NGS and Genomes assemblies

**Bioinformatics teachings** 

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months / years

### Data release

### Application and Challenge of 3rd Generation Sequencing for Clinical Bacterial Studies

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#### Abstract

Over the past 25 years, the powerful combination of genome sequencing and bioinformatics analysis has played a crucial role in interpreting information encoded in bacterial genomes. High-throughput sequencing technologies have paved the way towards understanding an increasingly wide range of biological questions. This revolution has enabled advances in areas ranging from genome composition to how proteins interact with nucleic acids. This has created unprecedented opportunities through the integration of genomic data into clinics for the diagnosis of genetic traits associated with disease. Since then, these technologies have continued to evolve, and recently, long-read sequencing has overcome previous limitations in terms of accuracy, thus expanding its applications in genomics, transcriptomics and metagenomics. In this review, we describe a brief history of the bacterial genome sequencing revolution and its application in public health and molecular epidemiology. We present a chronology that encompasses the various technological developments: whole-genome shotgun sequencing, high-throughput sequencing, long-read sequencing. We mainly discuss the application of next-generation sequencing to decipher bacterial genomes. Secondly, we highlight how long-read sequencing technologies go beyond the limitations of traditional short-read sequencing. We intend to provide a description of the guiding principles of the 3rd generation sequencing applications and ongoing improvements in the field of microbial medical research



### Data release

- Submission of the sequence on public databases
- Not always => publication

#### 3 main public databases:

- EMBL-EBI - ENA (European Nucleotide Archive) http://www.ebi.ac.uk/embl/

- GenBank (USA) – NCBI http://www.ncbi.nlm.nih.gov/Genbank/

- DDBJ (DNA DataBank of Japon) – CIB http://www.ddbj.nig.ac.jp/





They are associated (International Nucleotide Sequence Database Collaboration) and exchange the same data which is periodically duplicated together

#### Contain:

- Sequences of DNA or RNA from various sequencers technologies and from many labs
  - \* Some genome fragments : one or more genes, intergenic sequences, parts of a genome
  - \* Completed genomes
  - \* mRNA, tRNA, rRNA (ie. 16s)
- Annotations

### Aims

Molecular detection and identification



Phenotype prediction

### From the bench to the publication



### Quality level of genomes

# - The genome sequence must be completed with a high quality and annotated before the release

Of course the best, but very time consuming. Actually, 90-95 % of a microorganism genome could be easy covered without finishing, but the 5-10 % remained can take many weeks or months to be ended => now easier using long reads sequening



## - The sequence should be uncompleted with a draft quality, whether we suppose most of the genes are sequenced (and identified)

Many eukaryote genomes are only draft genomes, because of the complexity of finishing

=> In general, fundamental research usually performs high quality genomes and applicative research (industry, part of clinical) usually performs draft genomes

=> depending of the project : time and experience (bioinformatician), money (coverage of NGS), organisms, the question to answer

### Sequencing technologies



### Principle of sequencing and assembly



### Sequencing step: reads have heterogeneous distribution







Fragmentation + sequencing => sets of reads

Build of contigs with overlapping regions

Assembly :

=> alignements of reads + consensus

### Principle of sequencing and assembly

Search for best pairings by comparing each sequence (and its reverse complement) against every others sequences to find the best overlapping

=> list of best candidates with similarities criteria

Best candidate is a compromise between :

- maximum overlap length region of similarity between regions
- minimum overhang length unaligned ends of the sequences
- maximum % identity in overlap region
- minimum repeat length

overlap (19 bases) overhang (6 bases) ...AGCCTAGACCTACAGGATGCGCGGACACGTAGCCAGGAC CAGTACTTGGATGCGCTGACACGTAGCTTATCCGGT... overhang % identity = 18/19 % = 94.7%

=> Many **assemblers** tools existed (depending of sequencings technologies, libraries, genomes size, etc..)



### Constructions of library from genomes fragments

\* single - end (= shotgun) : reads sequenced independently

\* paired-end (similar to mate-pair) : reads are sequenced by pairs

- The distance between the reads is known (length of the insert), with some experimental uncertainty

- Distance of insert depends of technology (ie. Illumina ~150 nt for paired-end, ~1-5 kb mate-paired)



Why using PE/MP ? length of reads is limited => assemble repeatitive regions by using reads as "anchors"



**Contigs** : group of overlapping reads, without gap

**Scaffold** : group of contigs order and in the same sens. Gap ("NNN") could existed and their length are known. Scaffolds exists only if a paired-ends (or mate pairs) sequencing was performed !

**Mapped scaffolds** : scaffolds mapped along a reference. Order, orientation and length of gaps are estimated, but not sure !

### Main remaining problems



- Bad assembly of reads
- Low coverage of reads
- Bad insert size estimation
- Different orientation of contigs
- Error of sequencing
- Repeat sequence ambiguities

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11_2_48_792_121		=> P=33	Q=0   286 128 127 294		
5025/5024 5050/5045	5075/5074	5100/5099	5125/5124	5150/5149	5175/5174
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TCTTTTTTGTATTACATTTTTTATTGCCGT	GAOCTATTCTAGTTATTGTATTGGGGCATTTA	ATTTTTTACACAGAACTGGTAGAACAT	CTAGGTTATATTGCCACOGT	CATTATGATAAAGTTTGTTTTAGACCTGAGTA	CTTAOGTAAAGATGCTTATAA
CITITITGIATIACATITI TOCCGI	GAGCTATTCTAGTTATTGTATTGGGAC TTTA	ATTTTTTACACAGAACTGGTAGAACAT	C GGTTATATTGCCACGGT	GATTATGATAAAGTTT TTAGACCTGAGTA	CTTAGGTAAAGATGCTTATA
CITIT	GAGGTATICTAGTTATIGTATIGGGACATTT	TTTACACAGAACTGGTAGAACAT	TAGGTTATA OGT	CATTATGATAAAGTTTGTTTTAGACCTGAG	TAGGTAAAGATGCTTATA
CTITT GT	GAGATATTCTAGTTATTGTATTGGGAGAAATTT	CACAGAACTGGTAGAACAT	CTAOGTTATATATICC GGT	CATTATGATAAAGTTTGTTTTAGACCTGAG	OGTANAGATOCTTATA
CTITIT	GAGCTATTCTAGTTATTGTATTGGGACATTTA	A AGAACTGETAEAECAT	CTAGGTTATATTGCCAC GT	GATTATGATAAAGTTTGTTTTAGACCTGATC	<b>GOTAJAGATOCITATA</b>
CITITI	GAGCTATTCTAGTTATTGTATTGGGLCATTTA	AGAACTGGTAGACCAT	CTAGGTTATATTGCCAC	GATTATGATAAAGTTTGTTTTAGACCTGAGT	GOTAMAGATOCTTATA
CTITITIGTATTACAT	GCTATTCTAGTTATTGTATTGGGLCATTTA	GAACTGGTAGAACAT	CTAGGTTATATTGCCACG	GATTATGATAAAGTTTGTTTTAGACCTGAGTA	C GTAAAGATGCTTATA
CTITTTGTATTACATTTTTATTGC	TATTCTAGTTATTGTATTGGGGCACATTTA	GAACAT	CTAGGTTATATTGCCACGGT	GATTATG GTTTGTTTTAGACCTGAGTA	CTTAGGTAAAGAT
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CITITI	GTATTGGGACATTTA	ATTTTTTACACAGAACT	GGTTATATTGCCAOGGT	GATTATGATAAAGTTT AGACCTGAGTA	CTTAGGTAAAGATGCTTATA
	GTATTGGGLCATTTA	ATTTTTTACACAGAACT	GTTATATTGCCACGGT	GATTATGATAAAGTTIG AGACCTGAGTA	CTTROSTARRATICTTATA
CITITIG	GTATTGGGACATTTA	ATTTTTTACACAGAACT	GTTATATTGCCACGGT	GATTATGATAAAGTTTG GACCTGAGTA	CTTROCTARREATOCTTATA
CTTTTTTG	TATTGGGACATTTA	ATTTTTTACACAGAACTG	GTTATATTGCCACGGT	CATTATGA CAAAGTTTG TGAGTA	CTTROGTARAGATOCITATA
CTTTTTTGTATTACATTTTTTATTGC	GACATTTA	ATTTTTTACACAGAACTGGTAGAA	GTTATATTGCCACGGT	CATTATGATAAAGTTIG GAGTA	CTTAGGTAAAGATGCTTATA
CITITITGTATTACATTITITATTGCC	ACATTTA	ATTTTTTACACAGAACTGGTAGAAC	TTATATTGCCACGGT	GATTATGATAAAGTTTGT GAGT	CITAGGTAAAGATGCTTATA
CTITITIGIATTACATTTTTTATTGCCGT	ACATTTA	ATTTTTTACACAGAACTGGTAGAAC	ATTOCCACGGT	GATTATGATAAAGTTTGTTTTA GAGTA	CTTAGGTAAAGATGCTTATA
CTTTTTTGTATTACATTTTTTATTGCCGT	ATTTA	ATTTTTTACACAGAACTGGTAGAACAT	ATTGCCACGGT	COTTATGATAAAGTTTGTTTTA GAGTA	CTTAGGTAAAGATGCTTATA
CTTTTTTGTATTACATTTTTTATTGCCGT		ATTTTTTACACAGAACTGGTAGAACAT	CTAG GGT	GATTATGATAAAGTTTGTTTTAGACCTGAG	GTAAAGATGCTTATA
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				GATAAAGT TGTTTTAGACCTGAGTA	CITAGGT GATGCTTATA
				GATAAAGTTTGTTTTAGACCTGAGTA	CITAOGT GATOCITATA
				GTTTGTTTTAGACCTGAGTA	CTTAGGTAAAGAT
				GTTTTAGACCTGAGTA	CTTAGGTAAAGATGCTT
				GAGTA	CITAGGTAAAGATGCTT
				GAGTA	CTTAGGTAAAGATGCTTATA
				GAGTA	CTTAGGTANAGATGCTTATA
				GAGTA	CTTAGGTANAGATGCTTATA
				GAGTA	CITROGTANAGATOCITATA
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Re-Mapping, to see coverage, SNP and potential errors



### Finishing

- (re)Mapping of reads along the assembled genome (or/and a reference)
- help to correct the low quality/coverage areas
- Check the order of contigs
- Check the redundancy of contigs (false contigs or true repeat contigs like rRNA operons)
- Compare syntheny between multiple assemblers (global alignement)
- Fill the gaps by extending the boundaries of each gap using ends of mapping reads (or use PCR)
- Order (or reorder) contigs
- Desassemble some areas if they seem to be false



Bacillus cereus assemblies using 3 assemblers tools.2 first genomes are very similar, the third show many differences

=> High improvement with new long-reads technology (MinIon Nanopore, PacBio)