Graphs for reference-free analysis of sequencing data

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Motivation

Reference-free sequencing data analysis: no genome sequence, or choosing not to use it

Some applications:

- 1. Read error correction
- 2. DNA/RNA/metaDNA assembly
- 3. Alternative splicing detection
- 4. DNA variants detection
- 5. Transcript quantification

Presentation will be application-agnostic. We'll talk about the overlap graph, string graph, de Bruijn graph, A-Bruijn graph, superstring graph

This talk

.. will be technical!

Computer Scientists: sit back and relax, it gets challenging in the second part

(as Pierre would say,) *Biologists*: the first part will possibly be helpful to understand other talks

Vocabulary

A graph is:

- a set of nodes, and
- a set of edges (directed or not)



k-mers are strings of length *k*

Graphs for sequencing data

Graphs represent overlaps between sequences in reads.

Two widely used families of graphs for sequencing data:

- de Bruijn graphs generally for Illumina data
- string graphs generally for Sanger/PacBio/Nanopore data

Overlap graphs

First, agree on some definition of "what is an overlap".

- 1. Nodes = reads.
- 2. Edges = overlap between two reads.

In this example, let's say that an overlap needs to be:

exact

- and over at least 3 characters.

ACTGCT

CTGCT (overlap of length 5) GCTAA (overlap of length 3)



String graphs

A **string graph** is obtained from an overlap graph by removing redundancy:

- redundant reads (those fully contained in another read)
- transitively redundant edges (if $a \rightarrow c$ and $a \rightarrow b \rightarrow c$, then remove $a \rightarrow c$)

Example:	Let's have inexact overlaps here
ACTGCT	ACTGCT
CTGCT (overlap length 5)	CTACT
GCTAA (overlap length 3)	GCTAA

String graph / de Bruijn graph (4)

So, which is better?

- String graphs capture whole read information
- de Bruijn graphs are conceptually simpler:
 - single node length
 - single overlap definition

string graphs have been mostly used for long reads and de Bruijn graphs mostly for short reads.

String graphs are also known as the **Overlap Layout Consensus** (OLC) method.

de Bruijn graphs

A **de Bruijn** graph for a fixed integer *k*:

- 1. **Nodes** = all *k*-mers (substrings of length *k*) in the reads.
- 2. There is an **edge** between x and y if the (k 1)-mer prefix of y matches exactly the (k 1)-mer suffix of x.

Example for k = 3 and a single read:

ACTG

ACT 🔶 CTG

de Bruijn graphs

Example for many reads and still k = 3.

ACTG CTGC TGCC



de Bruijn graphs: redundancy

What happens if we add redundancy?

ACTG ACTG CTGC CTGC CTGC TGCC TGCC

dBG, *k* = 3:

ACT \rightarrow CTG \rightarrow TGC \rightarrow GCC

de Bruijn graphs: errors

How is a sequencing error (at the end of a read) impacting the de Bruijn graph?

ACTG

CTGC

CTGA

TGCC

dBG, *k* = 3:



de Bruijn graphs: SNPs

What is the effect of a SNP (or a sequencing error inside a read) on the graph?

AGCCTGA AGCATGA

dBG, *k* = 3:



de Bruijn graphs: repeats

What is the effect of a small repeat on the graph?

ACTG CTGC TGCT GCTG CTGA TGAT

dBG, *k* = 3:



Comparison string graph / de Bruijn graph

On the same example, compare the de Bruijn graph with the string graph:

AGTGCT GTGCTA GCTAA

String graph, overlap threshold of 3:

AGTGCT -----> GTGCTA ----> GCTAA

de Bruijn graph, k = 3:

 $AGT \rightarrow GTG \rightarrow TGC \rightarrow GCT \rightarrow CTA \rightarrow TAA$

Short note on reverse complements

Because sequencing is generally not strand-specific:

We always consider that reads (and k-mers) are equal to their reverse complements.

E.g: AAA = TTT ATG = CAT

Short note on dBG definitions

Node-centric dBG:

what we saw. nodes = k-mers edges = (k - 1)-overlaps

Edge-centric dBG:

nodes = k-mers edges = (k + 1)-mers

Also called dBG^- and dBG^+ in the south of France.

They're related but not strictly equivalent.

Space needed to represent the dBG

With a hash table: $\geq 2k$ bits / node



Memory-efficient dBG data structures:

khmerBloom filter[Pell et al. 11]MiniaBF \ false positives[Chikhi, Rizk 12], [Salikhov et al. 13]SDBGXBW + rank/select[Bowe et al. 12]Why are they so efficient? \rightarrow not all operations are supported

Navigational data structures
Membership
(e.g. hash table)Traverse dBG from known nodes \checkmark \checkmark Query membership of arbitrary nodesx \checkmark Enumerate nodesx \checkmark

NDS has undefined behavior if query node not present.



Software

To just construct the graph:

String graph

Illumina: SGA¹ or SlideSort² PacBio: Minimap + Miniasm³

de Bruijn graph

GATB-Core⁴ (for developers, navigational index), BCALM 2⁵ (stand-alone, graph on disk), ABySS⁶

BCALM 2: 3 GB RAM / 2 wall-clock hours for mammalian genome, outputs unitigs.

¹github.com/jts/sga ²github.com/iskana/SlideSort ³github.com/lh3/miniasm ⁴github.com/GATB/gatb-core ⁵github.com/GATB/bcalm ⁶github.com/bcgsc/abyss

Focus on the Bloom Filter dBG representation (prelude to GATB)



Adapted from slides by G. Collet

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Compacted de Bruijn graph



Each non-branching path becomes a single node (unitig).

- no loss of information
- less space

Can be constructed using BCALM.

Some assemblers use it as intermediate representation.

A-Bruijn graphs

Introduced by [Pevzner 2004]. Generalize de Bruijn graphs. Recently used in ABruijn assembler (PacBio) [Lin 2016].

Principle: **nodes** = an arbitrary set of words V. To construct **edges**, given a set of strings (the reads):

- For each string *r*, find **substrings** that are elements of *V*.
- Add $v_1 \rightarrow v_2$ when v_1 is just before v_2 in r.
- Label edge with $(pos(v_2, r) pos(v_1, r))$.

Observe that $dBG(Reads) = ABruijn(V = \Sigma^k, E = Reads)$

Ex: $ABruijn(V = \{ab, dca, bcd\}, abcdcab)$:



de Bruijn vs A-Bruijn



String=CATCAGATAGGA. For dBG: k = 2, edge-centric.

For A-Bruijn, $V = \{CA, AT, TC, AGA, TA, AGG, AC\}$ (*V* is arbitrary here)

A-Bruijn usages

- Original paper Sanger, handling repeats in EULER [Pevzner 2004]
- SPAdes (early vers)

Illumina DNA/metaDNA assembler [Bankevich,...,Pevzner 2012]

- ABruijn

Pacbio assembler [Lin,...,Pevzner 2016]

Superstring graph

[Cazaux, Sacomoto, Rivals 2016]

Consider a set of reads \mathbf{P} (none is contained in another). Consider all maximal overlaps between these reads, $\mathbf{Ov}(\mathbf{P})$.

Given a set of strings **F** (also none contained), remove all overlaps that are substrings of *F*. Call this set $Ov^*(P, F)$.

E.g.

$$P = \{abcc, ccd, cde\},\$$

 $Ov(P) = \{cc, cd, c\}.$

Now if $F = \{cdef\},\$

$$Ov^*(P,F) = \{cc\}.$$

Superstring graph (2)

Recall, $Ov^*(P, F)$ are maximal overlaps between words in *P*, except substrings in *F*. Typically $F = \Sigma^{k_{min}-1}$.

The Truncated Hierarchical Overlap Graph, $\mathbf{THOG}(\mathbf{P}, \mathbf{F})$ is a graph, where:

- nodes = P and $Ov^*(P, F)$
- edges = whenever a node is the longest suffix or the longest prefix of another node

E.g. $THOG(\{aacbb, bbdaa, aeb, bfa\}, \emptyset)$



Superstring graph (3)

The superstring graph is defined as a subgraph of THOG.

It consists of a set of paths in *THOG* that correspond to a greedy solution of a computational problem (Constrained Shortest Cyclic Cover of Strings).



Applications:

- superstring problems (CompSci)
- multi-k assembly

How an assembler works

[SPAdes, Velvet, ABySS, SOAPdenovo, SGA, Megahit, Minia, .., HGAP, FALCON]

- 1) Maybe correct the reads. (SPAdes, HGAP, SGA, FALCON)
- 2) Construct a graph from the reads.

Assembly graph with variants & errors



3) Likely sequencing errors are removed. (not in FALCON)



3) Known biological events are removed. (not in FALCON)

4) Finally, simple paths (i.e. contigs) are returned.



The choice of k

Choice of k is critical in dBG applications:

- k-mers with sequencing errors are noise
- only non-erroneous k-mers matter
- $k < \log_4(|\text{genome}|)$: nearly complete graph, uninformative
- small k: collapses repeats, more non-erroneous k-mers
- large k: less repeat collapsing, less non-erroneous k-mers (due to error and shortness of reads)

Generally, $k \ge 20$.

(Compare 4^k to the genome size.)

Higher sequencing coverage means larger k values can be used.

Graph formats

- FASTG
- GFA
- GFA2

Η	VN:Z:1.0					
S	11	ACCTT				
S	12	TCAAGG				
S	13	CTT	GATT			
L	11	+	12	-	4M	
L	12	-	13	+	5M	
L	11	+	13	+	ЗM	
Р	14	11+	,12-	,13+	4M,5M	



Conclusion

We saw many graphs

- Overlap graph
- String graph
- De Bruijn graph
- A-Bruijn graph
- Superstring graph

Those slides will be available at http://rayan.chikhi.name

Enjoy the workshop!