

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

Plateforme Bio

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 $L < N$

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)
- ψ much more difficult to predict with transcriptome data

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-4.15 assemblathon stats.pl Usage: assemblathon_stats.pl <assembly_scaffolds_file> options:

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

produce a CSV output file of all results $-csv$

produce a CSV output file of NG(X) values (NG1 through to NG99), suitable for graphing -graph

specify how many consecutive N characters should be used to split scaffolds into contigs $-n *int*$ -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' -

Transcript length histogram

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

Transcript length histogram

Transcript length histogram

Zebrafish tissue specific assembled transcriptomes not so different

Length histogram

python /usr/local/bioinfo/Scripts/bin/length_histogram.py -1 -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.qz

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 \mathbb{O}

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**
	- $A G A T$
	- G A T C
	- A T C C
		- T C C G
			- C C G T

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.

 $BIOINFORMA TICS$ **APPLICATIONS NOTE** Vol. 22 no. 13 2006, pages 1658-1659

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 \rightarrow grouped into that cluster
- if not, a new cluster is defined

Cd-hit: command line

bash-4.1\$ cd-hit-est

 $=$ $=$ $=$ $=$ E . 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 sequence identity threshold, default 0.9 $-C$ 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 band_width of alignment, default 20 $-b$ $-M$ 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 guide for choosing it $-n$ -1 length of throw_away_sequences, default 10 -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% $\mathbf{1}$ 16479nt, >k43 Locus 3102 Transcript 1... at -/100.00% $\overline{2}$ 16528nt, >k49 Locus 3163 Transcript 1... at +/100.00% 3 16528nt, >k55 Locus 3311 Transcript 1... at -/100.00% \overline{A} 14091nt, >k61 Locus 3512 Transcript 1... at -/100.00% 5 6 2497nt, >k61 Locus 4997 Transcript 1... at -/100.00% $\overline{7}$ 14091nt, >k65 Locus 3683 Transcript 1... at -/100.00% 2497nt, >k65_Locus_5316_Transcript_1... at -/100.00% 8 $\overline{9}$ 11230nt, >k69 Locus 3750 Transcript 1... at -/100.00% 10 2497nt, >k69 Locus 5474 Transcript 1... at -/100.00% 2927nt, >k69 Locus 5643 Transcript 1... at -/100.00% 11 >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% 15675nt, >k37_Locus_3865_Transcript_3... * $\overline{2}$ 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 vertical and consideration at the OWS20

Genome analysis

Assembling millions of short DNA sequences using SSAKE

René L. Warren*, Granger G. Sutton¹, 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 '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 Editor: Alex Bateman

$BIOINFORMA TICS$ **APPLICATIONS NOTE** Vol. 23 no. 21 2007, pages 2942-2944

Genome analysis

Extending assembly of short DNA sequences to handle error

William R. Jeck^{1,*}, Josephine A. Reinhardt¹, David A. Baltrus¹, Matthew T. Hickenbotham², Vincent Magrini², Elaine R. Mardis², Jeffery L. Dangl^{1,3} and Corbin D. Jones^{1,3}

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Received on July 23, 2007; revised on August 21, 2007; accepted on August 24, 2007 Advance Access publication September 24, 2007 Associate Editor: Alex Baternan

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 bioinfo

Hash table keyed by num occurrences

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

SSAKE / VCAKE

SSAKE / VCAKE

SSAKE / VCAKE

TGTAGCGCTATCGTCAAG GTAGCGCTATCGTCAAG TAGCGCTATCGTCAAG AGCGCTATCGTCAAG GCGCTATCGTCAAG CGCTATCGTCAAG GCTATCGTCAAG CTATCGTCAAG TATCGTCAAG ATCGTCAAG TCGTCAAG CGTCAAG Min length = 8 Each 3′ most k-mer

 ${\bf TCCTATCGTGTA}$

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

- 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

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** 10.19 no. 5 2003, pages 65 1-652

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

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

TGICL

TGICL benefit:

- mgblast (MegaBLAST with new filtering options) \rightarrow 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

 $TGL(1)$

User Contributed Perl Documentation

 $TGL(1)$

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_overhang>]
    [-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 tgiuser, the password is tgipass, the requested database is of course mysql, the database is located on server dbserver.my.domain and the db is tgicldb.

tgicl $-F$ input.fa $-c$ 2 -1 60 $-p$ 96

TGICL: output

tgicl $-F$ all_seq.fa $-c$ 2 -1 60 $-p$ 96

TGICL: ACE format

AS 31 115

CO CL1Contig1 286 2 3 U

GAAGGAGGAGAGCGAAGACCTGAGTGAAGATGAGGAGAAACATCATGTCAAAACTGAAGG CAACCTGACGGTGAAGAAAGAGGAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCA AAGCGTTCAAACAGAAGTGAAGTCAAATACTTGCTCTTTGTGGAAAGACTTTTATAAA GCCATCGTATTTAAAACGACACCAGAGGACTCACACTGGAGCGAAA

B₀

 $\overline{10} \begin{array}{l} 10 \end{array} \begin{$

```
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
```
RD k25 Locus 18748 Transcript 2 154 0 0 AATCAGAGACTGAACATTTATCTTTTCTACAATCAAACACACATAACAATTACAATGTGA TCATGAAGGAGGAGAGTGAAG GATCATITATCTT

QA 82 154 82 154 **DS**

RD k25_Locus_14452_Transcript_9 286 0 0 GAAGGAGGAGAGCGAAGACGTGAGTGAGATGAGGAGAAACATCATGTCAAAACTGAAGG CAACCTGACGGTGAAGAAAGAGGAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCA AAGCGTTCAAACAGAAGTGAAGTCAAATACTTGCTCTTTGTGTGGAAAGACTTTTATAAA GCCATCGTATTTAAAACGACACCAGAGGACTCACACTGGAGCGAAA

QA 1 286 1 286 **DS**

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

TGICL: Align format

 \mathbb{R}^2

******************* CL1Contig13 ******************** k55 Locus 14801 Tran+ CGTTCATTCAGCGTCGAGTCTTCAGAGCTGAGCTCTGCCGCCATCAGTGAAAGCCAAAGA consensus CGTTCATTCAGCGTCGAGTCTTCAGAGCTGAGCTCTGCCGCCATCAGTGAAAGCCAAAGA k55 Locus 14801 Tran+ CAGAGATTATTGATGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAAAATCAAACAG k37 Locus 13099 Tran+ TTATTGATGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAAAATCAAACAG k69 Locus 1514 Trans+ ATTGAAGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAGAATTAAACAG CAGAGATTATTGATGGACAGTGAGAAGATGAGTGATCCAGAACCCTGCAAAATCAAACAG consensus **Contract Contract Contract** k55 Locus 14801 Tran+ GAAGACACTGAAGAACTAATAGATGTGAAGGAGGAGAGTGAAGAACTGAGTGAAGATGAG k37 Locus 13099 Tran+ GAAGACACTGAAGAACTAATAGATGTGAAG k69 Locus 1514 Trans+ GAAGAGACTGAAGAACTAATAGATGTGAAGCAGGAGAGTGAAGAACTGAGTGAAGATGAG GAAGACACTGAAGAACTAATAGATGTGAAGCAGGAGAGTGAAGAACTGAGTGAAGATGAG consensus $\verb|k55 Locus_14801 Tran+ 6AGAACATCAGGICA\\ k69 Locus_1514_Tras+ 6AGAACATCAGGTCAAAGTGAAAGAAAGTCAAGAACTCAATCAGACATTATCTTTT$ $k37$ Locus $-1690\overline{8}$ Tran+ GTCAAAAGTGGAGAAGAAACTCAATCAGAGACTGAACATTTATCTTTT GAGAAACATCAGGTCAAAAGTGAAGAAGAAACTCAATCAGAGACTGAACATTTATCTTTT consensus k69_Locus_1514_Trans+ CTACAATCAAACACACATAACAATTACAATGTGATCATGAAGGAGGAGAGTGAAGAACTG k37 Locus 16908 Tran+ CTACAATCAAACACACATAACAATTACAATGTGATCATGAAGGAGGAGAGTGAAGAACTG k55 Locus 11728 Tran+ GTGAAGGAGGAGAGTGAAGAACTG CTACAATCAAACACACATAACAATTACAATGTGATCATGAAGGAGGAGAGTGAAGAACTG consensus and the control of k69 Locus 1514 Trans+ AATGAAGATGAGGAGAAACATCAGGTCAAAAGTGAAGAAGAAACTCAATCAGAGACTGAA k37_Locus_16908_Tran+ AATGAAGATGAGGAGAAACATCAGGTCAAAAGTGAAGA k55 Locus 11728 Tran+ AGTGAAGATGAGGAGAAACATCAGGTCAAAAGTGGAGAAGAAACTCAATCAGAGACTGAA AATGAAGATGAGGAGAAACATCAGGTCAAAAGTGAAGAAACTCAATCAGAGACTGAA consensus consensus k69_Locus_1514_Trans+ GAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCAAAGCGTTCAAACAGAAGTGAAG k55 Locus 11728 Tran+ GAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCAAAGCGTTCAAACAGAAGTGAAG GAAAGTGAAGAACTGTGTAAAGACAAGAAAGTCTATCAAAGCGTTCAAACAGAAGTGAAG consensus k69_Locus_1514_Trans+ TCAAATACTTGCTCTTTGTGGAAAGACTTTTATAAAGCCATCGTATTTAAAACGACAC k55 Locus 11728 Tran+ TCAAATACTTGCTCTTTGTGGGAAAGACTTTTATAAAGCCATCGTATTTAAAACGACAC

consensus

TCAAATACTTGCTCTTTGTGTGGAAAGACTTTTATAAAGCCATCGTATTTAAAACGACAC

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Other OLC assemblers

Genome analysis

Aggressive assembly of pyrosequencing reads with mates

Jason R. Miller^{1,*}, Arthur L. Delcher², Sergey Koren¹, Eli Venter¹, Brian P. Walenz¹, Anushka Brownley¹, Justin Johnson¹, Kelvin Li¹, Clark Mobarry³ and Granger Sutton¹

ORIGINAL PAPER

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Received on June 20, 2008; revised on October 17, 2008; accepted on October 20, 2008 Advance Access publication October 24, 2008 Associate Editor: Dmitrij Frishman

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 Sentember 2005 doi:10.1038/nature03959

Newbler

ARTICLES

Genome sequencing in microfabricated high-density picolitre reactors

Marcel Margulies^{1*}, Michael Egholm^{1*}, William E. Altman¹, Said Attiva¹, Joel S. Bader¹, Lisa A. Bemben¹, Jan Berka¹, Michael S. Braverman¹, Yi-Ju Chen¹, Zhoutao Chen¹, Scott B. Dewell¹, Lei Du¹, Joseph M. Fierro¹, Xavier V. Gomes¹, Brian C. Godwin¹, Wen He¹, Scott Helgesen¹, Chun He Ho¹, Gerard P. Irzyk¹,

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

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

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 $CAGCG$ \rightarrow $AGCC$) CGCTT GCTTC CCGCC ACCGC **GCCCT** CCCTC CCTCA **CTCAG** TCAGC Deletion or intron **CTGGT** GGTCT **GTCTC** TCTCT **CTGCT GCTGG** TGGTC TCCTG **CCTGC** TGCTG $(\mathtt{T}\mathtt{T}\mathtt{CC}\mathtt{T}$ CTTCC TCCTC CCTCT CTCTT **GTTGG GTCGT** TCGTA CGTAG TCTTG CTTGT **TTGTT** TGTTG TTGGT TGGTC GGTCG c Collapse the De Bruijn graph TCCTGCTGGTCTCT GCCCACAGC CTCTTGTTGGTCGTAG CAGCGCTTCCT ACCGCCC GCCCTCAGO TCCTCT

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-Pyréne

DBG OLC

 \bigoplus Overlap computation is a very time and \bigotimes computationally intensive step

 \oplus More efficient ways to find Eulerian paths than \otimes Hamiltonian paths

 \otimes Overlaps can vary in length \oplus

 \otimes Very sensitive to repeats \oplus

 \otimes Very sensitive to sequencing errors \oplus

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

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

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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 \rightarrow 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

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Velvet – Correcting graph

Next steps

- graph simplification \rightarrow 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)

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

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

Oases

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)

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

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

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

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

 \rightarrow 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

 ψ smaller k values are susceptible to assembled lowabundance isoforms

Merged 19-35

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

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

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Compilation settings: $CATEGORIES = 2$ $MAXKMERLENGTH = 31$

Usage:

./velveth directory hash_length {[-file_format][-read_type][-separate|-interleaved] filename1 [filename2 ...]} {...} [options]

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

Sta

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: $CATEGORIES = 2$ $MAXKMERLENGTH = 64$

Usage: ./oases directory [options]

directory

: working directory name

Standard options:

mkdir oasesOutDir

velveth 70 LONG oasesOutDir 27 -shortPaired -fastq.qz \setminus $-s$ eparate $\overline{R}1$.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 \setminus $-min$ contig lgth 200 oases 70 LONG oasesOutDir/k61 $---$

Redo for all chosen k-mers

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 -1th --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 1rwxrwxrwx 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

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
toul
bioinfo

Pipeline:

- Inchworm
- Chrysalis
- Butterfly

Trinity

Trinity - Inchworm

Inchworm

- 1.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

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 lar](http://www.broadinstitute.org/videos?criteria=RNA-Seq)gest number of k-mers

Trinity Chrysalis

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

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

bioinfo

bash-4.1\$ ls -lth --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 0 4 déc. 14:01 iworm_scaffolds.txt.finished 458K 4 déc. 14:01 iworm scaffolds.txt 39M 4 déc. 14:01 scaffolding entries.sam 0 4 déc. 14:00 bowtie.nameSorted.sam.finished 360M 4 déc. 14:00 bowtie.nameSorted.sam 0 4 déc. 14:00 bowtie.out.finished 0 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 target.fa -> inchworm.K25.L25.DS.fa 0 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 0 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-4.1\$./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 -1 2479 cat chrysalis/quantifyGraph commands | wc -1 2479 cat chrysalis/butterfly commands | wc -1 2479 grep -c '^>' Trinity.fasta 4365

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 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) \leq T/C]
	- ⁕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%) \oplus
- significantly decrease #contigs (-10 to 15%) \oplus
- slightly decrease #rebuilt proteins (-3%) \otimes
- null or positive effect on remapping rate (0 to 10%) \oplus

See you tomorrow!