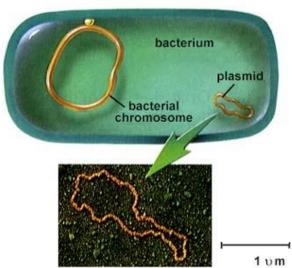
### Genomes

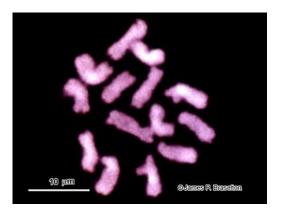
**Atlantic Salmon** 

• Genome = collection of chromosomes



Human

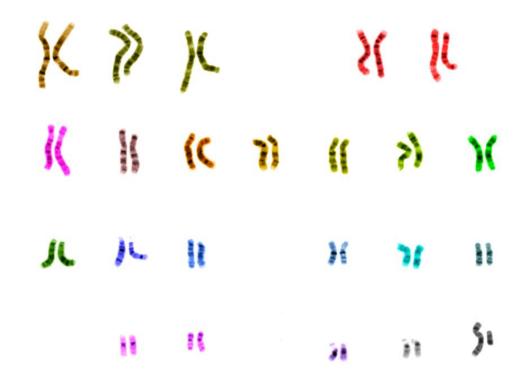




Pea plant

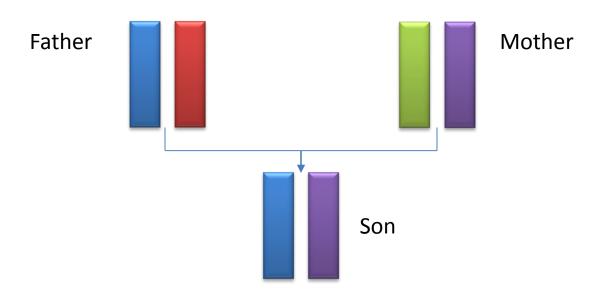
## **Diploid Organisms**

Diploid = two sets of homologous chromosomes



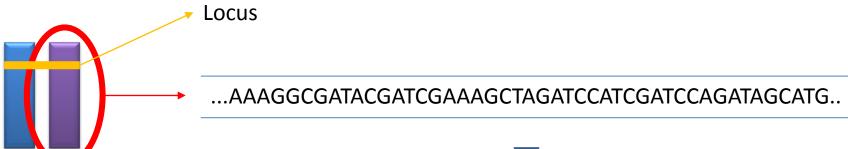
### Mendelian Laws

- For each chromosome there are 2 copies:
  - 1 inherited from the mother
  - 1 inherited from the father
- For now, ignore recombinations



## Haplotype

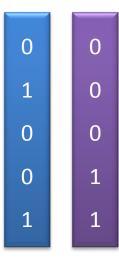
- Haplotype = One copy of a chromosome
- Locus = genome position
- Haplotype is represented as a binary vector (0 major/1 minor allele)





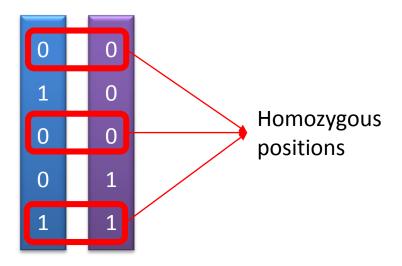
## Single Nucleotide Polymorphisms (SNPs)

- Each pair of homologous haplotypes exhibits differences in terms of Single Nucleotide Polymorphisms (SNPs)
- SNPs = Heterozygous positions



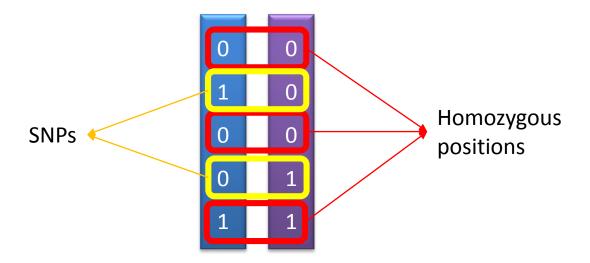
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### **Motivations**

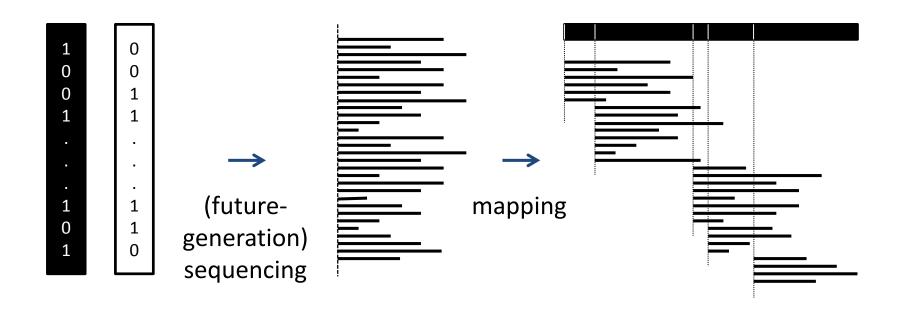
- The haplotypes are of fundamental importance for many applications (Bronwing and Browning, Nature Reviews, 2011):
  - genetic variations / gene function
  - genetic variations / disease susceptibility
  - Genetic variations / drug resistance
  - Etc...
- A whole-experimental reconstruction of the haplotypes is not cost effective => Reconstruction of haplotypes from easier collectable data is necessary:
  - Computational reconstruction (i.e., Statistical approach)
  - Experimental reconstruction throught combinatorial approaches

## Sequencing

- Read = A fragment of a single strand of DNA (i.e., chromosomes)
- Sequencing = collect reads from a copy of a chromosome



## Pipeline

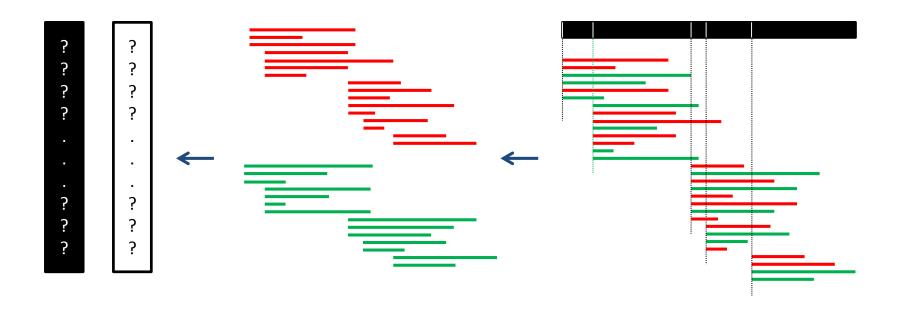


True complete haplotypes

Collection of reads

Aligned reads

## Pipeline



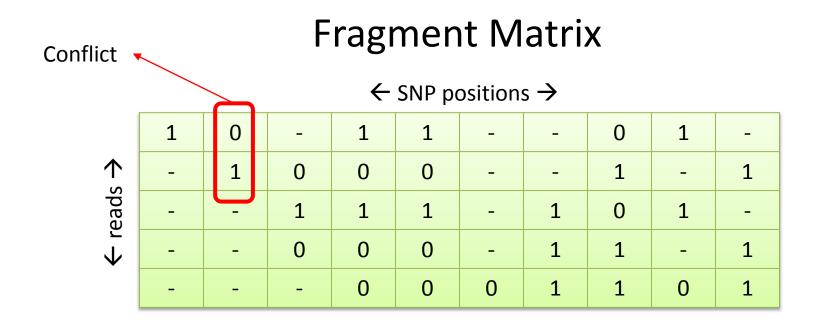
The reads are bipartite in order to reconstruct the two haplotypes

## Fragment Matrix

 $\leftarrow$  SNP positions  $\rightarrow$ 

					-					
	1	0	-	1	1	-	-	0	1	-
<b>↑</b>	-	1	0	0	0	-	-	1	-	1
reads	-	-	1	1	1	-	1	0	1	-
$\downarrow$	-	-	0	0	0	-	1	1	-	1
	-	-	-	0	0	0	1	1	0	1

- Each row corresponds to a read
- Each column correspond to a SNP position
- NOTICE: Homozygous positions may be ignored since they do not give information on the reconstruction/bipartition



- Conflict = different values on the same position
- The conflict may guide the reconstruction of the bipartition

## Fragment Matrix

← SNP positions →

	1	0	-	1	1	-	-	0	1	-
<b>↑</b> s	-	1	0	0	0	-	-	1	-	1
reads	-	-	1	1	1	-	1	0	1	-
<b>→</b>	-	-	0	0	0	-	1	1	-	1
	-	-	-	0	0	0	1	1	0	1

- Conflict = different values on the same position
- The conflict may guide the reconstruction of the bipartition

### **Errors**

← SNP positions →

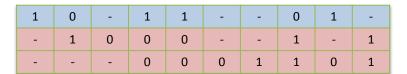
	1	0	-	1	1	-	-	0	1	-
↑ s	-	1	0	0	0	-	-	1	-	1
reads	-	-	1	0	1	-	1	0	1	-
<b>↓</b>	-	-	0	1	0	-	1	1	-	0
	-	-	-	0	0	0	1	1	0	1

- Sequencing or mapping errors => the reads cannot be unambigously bipartited
- The errors lead to an optimization problem

### MFR and MSR

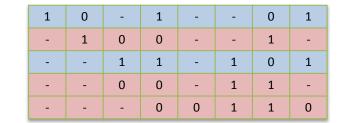
#### Minimum Fragment Removal (MFR)

1	0	-	1	1	-	-	0	1	-
-	1	0	0	0	-	-	1	-	1
-	-	1	0	1	-	1	0	1	-
-	-	0	1	0	-	1	1	-	0
-	-	-	0	0	0	1	1	0	1



#### Minimum SNP Removal (MSR)

1	0	-	1	1	-	-	0	1	-
-	1	0	0	0	-	-	1	-	1
-	-	1	0	1	-	1	0	1	-
-	-	0	1	0	-	1	1	-	0
-	-	-	0	0	0	1	1	0	1



## Minimum Error Correction (MEC)

$\leftarrow$ SNP	positions	$\rightarrow$
------------------	-----------	---------------

	1	0	-	1	1	-	-	0	1	-
<b>↑</b>	-	1	0	0	0	-	-	1	-	1
reads	-	-	1	0	1	-	1	0	1	-
↓ ↓	-	-	0	1	0	-	1	1	-	0
	-	-	-	0	0	0	1	1	0	1

Input: Fragment matrix

 Output: Minimum number of corrections that allow to unambiguously bipartite the reads

 A weighted variant (wMEC) assigns a weight to each element and minimize the total correcting weight => improve accuracy

## Minimum Error Correction (MEC)

	← SNP positions →										
	1	0	-	1	1	-	-	0	1	-	
<b>↑</b> s	-	1	0	0	0	-	-	1	-	1	
reads	-	-	1	0	1	-	1	0	1	-	
$\downarrow$	-	-	0	1	0	-	1	1	-	0	
	-	-	-	0	0	0	1	1	0	1	

• **Input**: Fragment matrix

 Output: Minimum number of corrections that allow to unambiguously bipartite the reads

 A weighted variant (wMEC) assigns a weight to each element and minimize the total correcting weight => improve accuracy

### **MEC Variants**

Binary MEC (no holes, no gaps)

Gapless MEC (holes, no gaps)

Gap MEC (holes, gaps)

0	0	1	0	1	1	0	1	0	1
0	0	0	1	0	1	0	1	0	1
0	0	1	1	1	0	1	0	1	0

0	0	1	0	1	1	-	-	-	-
-	0	0	1	0	1	0	-	-	-
-	-	-	-	1	0	1	0	1	0

0	0	-	-	1	1	-	-	-	-
-	0	-	-	0	-	0	-	-	_
-	1	1	-	-	1	0	-	-	-

### **MEC Variants**

Binary MEC (no holes, no gaps)

- Computational Complexity?
- Scheme of Approximation

Gapless MEC (holes, no gaps)

- NP-Hard
- Approximation?

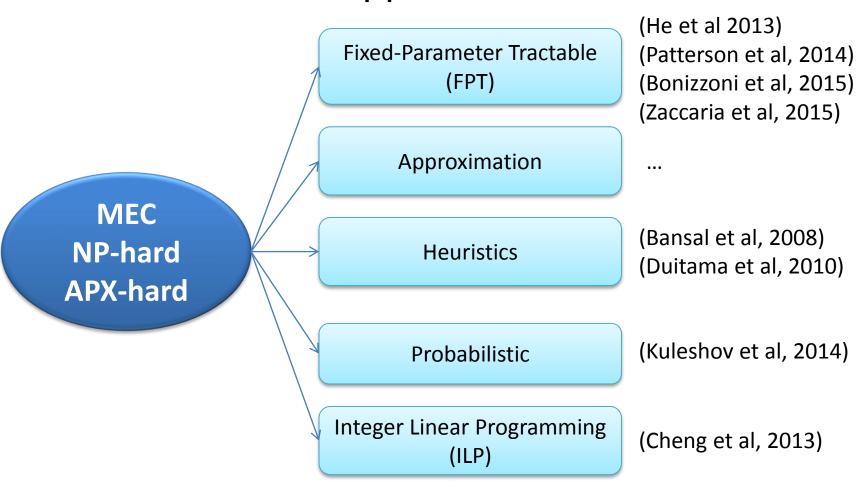
Gap MEC (holes, gaps)

- NP-Hard
- APX-Hard (recently, not in APX)

<sup>\*</sup>Cilibrasi et al., The Complexity of the Single Individual SNP Haplotyping Problem, Algorithmica, 2007.

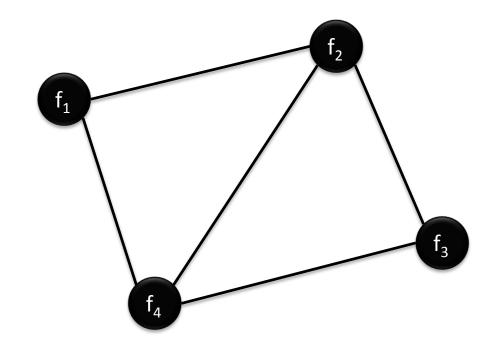
<sup>\*</sup>Bonizzoni et al., 2015

## **Approaches**



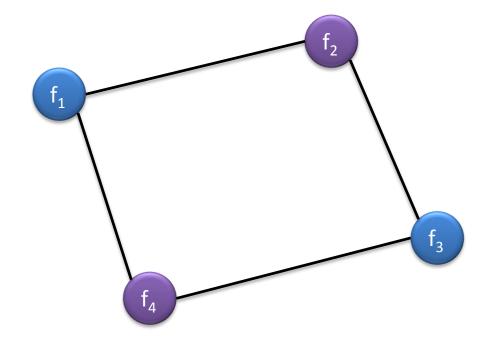
## Graph Model: Fragment Conflict Graph

	$p_1$	p <sub>2</sub>	$p_3$	$p_4$
$f_1$	0	1	-	-
$f_2$	-	0	1	1
$f_3$	-	-	0	1
$f_4$	1	-	-	0



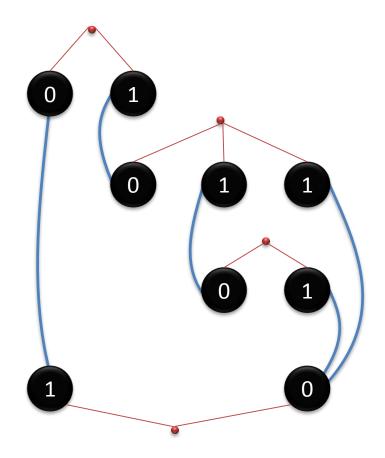
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$f_4$	1	-	-	0



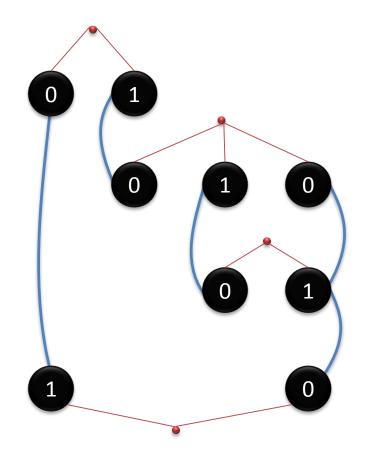
# Graph Model: SNP Conflict Graph

	$p_1$	p <sub>2</sub>	$p_3$	$p_4$
$f_1$	0	1	-	-
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$f_3$	-	-	0	1
$f_4$	1	-	-	0

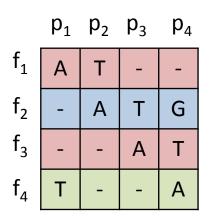


# Graph Model: SNP Conflict Graph

	$p_1$	p <sub>2</sub>	$p_3$	$p_4$
$f_1$	0	1	-	-
$f_2$	-	0	1	0
$f_3$	-	-	0	1
$f_4$	1	-	-	0



 K-ploid Haplotype Assembly (fundamental for k-organisms or tumors, etc...)



#### Theoretical Focus:

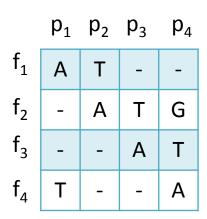
- Computational Complexity
- Fixed-Parameter Tractability
- Approximation Complexity

#### Algorithmic Focus:

- Fixed-Parameter Tractable algorithm
- Approximation algorithm
- Euristichs

#### Experimental Focus:

multiallelic Minimum Error Correction (mMEC)



#### Theoretical Focus:

- Computational Complexity
- Fixed-Parameter Tractability
- Approximation Complexity

#### Algorithmic Focus:

- Fixed-Parameter Tractable algorithm
- Approximation algorithm
- Euristichs

- Design of an extension of the FPT algorithm presented in (Bonizzoni et al., 2015) exponential in the read length and following a dynamic programming approach:
  - Extension in order to manage weights and extend the algorithm in order to deal with gaps in the most general MEC
  - Introduce additional constraints in order to improve performance and bound the searching space
  - Implement the algorithm in a tool and experiment it on real and realistically simulated data, comparing with other current state-of-the-art approaches

 Implement the 2-APX algorithm presented in (Bonizzoni et al., 2015), implement the other approaches proposed in literature and compare their performance and accuracy results.

 Analyze the possibility to extend the 2-APX algorithm presented in (Bonizzoni et al, 2015) to the gapless variant of the MEC problem

- Carry on the study of the computational complexity,
  approximability, and tractability of the variants of MEC problem:
  - Computational complexity of Binary MEC
  - Approximability of Gapless MEC
  - FPT of the variants in the remaning parameters (MEC on the read length)
- Search for:
  - New models (es. graph models) for MEC problems
  - New heuristics for the MEC problem and compare it with the existing ones

#### Good Luck!

For information or question, correspondence to:

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