

# RNA-Seq de novo assembly training Day 2





# **Session organisation: Day 2**

#### Morning:

Assembly quality assessment

- \* Assemblathon stats
- \* Read mapping stats
- Clustering
  - \* CD-HIT
- Greedy assembly
  - \* SSAKE/VCAKE
- Overlap Layout Consensus
  - \* CAP3

#### \* TGICL

#### Afternoon:

- de Bruijn graph
  - \* Velvet/Oases
  - \* Trinity
- Pretreatments



# **Objectives for this second day**

Answer the following questions:

- Which assembler should I choose to process my data?
- Which procedure should I use for my assembler?
- Which computer do I need to run my assembly?





#### **Genome vs transcriptome** assembly

#### Differ mainly in:

- matrices coverage
  - \* generally uniform vs highly variable
- combinations of sequences
  - \* repeats and allelic variations vs alternative splicing

Two steps using genome assembler:

- use genome assembler to assemble transcriptome
- develop pipelines to postprocess the output of genome assemblers



# **Objectives of the assembly**

An assembly is a sum-up of the matrices from which the reads have been produced:

- complete (all represented transcripts)
- compact (one contig for a transcript)
- independent of the expression level
- not affected by the random errors of the sequencing technology

Two possible level-analysis:

- transcripts
- genes



# How do you assess the quality of an transcriptome assembly?



#### **Assembly quality assessment**

#### 3 tracks

#### Assembly metrics



#### Shape of contig length histogram



# Reads mapping back rate







#### Vocabulary

- **Contig**: a set of overlapping segments that together represent a consensus sequence
- Scaffold: a series of contigs that are in the right order but not necessarily connected in one continuous stretch of sequence
- N50: given a set of contigs of varying lengths, the N50 length is defined as the length N for which 50% of all bases in the contigs are in contigs of length L < N</li>

contig size list L = (2, 2, 2, 3, 3, 4, 8, 8)

we have 50% of total length (16/32) above 4

**N50** is equal to 4+8/2 = 6

 L50: number of contigs that are greater than, or equal to, the N50 length



## **Contig metrics**

The possible metrics derived from genome assembly:

- Idea of global size (# bases)
- Idea of number of elements (# contigs/scaffolds)
- Idea of compactness (N50)
- where the second sec



# **Assemblathon statistics**

Script which calculates many of the basic contig and scaffold level statistics

- N50
- Longest/shortest contig/scaffold
- Median size of contigs/scaffolds
- Mean size of contigs/scaffolds
- Total size of contigs/scaffolds
- % N, A, T, G, C

Assemblathon 1: a competitive assessment of de novo short read assembly methods. Earl D & al. Genome Res. 2011 Dec;21(12):2224-41

Assemblathon 2: evaluating de novo methods of genome assembly in three vertebrate species. Bradnam KR & al. Gigascience. 2013 Jul 22;2(1):10





#### **Assemblathon: command line**

bash-4.1\$ assemblathon\_stats.pl
Usage: assemblathon\_stats.pl <assembly\_scaffolds\_file>
options:

-limit <int> limit analysis to first <int> sequences (useful for testing)

-csv produce a CSV output file of all results

-graph produce a CSV output file of NG(X) values (NG1 through to NG99), suitable for graphing -n <int> specify how many consecutive N characters should be used to split scaffolds into contigs

-genome\_size <int> estimated or known genome size

-n distinguish scaffolds and contigs (default 25)
-csv output in csv format



#### **Assemblathon statistics**

#### assemblathon stats.pl contigs.fa

- Information for assembly 'F\_Dr\_1/merge/cap3/all\_contigs\_singlets.filtered.tfa' -

Number of scaffolds	16164	
Total size of scaffolds	38963808	
Longest scaffold	17142	
Shortest scaffold	201	
Number of scaffolds > 1K nt	13358	82.69
Number of scaffolds > 10K nt	72	0.49
Number of scaffolds > 100K nt	Θ	0.09
Number of scaffolds > 1M nt	Θ	0.09
Number of scaffolds > 10M nt	Θ	0.09
Mean scaffold size	2411	
Median scaffold size	1922	
N50 scaffold length	3126	
L50 scaffold count	4028	
scaffold %A	27.16	
scaffold %C	22.96	
scaffold %G	23.25	
scaffold %T	26.62	
scaffold %N	0.00	
scaffold %non-ACGTN	0.00	
Number of scaffold non-ACGTN nt	Θ	



## **Transcript length histogram**

Transcript lengths are not randomly distribute We should get a known distribution shape





#### **Transcript length histogram**





# **Transcript length histogram**

#### 





#### Length histogram

python /usr/local/bioinfo/Scripts/bin/length\_histogram.py -l -i transcripts.fa



#### **Complete transcriptomes**

#### **Tissue specific assembled transcriptomes**





#### **Genome histogram**

Comparison with the panda genome assembly (v1, 2009)



81467 scaffolds Total 2,3 Gb Longest 6Mb Shortest 100 b N50 1,3 Mb L50 521



#### **Zebrafish transcriptome**



47880 contigs Total 93 Mb Longest 94 kb Shortest 10 b N50 2622 L50 10495





#### **Realignment metrics**

The assembly is a sum-up. The realignment rate gives how much of the initial information is inside the contigs. Reads mapped back to transcripts (RMBT)

- align reads against assembly generated transcripts
- compute percentage of reads mapped





## **Realignment metrics**

#### Factors affecting realignment rate:

- Presence of highly expressed genes
- Contamination by building blocks (adaptors)
- Reads quality







#### **RMBT** with **BWA**

#### The realignment steps are:

- indexing the reference (bwa)
- aligning the reads, producing a sam file (bwa)
- compressing, sorting and indexing the sam file in a bam file (samtools)
- counting the aligned reads (samtools):
  - \* global alignement rate
  - \* num reads / contig



#### **RMBT: command lines**

Index the reference

bwa index -a [is bwtsw] reference.fa

Align the reads

bwa aln -f R1.sai reference.fa R1.fastq.gz

bwa aln -f R2.sai reference.fa R2.fastq.gz

bwa sampe -f output.sam reference.fa R1.sai R2.sai R1.fastq.gz R2.fastq.gz

Compress, sort and index

samtools view -bS output.sam | samtools sort - output.sorted

samtools index output.sorted.bam

Count reads

samtools flagstat output.sorted.bam

samtools idxstats output.sorted.bam

Or use /home/sigenae/bin/runSampe 😇





# **Exercise n°1**



# Lets assemble!





# **Assembly: vocabulary**

- Graph: a data structure which consists of a finite set of ordered pairs, called edges or arcs, of certain entities called nodes or vertices
- a simple grap example:
  - \*6 nodes
  - \*7 edges



• **Path**: in a graph, a path is a sequence of edges which connect a sequence of vertices





## **Assembly: vocabulary**

- K-mer: a k-mer is a sub-string of length k
- A string of length L has (L-k+1) k-mers
- Example: a read with L = 8 has 5 k-mers when k = 4
  - A G A T C C G T
  - AGAT
  - GATC
  - ATCC
    - TCCG
      - СССТ



#### **Sequence clustering**

**Inclusion** reduces the overall size of the dataset without removing any sequence information by only removing "redundant" (or highly similar) sequences.

Tools:

- CD-HIT
- BLASTClust
- UCLUST
- UICluster



#### **CD-HIT**

# CD-HIT is a widely used program for clustering and comparing protein or nucleotide sequences.

BIOINFORMATICS APPLICATIONS NOTE Vol. 22 no. 13 2006, pages 1658-1659 doi:10.1093/bioinformatics/bt/158

Sequence analysis

Cd-hit: a fast program for clustering and comparing large sets of protein or nucleotide sequences

Weizhong Li\* and Adam Godzik Bumham Institute for Medical Research, La Jolla, CA 92037, USA

Received on March 23, 2006; revised and accepted on April 20, 2006 Advance Access publication May 26, 2006 Associate Editor: Golan Yona

Overview: CD-HIT clusters included sequences.







#### **CD-HIT**

CD-HIT uses greedy incremental clustering algorithm method:

- sequences are first sorted in order of decreasing length
- the longest one becomes the representative of the first cluster
- each remaining sequence is compared to the representatives of existing clusters
- if the similarity with any representative is above a given threshold → grouped into that cluster
- if not, a new cluster is defined



#### **Cd-hit: command line**

bash-4.1\$ cd-hit-est

====== CD-HIT version 4.6 (built on May 2 2012) ======

Usage: cd-hit-est [Options]

#### **Options**

-i input filename in fasta format, required -o output filename, required -c sequence identity threshold, default 0.9 this is the default cd-hit's "global sequence identity" calculated as: number of identical amino acids in alignment divided by the full length of the shorter sequence use global sequence identity, default 1 -G if set to 0, then use local sequence identity, calculated as : number of identical amino acids in alignment divided by the length of the alignment NOTE !!! don't use -G 0 unless you use alignment coverage controls see options -aL, -AL, -aS, -AS -b band width of alignment, default 20 -M memory limit (in MB) for the program, default 800; 0 for unlimitted; number of threads, default 1; with 0, all CPUs will be used -T word length, default 10, see user's quide for choosing it -n length of throw\_away\_sequences, default 10 -1 -d length of description in .clstr file, default 20 if set to 0, it takes the fasta defline and stops at first space -s length difference cutoff, default 0.0 if set to 0.9, the shorter sequences need to be at least 90% length of the representative of the cluster

cd-hit-est -i input.fa -o output.fa -M 0 -d 0 -c 0.98 -T 8 > output.log



#### **Cd-hit: output**

cd-hit-est -i transcripts.fa -o all\_seq.fa -M 0 -d 0 -c 0.98 -T 8 > transcripts.fa.cd-hit.log

bash-4.1\$ grep -c '^>' transcripts.fa all\_seq.fa
transcripts.fa:177687
all seq.fa:35450

bash-4.1\$ sed -n '60,\$p' all seq.fa.clstr | head -20 >Cluster 1 16528nt, >k37 Locus 3131 Transcript 3... \* Θ 16528nt, >k31 Locus 3279 Transcript 3... at +/100.00% 1 2 16479nt, >k43\_Locus\_3102\_Transcript\_1... at -/100.00% 3 16528nt, >k49 Locus 3163 Transcript 1... at +/100.00% 16528nt, >k55 Locus 3311 Transcript 1... at -/100.00% 4 14091nt, >k61 Locus 3512 Transcript 1... at -/100.00% 5 6 2497nt, >k61 Locus 4997 Transcript 1... at -/100.00% 7 14091nt, >k65 Locus 3683 Transcript 1... at -/100.00% 8 2497nt, >k65 Locus 5316 Transcript 1... at -/100.00% 9 11230nt, >k69\_Locus\_3750\_Transcript\_1... at -/100.00% 10 2497nt, >k69 Locus 5474 Transcript 1... at -/100.00% 11 2927nt, >k69 Locus 5643 Transcript 1... at -/100.00% >Cluster 2 15727nt, >k31\_Locus\_124\_Transcript\_1... \* Θ >Cluster 3 15481nt, >k25 Locus 2766 Transcript 1... at -/99.97% Θ 1 181nt, >k31 Locus 10917 Transcript 1... at -/100.00% 2 15675nt, >k37 Locus 3865 Transcript 3... \* 3 15494nt, >k31 Locus 4017 Transcript 3... at +/99.94%





# **Assembly introduction**

#### Three classes of methods:

- Greedy method
- Overlap Layout Consensus (OLC) method
- de Bruijn graph (DBG) method



#### **Greedy method**

Greedy method joins a read with another read that has the best overlap score until no more reads can be joined.

Overview:

- calculate pairwise alignments of all reads
- score and sort alignments (length/matching)
- merge the two reads with the highest scoring overlap and add the resulting "contig" to the pool of sequences
- continue to extend a contig until no more quality overlaps exist Greedy:
  - optimize a local objective function (quality of the overlap)
  - approach that may not lead to a globally optimal solution



#### **SSAKE / VCAKE**

# VCAKE is a modification of simple k-mer extension (SSAKE) that overcomes error by using high depth coverage

#### **BIOINFORMATICS APPLICATIONS NOTE**

Vol. 23 no. 4 2007, pages 500–501 dol:10.1093/bioinformatics/bt/629

#### Genome analysis

#### Assembling millions of short DNA sequences using SSAKE

René L. Warren\*, Granger G. Sutton<sup>1</sup>, Steven J. M. Jones and Robert A. Holt British Columbia Cancer Agency, Genome Sciences Centre, 675 West 10th Avenue, Vancouver, BC V5Z 1L3, Canada and <sup>1</sup>J. Craig Venter Institute, 9704 Medical Center Drive, Rockville, MD 20850, USA

Received on October 6, 2006; revised on November 15, 2006; accepted on December 5, 2006 Advance Access publication December 8, 2006 Associate Filter: Alex Externan BIOINFORMATICS APPLICATIONS NOTE Vol. 23 no. 21 2007, pages 2942-2944 doi:10.1093/bioinformatics/btm451

#### Genome analysis

#### Extending assembly of short DNA sequences to handle error

William R. Jeck<sup>1,\*</sup>, Josephine A. Reinhardt<sup>1</sup>, David A. Baltrus<sup>1</sup>, Matthew T. Hickenbotham<sup>2</sup>, Vincent Magrini<sup>2</sup>, Elaine R. Mardis<sup>2</sup>, Jeffery L. Dangl<sup>1,3</sup> and Corbin D. Jones<sup>1,3</sup>

<sup>1</sup>Department of Biology, University of Carolina-Chapel Hill, Chapel Hill, NC 27599, <sup>2</sup>Department of Genetics, Washington University School of Medicine, St. Louis, MO 63108 and <sup>3</sup>Carolina Center for Genome Sciences, University of Carolina-Chapel Hill, Chapel Hill, NC 27599, USA

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Each possible 3' most k-mer from seed reads are used to browse a prefix tree organizing reads by their first eleven 5' end bases.

aeno toul 4

#### Hash table keyed by num occurrences

96	ATGTCTTCCTATCGTGTA
89	TGTAGCGCTATCGTCAAG
67	GTCATGTCGTATTTTGTA
42	CGATCGATGCTAGTATAT

Each 3' most k-mer ATGTCTTCCTATCGTGTA **TGTCTTCCTATCGTGTA GTCT**TCCTATCGTGTA **TCTTCCTATCGTGTA CTTCCTATCGTGTA TTCCTATCGTGTA** TCCTATCGTGTA **CCTATCGTGTA CTATCGTGTA TATCGTGTA ATCGTGTA** Min length = 8

#### **SSAKE / VCAKE**





#### **SSAKE / VCAKE**




# **SSAKE / VCAKE**



Each 3' most k-mer TGTAGCGCTATCGTCAAG GTAGCGCTATCGTCAAG AGCGCTATCGTCAAG AGCGCTATCGTCAAG GCGCTATCGTCAAG GCTATCGTCAAG GCTATCGTCAAG TATCGTCAAG ATCGTCAAG ATCGTCAAG CGTCAAG Min length = 8





# **OLC method**

OLC method generates a graph using reads and overlaps. The assembly process becomes synonymous with finding a pathway through the graph that visits every node at exactly once.

#### Tree steps:

- Overlap
- Layout
- Consensus







# **Overlap**

- Each read is compared to every other reads in both the forward and reverse complement orientations
- Different OLC algorithms have different criteria for OLCquality overlaps
- In the assembly graph, the nodes represent actual reads, the edges represent overlaps between these reads







# Layout: simplify graph

- In order to decrease the size of the graph, the OLC assembly graph is simplified in the layout stage
- There is one path that visits every read, highlighted in red
- Nodes are compressed into contigs until a fork is reached







# **Layout: resolve repeats**

Forks typically signify the boundary between repeats and unrepeated segments

- a fork is formed because the reads that link the R1+R2 contig to Y and Z do not overlap on the suffix end
- both repeat sections R1 and R2 are compressed into a repeat contig. X, Y and Z are compressed into unique contigs
   A x R<sub>1</sub> Y R<sub>2</sub> Z



http://gcat.davidson.edu/phast/olc.html





## Consensus

- After contig generation, consensus sequences are derived
- Starting from the left most read of each contig, the OLC algorithm computes the consensus of all of the reads composing each contig



# **Greedy vs OLC**

- Both begin with overlap generation
- Steps of OLC enable a global analysis of the assembly problem
- Local analysis of greedy algorithms is a limit
- For this repeat example:
  - \* the proper reconstruction X-R1+R2-Y-R1+R2-Z is easily inferred using the OLC method
  - \* a greedy extension would produce
    - fragmented assembly (X-R1+R2; Y; Z)
    - misassembly (X-R1+R2-Z ; Y)







CAP3 (CONTIG ASSEMBLY PROGRAM Version 3) is a sequence assembly program for small-scale assembly with or without quality



#### CAP3: A DNA Sequence Assembly Program

Xiaoqiu Huang and Anup Madan

Genome Res. 1999 9:868-877 Access the most recent version at doi:10.1101/gr.9.9.868





# TGICL

#### TGICL is a pipeline for analysis of large Expressed Sequence Tags (EST) and mRNA databases

- MegaBLAST
- Clustering
- Large clusters splitting
- CAP3



#### BIOINFORMATICS APPLICATIONS NOTE Vol. 19 no. 5 2003, pages 651–652 DOI: 10.1093/bioinformatics/btg034

#### TIGR Gene Indices clustering tools (TGICL): a software system for fast clustering of large EST datasets

Geo Pertea<sup>1,\*</sup>, Xiaoqiu Huang<sup>2</sup>, Feng Liang<sup>1, 3</sup>, Valentin Antonescu<sup>1</sup>, Razvan Sultana<sup>1</sup>, Svetlana Karamycheva<sup>1</sup>, Yuandan Lee<sup>1</sup>, Joseph White<sup>1</sup>, Foo Cheung<sup>1</sup>, Babak Parvizi<sup>1</sup>, Jennifer Tsai<sup>1</sup> and John Quackenbush<sup>1,\*</sup>

<sup>1</sup>The Institute for Genomic Research, Rockville, MD 20850 USA, <sup>2</sup>Department of Computer Science, Iowa State University, Ames, IA 50011 USA and <sup>3</sup>Current address Invitrogen Corporation, Carlsbad, CA 92008 USA

Received on July 10, 2002; revised on August 12, 2002; accepted on August 16, 2002





# **TGICL**

## TGICL benefit:

- mgblast (MegaBLAST with new filtering options)
   → compressed sorted file of decreasing pairwise
   alignment score
- clustering (tclust, sclust, nrcl) → generally a very large connected component
- splitting  $\rightarrow$  partitioning attempt based on full-length transcripts



# **TGICL: command line**

TGICL(1)

User Contributed Perl Documentation

TGICL(1)

#### TGICL

The Gene Index Clustering Tool (TGICL) is a software package that tries to efficiently cluster and create assemblies (contigs) from a set of input DNA sequences given in a fasta file. The "clustering" phase is intended to partition the input data set into smaller groups of sequences (clusters) that due to stringent similarity have a greater probability to originate from the same longer sequence. However, the clustering phase does not perform any multiple alignment but only fast pairwise alignments (using megablast), which are then filtered and used to build subsets of sequences by a transitive closure approach. In the assembly phase each such cluster is sent to the assembly program (cap3) which attempts the multiple alignment of the sequences in the cluster and creates one or more contigs (consensus sequences). Both clustering and assembly phases can be executed in parallel on multiple CPU machines or in a PVM environment.

#### VERSION

Version 2.1

#### SYNOPSIS

```
tgicl -F <fasta_db> [-q <qualdb>] [-d <refDb>] [-c {<num_CPUs>|<PVM_nodefile>}]
    [-m <user>] [-0 'cap3_options'] [-1 <min_overlap>] [-v <max_overlang>]
    [-p <pid>] [-n slicesize] [-s <maxsize>] [-a <cluster_file>] [-M] [-K]
    [-L] [-X] [-I] [-C] [-G] [-R] [-W <pairwise_script.psx>] [-A <asm_program.psx>]
    [-b user:pass:driver:server:schema] [-P <param_file>] [-u <seq_list>] [-f <prefix_filter>] [-D]
```

To start clustering and assembling all the sequences from fasta file myseq.fasta using 2 CPUs :

tgicl -F myseq.fasta -c 2

To start the same process loading the results into a MySQL database and generating some static HTML file reports.

tgicl -F myseq.fasta -c 2 -b tgiuser:tgipass:mysql:dbserver.my.domain:tgicldb -R

In the previous example the database username is <u>tgiuser</u>, the password is <u>tgipass</u>, the requested database is of course <u>mysql</u>, the database is located on server <u>dbserver.my.domain</u> and the db is <u>tgicldb</u>.

#### tgicl -F input.fa -c 2 -l 60 -p 96



# **TGICL: output**

#### tgicl -F all\_seq.fa -c 2 -l 60 -p 96

bash-4.1\$ 11								
total 13392								
-rw-rw-r 1	sigenae	sigenae	10312464	1 30	oct.	. 16:	00	all_seq.fa
-rw-rw-r 1	sigenae	sigenae	297172	2 30	oct.	. 16:	00	all_seq.fa.cidx
-rw-rw-r 1	sigenae	sigenae	4003	30	oct.	. 16:	01	all_seq.fa_cl_clusters
-rw-rw-r 1	sigenae	sigenae	1926	i 30	oct.	. 16:	01	all_seq.fa_cl_tabhits_001.Z
-rw-rw-r 1	sigenae	sigenae	228976	i 30	oct.	. 16:	00	all_seq.fa.nhr
-rw-rw-r 1	sigenae	sigenae	60064	1 30	oct.	. 16:	00	all_seq.fa.nin
-rw-rw-r 1	sigenae	sigenae	2502693	30	oct.	. 16:	00	all_seq.fa.nsq
-rw-rw-r 1	sigenae	sigenae	136851	30	oct.	. 16:	01	all_seq.fa.singletons
drwxr-x 2	sigenae	sigenae	16384	1 30	oct.	. 16:	01	asm_1
drwxr-x 2	sigenae	sigenae	16384	1 30	oct.	. 16:	01	asm_2
-rw-rw-r 1	sigenae	sigenae	2598	30	oct.	. 16:	01	err_tgicl_all_seq.fa.log
-rw-rw-r 1	sigenae	sigenae	163	30	oct.	. 16:	00	formatdb.log
-rw-rw-r 1	sigenae	sigenae	6	30	oct.	. 16:	01	masked.lst
-rw-rw-r 1	sigenae	sigenae	2238	30	oct.	. 16:	01	tgicl_all_seq.fa.log
-rw-rw-r 1	sigenae	sigenae	3357	30	oct.	. 16:	00	tgicl.cfg
bash-4.1\$ 11	asm 1							
total 736								
-rw-rw-r 1	sigenae	sigenae	297943 3	30 o	ct.	16:01	AC	DE CONTRACTOR OF
-rw-rw-r 1	sigenae	sigenae	335080 3	30 o	ct.	16:01	al	Lign
-rw-rw-r 1	sigenae	sigenae	53919 3	30 o	ct.	16:01	С	ontigs
-rw-rw-r 1	sigenae	sigenae	03	30 o	ct.	16:01	er	r_log
-rw-rw-r 1	sigenae	sigenae	1902 3	30 o	ct.	16:01	10	og_std
-rw-rw-r 1	sigenae	sigenae	781 3	30 o	ct.	16:01	si	inglets



# **TGICL: ACE format**

#### AS 31 115

CO CL1Contig1 286 2 3 U

#### ΒQ

```
AF k25_Locus_18748_Transcript_2 U -61
AF k25_Locus_14452_Transcript_9 U 1
BS 1 19 k25_Locus_14452_Transcript_9
BS 20 20 k25_Locus_18748_Transcript_2
BS 21 286 k25_Locus_14452_Transcript_9
```

QA 82 154 82 154 DS

QA 1 286 1 286 DS

http://bozeman.mbt.washington.edu/consed/distributions/README.14.0.txt



# **TGICL: Align format**

k55\_Locus\_14801\_Tran+ CGTTCATTCAGCGTCGAGTCTTCAGAGCTCGAGCTCGCCGCCATCAGTGAAAGCCAAAGA consensus CGTTCATTCAGCGTCGAGTCTTCAGAGCTGAGCTCTGCCGCCATCAGTGAAAGCCAAAGA k55 Locus 14801 Tran+ CAGAGATTATTGATGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAAAATCAAACAG k37 Locus 13099 Tran+ TTATTGATGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAAAATCAAACAG k69 Locus 1514 Trans+ ATTGAAGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAGAATTAAACAG CAGAGATTATTGATGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAAAATCAAACAG consensus k55 Locus 14801 Tran+ GAAGACACTGAAGAACTAATAGATGTGAAGGAGGAGAGTGAAGAACTGAGTGAAGATGAG k37 Locus 13099 Tran+ GAAGACACTGAAGAACTAATAGATGTGAAG k69 Locus 1514 Trans+ GAAGAGACTGAAGAACTAATAGATGTGAAGCAGGAGGAGTGAAGAACTGAGTGAAGATGAG GAAGACACTGAAGAACTAATAGATGTGAAGCAGGAGAGTGAAGAACTGAGTGAAGATGAG consensus k55\_Locus\_14801\_Tran+ GAGAAACATCAGGTCA k69\_Locus\_1514\_Trans+ GAGAAACATCAGGTCAAAAGTGAAGAAGAAACTCAATCAGAGACTGAACATTTATCTTTT k37 Locus 16908 Tran+ GTCAAAAGTGGAGAAGAAACTCAATCAGAGACTGAACATTTATCTTTT GAGAAACATCAGGTCAAAAGTGAAGAAGAAGAACTCAATCAGAGACTGAACATTTATCTTTT consensus k69\_Locus\_1514\_Trans+\_CTACAATCAAACACACATAACAATTACAATGTGATCATGAAGGAGGAGGAGGAGGAGAGTGAAGAACTG k37 Locus 16908 Tran+ CTACAATCAAACACACATAACAATTACAATGTGATCATGAAGGAGGAGGAGGAGGAGGAGAACTG k55\_Locus\_11728\_Tran+ GTGAAGGAGGAGAGTGAAGAACTG CTACAATCAAACACACATAACAATTACAATGTGATCATGAAGGAGGAGAGTGAAGAACTG consensus k69 Locus 1514 Trans+ AATGAAGATGAGGAGAAACATCAGGTCAAAAGTGAAGAAGAAACTCAATCAGAGACTGAA k37 Locus 16908 Tran+ AATGAAGATGAGGAGAAACATCAGGTCAAAAGTGAAGA k55 Locus 11728 Tran+ AGTGAAGATGAGGAGAAACATCAGGTCAAAAGTGGAGAAGAAACTCAATCAGAGACTGAA AATGAAGATGAGGAGAAACATCAGGTCAAAAGTGAAGAAGAAACTCAATCAGAGACTGAA consensus consensus k69 Locus 1514 Trans+ GAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCAAAGCGTTCAAACAGAAGTGAAG k55\_locus\_11728\_tran+ GAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCAAAGCGTTCAAACAGAAGTGAAG GAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCAAAGCGTTCAAACAGAAGTGAAG consensus k69 Locus 1514 Trans+ TCAAATACTTGCTCTTTGTGGGAAAGACTTTTATAAAGCCATCGTATTTAAAACGACAC k55 Locus 11728 Tran+ TCAAATACTTGCTCTTTGTGTGGAAAGACTTTTATAAAGCCATCGTATTTAAAACGACAC

Plateforme Bioinformatique Midi-Pyrén<del>ée</del>s

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# **Other OLC assemblers**

#### Genome analysis

#### Aggressive assembly of pyrosequencing reads with mates

Jason R. Miller<sup>1,\*</sup>, Arthur L. Delcher<sup>2</sup>, Sergey Koren<sup>1</sup>, Eli Venter<sup>1</sup>, Brian P. Walenz<sup>1</sup>, Anushka Brownley<sup>1</sup>, Justin Johnson<sup>1</sup>, Kelvin Li<sup>1</sup>, Clark Mobarry<sup>3</sup> and Granger Sutton<sup>1</sup>

**ORIGINAL PAPER** 

<sup>1</sup>The J. Craig Venter Institute, 9712 Medical Center Drive, Rockville MD 20850, <sup>2</sup>Center for Bioinformatics & Computational Biology, University of Maryland, College Park, MD 20742 and <sup>3</sup>White Oak Technologies Inc, 1300 Spring St., Ste 320, Silver Spring, MD 20910, USA

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#### CABOG Celera Assembler with the Best Overlap Graph

nature

Vol. 24 no. 24 2008, pages 2818-2824

doi:10.1093/bioinformatics/btn548

Vol 437|15 September 2005|doi:10.1038/nature03959

#### Newbler

#### ARTICLES

# Genome sequencing in microfabricated high-density picolitre reactors

Marcel Margulies<sup>1</sup>\*, Michael Egholm<sup>1</sup>\*, William E. Altman<sup>1</sup>, Said Attiya<sup>1</sup>, Joel S. Bader<sup>1</sup>, Lisa A. Bemben<sup>1</sup>, Jan Berka<sup>1</sup>, Michael S. Braverman<sup>1</sup>, Yi-Ju Chen<sup>1</sup>, Zhoutao Chen<sup>1</sup>, Scott B. Dewell<sup>1</sup>, Lei Du<sup>1</sup>, Joseph M. Fierro<sup>1</sup>, Xavier V. Gomes<sup>1</sup>, Brian C. Godwin<sup>1</sup>, Wen He<sup>1</sup>, Scott Helgesen<sup>1</sup>, Chun He Ho<sup>1</sup>, Gerard P. Irzyk<sup>1</sup>,



#### Using the miraEST Assembler for Reliable and Automated mRNA Transcript Assembly and SNP Detection in Sequenced ESTs

Bastien Chevreux, Thomas Pfisterer, Bernd Drescher, et al.

Genome Res. 2004 14: 1147-1159 Access the most recent version at doi:10.1101/gr.1917404

#### **MIRA**



# But we have billions of reads!





The de Bruijn graph (k-mer graph) approach is more appropriate for the large volumes of reads associated with short-read sequencing:

- avoids the computationally expensive all-against-all pairwise read comparisons
- avoids loading all the replicate sequences associated with high-coverage sequencing

Miller JR, Koren S, Sutton G. Assembly algorithms for next-generation sequencing data. Genomics. 2010

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# De Bruijn graph

## The DBG:

- directed graph
- edges are unique k-mers
- nodes are overlaps of length k-1
- an edge connects two nodes if the suffix of the source node shares an exact match of length k-2 with the prefix of the destination node
- the assembly algorithm becomes finding a path in the graph that visits every edge at least once





## Overview of the assembly strategy

#### a Generate all substrings of length k from the reads

ACAGC TCCTG GTCTC	AGCGC CTCTT GGTCG	]
CACAG TTCCT GGTCT	CAGCG CCTCT TGGTC	
CCACA CTTCC TGGTC TGTTG	TCAGC TCCTC TTGGT	
CCCAC GCTTC CTGGT TTGTT	CTCAG TTCCT GTTGG	– k-mers (k=5)
GCCCA CGCTT GCTGG CTTGT	CCTCA CTTCC TGTTG	
CGCCC GCGCT TGCTG TCTTG	CCCTC GCTTC TTGTT CGTAG	
CCGCC AGCGC CTGCT CTCTT	GCCCT CGCTT CTTGT TCGTA	
ACCGC CAGCG CCTGC TCTCT	CGCCC GCGCT TCTTG GTCGT	
ACCGCCCACAGCGCTTCCTGCTGGTCTCTTGTTG	CGCCCTCAGCGCTTCCTCTTGTTGGTCGTAG	– Reads

Martin JA, Wang Z. Next-generation transcriptome assembly. Nat Rev Genet. 2011

a Generate all substrings of length k from the reads



**b** Generate the De Bruijn graph



Martin JA, Wang Z. Next-generation transcriptome assembly. Nat Rev Genet. 2011





## Overview of the assembly strategy

**b** Generate the De Bruijn graph Sequencing error or SNP GCCCA CCCAC CCACA - CACAG ACAGC GCGCT CGCCC CAGCG) > (AGCGC) > GCTTC ACCGC >> GCCCT CCCTC CCTCA ) CTCAG TCAGC Deletion or intron GCTGG CTGGT GGTCT GTCTC) TCTCT CCTGC CTGCT TGCTG TGGTC TCCTG CTTCC) > (TTCCT CCTCT TCCTC TTGTT GTTGG TTGGT TGGTC GGTCG GTCGT TCGTA CGTAG CTCTT TCTTG CTTGT TGTTG c Collapse the De Bruijn graph TCCTGCTGGTCTCT GCCCACAGC CTCTTGTTGGTCGTAG CAGCGCTTCCT ACCGCCC GCCCTCAGC TCCTC



## Overview of the assembly strategy



Martin JA, Wang Z. Next-generation transcriptome assembly. Nat Rev Genet. 2011



## Overview of the assembly strategy



Martin JA, Wang Z. Next-generation transcriptome assembly. Nat Rev Genet. 2011



# **DBG vs OLC**

Plateforme Bioinformatique Midi-Pyrér

## OLC

⊕ Overlap computation is a very time and ⊗ computationally intensive step

More efficient ways to find Eulerian paths than 
 Mamiltonian paths

 $\otimes$  Overlaps can vary in length  $\oplus$ 

⊗ Very sensitive to repeats ⊕

⊗ Very sensitive to sequencing errors ⊕

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# **Velvet / Oases**

Velvet: sequence assembler for very short reads (2008)



Velvet: Algorithms for de novo short read assembly using de Bruijn graphs

Daniel R. Zerbino and Ewan Birney

Genome Res. 2008 18: 821-829 originally published online March 18, 2008 Access the most recent version at doi:10.1101/gr.074492.107

## Oases: de novo transcriptome assembler (2012)

BIOINFORMATICS ORIGIN

ORIGINAL PAPER

Vol. 28 no. 8 2012, pages 1086–1092 doi:10.1093/bioinformatics/bts094

Sequence analysis

Advance Access publication February 24, 2012

Oases: robust de novo RNA-seq assembly across the dynamic range of expression levels

Marcel H. Schulz<sup>1,2,3,†</sup>, Daniel R. Zerbino<sup>3,4,\*,†</sup>, Martin Vingron<sup>1</sup> and Ewan Birney<sup>3</sup>

<sup>1</sup>Department of Computational Molecular Biology, Max Planck Institute for Molecular Genetics, Ihnestraße 63-73, D-14195 Berlin, Germany, <sup>2</sup>Lane Center for Computational Biology, Carnegie Mellon University, Pittsburgh, PA 15213, USA, <sup>3</sup>European Bioinformatics Institute, Wellcome Trust Genome Campus, CBS 10 SD, Hinxton, Cambridgeshire, UK and <sup>4</sup>Center for Biomolecular Science and Engineering, University of California Santa Cruz, Santa Cruz, CA 95064, USA

Associate Editor: No Hofacker



# **Velvet - Graph structure**

#### Graph structure:

- each node represents a series of overlapping k-mers
- sequence of the final nucleotides is called the sequence of the node
- each node is attached to a twin node (rev. comp k-mers → k must be odd)
- nodes are connected by a directed arc. The last k-mer of an arc's source node shares an overlap of length k-1 with the first of the destination node
- reads are mapped as "paths" traversing the graph



Zerbino DR, Birney E. Velvet: algorithms for de novo short read assembly using de Bruijn graphs. Genome Res. 2008





# **Velvet - Graph construction**

## Graph construction:

- reads are first hashed according to predefined k-mer length
- build an hash table storing for each k-mer, the ID of the first read encountered containing this k-mer and the position
- build another table storing for each read which of its original k-mers are overlapped by subsequent reads
- ordered set of original k-mers of a read is cut each time an overlap with another read begins or ends
- for each uninterrupted sequence of original k-mers, a node is created

Zerbino DR, Birney E. Velvet: algorithms for de novo short read assembly using de Bruijn graphs. Genome Res. 2008





# Velvet – Correcting graph

## Next steps

- graph simplification → chains of blocks or linear connected subgraph are merged (no loss of information)
- error removal  $\rightarrow$  three major problems:
  - \* "false positive graphs" due to errors in reads
  - \* "gap problems" due to non-uniform or low coverage  $\rightarrow$  short "dead-end" paths ( $\leftarrow$  larger k)
  - \* "branching problems": k-mers connected to multiple kmers due to repeat regions or erroneous reads introduce branches in the graph ( $\leftarrow$  smaller k)



# **Velvet - Removing errors**

## **Removing tips**

Tip is a chain of nodes that is disconnected on one end. Tips to remove are recognized on the base of two criteria: length and minority count.





# **Velvet - Removing errors**

## Removing bubbles (Tour Bus algorithm)

Two paths that start and end at the same nodes and contain similar sequences defined a "bubble". If judged "similar" enough, the paths are merged. The path that reaches the end node first\* is used as the consensus path.



\* according to the notion of distance, that considers the path-length and the related multiplicity  $\rightarrow$  gives priotity to higher coverage paths

http://hpc.isti.cnr.it/~rossano/ReadingDaySlides/SemCap.pdf



# **Velvet - Removing errors**

#### **Removing erroneous connections**

These unwanted connections do not create any recognizable loop or structure. Remove them with a basic coverage cutoff.

Genome assembly process in Velvet is stopped here  $\rightarrow$  fork due to transcriptome assembly singularities

Zerbino DR, Birney E. Velvet: algorithms for de novo short read assembly using de Bruijn graphs. Genome Res. 2008

Schulz MH & al. Oases: robust de novo RNA-seq assembly across the dynamic range of expression levels. Bioinformatics. 2012

#### Overview

- individual reads are sequenced from an RNA sample
- contigs are built from those reads, some of them are labeled as long (clear), others short (dark)
- long contigs, connected by single reads or read-pairs are grouped into connected components called loci
- short contigs are attached to the loci
- the loci are transitively reduced. Transfrags are then extracted from the loci







# **Oases - Contig construction**

Contig construction (Velvet preprocess)

- the Oases pipeline receives as input a preliminary assembly produced by Velvet
- initial stages (hashing and graph construction) can be used indifferently on transcriptomic data
- contig correction performed with a modified version of the Tour Bus algorithm (fitted for coverage disparity and complexity of graphs)

Schulz MH & al. Oases: robust de novo RNA-seq assembly across the dynamic range of expression levels. Bioinformatics. 2012





# **Oases - Contig correction**

### **Contig correction**

- local edge removal  $\rightarrow$  for each node, an outgoing edge is removed if its coverage represents less than 10% of the sum of coverages of all outgoing edges)
- static coverage cutoff  $\rightarrow$  contigs with less than the cutoff are removed from the assembly (3x by default)
- contigs longer than a given threshold (by default > 50+k-1) are labeled as long contig and treated as if unique
- other contigs are labeled as short

Schulz MH & al. Oases: robust de novo RNA-seg assembly across the dynamic range of expression levels. Bioinformatics. 2012



# **Oases - contigs to scaffolds**

#### Scaffold construction

- connexion between contigs can be supported by both spanning single reads (direct) or paired-end reads (indirect)
- the total number of spanning reads and pair-reads confirming a connection between 2 contigs is called its support

Scaffold filtering

- based on static (very low) and dynamic (vs local) coverage thresholds
- connections with low support are removed
- short contigs can only be joined by direct connections

Schulz MH & al. Oases: robust de novo RNA-seq assembly across the dynamic range of expression levels. Bioinformatics. 2012

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# **Oases - Locus resolution**



#### Locus construction

- contigs are organized into clusters called loci
- two steps:
  - \* long contigs are first clustered into connected components
  - \* to each locus are added short nodes connected to one of the long nodes in the cluster

Transitive reduction of the loci

- remove redundant long distance connexions
- example: two contigs which are not consecutive in a locus are frequently connected by a paired-end read

Schulz MH & al. Oases: robust de novo RNA-seq assembly across the dynamic range of expression levels. Bioinformatics. 2012


### **Oases - Transcripts extraction**

#### Extracting transcript assemblies Loci are divided into four categories

- chains
- bubbles
- forks
- complex



Schulz MH & al. Oases: robust de novo RNA-seq assembly across the dynamic range of expression levels. Bioinformatics. 2012



### **Oases - Transcripts extraction**

#### Extracting transcript assemblies

- trivial locus topologies (chains, forks and bubbles) are straightforward to resolve (if isolated from any other branching point)
- complex loci are processed using an additional heuristic method\* which produces a parsimonious set of putative highly expressed transcripts



\* Lee C. Generating consensus sequences from partial order multiple sequence alignment graphs. Bioinformatics. 2003

Schulz MH & al. Oases: robust de novo RNA-seq assembly across the dynamic range of expression levels. Bioinformatics. 2012



## Which k-mer size to choose?

K-mer size is the parameter the most influent upon assembly results.

It's a trade-off between specificity and sensitivity.

Longer k-mers bring you more specificity:

- inherently rarer
- give graphs with lower coverage over all
- the longer the k-mer the more likely it is that it includes an error

 $\clubsuit$  larger k values bias your results towards more abundant isoforms





## Which k-mer size to choose?

Shorter k-mers bring you more sensitivity:

- high coverage graph
- at the cost of a more complex graph
- due to more spurious overlaps

smaller k values are susceptible to assembled lowabundance isoforms





Η	M	k	transfrags	95%
			$(\geq 100 \text{bps})$	aligned
~		19	67319	81
~		25	53504	85
~		29	50936	90
1		35	34012	90
	~	21	57095	86
	1	25	56473	88
	1	29	59503	88
	1	35	61939	90

Schulz MH & al. Oases: robust de novo RNAseq assembly across the dynamic range of expression levels. Bioinformatics. 2012

http://ivory.idyll.org/blog/the-k-parameter.html





### The multiple k-mer strategy



#### Optimization of de novo transcriptome assembly from next-generation sequencing data

Yann Surget-Groba and Juan I. Montoya-Burgos

Genome Res. 2010 20: 1432-1440 originally published online August 6, 2010 Access the most recent version at doi:10.1101/gr.103846.109

#### Overview

- independent assemblies which vary by k-mer length
- assemblies are then merged into a final assembly



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#### **Velvet / Oases: command line**

bash-4.1\$ velveth velveth - simple hashing program Version 1.2.07

Copyright 2007, 2008 Daniel Zerbino (zerbino@ebi.ac.uk) This is free software; see the source for copying conditions. There is NO warranty; not even for MERCHANTABILITY or FITNESS FOR A PARTICULAR PURPOSE.

Compilation settings: CATEGORIES = 2 MAXKMERLENGTH = 31

Usage:

./velveth directory hash\_length {[-file\_format][-read\_type][-separate|-interleaved] filename1 [filename2 ...]} {...} [options]

	directory hash_length filename	<pre>: directory name for output files : EITHER an odd integer (if even, it will be decremented) &lt;= 31 (if above, will be reduced) : OR: m,M,s where m and M are odd integers (if not, they will be decremented) with m &lt; M &lt;= 31 (if above, will be reduced)</pre>
ptions:	-strand_specific -reuse_Sequences -noHash -create_binary	: for strand specific transcriptome sequencing data (default: off) : reuse Sequences file (or link) already in directory (no need to provide original filenames in this case (default: off) : simply prepare Sequences file, do not hash reads or prepare Roadmaps file (default: off) : create binary CnyUnifiedSeq file (default: off)

bash-4.1\$ velvetg velvetg - de Bruijn graph construction, error removal and repeat resolution Version 1.2.07 Compilation settings: CATEGORIES = 2 MAXKMERLENGTH = 31

Usage: ./velvetg directory [options]

directory

: working directory name

Standard options:

iuaru options.	
<pre>-cov_cutoff <floating-point aut< pre=""></floating-point aut<></pre>	> : removal of low coverage nodes AFTER tour bus or allow the system to infer it
(default: no removal)	
-ins_length <integer></integer>	: expected distance between two paired end reads (default: no read pairing)
-read_trkg <yes no></yes no>	: tracking of short read positions in assembly (default: no tracking)
<pre>-min_contig_lgth <integer></integer></pre>	: minimum contig length exported to contigs.fa file (default: hash length * 2)





### **Velvet / Oases: command line**

bash-4.1\$ oases oases - De novo transcriptome assembler for the Velvet package Version 0.2.06

Copyright 2009,2010 Daniel Zerbino (dzerbino@soe.ucsc.edu) This is free software; see the source for copying conditions. There is NO warranty; not even for MERCHANTABILITY or FITNESS FOR A PARTICULAR PURPOSE.

Compilation settings: CATEGORIES = 2MAXKMERLENGTH = 64

Usage: ./oases directory [options]

directory

: working directory name

Standard options:

a operono.	
-ins_length2 <integer> :</integer>	: expected distance between two paired-end reads in the second short-read dataset (default: no read pairing)
<pre>-ins_length_long <integer> :</integer></pre>	: expected distance between two long paired-end reads (default: no read pairing)
-ins_length*_sd <integer> :</integer>	est. standard deviation of respective dataset (default: 10% of corresponding length)
[replace '*' by nothing,	'2' or '_long' as necessary]
-unused_reads <yes no> :</yes no>	: export unused reads in UnusedReads.fa file (default: no)
-amos_file <yes no> :</yes no>	: export assembly to AMOS file (default: no export)
-alignments <yes no> :</yes no>	export a summary of contig alignment to the reference sequences (default: no)
help :	: this help message





### **Velvet / Oases: command line**

mkdir oasesOutDir

velveth\_70\_LONG oasesOutDir 27 -shortPaired -fastq.gz \
-separate R1.fastq.gz R2.fastq.gz -noHash

mkdir oasesOutDir/k61 ln -s ../Sequences oasesOutDir/k61/Sequences velveth\_70\_LONG oasesOutDir/k61 61 -reuse\_Sequences velvetg\_70\_LONG oasesOutDir/k61 -read\_trkg yes \ -min\_contig\_lgth 200 oases\_70\_LONG oasesOutDir/k61

Redo for all chosen k-mers



### **Velvet / Oases: command line**

mkdir oasesOutDir/merge

velveth\_70\_LONG oasesOutDir/merge 27 -long \
oasesOutDir/k\*/transcripts.fa

velvetg\_70\_LONG oasesOutDir/merge -read\_trkg yes \
-conserveLong yes

oases\_70\_LONG oasesOutDir/merge -merge yes



#### **Velvet / Oases: output**

bash-4.1\$ ls -lth --color k69 total 81M -rw-rw-r-- 1 sigenae sigenae 300K 4 déc. 11:31 contig-ordering.txt -rw-rw-r-- 1 sigenae sigenae 301K 4 déc. 11:31 transcripts.fa -rw-rw-r-- 1 sigenae sigenae 9,7M 4 déc. 11:31 LastGraph -rw-rw-r-- 1 sigenae sigenae 42K 4 déc. 11:31 stats.txt -rw-rw-r-- 1 sigenae sigenae 1,3K 4 déc. 11:31 Log -rw-rw-r-- 1 sigenae sigenae 272K 4 déc. 11:31 contigs.fa -rw-rw-r-- 1 sigenae sigenae 9,7M 4 déc. 11:31 Graph2 -rw-rw-r-- 1 sigenae sigenae 584K 4 déc. 11:31 PreGraph -rw-rw-r-- 1 sigenae sigenae 60M 4 déc. 11:31 Roadmaps lrwxrwxrwx 1 sigenae sigenae 12 4 déc. 11:31 Sequences -> ../Sequences

> bash-4.1\$ ../count.sh grep -c '^>' Sequences 1700490 grep -c ^NODE PreGraph 4177 grep -c ^NODE Graph2 719 grep -c '^>' contigs.fa 609 grep -c ^NODE LastGraph 646 grep -c '^>' transcripts.fa 586





## Exercise n°2





# Trinity: a novel method for the efficient and robust de novo reconstruction of transcriptomes from RNA-seq data

NATURE BIOTECHNOLOGY | RESEARCH | ARTICLE

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日本語要約

## Full-length transcriptome assembly from RNA-Seq data without a reference genome

Manfred G Grabherr, Brian J Haas, Moran Yassour, Joshua Z Levin, Dawn A Thompson, Ido Amit, Xian Adiconis, Lin Fan, Raktima Raychowdhury, Qiandong Zeng, Zehua Chen, Evan Mauceli, Nir Hacohen, Andreas Gnirke, Nicholas Rhind, Federica di Palma, Bruce W Birren, Chad Nusbaum, Kerstin Lindblad-Toh, Nir Friedman & Aviv Regev

Affiliations | Contributions | Corresponding authors

Nature Biotechnology 29, 644–652 (2011) | doi:10.1038/nbt.1883 Received 03 December 2010 | Accepted 28 April 2011 | Published online 15 May 2011 geno toulΣ bioinfo

#### Pipeline:

- Inchworm
- Chrysalis
- Butterfly



#### **Trinity**



## **Trinity - Inchworm**

#### Inchworm

- build a k-mer dictionary from reads set (count occurrences)
- 2.removes likely error-containing k-mer based on occurrence
- 3.selects the most frequent k-mer to seed contig assembly
- 4.extends the seed in each direction by finding highest occurring k-mer with a k-1 overlap
- 5. extends until it cannot be extended and report contig ; removes assembled k-mers
- 6.repeats steps 3-5, starting with the most abundant k-mer until the entire dictionary has been depleted





### **Trinity - Inchworm**

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## **Trinity - Chrysalis**

#### Chrysalis

- recursively groups inchworm contigs into connected components
  - \*perfect overlaps of k-1 bases
  - \* minimal number of reads span the junction across both contigs
- builds a de Bruijn graph per component
   \*nodes are word size of k-1
  - \*edges are word size of k
  - \*weights each edge with the number of (k-1)-mers in the original reads that support it
- assigns each read to the component with which it shares the largest number of k-mers





Expressed isoforms

#### Trinity Chrysalis



Expression



## **Trinity - Butterfly**

**Butterfly** reconstructs distinct transcripts for splice isoforms and paralogous genes

- Graph simplification
  - \*merge consecutive nodes in linear paths
  - \* prune edges that represent minor deviations
    (supported by comparatively few reads)
- Plausible path scoring
  - identifies those paths that are supported by actual reads and read pairs
  - \* a dynamic programming procedure resolves ambiguities and reduce the combinatorial number of paths
  - \*enumerate linear sequences





#### **Trinity - Butterfly**





## **Trinity: command line**

# # # • XI II I I I I # # # Required: # # --seqType <string> :type of reads: ( cfa, cfq, fa, or fq ) --JM <string> :(Jellyfish Memory) number of GB of system memory to use for k-mer counting by jellyfish (eq. 10G) \*include the 'G' char If paired reads: --left <string> :left reads, one or more (separated by space) --right <string> :right reads, one or more (separated by space) Or, if unpaired reads: ±± --single <string> :single reads, one or more (note, if single file contains pairs, can use flag: --run\_as\_paired ) Ħ # # --SS lib type <string> :Strand-specific RNA-Seg read orientation. if paired: RF or FR, if single: F or R. (dUTP method = RF) See web documentation. --output <string> :name of directory for output (will be created if it doesn't already exist) default( "/work/sigenae/temp/RNASeg de novo 2014/trinity out dir" ) # --CPU <int> :number of CPUs to use, default: 2 --min\_contig\_length <int> :minimum assembled contig length to report (def=200)

Trinity.pl --output trinityOutDir --seqType fq --JM 64G --left R1.fastq.gz --right R2.fastq.gz --CPU 4

Mem usage 1G RAM per 1M ~76 base Illumina paired reads

geno toulΣ bioinfo

bash-	4.1	\$ ls	-1th	color
total 913M				
617	4	déc.	14:19	Trinity.timing
3,8M	4	déc.	14:19	Trinity.fasta
16K	4	déc.	14:07	chrysalis
291K	4	déc.	14:02	iworm_bundle_file_listing.txt
Θ	4	déc.	14:01	iworm_scaffolds.txt.finished
458K	4	déc.	14:01	iworm_scaffolds.txt
39M	4	déc.	14:01	scaffolding_entries.sam
Θ	4	déc.	14:00	bowtie.nameSorted.sam.finished
360M	4	déc.	14:00	bowtie.nameSorted.sam
Θ	4	déc.	14:00	bowtie.out.finished
Θ	4	déc.	13:59	target.fa.finished
8,3M	4	déc.	13:59	target.rev.1.ebwt
728K	4	déc.	13:59	target.rev.2.ebwt
8,3M	4	déc.	13:59	target.1.ebwt
728K	4	déc.	13:59	target.2.ebwt
570K	4	déc.	13:59	target.3.ebwt
1,5M	4	déc.	13:59	target.4.ebwt
99	4	déc.	13:59	<pre>target.fa -&gt; inchworm.K25.L25.DS.fa</pre>
Θ	4	déc.	13:59	inchworm.K25.L25.DS.fa.finished
7,5M	4	déc.	13:59	inchworm.K25.L25.DS.fa
9	4	déc.	13:58	inchworm.kmer_count
Θ	4	déc.	13:58	jellyfish.1.finished
317M	4	déc.	13:58	jellyfish.kmers.fa
8	4	déc.	13:54	both.fa.read_count
160M	4	déc.	13:54	both.fa

### **Trinity: output**

bash-4	4.1	\$ ls	-1th	color chrysalis/
total	22	M		
801K	4	déc.	14:19	butterfly_commands.adj.completed
128K	4	déc.	14:19	RawComps.0
256K	4	déc.	14:19	RawComps.1
16K	4	déc.	14:19	RawComps.3
64K	4	déc.	14:19	RawComps.2
16K	4	déc.	14:18	RawComps.5
16K	4	déc.	14:18	RawComps.7
16K	4	déc.	14:18	RawComps.4
16K	4	déc.	14:17	RawComps.9
16K	4	déc.	14:17	RawComps.6
16K	4	déc.	14:17	RawComps.8
Θ	4	déc.	14:07	quantifyGraph_commands.run.finished
1,1M	4	déc.	14:07	quantifyGraph_commands.completed
801K	4	déc.	14:06	butterfly_commands.adj
Θ	4	déc.	14:06	chrysalis.finished
1,1M	4	déc.	14:05	quantifyGraph_commands
559K	4	déc.	14:05	butterfly_commands
259K	4	déc.	14:05	component_file_listing.txt
Θ	4	déc.	14:05	readsToTranscripts.finished
7	4	déc.	14:05	readcounts.out
3,3M	4	déc.	14:02	bundled.fasta
Θ	4	déc.	14:02	GraphFromIwormFasta.finished
14M	4	déc.	14:02	GraphFromIwormFasta.out

bash-4.1\$ ./count.sh grep -c '^>' both.fa 1700490 grep -c '^>' jellyfish.kmers.fa 11380211 grep -c '^>' inchworm.K25.L25.DS.fa 64776 cat chrysalis/component\_file\_listing.txt | wc -l 2479 cat chrysalis/quantifyGraph\_commands | wc -l 2479 cat chrysalis/butterfly\_commands | wc -l 2479 grep -c '^>' Trinity.fasta 4365

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## **Exercise n°3**



## What to do before assembling?

#### The aim is to simplify graphs:

- Cleaning
- Sampling
- Unicity
- Normalization



#### Sampling

A comparison across non-model animals suggests an optimal sequencing depth for *de novo* transcriptome assembly \$\overline\$ 20 (tissue) to 30 (whole-animal) millions reads



• contigs with reliable hit against KOGs

□ contigs which the translated protein was within the expected size range of the conserved gene



### Sampling

#### Misassembly examples







#### Normalization

#### TRINITY\_RNASEQ\_ROOT/util/insilico\_read\_normalization.pl

- Build a catalog of k-mers and compute abundance
- Compute the k-mer abundance profile for each read
   \* median k-mer abundance (C)
  - \* standard deviation for the k-mer coverage
- Retain reads with probability min(1, T/C) [Perl: rand(1) <= T/C]</li>
  - \* captures all reads falling below the targeted cov. level (T)
  - \* down-samples reads occurring at higher coverage than T
- Discard reads with aberrant k-mer abundance profile (std-dev k-mer cov > median k-mer abundance)



#### Normalization

#### Normalization effects (our experience):

- drastically decrease #reads or #pairs (-50 to -90%) ⊕
- significantly decrease #contigs (-10 to 15%) ⊕
- slightly decrease #rebuilt proteins (-3%) ⊗
- null or positive effect on remapping rate (0 to 10%) ⊕



## See you tomorrow!