Details of Whole Genome Sequencing Details

Principals of a genome project for a species

- Identify the reference line
 - Typically the line has been used historically in genetic research
 - Or, a resource is available for the line
 - Clone library
 Autorat lines
 Obsolete. Reason.
 Clones not used now!!!

Or Hi-C reads!!!

- Mutant lines
- Mapping parent
- Isolate high quality DNA
- Apply some sequencing approach that fits the budget
- Collect DNA reads
- Assembly the reads into
 - \circ Contigs
 - Scaffolds

• Order scaffolds using a genetic map into pseudochromosomes

- Annotate the genes
- Use the reference genome in research
 - Discover candidate gene(s) that control a phenotype
 - Develop markers to tract important genes/regions of the genome

REFERENCE
SEQUENCE

Approach to the Actual Genome Sequencing

- Fragment the genomic DNA
- Old school
 - Clone those fragments into a cloning vector
 - o Isolate many clones
 - Sequence each clone
- Today
 - Massively parallel sequencing

How is Whole Genomic Sequencing Different than Gene Sequencing?

- The scale of the effort
 - \circ Example
 - Public draft of human genome
 - Hierarchical sequencing
 - Based on 23 billon bases of data
 - Private project (Celera Genomics) draft of human genome
 - Whole genome shotgun sequencing approach
 - Based on 27.2 billon clones
 - 14.8 billion bases

Result:

• Human Genome = 2.91 billion bases

Changes That Facilitated Genomic Sequencing

- Sequencing
 - Massively parallel sequencing is universally adopted
- Major changes over time
 - Thermostable polymerase enzymes
 - Improves quality of sequencing products
 - Fluorescently labeled nucleotides for the reaction
 - Allows for laser detection
 - Massively parallel sequencers
 - Millions to billions of simultaneous reads

• Robotics (early on a major innovation)

- Key addition to genomic sequencing
- Human hand rarely touches the clone that is being sequenced
- o Robots
 - Pick subclones
 - Distribute clones into reaction plates
 - Create the sequencing reaction
 - Load the plates onto the capillary detection system
- o Result
 - Increased the quality and quantity of the data
 - Decreasing the cost
 - Dropped over 100 fold since 1990
 - Improvements felt in small research lab
- Not as important for massively parallel sequencing

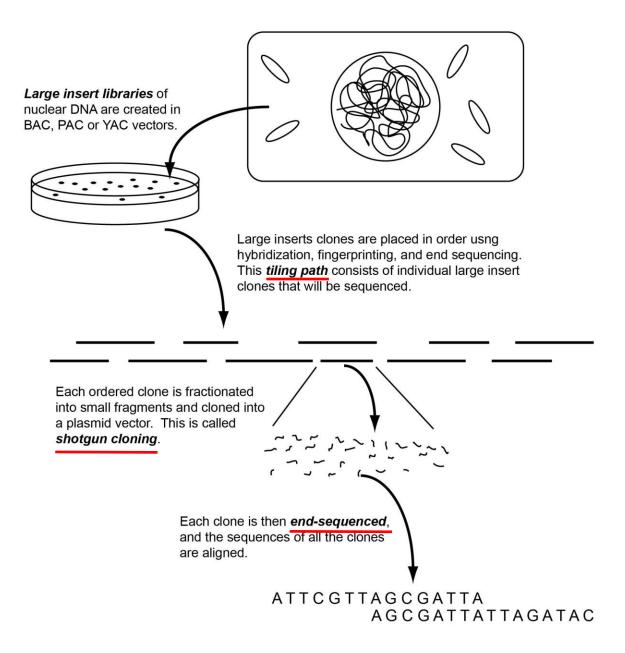
Sequencing Methods

- Two major sequencing approaches
 - Hierarchical shotgun sequencing
 - Whole genome shotgun sequencing
- Hierarchical shotgun sequencing
 - Historically
 - First approach
 - o Why???
 - Techniques for massively parallel sequencing not developed
 - Sophisticated sequence assembly software not availabile
- Concept of the approach
 - Necessary to carefully develop physical map of overlapping clones
 - Clone-based contig (*contig*uous sequence)
 - Assembly of final genomic sequence easier
 - Contig provides fixed sequence reference point
- But
 - Advent of sophisticated software permitted
 - Assembly of a large collection of unordered small, random sequence reads might be possible
 - o Lead to Whole Genome Shotgun approach

Hieracrchical Shotgun Sequencing of Genomes

A. The Concept

Hierarchical shotgun sequencing requires that large insert libraries be constructed. A series of these clones are ordered by several techniques. Once these clones are ordered, each clone is separately fractionated into small fragments and cloned into plasmid vectors. The plasmid clones are sequenced, and the sequence is assembled. This is the procedure used to sequence the *Arabidoposis* genome, and by the public project to sequence the human genome.



Steps Of Hierarchical Shotgun Sequencing

- Requires large insert library
 - Clone types
 - YAC (yeast artificial chromosomes)
 - Megabases of DNA
 - Few (several thousand) overlapping clones necessary for contig assembly
 - But
 - YACs are difficult to manipulate
 - Most research skilled with bacteria but not yeast culture
 - Rarely, if ever, used today
 - BAC or P1 (bacterial artificial chromosomes)
 - Primary advantages
 - Contained reasonable amounts of DNA
 - about 75-150 kb (100,000 200,000) bases
 - Do not undergo rearrangements (like YACs)
 - Could be handled using standard bacterial procedures

Developing The Ordered Array of Clones

• Using a Molecular Map

- DNA markers
- Aligned in the correct order along a chromosome
- *Genetic terminology*
 - Each chromosome is defined as a linkage group
- Map:
 - Is reference point to begin ordering the clones
 - Provides first look at sequence organization of the genome

BAC clones

Tomato High Density Marker Collection

Tomato Hig	h Density	Marker Collection	
Chromosome	# Markers		
1	363		
2	310		Genetic approach Minimum Tiling path \$\$\$\$
3	242		μπητατή τητης ματη φφφφ
4	238	Marker Map	
5	158		
6	202		
7	191		
8	173		
9	184		
10	160		
11	149		
12	136		

Marker Type Early map; few SNPs

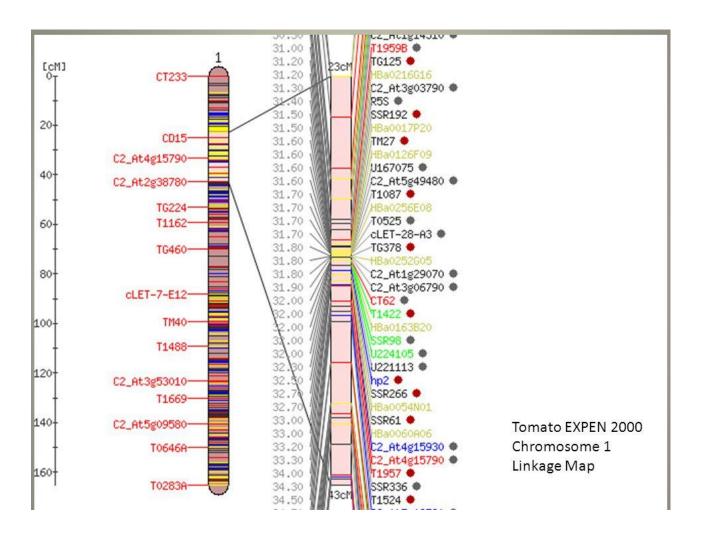
• CAPS (1088), RFLP (1342), SNP (19), SSR (155)

Mapping population

• 88 F2 individuals

Today

- Exclusively SNP markers (n>6,000 SNPs)
- Populations larger (n>200 individuals)



Developing a Minimal Tiling Path

- Definition
 - Fewest clones necessary to obtain complete sequence
- Caution is needed
 - Clones must be authentic
 - Cannot contain chimeric fragments
 - Fragments ligated together from different (noncontiguous) regions of the genome
 - How to avoid chimeras and select the minimum path
 - Careful fingerprinting
- Overlapping the clones

Physical Minimum Tiling Path

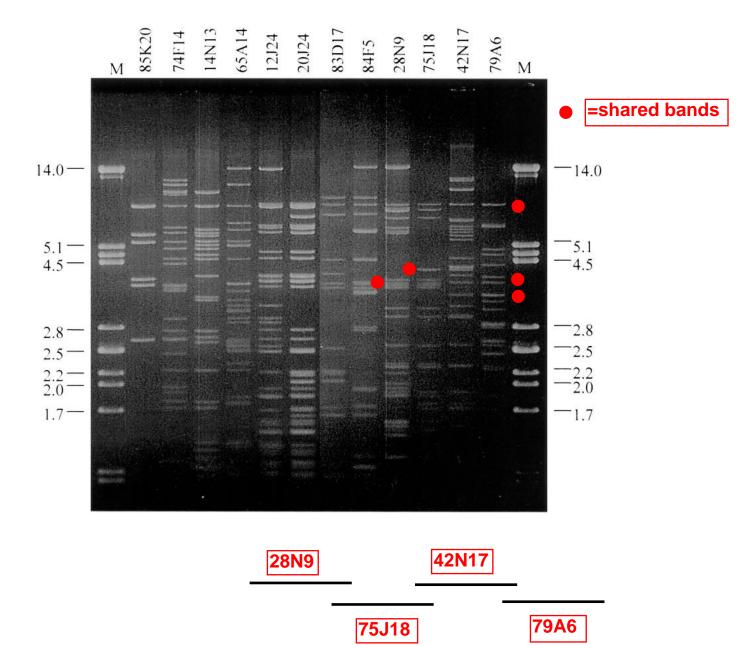
- Maps not dense enough to provide overlap
- Fingerprinting clones
 - Cut each with a restriction enzyme (*HindIII*)
 - Pattern is generally unique for each clone
 - Overlapping clones defined by
 - Partially share fingerprint fragments
- Overlapping define the *physical map* of the genome

BAC clone fingerprinting

- Restriction enzyme digestion
- Digital, imaging, scoring and aligning

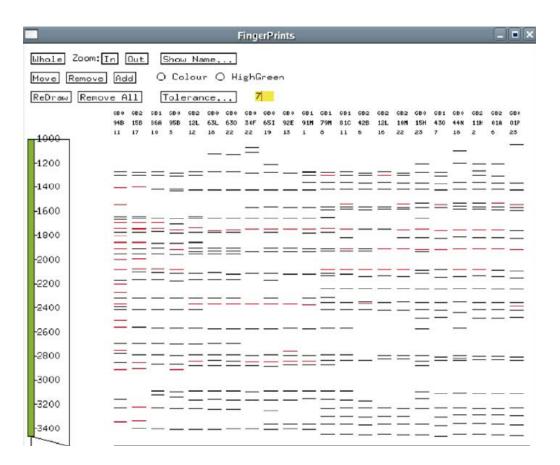
Gel Photograph of digested BAC clones

(https://www.researchgate.net/profile/T_Mirkov/publication/11896126/figure/fig3/AS:394567 875612700@1471083720903/Fig-3-HindIII-fingerprinting-gel-of-the-major-BAC-clones-in-the-12-Mb-contig-that.ppm)



Digital Image of Clones

(https://www.researchgate.net/figure/6535067_fig5_Fig-5-Example-of-the-clone-order-fingerprints-of-a-BAC-contig-of-the-apple-physical)



Genomic Physical Maps

- Human
 - o 29,298 large insert clones sequenced
 - More than necessary
 - Why???
 - Genomic sequencing began before physical map developed
 - Physical map was suboptimal
- Arabidopsis
 - o 1,569 large insert clones defined ten contigs
 - Map completed before the onset of sequencing
 - Smaller genome
 - about 125 megabases
- Yeast

Cosmid clones = 50 kb

- 493 cosmid (smaller insert clones) clones
 - Relatively high number of clones for genome size

Sequencing Clones Of The Minimal Tiling Path

- Steps
 - Physically fractionate clone in small pieces and clone into plasmid
 - Sequence data can be collected from both ends of insert
 - Read pairs or mate pairs
 - Sequence data from both ends of insert DNA
 - Simplifies assembly
 - Sequences are known to reside near each other

Assembly of Hierarchical Shotgun Sequence Data

- Process
 - Data collected
 - Analyzed using computer algorithms
 - $\circ~$ Overlaps in data looked for

Accuracy levels

- Analyzing full shotgun sequence data for a BAC clone
 - Goal: 99.9% accuracy
 - 100 kb BAC clone
 - 2000 sequence reads
 - Equals 8-10x coverage of clone
 - Typical level of accuracy that is sought
 - Primary software used is Phrap
 - Phrap = f(ph)ragment <u>a</u>ssembly program
 - Efficient for a "small" number of clones
 - Small relative to number from a whole genome shotgun approach
 - Each sequence read is assesses for quality by the companion software Phred
 Pfred = q>30;
 - Assembles sequence contigs only from high
 quality reads
- Working draft sequence
 - 93-95% accuracy
 - 3-5 x coverage of 100 kb BAC clone

Viewing the Quality Score Data

Here the Phred scores are overlaid on the chromatogram of a Sanger sequencing output.

- This is just one format the data can be visualized.
- The visualization comes from a quality score data file generated by base-call machine.

From:

http://assets.geneious.com/manual/8.1/GeneiousManualse29.html



Finishing The Sequence

• Gaps need to be filled

- Study more clones
 - Additional subclones sequenced
 - Tedious and expensive
- $\circ~$ Directed sequencing of clones or genomic DNA used
 - Create primers near gaps
 - Amplify BAC clone DNA
 - Sequence and analyze
 - Amplify genomic DNA
 - Sequence and analyze

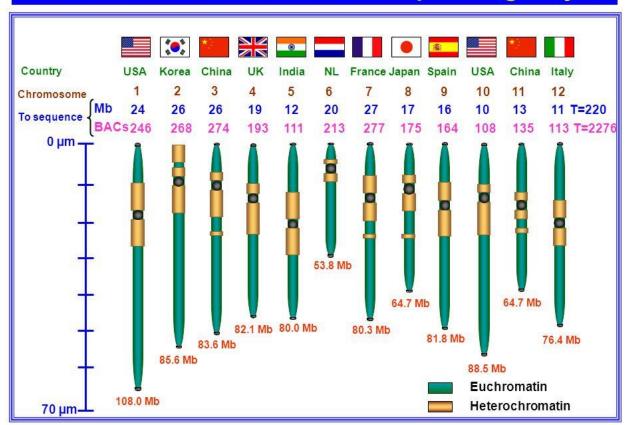
Confirming the Sequence

Genetic maps are the proper order of the genome

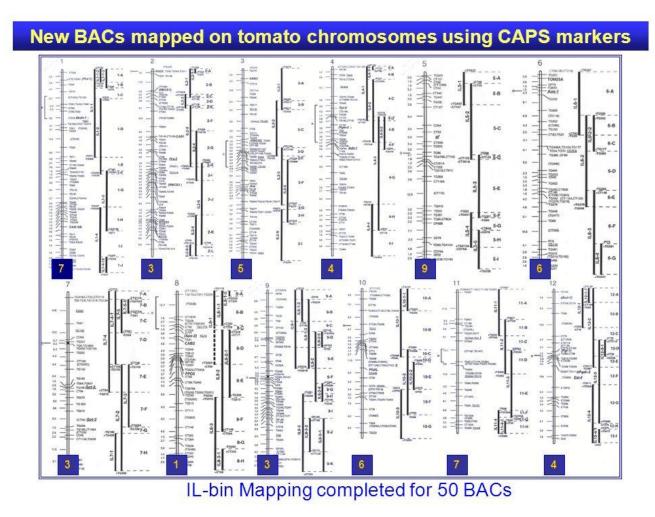
- Molecular map data
 - Molecular markers should be in proper location
- Fingerprint data
 - Fragment sizes should readily recognized in sequence data

Tomato Genome Project

• Distribution of BAC sequencing effort across the genome

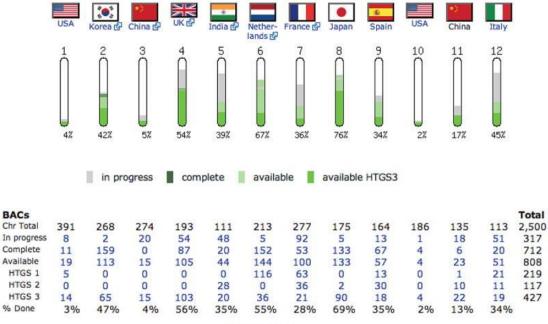


International Tomato Genome Sequencing Project



• BAC selection for the tomato genome project

Visualizing the progress in the Tomato Genome Project



Overall Stats

30% of sequencing is complete

29% of BACs are reported finished

32% of BACs have downloadable sequence

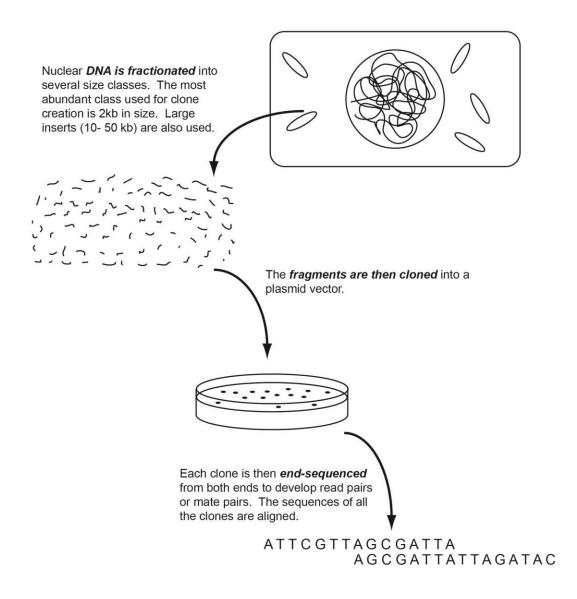
Headlines of Human Genome Sequencing Project

- February 2001
 - Working draft announced
 - Major worldwide news event
- April 2003
 - **o** Finished draft announced
 - Little fanfare
 - Data more useful
- Many more \$\$s will be spent on projects to improve and understand the genome after the reference genome is released to the public

Whole Genome Shotgun Sequencing

A. The Concept

Shotgun sequencing requires that random, small insert libraries are created from the total nuclear DNA of the species of interest. A plasmid cloning vector is used for this step. These clones are then sequenced. This step is analogous to the shotgun cloning and sequencing step used for each large-insert clone used in hierarchial shotgun. The sequences of the clones are then aligned. This is the procedure used to sequence the **Drosophila** genome, and by Celera to sequence the human genome.



Whole Genome Shotgun Sequencing (WGS)

Original WGS approach

- Isolate and shear nuclear DNA
 - Modify DNA by adding restriction site adaptors
- o Clone into plasmids
- Plasmids are then directly sequenced
 - Approach requires read-pairs
 - Especially true because of the repetitive nature of complex genomes
- Modern WGS approach
 - Isolate and shear nuclear DNA
 - Create massively parallel sequencing library
 - Sequencing billions of bases
 - NO plasmid cloning

WGS

- Proven very successful for smaller genomes
 - Essentially the only approach used to sequence smaller genomes like bacteria
- Is WGS useful for large, complex genomes?
 - o Initially considered a bold suggestion
 - Large public effort dedicated to hierarchical approach
 - o Drosophila (2002)
 - Sequenced using the WGS approach

WGS – Major Challenge 1 Assembly of repetitive DNA is difficult

- Difficult to place in the correct genomic position
- o Many types
 - Retrotransposons (RNA mobile elements)
 - DNA transposons
 - Alu repeats (human)
 - Long and Short Interspersed Repeat (LINE and SINE) elements
 - Microsatellites
- Solution
 - Use sequence data from 2, 10 and 50 kb clones
 - Data from fragments containing different types of sequences can be collected
 Recent solution
 - Paired-end reads collected

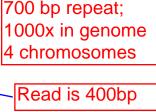
• Assembly Process

- Repeat sequences are initially masked
- Overlaps of non-repeat sequences detected
- Contigs overlapped to create supercontigs
- Software available but wass mostly useful to the developers
 - Examples: Celera Assembler, Arcane, Phusion, Atlas

WGS – Major Challenge 2

- For the two sequences approaches, assembly is a scale issue
 - WGS approact bases
 - Gigabytes of sequence data
 - Hierarchical approach
 - Magnitudes less
 - On-going research focuses on developing new algorithms to handle and assembly the huge data sets generated by WGS

Recent solution **PacBio Reads **Span the full repeat



To which chromosome should the repeat be assigned?

Mouse WGS Data

- 29.7 million reads
- 7.4x coverage
- Newer software
- Assembled without mapping or clone data
 - $\circ~$ Human WGS had access to this data from the public project
 - o 225,000 contigs
 - Mean length (Not Contig L50) = 25 kilobases in length

Super contig subset

- Mean length (Not Scaffold L50) = 16.9 megabases
- **o 200 largest supercontigs**
 - Anchored using mapping data
 - Represents 96% (9187 Mb) of the euchromatic region of genome

Illumina Short-read Sequencing Projects

The Panda Project (genome = 2.4 Gb)

- First genome assembled fully from short reads
- 2010

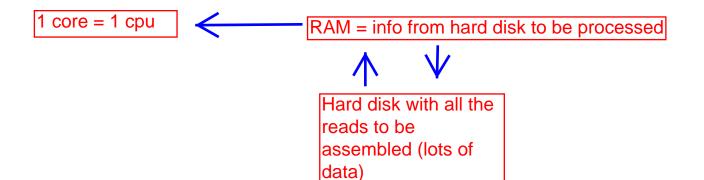
Sequencing

Human genome released in 2001

- 37 paired-end libraries
 - o 150 bp, 500 bp, 2 kb, 5 kb, 10 kb in size
- 176 Gb usable sequence data
 - o 73x coverage

Assembly

- SOAPdenovo used for assembly
 - Part of SOAP software package
 - Short Oligonucleotide Analysis Package
 - "SOAPdenovo uses the de Bruijn graph algorithm and applies a stepwise strategy to make it feasible to assemble the panda genome using a supercomputer (32 cores and 512 Gb random access memory (RAM)."
- Poor library and low-quality reads excluded
 - \circ 134 Gb sequence data used
 - 56x coverage



100-150 bp reads???

Step 1. Contig building

- Data from 500 bp or smaller libraries used first
- Assembly halted when repeat region encountered
 - **39X coverage achieved**
 - **N50 = 1.5 kb**
 - \circ Length = 2.0 Gb

Step 2. Scaffold building

- Paired-end data from all libraries used
 - N50 = 1.3 Mb
 - \circ Total length = 2.3 Gb

Step 3. Closing the gaps

- Local assembly (within a specific gap) using paired end read with one end in a contig and the other in a gap
 - o 223.7 Mb gaps closed
 - \circ 54.2 remained unclosed

Step 4. Compare with other carnivores

• Determined that gaps most likely repetitive elements

Genome Assembly

Goal of assembly

• Create contigs based on similarity of sequence reads

Assembly problem

- Finding the shortest supersting (T) from a set of strings (s₁, s₂,...s_n)
 - Strings = Reads
 - Superstring = Contig

Issues that make assembly difficult

- Sequencing errors
 - Hard to ascertain, so ignored during assembly
- Repetitive sequences
 - Some found 100,000 times
 - Repeats will lead to incorrect assembly
 - Hard to know which sequences overlap
 - Brings two regions together that are not in fact together
 - Resolving some repeats
 - If repeat is shorter than read length, assembly is possible

Features of original assembly algorithms

- Greedy Algorithm Approach
 - Search for the best solution at each step
 - Does not go back and correct previous steps
 - Compute all possible overlaps between strings and assign a quality score
 - Merge strings with highest score
 - Continue until no other strings can be merged

• Fastest method to a solution

- Doesn't guarantee optimum solution
 - Approach doesn't work for large genomes
 - Large RAM memory requirements

Next generation assembly algorithms

• Graph theory approach

- Graph definition
 - A mathematical structure that models pairs of objects from a collection of objects
 - For sequencing, the objects are sequence reads

Overlap-Layout Consensus approach

- o Early algorithm for used small genomes
 - Set a sequence as a node
 - Overlaps are edges that connect nodes
 - Contig is a path of nodes and edges

Process

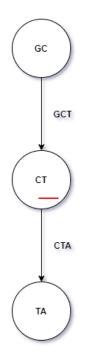
- 1. Find all possible alignments
- 2. Remove overlap duplications
- 3. Construct consensus node/edge combination to create contig

• DeBruijn Graph

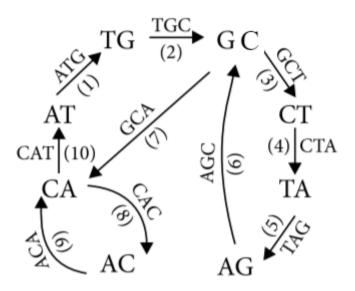
- For perfect sequencing
 - 1. Break reads in k-mer length (often k=24)
 - 2. Form left and right k-1-mers for each k-mer that are **nodes**
 - 3. Draw **directed edge** from left k-1mer to right k-1 mer
 - 4. Add new k-1mer as node if not yet included
 - 5. Add directed edge to new node
 - 6. Repeat until k-mers for all reads are mapped as edges
- Challenges for De Bruijn assemblers
 - 1. Sequencing errors
 - 2. Repetitive sequences

k-mers: GCT, CTA

This is the multigraph for the two k-mers.



Multigraph for the sequence; ATGCTAGCAC



Arachne Whole Genome Assembler

Genome Research 12:177 (2002)

- 1. Breaks 600 nt read into 24 nt sequences and note read origin of the sequence
- 2. Create database with each sequence as main entry
 - Each sequence entry contains frequency and read identifier data
 - Read 1 = ATGCATTATTCGAGATT
 - k-mer = 4 nucleotides

K-MER	Read	Position
ATGC	1	1
TGCA	1	2
GCAT	1	3
CATT	1	4
ATTA	1	5
TTAT	1	6
TATT	1	7
ATTC	1	8
TTCG	1	9
TCGA	1	10
CGAG	1	11
AGAT	1	12
GATT	1	13

• File with k-mer data

k-mer	read	start position	direction
ATGC	1,2,21,24	1,18,30,10	f,f,f,r
TGCA	1,8,31	2,4,51	r,r,f
GCAT	1,2	3,81	f,r
CATT	1,41,51,52,53	4,6,10,15,67	f,f,f,r,r
Etc			

- 4. Discard high copy reads (these are repeats)
- 5. Align reads from low frequency sequences
- 6. Discover mate pairs represented in two plasmids of same length
 - These are paired pairs
- 7. Find a mate pair that matches only one end of the paired pair
 - Sequences are considered to be a single large read
- 8. Process continues until a repeat is encountered
- 9. Assembly stops and a unique contig is declared

Assembly of PacBio reads using hifiasm. From: Cheng et al. 2021. Haplotype-resolved de novo assembly using phased assembly graphs with hifiasm. Nature Methods 18:170-175.

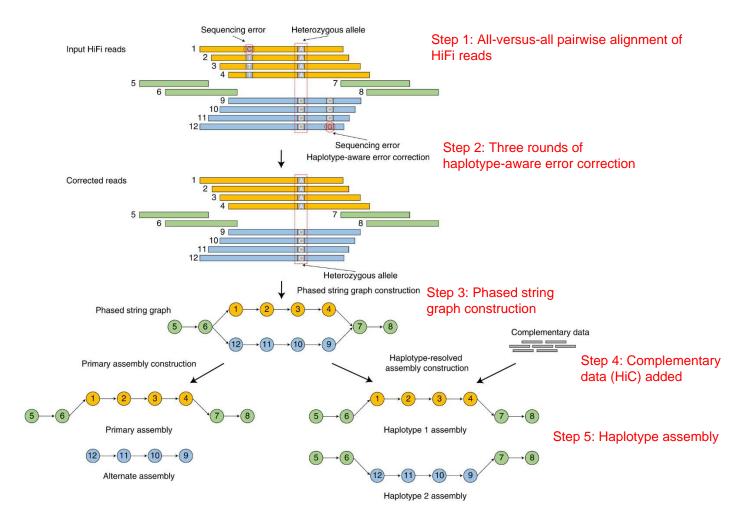


Fig. 1 | Outline of the hifiasm algorithm. Orange and blue bars represent the reads with heterozygous alleles carrying local phasing information, while green bars come from the homozygous regions without any heterozygous alleles. In the phased string graph, a vertex corresponds to the HiFi read with the same ID, and an edge between two vertices indicates that their corresponding reads are overlapped with each other. Hifiasm first performs haplotype-aware error correction to correct sequence errors but keep heterozygous alleles, and then builds a phased assembly graph with local phasing information from the corrected reads. Only the reads coming from the same haplotype are connected in the phased assembly graph. With complementary data providing global phasing information, hifiasm generates a completely phased assembly for each haplotype from the graph. Hifiasm also can generate an unphased primary assembly only with HiFi reads. This unphased primary assembly represents phased blocks (regions) that are resolvable with HiFi reads, but does not preserve phasing information between two phased blocks.

Notes on hifiasm:

**Minimum coverage for hifiasm: >13x per haplotype

**hifiasm also incorporates Hi-C data in the assembly

Common Bean: 454-based Project

Sequencing Libraries

Majority of sequencing data used for assembly

Library	Sequencing Platform	Average Read/Insert Size	Read Number	Assembled Sequence Coverage
Linear	454 XLR & FLX+	362	38,107,155	18.64x
GPNB	454 XLR paired	2,798 ± 1,047	589,346	0.11x
GGAS	454 XLR paired	3,922 ± 643	1,940,576	0.41x
GXSF	454 XLR paired	3,991 ± 337	467,414	0.07x
HYFA	454 XLR paired	4,729 ± 497	1,648,022	0.25x
HYFC	454 XLR paired	4,736 ± 504	1,491,648	0.24x
HYFB	454 XLR paired	4,759 ± 528	1,196,104	0.17x
НХТІ	454 XLR paired	8,022 ± 1,016	1,364,808	0.22x
GXNX	454 XLR paired	9,192 ± 1,058	878,832	0.16x
HXWF	454 XLR paired	11,903 ± 1,928	724,196	0.13x
HXWH	454 XLR paired	12,231 ± 1,902	413,396	0.08x
VUK (Fosmid-end)	Sanger	34,956 ± 4,536	240,384	0.20x
VUL (Fosmid-end)	Sanger	36,001 ± 4,632	88,320	0.08x
PVC (BAC-end)	Sanger	121,960 ± 16,572	81,408	0.08x
PVA (BAC-end)	Sanger	126,959 ± 25,658	89,017	0.09x
PVB (BAC-end)	Sanger	135,292 ± 21,487	92,160	0.09x
Total		N/A	49,412,786	21.02x

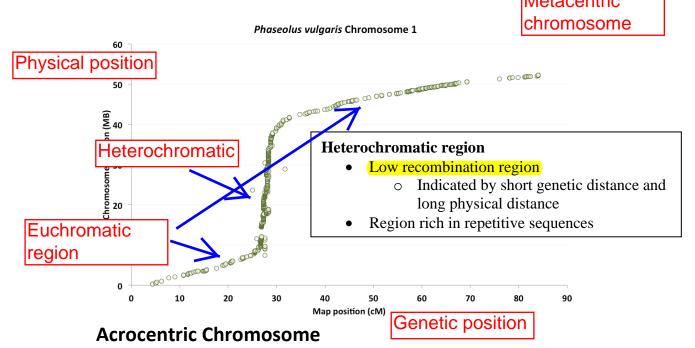
Genome Assembly Statistics

<u>Comparison of Two Legume Species and Sanger and</u> <u>Short Read Sequence Data Collection</u>

Statistic	Soybean	Common Bean
Sequencing method	WGS, Sanger	WGS, 454 & Illumina
Genome size (contig)	955 Mb (1.9% gap)	473 Mb (9.3% gap)
Genome size (scaffold)	973 Mb	521 Mb
Contig number	16,311	41,391
Contig N50	1,492	3.273
Contig L50	189 kb	39.5 kb
Scaffold number	1,168	708
Scaffold N50	10	5
Scaffold L50	47.8 Mb	50.4 Mb
Genetic map loci	1.536 SNPs	7,015 SNPs

Typical Chromosome

- High recombination on ends of chromosome
- Low recombination in the center (heterochromatic) region of the genome
 Metacentric



• Heterochromatic repeat rich region at end of chromosome

