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Canu specializes in assembling PacBio or Oxford Nanopore sequences. Canu operates in three phases: correction, trimming and assembly. The correction phase will improve the accuracy of bases in reads. The trimming phase will trim reads to the portion that appears to be high-quality sequence, removing suspicious regions such as remaining SMRTbell adapter. The assembly phase will order the reads into contigs, generate consensus sequences and create graphs of alternate paths.

For eukaryotic genomes, coverage more than 20x is enough to outperform current hybrid methods, however, between 30x and 60x coverage is the recommended minimum. More coverage will let Canu use longer reads for assembly, which will result in better assemblies.

Input sequences can be FASTA or FASTQ format, uncompressed or compressed with gzip (.gz), bzip2 (.bz2) or xz (.xz). Note that zip files (.zip) are not supported.

Canu can resume incomplete assemblies, allowing for recovery from system outages or other abnormal terminations. On each restart of Canu, it will examine the files in the assembly directory to decide what to do next. For example, if all but two overlap tasks have finished, only the two that are missing will be computed. For best results, do not change Canu parameters between restarts.

Canu will auto-detect computational resources and scale itself to fit, using all of the resources available and are reasonable for the size of your assembly. Memory and processors can be explicitly limited with with parameters maxMemory and maxThreads. See section Execution Configuration for more details.

Canu will automatically take full advantage of any LSF/PBS/PBSPro/Torque/Sshum/SGE grid available, even submitting itself for execution. Canu makes heavy use of array jobs and requires job submission from compute nodes, which are sometimes not available or allowed. Canu option useGrid=false will restrict Canu to using only the current machine, while option useGrid=remote will configure Canu for grid execution but not submit jobs to the grid. See section Execution Configuration for more details.

The Canu Tutorial has more background, and the Canu FAQ has a wealth of practical advice.

1.1 Assembling PacBio CLR or Nanopore data

Pacific Biosciences released P6-C4 chemistry reads for Escherichia coli K12. You can download them from their original release, but note that you must have the SMRTpipe software installed to extract the reads as FASTQ. Instead,
use a FASTQ format 25X subset (223MB). Download from the command line with:

```
curl -L -o pacbio.fastq http://gembox.cbcb.umd.edu/mhap/raw/ecoli_p6_25x.filtered.fastq
```

Confirm the MD5SUM matches:

```
9bb4c10c41c5442d630af8b504042334 pacbio.fastq
```

There doesn’t appear to be any “official” Oxford Nanopore sample data, but the Albertsen Lab released a run, also for Escherichia coli K12. Download the R10 data from FigShare

Confirm the MD5SUM matches:

```
e2688fd8b3fba61aabc6b65b787c72a4 ecolk12mg1655_R10_3_guppy_345_HAC.fastq
```

By default, Canu will correct the reads, then trim the reads, then assemble the reads to unitigs. Canu needs to know the approximate genome size (so it can determine coverage in the input reads) and the technology used to generate the reads.

For PacBio:

```
canu \
  -p ecoli -d ecoli-pacbio \ 
genomeSize=4.8m \ 
  -pacbio pacbio.fastq
```

For Nanopore:

```
canu \
  -p ecoli -d ecoli-oxford \ 
genomeSize=4.8m maxInputCoverage=100 \ 
  -nanopore ecolk12mg1655_R10_3_guppy_345_HAC.fastq
```

Output and intermediate files will be in directories ‘ecoli-pacbio’ and ‘ecoli-nanopore’, respectively. Intermediate files are written in directories ‘correction’, ‘trimming’ and ‘unitigging’ for the respective stages. Output files are named using the ‘-p’ prefix, such as ‘ecoli.contigs.fasta’, ‘ecoli.unitigs.gfa’, etc. See section Outputs for more details on outputs (intermediate files aren’t documented).

### 1.2 Assembling PacBio HiFi with HiCanu

HiCanu has support for PacBio HiFi data by compressing homopolymers, correcting isolated errors, and masking systematic errors. We will now assemble and E. coli K12 HiFi dataset sequenced by PacBio available at NCBI SRA (3 GB). When assembling, we use `-pacbio-hifi` to specify the input reads:

```
canu \
  -p asm -d ecoli_hifi \ 
genomeSize=4.8m \ 
  -pacbio-hifi ecoli.fastq
```

Retrieve the E. coli fastq file from NCBI SRA:

```
curl -L -o ecoli.fastq https://sra-pub-src-1.s3.amazonaws.com/SRR10971019/m54316_\ 
  180808_005743.fastq.1
```
1.3 Trio Binning Assembly

Canu has support for using parental short-read sequencing to classify and bin the F1 reads (see Trio Binning manuscript for details). This example demonstrates the functionality using a synthetic mix of two Escherichia coli datasets. First download the data

```bash
curl -L -o K12.parental.fasta https://gembox.cbcb.umd.edu/triobinning/example/k12.12.fasta
curl -L -o O157.parental.fasta https://gembox.cbcb.umd.edu/triobinning/example/o157.12.fasta
curl -L -o F1.fasta https://gembox.cbcb.umd.edu/triobinning/example/pacbio.fasta
```

Confirm the MD5SUM matches:

```
69920456a2ef25fc3e89cdcb604861ed K12.parental.fasta
792d0af43740b3534516e8f73ead8a35 O157.parental.fasta
64c8befea83d043344bdff7c43b04a71 F1.fasta
```

and run Canu:

```bash
canu \
-p asm -d ecoliTrio \
genomeSize=5m \
-haplotypeK12 K12.parental.fasta \
-haplotypeO157 O157.parental.fasta \
-pacbio F1.fasta
```

The run will first bin the reads into the haplotypes (ecoliTrio/haplotype/haplotype-*.fasta.gz) and provide a summary of the classification in ecoliTrio/haplotype/haplotype.log:

```
-- 119848 reads 378658103 bases written to haplotype file ./haplotype-K12.fasta.gz.
-- 308353 reads 1042955878 bases written to haplotype file ./haplotype-O157.fasta.gz.
-- 4114 reads 6520294 bases written to haplotype file ./haplotype-unknown.fasta.gz.
```

Next, the haplotypes are assembled in ecoliTrio/asm-haplotypeK12/asm-haplotypeK12.contigs.fasta and ecoliTrio/asm-haplotypeO157/asm-haplotypeO157.contigs.fasta. By default, if the unassigned bases are > 5% of the total, they are included in both haplotypes. This can be controlled with the `hapUnknownFraction` option.

As comparison, you can try co-assembling the datasets instead:

```bash
canu \
-p asm -d ecoliHap \
genomeSize=5m \
corOutCoverage=200 "batOptions=-dg 3 -db 3 -dr 1 -ca 500 -cp 50" \
-pacbio F1.fasta
```

and compare the continuity/accuracy.

Please note, trio binning is designed to work with raw sequences prior to correction. Do not correct the reads together and then run trio-binning, this will not work and Canu will give an error.
Trio binning does not yet support inputting PacBio HiFi reads for binning as they get flagged as “corrected” and the same error as above is given. As a workaround, run canu -haplotype specifying the HiFi reads as -pacbio-raw. This will bin the data and create shell scripts to start the assembly. Edit the shell scripts to replace -pacbio-raw with -pacbio-corrected or -pacbio-hifi and run the assemblies manually.

### 1.4 Assembling With Multiple Technologies and Multiple Files

Canu can use reads from any number of input files, which can be a mix of formats and technologies. Note that current combining PacBio HiFi data with other datatypes it not supported. We’ll assemble a mix of 10X PacBio CLR reads in two FASTQ files and 10X of Nanopore reads in one FASTA file:

```bash
curl -L -o mix.tar.gz http://gembox.cbcb.umd.edu/mhap/raw/ecoliP6Oxford.tar.gz
tar xvzf mix.tar.gz

canu \
  -p ecoli -d ecoli-mix \
  genomeSize=4.8m \
  -pacbio pacbio.part?.fastq.gz \
  -nanopore oxford.fasta.gz
```

#### 1.4.1 Correct, Trim and Assemble, Manually

Sometimes, however, it makes sense to do the three top-level tasks by hand. This would allow trying multiple unitig construction parameters on the same set of corrected and trimmed reads, or skipping trimming and assembly if you only want corrected reads.

We’ll use the PacBio reads from above. First, correct the raw reads:

```bash
canu -correct \
  -p ecoli -d ecoli \n  genomeSize=4.8m \n  -pacbio pacbio.fastq
```

Then, trim the output of the correction:

```bash
canu -trim \
  -p ecoli -d ecoli \n  genomeSize=4.8m \n  -corrected -pacbio ecoli/ecoli.correctedReads.fasta.gz
```

And finally, assemble the output of trimming, twice, with different stringency on which overlaps to use (see `correctedErrorRate`):

```bash
canu \
  -p ecoli -d ecoli-erate-0.039 \n  genomeSize=4.8m \n  correctedErrorRate=0.039 \n  -trimmed -corrected -pacbio ecoli/ecoli.trimmedReads.fasta.gz

canu \
  -p ecoli -d ecoli-erate-0.075 \n  genomeSize=4.8m \n  correctedErrorRate=0.075 \n  -trimmed -corrected -pacbio ecoli/ecoli.trimmedReads.fasta.gz
```
Note that the assembly stages use different `-d` directories. It is not possible to run multiple copies of canu with the same work directory.

You can also try uncorrected ONT assembly which works for higher quality (95% accuracy) data, though this mode should be considered experimental:

```
canu \
-p ecoli -d ecoli-oxford-uncorrected \\n genomeSize=4.8m \\
 -untrimmed correctedErrorRate=0.12 maxInputCoverage=100 'batOptions=-eg 0.10 -sb 0.01 -dg 2 -db 1 -dr 3' -pacbio-hifi ecolk12mg1655_R10_3_guppy_345_HAC.fastq
```

### 1.5 Assembling Low Coverage Datasets

We claimed Canu works down to 20X coverage, and we will now assemble a 20X subset of *S. cerevisae* (215 MB). When assembling, we adjust `correctedErrorRate` to accommodate the slightly lower quality corrected reads:

```
curl -L -o yeast.20x.fastq.gz http://gembox.cbcb.umd.edu/mhap/raw/yeast_filtered.20x.fastq.gz

canu \
-p asm -d yeast \\n genomeSize=12.1m \\
 correctedErrorRate=0.105 \\
 -pacbio yeast.20x.fastq.gz
```

### 1.6 Consensus Accuracy

HiCanu consensus sequences using PacBio HiFi data are typically well above 99.99% We discourage any post-processing/polishing of these assemblies as mis-mapping within repeats can introduce errors.

Canu consensus sequences are typically well above 99% identity for PacBio datasets. Nanopore accuracy varies depending on pore and basecaller version, but is typically above 99% for recent data. Accuracy can be improved by polishing the contigs with tools developed specifically for that task. We recommend Arrow for PacBio and Nanopolish or Medaka for Oxford Nanopore data. When Illumina reads are available, FreeBayes can be used to polish either PacBio or Oxford Nanopore assemblies.
Canu FAQ

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• What parameters should I use for my reads?
• Can I assemble RNA sequence data?
• Can I assemble amplicon sequence data?
• My assembly is running out of space, is too slow?
• My assembly continuity is not good, how can I improve it?
• What parameters can I tweak?
• My asm.contigs.fasta is empty, why?
• Why is my assembly is missing my favorite short plasmid?
• Why do I get less corrected read data than I asked for?
• What is the minimum coverage required to run Canu?
• Can I use Illumina data too?
• My circular element is duplicated/has overlap?
• My genome is AT (or GC) rich, do I need to adjust parameters? What about highly repetitive genomes?
2.1 What resources does Canu require for a bacterial genome assembly? A mammalian assembly?

Canu will detect available resources and configure itself to run efficiently using those resources. It will request resources, for example, the number of compute threads to use, based on the genome size being assembled. It will fail to even start if it feels there are insufficient resources available.

A typical bacterial genome can be assembled with 8GB memory in a few CPU hours - around an hour on 8 cores. It is possible, but not allowed by default, to run with only 4GB memory.

A well-behaved large genome, such as human or other mammals, can be assembled in 10,000 to 25,000 CPU hours, depending on coverage from raw data. A well-behaved large genome, such as human or other mammals with HiFi data can be assembled in 1,000 to 2,000 CPU hours. A grid environment is strongly recommended, with at least 16GB available on each compute node, and one node with at least 64GB memory. You should plan on having 3TB free disk space (200 GB for HiFi), much more for highly repetitive genomes.

Our compute nodes have 48 compute threads and 128GB memory, with a few larger nodes with up to 1TB memory. We develop and test (mostly bacteria, yeast and drosophila) on laptops and desktops with 4 to 12 compute threads and 16GB to 64GB memory.

2.2 How do I run Canu on my SLURM / SGE / PBS / LSF / Torque system?

Canu will detect and configure itself to use on most grids. You do not need to submit Canu to the grid in this case. Canu will query the system for grid support, configure itself for the machines available in the grid, then submit itself to the grid for execution. You can then monitor the run using your grid.

Canu will NOT request explicit time limits or queues/partitions. You can supply your own grid options, such as a partition on SLURM, an account code on SGE, and/or time limits with gridOptions="<your options list>" which will passed to every job submitted by Canu. Similar options exist for every stage of Canu, which could be used to, for example, restrict overlapping to a specific partition or queue.

To disable grid support and run only on the local machine, specify useGrid=false

It is possible to limit the number of grid jobs running at the same time, but this isn’t directly supported by Canu. The various gridOptions parameters can pass grid-specific parameters to the submit commands used; see Issue #756 for Slurm and SGE examples.

2.3 My run stopped with the error 'Mhap precompute jobs failed'

Several package managers make a mess of the installation causing this error (conda and ubuntu in particular). Package managers don’t add much benefit to a tool like Canu which is distributed as pre-compiled
binaries compatible with most systems so our recommended installation method is downloading a binary release. Try running the assembly from scratch using our release distribution and if you continue to encounter errors, submit an issue.

### 2.4 My run stopped with the error 'Failed to submit batch jobs'

The grid you run on must allow compute nodes to submit jobs. This means that if you are on a compute host, `qsub/bsub/sbatch/etc` must be available and working. You can test this by starting an interactive compute session and running the submit command manually (e.g. `qsub` on SGE, `bsub` on LSF, `sbatch` on SLURM).

If this is not the case, Canu WILL NOT work on your grid. You must then set `useGrid=false` and run on a single machine. Alternatively, you can run Canu with `useGrid=remote` which will stop at every submit command, list what should be submitted. You then submit these jobs manually, wait for them to complete, and run the Canu command again. This is a manual process but currently the only workaround for grids without submit support on the compute nodes.

### 2.5 My run of Canu was killed by the sysadmin; the power going out; my cat stepping on the power button; et cetera. Is it safe to restart? How do I restart?

Yes, perfectly safe! It’s actually how Canu runs normally: each time Canu starts, it examines the state of the assembly to decide what it should do next. For example, if six overlap tasks have no results, it’ll run just those six tasks.

This also means that if you want to redo some step, just remove those results from the assembly directory. Some care needs to be taken to make sure results computed after those are also removed.

Short answer: just rerun the _exact_ same command as before. It’ll do the right thing.

### 2.6 My genome size and assembly size are different, help!

The difference could be due to a heterozygous genome where the assembly separated some loci. It could also be because the previous estimate is incorrect. We typically use two analyses to see what happened. First, a BUSCO analysis will indicate duplicated genes. For example this assembly:

- INFO C:98.5%[S:97.9%,D:0.6%],F:1.0%,M:0.5%,n:2799
- INFO 2756 Complete BUSCOs (C)
- INFO 2740 Complete and single-copy BUSCOs (S)
- INFO 16 Complete and duplicated BUSCOs (D)

does not have much duplication but this assembly:

- INFO C:97.6%[S:15.8%,D:81.8%],F:0.9%,M:1.5%,n:2799
- INFO 2732 Complete BUSCOs (C)
- INFO 443 Complete and single-copy BUSCOs (S)
- INFO 2289 Complete and duplicated BUSCOs (D)
does. We have had success using `purge_dups` to remove duplication. Purge dups will also generate a coverage histogram which will usually have two peaks when assemblies have separated some loci, make sure to inspect it to make sure the cutoffs selected are reasonable.

### 2.7 What parameters should I use for my reads?

Canu is designed to be universal on a large range of PacBio CLR, PacBio HiFi, Oxford Nanopore (R6 through R10) data. Assembly quality and/or efficiency can be enhanced for specific datatypes:

**Nanopore R7 1D and Low Identity Reads** With R7 1D sequencing data, and generally for any raw reads lower than 80% identity, five to ten rounds of error correction are helpful:

```bash
# Round 1
canu -p r1 -d r1 -correct corOutCoverage=500 corMinCoverage=0
  → corMhapSensitivity=high -nanopore-raw your_reads.fasta

# Round 2
canu -p r2 -d r2 -correct corOutCoverage=500 corMinCoverage=0
  → corMhapSensitivity=high -nanopore-raw r1/r1.correctedReads.fasta.gz

# Round 3
canu -p r3 -d r3 -correct corOutCoverage=500 corMinCoverage=0
  → corMhapSensitivity=high -nanopore-raw r2/r2.correctedReads.fasta.gz

# Round 4
canu -p r4 -d r4 -correct corOutCoverage=500 corMinCoverage=0
  → corMhapSensitivity=high -nanopore-raw r3/r3.correctedReads.fasta.gz

# Round 5
canu -p r5 -d r5 -correct corOutCoverage=500 corMinCoverage=0
  → corMhapSensitivity=high -nanopore-raw r4/r4.correctedReads.fasta.gz
```

Then assemble the output of the last round, allowing up to 30% difference in overlaps:

```bash
canu -p asm -d asm correctedErrorRate=0.3 utgGraphDeviation=50 -nanopore-
  → corrected r5/r5.correctedReads.fasta.gz
```

**Nanopore R7 2D and Nanopore R9 1D** The defaults were designed with these datasets in mind so they should work. Having very high coverage or very long Nanopore reads can slow down the assembly significantly. You can try the `-fast` option which is much faster but may produce less contiguous assemblies on large genomes.

**Nanopore flip-flop R9.4 or R10.3** Based on a human dataset, the flip-flop basecaller reduces both the raw read error rate and the residual error rate remaining after Canu read correction. For this reason you can reduce the error tolerated by Canu. If you have over 30x coverage add the options: `corMhapOptions=--threshold 0.8 --ordered-sketch-size 1000 --ordered-kmer-size 14` correctedErrorRate=0.105. This is primarily a speed optimization so you can use defaults, especially if your genome’s accuracy is not improved by the flip-flop caller.

**PacBio Sequel**

Based on an *A. thaliana* dataset, and a few more recent mammalian genomes, slightly increase the maximum allowed difference from the default of 4.5% to 8.5% with correctedErrorRate=0.085 corMhapSensitivity=normal.

Only add the second parameter (corMhapSensitivity=normal) if you have >50x coverage.

**PacBio Sequel II** The defaults for PacBio should work on this data. You could speed up the assembly by decreasing the error rate from the default, especially if you have high (>50x) coverage via correctedErrorRate=0.035 utgOvlErrorRate=0.065 trimReadsCoverage=2 trimReadsOverlap=500

**PacBio HiFi** The defaults for -pacbio-hifi should work on this data. There is still some variation in data quality between samples. If you have poor continuity, it may be because the data is lower quality than expected. Canu will try to auto-adjust the error thresholds for this (which will be included in the
report). If that still doesn’t give a good assembly, try running the assembly with \texttt{-untrimmed}. You will likely get a genome size larger than you expect, due to separation of alleles. See \textit{My genome size and assembly size are different, help!} for details on how to remove this duplication.

2.8 Can I assemble RNA sequence data?

Canu will likely mis-assemble, or completely fail to assemble, RNA data. It will do a reasonable job at generating corrected reads though. Reads are corrected using (local) best alignments to other reads, and alignments between different isoforms are usually obviously not ‘best’. Just like with DNA sequences, similar isoforms can get ‘mixed’ together. We’ve heard of reasonable success from users, but do not have any parameter suggestions to make.

Note that Canu will silently translate ‘U’ bases to ‘T’ bases on input, but \textbf{NOT} translate the output bases back to ‘U’.

2.9 Can I assemble amplicon sequence data?

In short, yes. Typically these have very high coverage so we recommend randomly downsampling (\texttt{readSamplingCoverage=100}) and turning off filtering of short contigs \texttt{contigFilter="2 0 1.0 0.5 0"}.

2.10 My assembly is running out of space, is too slow?

We don’t have a good way to estimate of disk space used for the assembly. It varies with genome size, repeat content, and sequencing depth. A human genome sequenced with PacBio or Nanopore at 40-50x typically requires 1-2TB of space at the peak. Plants, unfortunately, seem to want a lot of space. 10TB is a reasonable guess. We’ve seen it as bad as 20TB on some very repetitive genomes.

The most common cause of high disk usage is a very repetitive or large genome. There are some parameters you can tweak to both reduce disk space and speed up the run. Try adding the options \texttt{corMhapFilterThreshold=0.0000000002 corMhapOptions="--threshold 0.80 --num-hashes 512 --num-min-matches 3 --ordered-sketch-size 1000 --ordered-kmer-size 14 --min-olap-length 2000 --repeat-idf-scale 50" mhapMemory=60g mhapBlockSize=500 ovlMerDistinct=0.975}. This will suppress repeats more than the default settings and speed up both correction and assembly.

It is also possible to clean up some intermediate outputs before the assembly is complete to save space. If you already have a \texttt{*.ovlStore.BUILDING/1-bucketize.successs} file in your current step (e.g. \texttt{correct}), you can clean up the files under \texttt{1-overlapper/blocks}. You can also remove the \texttt{ovlStore} for the previous step if you have its output (e.g. if you have \texttt{asm.trimmedReads.fasta.gz}, you can remove \texttt{trimming/asm.ovlStore}).

2.11 My assembly continuity is not good, how can I improve it?

The most important determinant for assembly quality is sequence length, followed by the repeat complexity/heterozygosity of your sample. The first thing to check is the amount of corrected bases output by the correction step. This is logged in the stdout of Canu or in canu-scripts/canu.*.out if you are running in a grid environment. For example on a \texttt{haploid H. sapiens} sample:
Canu Documentation, Release 2.2

2.12 What parameters can I tweak?

For all stages:

- `rawErrorRate` is the maximum expected difference in an alignment of two _uncorrected_ reads. It is a meta-parameter that sets other parameters.

- `correctedErrorRate` is the maximum expected difference in an alignment of two _corrected_ reads. It is a meta-parameter that sets other parameters. (If you’re used to the `errorRate` parameter, multiply that by 3 and use it here.)

- `minReadLength` and `minOverlapLength`. The defaults are to discard reads shorter than 1000bp and to not look for overlaps shorter than 500bp. Increasing `minReadLength` can improve run time, and increasing `minOverlapLength` can improve assembly quality by removing false overlaps. However, increasing either too much will quickly degrade assemblies by either omitting valuable reads or missing true overlaps.

For correction:

- `corOutCoverage` controls how much coverage in corrected reads is generated. The default is to target 40X, but, for various reasons, this results in 30X to 35X of reads being generated.

- `corMinCoverage`, loosely, controls the quality of the corrected reads. It is the coverage in evidence reads that is needed before a (portion of a) corrected read is reported. Corrected reads are generated as a consensus of other reads; this is just the minimum coverage needed for the consensus sequence to be reported. The default is based on input read coverage: 0x coverage for less than 30X input coverage, and 4x coverage for more than that.

For assembly:

- `utgOvlErrorRate` is essentially a speed optimization. Overlaps above this error rate are not computed. Setting it too high generally just wastes compute time, while setting it too low will degrade assemblies by missing true overlaps between lower quality reads.

- `utgGraphDeviation` and `utgRepeatDeviation` what quality of overlaps are used in contig construction or in breaking contigs at false repeat joins, respectively. Both are in terms of a deviation from the mean error rate in the longest overlaps.

- `utgRepeatConfusedBP` controls how similar a true overlap (between two reads in the same contig) and a false overlap (between two reads in different contigs) need to be before the contig is
split. When this occurs, it isn’t clear which overlap is ‘true’ - the longer one or the slightly shorter one - and the contig is split to avoid misassemblies.

For polyploid genomes:

Generally, there’s a couple of ways of dealing with the ploidy. This is not applicable to HiFi data as the default parameters already separate haplotypes down to 0.01% divergence and thus it always produces assemblies which avoid collapsing the genome. See My genome size and assembly size are different, help! for details on how to remove this duplication.

1) **Avoid collapsing the genome** so you end up with double (assuming diploid) the genome size as long as your divergence is above about 2% (for PacBio data). Below this divergence, you’d end up collapsing the variations. We’ve used the following parameters for polyploid populations (PacBio data):

```
corOutCoverage=200 "batOptions=-dg 3 -db 3 -dr 1 -ca 500 -cp 50"
```

This will output more corrected reads (than the default 40x). The latter option will be more conservative at picking the error rate to use for the assembly to try to maintain haplotype separation. If it works, you’ll end up with an assembly >= 2x your haploid genome size. Post-processing using gene information or other synteny information is required to remove redundancy from this assembly.

2) **Smash haplotypes together** and then do phasing using another approach (like HapCUT2 or whatshap or others). In that case you want to do the opposite, increase the error rates used for finding overlaps:

```
corOutCoverage=200 correctedErrorRate=0.15
```

When trimming, reads will be trimmed using other reads in the same chromosome (and probably some reads from other chromosomes). When assembling, overlaps well outside the observed error rate distribution are discarded.

We strongly recommend option 1 which will lead to a larger than expected genome size. See My genome size and assembly size are different, help! for details on how to remove this duplication.

For metagenomes:

The basic idea is to use all data for assembly rather than just the longest as default. The parameters we’ve used recently are:

```
maxInputCoverage=10000 corOutCoverage=10000
corMhapSensitivity=high corMinCoverage=0 redMemory=32 oeaMemory=32 batMemory=200
```

For low coverage:

- This is not applicable to HiFi data as the defaults are designed for 20-30x coverage. For less than 30X raw PacBio or Nanopore coverage, increase the allowed difference in overlaps by a few percent (from 4.5% to 8.5% (or more) with `correctedErrorRate=0.105` for PacBio and from 14.4% to 16% (or more) with `correctedErrorRate=0.16` for Nanopore), to adjust for inferior read correction. Canu will automatically reduce `corMinCoverage` to zero to correct as many reads as possible.

For high coverage:

- For HiFi data, Canu automatically downsamples to a random 50x subset so this section is not applicable. For more than 60X raw PacBio or Nanopore coverage, decrease the allowed difference in overlaps (from 4.5% to 4.0% with `correctedErrorRate=0.040` for PacBio, from 14.4% to 12% with `correctedErrorRate=0.12` for Nanopore), so that only the better corrected reads.
are used. This is primarily an optimization for speed and generally does not change assembly continuity.

2.13 My asm.contigs.fasta is empty, why?

Canu creates three assembled sequence output files: `<prefix>.contigs.fasta`, `<prefix>.unitigs.fasta`, and `<prefix>.unassembled.fasta`, where contigs are the primary output, unitigs are the primary output split at alternate paths, and unassembled are the leftover pieces.

The `contigFilter` parameter sets several parameters that control how small or low coverage initial contigs are handled. By default, initial contigs with more than 50% of the length at less than 3X coverage will be classified as ‘unassembled’ and removed from the assembly, that is, `contigFilter="2 0 1.0 0.5 3"`. The filtering can be disabled by changing the last number from ‘3’ to ‘0’ (meaning, filter if 50% of the contig is less than 0X coverage).

2.14 Why is my assembly is missing my favorite short plasmid?

In Canu v1.6 and earlier only the longest 40X of data (based on the specified genome size) is used for correction. Datasets with uneven coverage or small plasmids can fail to generate enough corrected reads to give enough coverage for assembly, resulting in gaps in the genome or even no reads for small plasmids. Set `corOutCoverage=1000` (or any value greater than your total input coverage) to correct all input data.

An alternate approach is to correct all reads (`-correct corOutCoverage=1000`) then assemble 40X of reads picked at random from the `<prefix>.correctedReads.fasta.gz` output.

More recent Canu versions dynamically select poorly represented sequences to avoid missing short plasmids so this should no longer happen.

2.15 Why do I get less corrected read data than I asked for?

Some reads are trimmed during correction due to being chimeric or because there wasn’t enough evidence to generate a quality corrected sequence. Typically, this results in a 25% loss. Setting `corMinCoverage=0` will report all bases, even low those of low quality. Canu will trim these in its ‘trimming’ phase before assembly.

2.16 What is the minimum coverage required to run Canu?

For HiFi data, we recommend 20-25x or higher. For raw data, coverage more than 20X is typically enough to outperform current hybrid methods. Below that, you will likely not assemble the full genome. The following two papers have several examples.

- Koren et al. (2013) Reducing assembly complexity of microbial genomes with single-molecule sequencing
- Koren and Walenz et al. (2017) Canu: scalable and accurate long-read assembly via adaptive k-mer weighting and repeat separation
2.17 Can I use Illumina data too?

No. We’ve seen that using short reads for correction will homogenize repeats and mix up haplotypes. Even though the short reads are very high quality, their length isn’t sufficient for the true alignment to be identified, and so reads from other repeat instances are used for correction, resulting in incorrect corrections.

2.18 My circular element is duplicated/has overlap?

This is expected for any circular elements. They can overlap by up to a read length due to how Canu constructs contigs. Canu provides an alignment string in the GFA output which can be converted to an alignment to identify the trimming points.

An alternative is to run MUMmer to get self-alignments on the contig and use those trim points. For example, assuming the circular element is in `tig00000099.fa`. Run:

```
nucmer -maxmatch -nosimplify tig00000099.fa tig00000099.fa
show-coords -lrcTH out.delta
```

to find the end overlaps in the tig. The output would be something like:

```
1 1895 48502 50400 1895 1899 99.37 50400 50400 3.76 3.
   \rightarrow 77 tig00000001 tig00000001
48502 50400 1 1895 1899 1895 99.37 50400 50400 3.76 3.
   \rightarrow 77 3.76 tig00000001 tig00000001
```

means trim to 1 to 48502. There is also an alternate writeup.

2.19 My genome is AT (or GC) rich, do I need to adjust parameters? What about highly repetitive genomes?

On bacterial genomes, no adjustment of parameters is (usually) needed. See the next question.

On repetitive genomes with a significantly skewed AT/GC ratio, the Jaccard estimate used by MHAP is biased. Setting `corMaxEvidenceErate=0.15` is sufficient to correct for the bias in our testing.

In general, with high coverage repetitive genomes (such as plants) it can be beneficial to set the above parameter anyway, as it will eliminate repetitive matches, speed up the assembly, and sometimes improve unitigs.

2.20 How can I send data to you?

FTP to `ftp://ftp.cbcb.umd.edu/incoming/sergek`. This is a write-only location that only the Canu developers can see.

Here is a quick walk-through using a command-line ftp client (should be available on most Linux and OSX installations). Say we want to transfer a file named `reads.fastq`. First, run `ftp ftp.cbcb.umd.edu`, specify anonymous as the user name and hit return for password (blank). Then `cd incoming/sergek`, `put reads.fastq`, and `quit`.

That’s it, you won’t be able to see the file but we can download it.
Canu assembles reads from PacBio RS II or Oxford Nanopore MinION instruments into uniquely-assemblable contigs, unitigs. Canu owes lots of its design and code to Celera Assembler.

Canu can be run using hardware of nearly any shape or size, anywhere from laptops to computational grids with thousands of nodes. Obviously, larger assemblies will take a long time to compute on laptops, and smaller assemblies can’t take advantage of hundreds of nodes, so what is being assembled plays some part in determining what hardware can be effectively used.

Most algorithms in canu have been multi-threaded (to use all the cores on a single node), parallelized (to use all the nodes in a grid), or both (all the cores on all the nodes).

3.1 Canu, the command

The canu command is the ‘executive’ program that runs all modules of the assembler. It oversees each of the three top-level tasks (correction, trimming, unitig construction), each of which consists of many steps. Canu ensures that input files for each step exist, that each step successfully finished, and that the output for each step exists. It does minor bits of processing, such as reformatting files, but generally just executes other programs.

```bash
canu [-haplotype|-correct|-trim] \ 
    [-s <assembly-specifications-file>] \ 
    -p <assembly-prefix> \ 
    -d <assembly-directory> \ 
    genomeSize=<number>[g|m|k] \ 
    [other-options] \ 
    [-trimmed|--untrimmed|--raw|--corrected] \ 
    [-pacbio|--nanopore|--pacbio-hifi] *fastq
```

The -p option, to set the file name prefix of intermediate and output files, is mandatory. If -d is not supplied, canu will run in the current directory, otherwise, Canu will create the assembly-directory and run in that directory. It is _not_ possible to run two different assemblies in the same directory.

The -s option will import a list of parameters from the supplied specification (‘spec’) file. These parameters will be applied before any from the command line are used, providing a method for setting commonly used parameters, but
overriding them for specific assemblies.

By default, all needed top-level tasks are performed (-pacbio and -nanopore are assumed to be raw and untrimmed while -pacbio-hifi are assumed to be corrected and trimmed). It is possible to run exactly one task by specifying your read characteristics and a step name. These options can be useful if you want to correct reads once and try many different assemblies. We do exactly that in the Canu Quick Start. Additionally, supplying pre-corrected reads with -pacbio -corrected or -nanopore -corrected will run only the trimming and assembling stages. Specifying reads as -corrected -untrimmed will run only the assembly step.

Parameters are key=value pairs that configure the assembler. They set run time parameters (e.g., memory, threads, grid), algorithmic parameters (e.g., error rates, trimming aggressiveness), and enable or disable entire processing steps (e.g., don’t correct errors, don’t search for subreads). They are described later. One parameter is required: the genomeSize (in bases, with common SI prefixes allowed, for example, 4.7m or 2.8g; see genomeSize). Parameters are listed in the Canu Parameter Reference, but the common ones are described in this document.

Reads are supplied to canu by options that describe how the reads were generated, and what level of quality they are, for example, -pacbio-raw indicates the reads were generated on a PacBio RS II instrument, and have had no processing done to them. Each file of reads supplied this way becomes a ‘library’ of reads. The reads should have been (physically) generated all at the same time using the same steps, but perhaps sequenced in multiple batches. In canu, each library has a set of options setting various algorithmic parameters, for example, how aggressively to trim. To explicitly set library parameters, a text ‘gkp’ file describing the library and the input files must be created. Don’t worry too much about this yet, it’s an advanced feature, fully described in Section gkp-files.

The read-files contain sequence data in either FASTA or FASTQ format (or both! A quirk of the implementation allows files that contain both FASTA and FASTQ format reads). The files can be uncompressed, gzip, bzip2 or xz compressed. We’ve found that “gzip -1” provides good compression that is fast to both compress and decompress. For ‘archival’ purposes, we use “xz -9”.

3.2 Canu, the pipeline

The canu pipeline, that is, what it actually computes, comprises of computing overlaps and processing the overlaps to some result. Each of the three tasks (read correction, read trimming and unitig construction) follow the same pattern:

- Load reads into the read database, gkpStore.
- Compute k-mer counts in preparation for the overlap computation.
- Compute overlaps.
- Load overlaps into the overlap database, ovlStore.
- Do something interesting with the reads and overlaps.
  - The read correction task will replace the original noisy read sequences with consensus sequences computed from overlapping reads.
  - The read trimming task will use overlapping reads to decide what regions of each read are high-quality sequence, and what regions should be trimmed. After trimming, the single largest high-quality chunk of sequence is retained.
  - The unitig construction task finds sets of overlaps that are consistent, and uses those to place reads into a multialignment layout. The layout is then used to generate a consensus sequence for the unitig.

3.3 Module Tags

Because each of the three tasks share common algorithms (all compute overlaps, two compute consensus sequences, etc), parameters are differentiated by a short prefix ‘tag’ string. This lets canu have one generic parameter that can
be set to different values for each stage in each task. For example, “corOvlMemory” will set memory usage for overlaps being generated for read correction; “obtOvlMemory” for overlaps generated for Overlap Based Trimming; “utgOvlMemory” for overlaps generated for unitig construction.

The tags are:

<table>
<thead>
<tr>
<th>Tag</th>
<th>Usage</th>
</tr>
</thead>
<tbody>
<tr>
<td>master</td>
<td>the canu script itself, and any components that it runs directly</td>
</tr>
<tr>
<td>cns</td>
<td>unitig consensus generation</td>
</tr>
<tr>
<td>cor</td>
<td>read correction generation</td>
</tr>
<tr>
<td>red</td>
<td>read error detection</td>
</tr>
<tr>
<td>oea</td>
<td>overlap error adjustment</td>
</tr>
<tr>
<td>ovl</td>
<td>the standard overlapper</td>
</tr>
<tr>
<td>corovl</td>
<td>the standard overlapper, as used in the correction phase</td>
</tr>
<tr>
<td>obtovl</td>
<td>the standard overlapper, as used in the trimming phase</td>
</tr>
<tr>
<td>utgovl</td>
<td>the standard overlapper, as used in the assembly phase</td>
</tr>
<tr>
<td>mhap</td>
<td>the mhap overlapper</td>
</tr>
<tr>
<td>cormhap</td>
<td>the mhap overlapper, as used in the correction phase</td>
</tr>
<tr>
<td>obtmhap</td>
<td>the mhap overlapper, as used in the trimming phase</td>
</tr>
<tr>
<td>utgmhap</td>
<td>the mhap overlapper, as used in the assembly phase</td>
</tr>
<tr>
<td>mmap</td>
<td>the minimap overlapper</td>
</tr>
<tr>
<td>cormmap</td>
<td>the minimap overlapper, as used in the correction phase</td>
</tr>
<tr>
<td>obtmmap</td>
<td>the minimap overlapper, as used in the trimming phase</td>
</tr>
<tr>
<td>utgmmap</td>
<td>the minimap overlapper, as used in the assembly phase</td>
</tr>
<tr>
<td>ovb</td>
<td>the bucketizing phase of overlap store building</td>
</tr>
<tr>
<td>ovs</td>
<td>the sort phase of overlap store building</td>
</tr>
</tbody>
</table>

We’ll get to the details eventually.

### 3.4 Execution Configuration

There are two modes that canu runs in: locally, using just one machine, or grid-enabled, using multiple hosts managed by a grid engine. LSF, PBS/Torque, PBSPro, Sun Grid Engine (and derivations), and Slurm are supported, though LSF has had limited testing. Section **Grid Engine Configuration** has a few hints on how to set up a new grid engine.

By default, if a grid is detected the canu pipeline will immediately submit itself to the grid and run entirely under grid control. If no grid is detected, or if option `useGrid=false` is set, canu will run on the local machine.

In both cases, Canu will auto-detect available resources and configure job sizes based on the resources and genome size you’re assembling. Thus, most users should be able to run the command without modifying the defaults. Some advanced options are outlined below. Each stage has the same five configuration options, and tags are used to specialize the option to a specific stage. The options are:

- **useGrid[tag]=boolean**  Run this stage on the grid, usually in parallel.
- **gridOptions[tag]=string** Supply this string to the grid submit command.
- **<tag>Memory=integer** Use this many gigabytes of memory, per process.
**<tag>Threads**  Use this many compute threads per process.

**<tag>Concurrency**  If not on the grid, run this many jobs at the same time.

Global grid options, applied to every job submitted to the grid, can be set with `gridOptions`. This can be used to add accounting information or access credentials.

A name can be associated with this compute using `gridOptionsJobName`. Canu will work just fine with no name set, but if multiple canu assemblies are running at the same time, they will tend to wait for each others jobs to finish. For example, if two assemblies are running, at some point both will have overlap jobs running. Each assembly will be waiting for all jobs named `ovl_asm` to finish. Had the assemblies specified job names, `gridOptionsJobName=apple` and `gridOptionsJobName=orange`, then one would be waiting for jobs named `ovl_asm_apple`, and the other would be waiting for jobs named `ovl_asm_orange`.

### 3.5 Error Rates

Canu expects all error rates to be reported as fraction error, not as percent error. We’re not sure exactly why this is so. Previously, it used a mix of fraction error and percent error (or both!), and was a little confusing. Here’s a handy table you can print out that converts between fraction error and percent error. Not all values are shown (it’d be quite a large table) but we have every confidence you can figure out the missing values:

<table>
<thead>
<tr>
<th>Fraction Error</th>
<th>Percent Error</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.01</td>
<td>1%</td>
</tr>
<tr>
<td>0.02</td>
<td>2%</td>
</tr>
<tr>
<td>0.03</td>
<td>3%</td>
</tr>
<tr>
<td></td>
<td></td>
</tr>
<tr>
<td>0.12</td>
<td>12%</td>
</tr>
</tbody>
</table>

Canu error rates always refer to the percent difference in an alignment of two reads, not the percent error in a single read, and not the amount of variation in your reads. These error rates are used in two different ways: they are used to limit what overlaps are generated, e.g., don’t compute overlaps that have more than 5% difference; and they are used to tell algorithms what overlaps to use, e.g., even though overlaps were computed to 5% difference, don’t trust any above 3% difference.

There are seven error rates. Three error rates control overlap creation (`corOvlErrorRate`, `obtOvlErrorRate` and `utgOvlErrorRate`), and four error rates control algorithms (`corErrorRate`, `obtErrorRate`, `utgErrorRate`, `cnsErrorRate`).

The three error rates for overlap creation apply to the `ovl` overlap algorithm and the reAlign option used to generate alignments from `mhap` or `minimap` overlaps. Since `mhap` is used for generating correction overlaps, the `corOvlErrorRate` parameter is not used by default. Overlaps for trimming and assembling use the `ovl` algorithm, therefore, `obtOvlErrorRate` and `utgOvlErrorRate` are used.

The four algorithm error rates are used to select which overlaps can be used for correcting reads (`corErrorRate`); which overlaps can be used for trimming reads (`obtErrorRate`); which overlaps can be used for assembling reads (`utgErrorRate`). The last error rate, `cnsErrorRate`, tells the consensus algorithm to not trust read alignments above that value.

For convenience, two meta options set the error rates used with uncorrected reads (`rawErrorRate`) or used with corrected reads, (`correctedErrorRate`). The default depends on the type of read being assembled.
In practice, only \textit{correctedErrorRate} is usually changed. The \textit{Canu FAQ} has specific suggestions on when to change this.

Canu v1.4 and earlier used the errorRate parameter, which set the expected rate of error in a single corrected read.

### 3.6 Minimum Lengths

Two minimum sizes are known:

- \textbf{minReadLength} Discard reads shorter than this when loading into the assembler, and when trimming reads.
- \textbf{minOverlapLength} Do not save overlaps shorter than this.

### 3.7 Overlap configuration

The largest compute of the assembler is also the most complicated to configure. As shown in the \textit{`module tags`} section, there are up to eight (!) different overlapper configurations. For each overlapper (\textit{`ovl`} or \textit{`mhap`}) there is a global configuration, and three specializations that apply to each stage in the pipeline (correction, trimming or assembly).

Like with \textit{`grid configuration`}, overlap configuration uses a \textit{`tag`} prefix applied to each option. The tags in this instance are `cor`, `obt` and `utg`.

For example:

- To change the k-mer size for all instances of the ovl overlapper, `merSize=23` would be used.
- To change the k-mer size for just the ovl overlapper used during correction, `corMerSize=16` would be used.
- To change the mhap k-mer size for all instances, `mhapMerSize=18` would be used.
- To change the mhap k-mer size just during correction, `corMhapMerSize=15` would be used.
- To use minimap for overlap computation just during correction, `corOverlapper=minimap` would be used. The minimap2 executable must be symlinked from the Canu binary folder (\textit{`Linux-amd64/bin`} or \textit{`Darwin-amd64/bin`} depending on your system).

### 3.8 Ovl Overlapper Configuration

\textbf{<tag>Overlapper} select the overlap algorithm to use, \textit{`ovl`} or \textit{`mhap`}.

### 3.9 Ovl Overlapper Parameters

- \textbf{<tag>ovlHashBlockLength} how many bases to reads to include in the hash table; directly controls process size
- \textbf{<tag>ovlRefBlockSize} how many reads to compute overlaps for in one process; directly controls process time
- \textbf{<tag>ovlRefBlockLength} same, but use \textit{`bases in reads`} instead of \textit{`number of reads`}
- \textbf{<tag>ovlHashBits} size of the hash table (SHOULD BE REMOVED AND COMPUTED, MAYBE TWO PASS)
<tag>ovlHashLoad</tag> how much to fill the hash table before computing overlaps (SHOULD BE REMOVED)

<tag>ovlMerSize</tag> size of kmer seed; smaller - more sensitive, but slower

The overlapper will not use frequent kmers to seed overlaps. These are computed by the ‘meryl’ program, and can be selected in one of three ways.

Terminology. A k-mer is a contiguous sequence of k bases. The read ‘ACTTA’ has two 4-mers: ACTT and CTTA. To account for reverse-complement sequence, a ‘canonical kmer’ is the lexicographically smaller of the forward and reverse-complemented kmer sequence. Kmer ACTT, with reverse complement AAGT, has a canonical kmer AAGT. Kmer CTTA, reverse-complement TAAG, has canonical kmer CTAA.

A ‘distinct’ kmer is the kmer sequence with no count associated with it. A ‘total’ kmer (for lack of a better term) is the kmer with its count. The sequence TCGTTTTTTTCGTCG has 12 ‘total’ 4-mers and 8 ‘distinct’ kmers.

<table>
<thead>
<tr>
<th>Sequence</th>
<th>Count</th>
<th>Type</th>
</tr>
</thead>
<tbody>
<tr>
<td>TCGTTTTTTTCGTCG</td>
<td>count</td>
<td></td>
</tr>
<tr>
<td>TCGT</td>
<td>2</td>
<td>distinct-1</td>
</tr>
<tr>
<td>CGTT</td>
<td>1</td>
<td>distinct-2</td>
</tr>
<tr>
<td>GTTT</td>
<td>1</td>
<td>distinct-3</td>
</tr>
<tr>
<td>TTTT</td>
<td>4</td>
<td>distinct-4</td>
</tr>
<tr>
<td>TTTT</td>
<td>4</td>
<td>copy of distinct-4</td>
</tr>
<tr>
<td>TTTT</td>
<td>4</td>
<td>copy of distinct-4</td>
</tr>
<tr>
<td>TTTT</td>
<td>4</td>
<td>copy of distinct-4</td>
</tr>
<tr>
<td>TTTC</td>
<td>1</td>
<td>distinct-5</td>
</tr>
<tr>
<td>TTTC</td>
<td>1</td>
<td>distinct-6</td>
</tr>
<tr>
<td>TCGT</td>
<td>2</td>
<td>copy of distinct-1</td>
</tr>
<tr>
<td>CGTC</td>
<td>1</td>
<td>distinct-7</td>
</tr>
<tr>
<td>GTGC</td>
<td>1</td>
<td>distinct-8</td>
</tr>
</tbody>
</table>

<tag>MerThreshold</tag> any kmer with count higher than N is not used

<tag>MerDistinct</tag> pick a threshold so as to seed overlaps using this fraction of all distinct kmers in the input. In the example above, fraction 0.875 of the k-mers (7/8) will be at or below threshold 2.

<tag>MerTotal</tag> pick a threshold so as to seed overlaps using this fraction of all kmers in the input. In the example above, fraction 0.667 of the k-mers (8/12) will be at or below threshold 2.

<tag>FrequentMers</tag> don’t compute frequent kmers, use those listed in this file

3.10 Mhap Overlapper Parameters

<tag>MhapBlockSize</tag> Chunk of reads that can fit into 1GB of memory. Combined with memory to compute the size of chunk the reads are split into.

<tag>MhapMerSize</tag> Use k-mers of this size for detecting overlaps.

<tag>ReAlign</tag> After computing overlaps with mhap, compute a sequence alignment for each overlap. For Canu versions 2.1.1 or earlier we recommend always setting utgReAlign to true. Newer versions default to off for correction and trimming, on for unitigging. Turning it off for unitigging is not recommended and can lead to crashes. Also applicable to Minimap overlaps.

<tag>MhapSensitivity</tag> Either ‘normal’, ‘high’, or ‘fast’.

Mhap also will down-weight frequent kmers (using tf-idf), but it’s selection of frequent is not exposed.
3.11 Minimap Overlapper Parameters

.getTag>MMapBlockSize Chunk of reads that can fit into 1GB of memory. Combined with memory to compute the size of chunk the reads are split into.

.getTag>MMapMerSize Use k-mers of this size for detecting overlaps
Minimap also will ignore high-frequency minimizers, but it’s selection of frequent is not exposed.

3.12 Outputs

As Canu runs, it outputs status messages, execution logs, and some analysis to the console. Most of the analysis is captured in <prefix>.report as well.

LOGGING

<prefix>.report Most of the analysis reported during assembly. This will report the histogram of read lengths, the histogram or k-mers in the raw and corrected reads, the summary of corrected data, summary of overlaps, and the summary of contig lengths.

You can use the k-mer corrected read histograms with tools like GenomeScope to estimate heterozygosity and genome size. In particular, histograms with more than 1 peak likely indicate a heterozygous genome. See the Canu FAQ for some suggested parameters.

The corrected read report gives a summary of the fate of all input reads. The first part:

<table>
<thead>
<tr>
<th></th>
<th>original</th>
<th>original</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>raw reads</td>
<td>raw reads</td>
</tr>
<tr>
<td></td>
<td>w/overlaps</td>
<td>w/o/overlaps</td>
</tr>
<tr>
<td>Number of Reads</td>
<td>250609</td>
<td>477</td>
</tr>
<tr>
<td>Number of Bases</td>
<td>2238902045</td>
<td>1896925</td>
</tr>
<tr>
<td>Coverage</td>
<td>97.344</td>
<td>0.082</td>
</tr>
<tr>
<td>Median</td>
<td>6534</td>
<td>2360</td>
</tr>
<tr>
<td>Mean</td>
<td>8933</td>
<td>3976</td>
</tr>
<tr>
<td>N50</td>
<td>11291</td>
<td>5756</td>
</tr>
<tr>
<td>Minimum</td>
<td>1012</td>
<td>0</td>
</tr>
<tr>
<td>Maximum</td>
<td>60664</td>
<td>41278</td>
</tr>
</tbody>
</table>

reports the fraction of reads which had an overlap. In this case, the majority had at least one overlap, which is good. Next:

<table>
<thead>
<tr>
<th></th>
<th>evidence</th>
<th>expected</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>reads</td>
<td>raw</td>
</tr>
<tr>
<td>Number of Reads</td>
<td>229397</td>
<td>48006</td>
</tr>
<tr>
<td>Number of Bases</td>
<td>2134291652</td>
<td>993586222</td>
</tr>
<tr>
<td>Coverage</td>
<td>92.795</td>
<td>43.199</td>
</tr>
<tr>
<td>Median</td>
<td>6842</td>
<td>15330</td>
</tr>
<tr>
<td>Mean</td>
<td>9303</td>
<td>20697</td>
</tr>
<tr>
<td>N50</td>
<td>11512</td>
<td>28066</td>
</tr>
<tr>
<td>Minimum</td>
<td>1045</td>
<td>10184</td>
</tr>
<tr>
<td>Maximum</td>
<td>60664</td>
<td>60664</td>
</tr>
</tbody>
</table>
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Reports that a total of 92.8x of raw bases are candidates for correction. By default, Canu only selects the longest 40x for correction. In this case, it selects 43.2x of raw read data which it estimates will result in 40x correction. Not all raw reads survive full-length through correction:

```
--- category       raw   corrected
---       Number of Reads       20030       20030
---       Number of Bases       90137165   61903752
---       Coverage             3.919       2.691
---       Median               3324        2682
---       Mean                 4500        3090
---       N50                  5529        3659
---       Minimum              1012        501
---       Maximum              41475       10179
```

The rescued reads are those which would not have contributed to the correction of the selected longest 40x subset. These could be short plasmids, mitochondria, etc. Canu includes them even though they’re too short by the 40x cutoff to avoid losing sequence during assembly. Lastly:

```
--- category       raw   corrected
---       Number of Reads       183050       183050
---       Number of Bases      1157075583  951438105
---       Coverage             50.308       41.367
---       Median               5729        5086
---       Mean                 6321        5197
---       N50                  7467        6490
---       Minimum              0           0
---       Maximum              50522       10183
```

are the reads which were deemed too short to correct. If you increase corOutCoverage, you could get up to 41x more corrected sequence. However, unless the genome is very heterozygous, this does not typically improve the assembly and increases the running time.

The assembly read error report summarizes how unitigging was run:

```
--- ERROR RATES
---
--- 21017 samples

--- command line (-eg) 30.00 0.0300%
--- command line (-eM) 1000.00 1.0000%
--- mean + std.dev 0.08 0.0043%
--- median + mad 0.00 0.0000%
--- 90th percentile 1.00 0.0010%
```

Canu selects multiple error rate thresholds and selects the most appropriate one based on how many reads end up without overlaps at each threshold. In this case, it used 0.001% or 1 error in 10 kbp after considering 1.0% and 0.0043%.

The assembly statistics (NG50, etc) are reported before and after consensus calling. Note that for HiFi data, the
pre-consensus statistics are in homopolymer-compressed space.

READS

`<prefix>.correctedReads.fasta.gz` The reads after correction.

`<prefix>.trimmedReads.fasta.gz` The corrected reads after overlap based trimming.

SEQUENCE

`<prefix>.contigs.fasta` Everything which could be assembled and is the full assembly, including both unique, repetitive, and bubble elements.

`<prefix>.unassembled.fasta` Reads and low-coverage contigs which could not be incorporated into the primary assembly.

The header line for each sequence provides some metadata on the sequence:

```
> tig#### len=<integer> reads=<integer> covStat=<float> class=
   -<contig|bubble|unassm> suggestRepeat=<yes|no> suggestCircular=<yes|no>

len
  Length of the sequence, in bp.

reads
  Number of reads used to form the contig.

class
  Type of sequence. Unassembled sequences are primarily low-coverage sequences, spanned by a single read.

suggestRepeat
  If yes, sequence was detected as a repeat based on graph topology or read overlaps to other sequences.

suggestBubble
  If yes, sequence is likely a bubble based on potential placement within longer sequences.

suggestCircular
  If yes, sequence is likely circular. The fasta line will have a trim=X-Y to indicate the non-redundant coordinates
```

GRAPHs

Canu versions prior to v1.9 created a GFA of the contig graph. However, as noted at the time, the GFA format cannot represent partial overlaps between contigs (for more details see the discussion of general edges on the GFA2 page). Because Canu contigs are not compatible with the GFA format, `<prefix>.contigs.gfa` has been removed. Prior to Canu v2.1, contigs split at overlap junctions were output as unitigs. However, these graphs often would be missing edges and be over-fragmented (split where there is no ambiguity). Thus `<prefix>.unitigs.fasta` and `<prefix>.unitigs.gfa` have been removed.

METADATA

The layout provides information on where each read ended up in the final assembly, including contig and positions. It also includes the consensus sequence for each contig.

`<prefix>.contigs.layout` (undocumented)

`<prefix>.contigs.layout.readToTig` The position of each read in a contig.

The file looks like:
In this case read ids 677083 and 2343812 ended up in tig00000004 and the coordinates are listed at the end (read 2343812 is reverse-complemented).

You need to do a bit of work to get the original id of 2343812, look in the gkpStore/readNames.txt file, there you should find:

```
2343812 m54033_180126_39780749/39781_51526 id=4778961_0 id=2354708
   →clr=181,11399
```

which gives you the original read (PacBio in this case) id.

<prefix>.contigs.layout.tigInfo A list of the contigs, lengths, coverage, number of reads and other metadata. Essentially the same information provided in the FASTA header line.
The pipeline is described in Koren S, Walenz BP, Berlin K, Miller JR, Phillippy AM. Canu: scalable and accurate long-read assembly via adaptive k-mer weighting and repeat separation. bioRxiv. (2016). Figure 1 of the paper shows the primary pipeline (below, top) and the supplement contains the sub-pipeline for building read and overlap databases (below, bottom).
To get the most up-to-date options, run

    canu -options

The default values below will vary based on the input data type and genome size.

Boolean options accept true/false or 1/0.

Memory sizes are assumed to be in gigabytes if no units are supplied. Values may be non-integer with or without a unit - ‘k’ for kilobytes, ‘m’ for megabytes, ‘g’ for gigabytes or ‘t’ for terabytes. For example, “0.25t” is equivalent to “256g” (or simply “256”).

### 5.1 Global Options

The catch all category.

**rawErrorRate <float=unset>** The allowed difference in an overlap between two uncorrected reads, expressed as fraction error. Sets `corOvlErrorRate` and `corErrorRate`. The `rawErrorRate` typically does not need to be modified. It might need to be increased if very early reads are being assembled. The default is 0.300 For PacBio reads, and 0.500 for Nanopore reads.

**correctedErrorRate <float=unset>** The allowed difference in an overlap between two corrected reads, expressed as fraction error. Sets `obtOvlErrorRate`, `utgOvlErrorRate`, `obtErrorRate`, `utgErrorRate`, and `cnsErrorRate`. The `correctedErrorRate` can be adjusted to account for the quality of read correction, for the amount of divergence in the sample being assembled, and for the amount of sequence being assembled. The default is 0.045 for PacBio reads, and 0.144 for Nanopore reads.

For low coverage datasets (less than 30X), we recommend increasing `correctedErrorRate` slightly, by 1% or so.

For high-coverage datasets (more than 60X), we recommend decreasing `correctedErrorRate` slightly, by 1% or so.

Raising the `correctedErrorRate` will increase run time. Likewise, decreasing `correctedErrorRate` will decrease run time, at the risk of missing overlaps and fracturing the assembly.
\textbf{minReadLength <integer=1000>}: Reads shorter than this are not loaded into the assembler. Reads output by correction and trimming that are shorter than this are discarded.

Must be no smaller than \texttt{minOverlapLength}.

[In Canu v1.9 and earlier] If set high enough, the gatekeeper module will claim there are errors in the input reads, as too many of the input reads have been discarded. As long as there is sufficient coverage, this is not a problem. See \texttt{stopOnReadQuality} and \texttt{stopOnLowCoverage}.

\textbf{minOverlapLength <integer=500>}: Overlaps shorter than this will not be discovered. Smaller values can be used to overcome lack of read coverage, but will also lead to false overlaps and potential misassemblies. Larger values will result in more correct assemblies, but more fragmented, assemblies.

Must be no bigger than \texttt{minReadLength}.

\textbf{readSamplingCoverage <integer=unset>}: After loading all reads into the sequence store, discard some reads so that this amount of coverage remains. Reads are discarded according to the score described in \texttt{readSamplingBias}.

\textbf{readSamplingBias <float=0.0>}: Adjust the sampling bias towards discarding longer (negative numbers) or shorter (positive numbers) reads. Reads are assigned a score equal to random\_number \times read\_length ^ bias and the lowest scoring reads are discarded, as described in \texttt{readSamplingCoverage}.

In the pictures below, green reads are kept, while purple reads are discarded. The reads are along the X axis, sorted by decreasing score. The Y axis is the length of each read.

A bias of 0.0 will retain random reads:

![readSamplingBias=0.0](image)

A negative bias will retain shorter reads:

![readSamplingBias=-1.0](image)
A positive bias will retain longer reads:

\[ \text{readSamplingBias} = 10.0 \]

\[ \text{readSamplingBias} = 1.0 \]

\[ \text{readSamplingBias} = -10.0 \]

`genomeSize <floatunset required>` An estimate of the size of the genome. Common suffices are allowed, for example, 3.7m or 2.8g.

The genome size estimate is used to decide how many reads to correct (via the `corOutCoverage` parameter) and how sensitive the mhap overlapper should be (via the `mhapSensitivity` parameter). It also impacts some logging, in particular, reports of NG50 sizes.

`fast <toggle>` This option uses MHAP overlapping for all steps, not just correction, making assembly significantly faster. It can be used on any genome size but may produce less continuous assemblies on genomes larger than 1 Gbp. It is recommended for nanopore genomes smaller than 1 Gbp or metagenomes.

The fast option will also optionally use `wtdbg` for unitigging if `wtdbg` is manually copied to the Canu binary folder. However, this is only tested with very small genomes and is NOT recommended.

`canuIteration <internal parameter, do not use>` Which parallel iteration is being attempted.
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**canuIterationMax <integer=2>** How many parallel iterations to try. Ideally, the parallel jobs, run under grid control, would all finish successfully on the first try. Sometimes, jobs fail due to other jobs exhausting resources (memory), or by the node itself failing. In this case, canu will launch the jobs again. This parameter controls how many times it tries.

**onSuccess <string=unset>** Execute the command supplied when Canu successfully completes an assembly. The command will execute in the <assembly-directory> (the -d option to canu) and will be supplied with the name of the assembly (the -p option to canu) as its first and only parameter.

**onFailure <string=unset>** Execute the command supplied when Canu terminates abnormally. The command will execute in the <assembly-directory> (the -d option to canu) and will be supplied with the name of the assembly (the -p option to canu) as its first and only parameter.

There are two exceptions when the command is not executed: if a ‘spec’ file cannot be read, or if canu tries to access an invalid parameter. The former will be reported as a command line error, and canu will never start. The latter should never occur except when developers are developing the software.

### 5.2 Process Control

**showNext <boolean=false>** Report the first major command that would be run, but don’t run it. Processing to get to that command, for example, checking the output of the previous command or preparing inputs for the next command, is still performed.

**stopOnReadQuality <string=false>** [In Canu v1.9 and earlier] If set, Canu will stop with the following error if there are significantly fewer reads or bases loaded into the read store than what is in the input data.

```
Gatekeeper detected potential problems in your input reads.

Please review the logging in files:
/assembly/godzilla/asm.gkpStore.BUILDING.err
/assembly/godzilla/asm.gkpStore.BUILDING/errorLog

If you wish to proceed, rename the store with the following command and restart canu.

mv /assembly/godzilla/asm.gkpStore.BUILDING \
/assembly/godzilla/asm.gkpStore.ACCEPTED

Option stopOnReadQuality=false skips these checks.
```

The missing reads could be too short (decrease minReadLength to include them), or have invalid bases or quality values. A summary of the files loaded and errors detected is in the asm.gkpStore.BUILDING.err file, with full gory details in the asm.gkpStore.BUILDING/errorLog.

To proceed, set stopOnReadQuality=false or rename the directory as shown.

Note that U bases are silently translated to T bases, to allow assembly of RNA sequences.

**stopOnLowCoverage <integer=10>** Stop the assembly if read coverage is too low to be useful. Coverage is checked whenever input sequences are initially loaded into the sequence store, when corrected reads are generated, and when read ends are trimmed off.

**stopAfter <string=undefined>** If set, Canu will stop processing after a specific stage in the pipeline finishes. Valid values are:
### 5.3 General Options

**shell <string="/bin/sh">** A path to a Bourne shell, to be used for executing scripts. By default, ‘/bin/sh’, which is typically the same as ‘bash’. C shells (csh, tsh) are not supported.

**java <string="java">** A path to a Java application launcher of at least version 1.8.

**minimap <string="minimap2">** A path to the minimap2 versatile pairwise aligner.

**gnuplot <string="gnuplot">** A path to the gnuplot graphing utility. Plotting is disabled if this is unset (gnuplot= or gnuplot=undef), or if gnuplot fails to execute, or if gnuplot cannot generate plots. The latter two conditions generate warnings in the diagnostic output of Canu.

**gnuplotImageFormat <string="png">** The type of image to generate in gnuplot. By default, canu will use png, svg or gif, in that order.

**preExec <string=undefined>** A single command that will be run before Canu starts in a grid-enabled configuration. Can be used to set up the environment, e.g., with ‘module’.

### 5.4 File Staging

The correction stage of Canu requires random access to all the reads. Performance is greatly improved if the gkpStore database of reads is copied locally to each node that computes corrected read consensus sequences. This ‘staging’ is enabled by supplying a path name to fast local storage with the `stageDirectory` option, and, optionally, requesting access to that resource from the grid with the `gridEngineStageOption` option.

**stageDirectory <string=undefined>** A path to a directory local to each compute node. The directory should use an environment variable specific to the grid engine to ensure that it is unique to each task.

For example, in Sun Grid Engine, `/scratch/$JOB_ID-$SGE_TASK_ID` will use both the numeric job ID and the numeric task ID. In SLURM, `/scratch/$SLURM_JOBID` accomplishes the same.
If specified on the command line, be sure to escape the dollar sign, otherwise the shell will try to expand it before Canu sees the option: `stageDirectory=/scratch/$JOB_ID-$SGE_TASK_ID`.

If specified in a specFile, do not escape the dollar signs.

**gridEngineStageOption <string=undefined>** This string is passed to the job submission command, and is expected to request local disk space on each node. It is highly grid specific. The string `DISK_SPACE` will be replaced with the amount of disk space needed, in gigabytes.

On SLURM, an example is `–gres=1scratch:DISK_SPACE`

## 5.5 Cleanup Options

**saveOverlaps <boolean=false>** If ‘true’, retain all overlap stores. If ‘false’, delete the correction and trimming overlap stores when they are no longer useful. Overlaps used for contig construction are never deleted.

**purgeOverlaps <string=normal>** Controls when to remove intermediate overlap results.

- ‘never’ removes no intermediate overlap results. This is only useful if you have a desire to exhaust your disk space.
- ‘false’ is the same as ‘never’.
- ‘normal’ removes intermediate overlap results after they are loaded into an overlap store.
- ‘true’ is the same as ‘normal’.
- ‘aggressive’ removes intermediate overlap results as soon as possible. In the event of a corrupt or lost file, this can result in a fair amount of suffering to recompute the data. In particular, overlapper output is removed as soon as it is loaded into buckets, and buckets are removed once they are rewritten as sorted overlaps.
- ‘dangerous’ removes intermediate results as soon as possible, in some cases, before they are even fully processed. In addition to corrupt files, jobs killed by out of memory, power outages, stray cosmic rays, et cetera, will result in a fair amount of suffering to recompute the lost data. This mode can help when creating ginormous overlap stores, by removing the bucketized data immediately after it is loaded into the sorting jobs, thus making space for the output of the sorting jobs.

Use ‘normal’ for non-large assemblies, and when disk space is plentiful. Use ‘aggressive’ on large assemblies when disk space is tight. Never use ‘dangerous’, unless you know how to recover from an error and you fully trust your compute environment.

For Mhap and Minimap2, the raw overlaps (in Mhap and PAF format) are deleted immediately after being converted to Canu ovb format, except when `purgeOverlaps=never`.

**saveReadCorrections <boolean=false>**. If set, do not remove raw corrected read output from correction/2-correction. Normally, this output is removed once the corrected reads are generated.

**saveIntermediates <boolean=false>** If set, do not remove intermediate outputs. Normally, intermediate files are removed once they are no longer needed.

NOT IMPLEMENTED.

**saveMerCounts <boolean=false>** If set, do not remove meryl binary databases.

**saveReads <boolean=false>** If set, save the corrected reads (in asm.correctedReads.fasta.gz) and trimmed reads (in asm.trimmedReads.fasta.gz). Both read sets are saved in the asm.gkpStore, and can be retrieved later.
5.6 Executive Configuration

The Canu ‘executive’ is responsible for controlling what tasks run and when they run. It doesn’t directly do any significant computations, rather it just examines the files that exist and decides which component to run next. For example, if overlaps exist but contigs do not, it would create contigs next.

When under grid control, some tasks can be run in the same job as the executive, if there is enough memory and threads reserved for the executive. The benefit of this is slight; on a heavily loaded grid, it would reduce the number of job scheduling iterations Canu needs to run.

executiveMemory <integer=4>

The amount of memory, in gigabytes, to reserve when running the Canu executive (and any jobs it runs directly). Increasing this past 4 GB can allow some tasks (such as creating an overlap store or creating contigs) to run directly, without needing a separate grid job.

executiveThreads <integer=1>

The number of threads to reserve for the Canu executive.

5.7 Overlapper Configuration

Overlaps are generated for three purposes: read correction, read trimming and unitig construction. The algorithm and parameters used can be set independently for each set of overlaps.

Two overlap algorithms are in use. One, mhap, is typically applied to raw uncorrected reads and returns alignment-free overlaps with imprecise extents. The other, the original overlapper algorithm ‘ovl’, returns alignments but is much more expensive.

There are three sets of parameters, one for the ‘mhap’ algorithm, one for the ‘ovl’ algorithm, and one for the ‘minimap’ algorithm. Parameters used for a specific type of overlap are set by a prefix on the option: ‘cor’ for read correction, ‘obt’ for read trimming (‘overlap based trimming’) or ‘utg’ for unitig construction. For example, ‘corOverlapper=ovl’ would set the overlapper used for read correction to the ‘ovl’ algorithm.

{prefix}Overlapper <string=see-below> Specify which overlap algorithm, ‘mhap’ or ‘ovl’ or ‘minimap’. The default is to use ‘mhap’ for ‘cor’ and ‘ovl’ for both ‘obt’ and ‘utg’.

5.7.1 Overlapper Configuration, ovl Algorithm

{prefix}OvlErrorRate <float=unset> Overlaps above this error rate are not computed. * corOvlErrorRate applies to overlaps generated for correcting reads; * obtOvlErrorRate applied to overlaps generated for trimming reads; * utgOvlErrorRate applies to overlaps generated for assembling reads. These limits apply to the ‘ovl’ overlap algorithm and when alignments are computed for mhap overlaps with mhapReAlign.

{prefix}OvlFrequentMers <string=undefined> Do not seed overlaps with these kmers, or, for mhap, do not seed with these kmers unless necessary (down-weight them).

For corFrequentMers (mhap), the file must contain a single line header followed by number-of-kmers data lines:

```
0 number-of-kmers
forward-kmer word-frequency kmer-count total-number-of-kmers
reverse-kmer word-frequency kmer-count total-number-of-kmers
```

Where kmer-count is the number of times this kmer sequence occurs in the reads, ‘total-number-of-kmers’ is the number of kmers in the reads (including duplicates; roughly the number of bases in the reads), and ‘word-frequency’ is ‘kmer-count’ / ‘total-number-of-kmers’.
For example:

<table>
<thead>
<tr>
<th>Sequence</th>
<th>Count</th>
<th>Id</th>
</tr>
</thead>
<tbody>
<tr>
<td>AAAATAATAGACTTATCGAGTC</td>
<td>0.0000382200</td>
<td>52</td>
</tr>
<tr>
<td>GACTCGATAAGCTATTATTTT</td>
<td>0.0000382200</td>
<td>52</td>
</tr>
<tr>
<td>AAAATAATAGACTTATCGAGTCA</td>
<td>0.0000382200</td>
<td>52</td>
</tr>
<tr>
<td>TGACTCGATAAGCTATTATTTT</td>
<td>0.0000382200</td>
<td>52</td>
</tr>
</tbody>
</table>

This file must be gzip compressed.

For obtFrequentMers and ovlFrequentMers, the file must contain a list of the canonical kmers and their count on a single line. The count value is ignored, but needs to be present. This file should not be compressed.

For example:

<table>
<thead>
<tr>
<th>Sequence</th>
<th>Count</th>
</tr>
</thead>
<tbody>
<tr>
<td>AAAATAATAGACTTATCGAGTC</td>
<td>52</td>
</tr>
<tr>
<td>AAAATAATAGACTTATCGAGTCA</td>
<td>52</td>
</tr>
</tbody>
</table>

**{prefix}OvlHashBits <integer=unset>** Width of the kmer hash. Width 22=1gb, 23=2gb, 24=4gb, 25=8gb. Plus 10b per ovlHashBlockLength.

**{prefix}OvlHashBlockLength <integer=unset>** Amount of sequence (bp to load into the overlap hash table.

**{prefix}OvlHashLoad <integer=unset>** Maximum hash table load. If set too high, table lookups are inefficient; if too low, search overhead dominates run time.

**{prefix}OvlMerDistinct <integer=unset>** K-mer frequency threshold; the least frequent fraction of distinct mers can seed overlaps.

**{prefix}OvlMerSize <integer=unset>** K-mer size for seeds in overlaps.

**{prefix}OvlMerThreshold <integer=unset>** K-mer frequency threshold; mers more frequent than this count are not used to seed overlaps.

**{prefix}OvlMerTotal <integer=unset>** K-mer frequency threshold; the least frequent fraction of all mers can seed overlaps.

**{prefix}OvlRefBlockLength <integer=unset>** Amount of sequence (bp to search against the hash table per batch.

**{prefix}OvlRefBlockSize <integer=unset>** Number of reads to search against the hash table per batch.

### 5.7.2 Overlapper Configuration, mhap Algorithm

**{prefix}MhapBlockSize <integer=unset>** For the MHAP overlapper, the number of reads to load per GB of memory (mhapMemory). When mhapSensitivity=high, this value is automatically divided by two.

**{prefix}MhapMerSize <integer=unset>** K-mer size for seeds in mhap.

**{prefix}ReAlign <boolean=false>** Compute actual alignments from mhap overlaps. uses either obtErrorRate or ovlErrorRate, depending on which overlaps are computed.

**{prefix}MhapSensitivity <string="normal">** Coarse sensitivity level: ‘low’, ‘normal’ or ‘high’. Based on read coverage (which is impacted by genomeSize), ‘low’ sensitivity is used if coverage is more than 60; ‘normal’ is used if coverage is between 60 and 30, and ‘high’ is used for coverages less than 30.

### 5.7.3 Overlapper Configuration, mmap Algorithm

**{prefix}MMapBlockSize <integer=unset>** Number of reads per 1GB block. Memory * size is loaded into memory per job.
5.8 Overlap Store

The overlap algorithms return overlaps in an arbitrary order, however, all other algorithms (or nearly all) require all overlaps for a single read to be readily available. Thus, the overlap store collects and sorts the overlapper outputs into a store of overlaps, sorted by the first read in the overlap. Each overlap is listed twice in the store, once in an “A vs B” format, and once in a “B vs A” format (that is, swapping which read is ‘first’ in the overlap description).

Two construction algorithms are supported. A ‘sequential’ method uses a single data stream, and is faster for small and moderate size assemblies. A ‘parallel’ method uses multiple compute nodes and can be faster (depending on your network disk bandwidth) for moderate and large assemblies. Be advised that the parallel method is less efficient than the sequential method, and can easily thrash consumer-level NAS devices resulting in exceptionally poor performance.

The sequential method load all overlapper outputs (.ovb files in 1-overlapper) into memory, duplicating each overlap. It then sorts overlaps, and creates the final overlap store.

The parallel method uses two parallel tasks: bucketizing (‘ovb’ tasks) and sorting (‘ovs’ tasks). Bucketizing reads the outputs of the overlap tasks (ovb files in 1-overlapper), duplicates each overlap, and writes these to intermediate files. Sorting tasks load these intermediate file into memory, sorts the overlaps, then writes the sorted overlaps back to disk. There will be one ‘bucketizer’ (‘ovb’ tasks) task per overlap task, and tens to hundreds of ‘sorter’ (‘ovs’ tasks). A final ‘indexing’ step is done in the Canu executive, which ties all the various files together into the final overlap store.

Increasing ovsMemory will allow more overlaps to fit into memory at once. This will allow larger assemblies to use the sequential method, or reduce the number of ‘ovs’ tasks for the parallel method.

Increasing the allowed memory for the Canu executive can allow the overlap store to be constructed as part of the executive job – a separate grid job for constructing the store is not needed.

ovsMemory <float> How much memory, in gigabytes, to use for constructing overlap stores. Must be at least 256m or 0.25g.

5.9 Meryl

The ‘meryl’ algorithm counts the occurrences of kmers in the input reads. It outputs a FASTA format list of frequent kmers, and (optionally) a binary database of the counts for each kmer in the input.

Meryl can run in (almost) any memory size, by splitting the computation into smaller (or larger) chunks.

merylMemory <integer> Amount of memory, in gigabytes, to use for counting kmers.

merylThreads <integer> Number of compute threads to use for kmer counting.

5.10 Overlap Based Trimming

obtErrorRate <float> Stringency of overlaps to use for trimming reads.

trimReadsOverlap <integer> Minimum overlap between evidence to make contiguous trim.

trimReadsCoverage <integer> Minimum depth of evidence to retain bases.
5.11 Trio binning Configuration

hapUnknownFraction <float=0.05> Fraction of unclassified bases to ignore for haplotype assemblies. If there are more than this fraction of unclassified bases, they are included in both haplotype assemblies.

5.12 Grid Engine Support

Canu directly supports most common grid scheduling systems. Under normal use, Canu will query the system for grid support, configure itself for the machines available in the grid, then submit itself to the grid for execution. The Canu pipeline is a series of about a dozen steps that alternate between embarrassingly parallel computations (e.g., overlap computation) and sequential bookkeeping steps (e.g., checking if all overlap jobs finished). This is entirely managed by Canu.

Canu has first class support for the various schedulers derived from Sun Grid Engine (Univa, Son of Grid Engine) and the Simple Linux Utility for Resource Management (SLURM), meaning that the developers have direct access to these systems. Platform Computing’s Load Sharing Facility (LSF) and the various schedulers derived from the Portable Batch System (PBS, Torque and PBSPro) are supported as well, but without developer access bugs do creep in. As of Canu v1.5, support seems stable and working.

useGrid <boolean=true> Master control. If ‘false’, no algorithms will run under grid control. Does not change the value of the other useGrid options.

If ‘remote’, jobs are configured for grid execution, but not submitted. A message, with commands to launch the job, is reported and canu halts execution.

Note that the host used to run canu for ‘remote’ execution must know about the grid, that is, it must be able to submit jobs to the grid.

It is also possible to enable/disable grid support for individual algorithms with options such as useGridBAT, useGrid-CNS, et cetera. This has been useful in the (far) past to prevent certain algorithms, notably overlap error adjustment, from running too many jobs concurrently and thrashing disk. Recent storage systems seem to be able to handle the load better – computers have gotten faster quicker than genomes have gotten larger.

There are many options for configuring a new grid (‘gridEngine*’) and for configuring how canu configures its computers to run under grid control (‘gridOptions*’). The grid engine to use is specified with the ‘gridEngine’ option.


5.12.1 Grid Engine Configuration

There are many options to configure support for a new grid engine, and we don’t describe them fully. If you feel the need to add support for a new engine, please contact us. That said, file src/pipeline/canu/Defaults.pm lists a whole slew of parameters that are used to build up grid commands, they all start with gridEngine. For each grid, these parameters are defined in the various src/pipeline/Grid_*.pm modules. The parameters are used in src/pipeline/canu/Execution.pm.

In Canu 1.8 and earlier, gridEngineMemoryOption and gridEngineThreadsOption are used to tell Canu how to request resources from the grid. Starting with snapshot v1.8 +90 changes (roughly January 11th), those options were merged into gridEngineResourceOption. These options specify the grid options needed to request memory and threads for each job. For example, the default gridEngineResourceOption for PBS/Torque is ‘-l nodes=1:ppn=THREADS:mem=MEMORY’, and for Slurm it is ‘--cpus-per-task=THREADS --mem-per-cpu=MEMORY’. Canu will replace “THREADS” and “MEMORY” with the specific values needed for each job.
5.12.2 Grid Options

To run on the grid, each stage needs to be configured - to tell the grid how many cores it will use and how much memory it needs. Some support for this is automatic (for example, overlapInCore and mhap know how to do this), others need to be manually configured. Yes, it’s a problem, and yes, we want to fix it.

The gridOptions* parameters supply grid-specific options to the grid submission command.

gridOptions <string=unset> Grid submission command options applied to all grid jobs
gridOptionsJobName <string=unset> Grid submission command jobs name suffix
gridOptionsBAT <string=unset> Grid submission command options applied to unitig construction with the bogart algorithm
gridOptionsGFA <string=unset> Grid submission command options applied to gfa alignment and processing
gridOptionsCNS <string=unset> Grid submission command options applied to unitig consensus jobs
gridOptionsCOR <string=unset> Grid submission command options applied to read correction jobs
gridOptionsExecutive <string=unset> Grid submission command options applied to master script jobs
gridOptionsOEA <string=unset> Grid submission command options applied to overlap error adjustment jobs
gridOptionsRED <string=unset> Grid submission command options applied to read error detection jobs
gridOptionsOVB <string=unset> Grid submission command options applied to overlap store bucketizing jobs
gridOptionsOVS <string=unset> Grid submission command options applied to overlap store sorting jobs
gridOptionsCORMHAP <string=unset> Grid submission command options applied to mhap overlaps for correction jobs
gridOptionsCOROVL <string=unset> Grid submission command options applied to overlaps for correction jobs
gridOptionsOBTMHAP <string=unset> Grid submission command options applied to mhap overlaps for trimming jobs
gridOptionsOBTOVL <string=unset> Grid submission command options applied to overlaps for trimming jobs
gridOptionsUTGMHAP <string=unset> Grid submission command options applied to mhap overlaps for unitig construction jobs
gridOptionsUTOVL <string=unset> Grid submission command options applied to overlaps for unitig construction jobs

5.13 Algorithm Selection

Several algorithmic components of canu can be disabled, based on the type of the reads being assembled, the type of processing desired, or the amount of compute resources available. Overlap

enableOEA <boolean=true> Do overlap error adjustment - comprises two steps: read error detection (RED and overlap error adjustment (OEA

5.13.1 Algorithm Execution Method

Canu has a fairly sophisticated (or complicated, depending on if it is working or not) method for dividing large computes, such as read overlapping and consensus, into many smaller pieces and then running those pieces on a grid or in parallel on the local machine. The size of each piece is generally determined by the amount of memory the task is allowed to use, and this memory size – actually a range of memory sizes – is set based on the genomeSize parameter,
but can be set explicitly by the user. The same holds for the number of processors each task can use. For example, a
genomeSize=5m would result in overlaps using between 4gb and 8gb of memory, and between 1 and 8 processors.

Given these requirements, Canu will pick a specific memory size and number of processors so that the maximum
number of jobs will run at the same time. In the overlapper example, if we are running on a machine with 32gb
memory and 8 processors, it is not possible to run 8 concurrent jobs that each require 8gb memory, but it is possible
to run 4 concurrent jobs each using 6gb memory and 2 processors.

To completely specify how Canu runs algorithms, one needs to specify a maximum memory size, a maximum number
of processors, and how many pieces to run at one time. Users can set these manually through the {prefix}Memory,
{prefix}Threads and {prefix}Concurrent options. If they are not set, defaults are chosen based on genomeSize.

{prefix}Concurrency <integer=unset> Set the number of tasks that can run at the same time, when running without
grid support.

{prefix}Threads <integer=unset> Set the number of compute threads used per task.

{prefix}Memory <integer=unset> Set the amount of memory, in gigabytes, to use for each job in a task.

Available prefixes are:

<table>
<thead>
<tr>
<th>Prefix</th>
<th>Algorithm</th>
</tr>
</thead>
<tbody>
<tr>
<td>cor</td>
<td>mhap</td>
</tr>
<tr>
<td></td>
<td>Overlap generation using the</td>
</tr>
<tr>
<td></td>
<td>‘mhap’ algorithm for</td>
</tr>
<tr>
<td></td>
<td>‘cor’=correction</td>
</tr>
<tr>
<td></td>
<td>‘obt’=trimming</td>
</tr>
<tr>
<td></td>
<td>‘utg’=assembly</td>
</tr>
<tr>
<td>obt</td>
<td></td>
</tr>
<tr>
<td>utg</td>
<td></td>
</tr>
<tr>
<td>cor</td>
<td>mmap</td>
</tr>
<tr>
<td></td>
<td>Overlap generation using the</td>
</tr>
<tr>
<td></td>
<td>‘minimap2’ algorithm for</td>
</tr>
<tr>
<td></td>
<td>‘cor’ = correction</td>
</tr>
<tr>
<td></td>
<td>‘obt’ = trimming</td>
</tr>
<tr>
<td></td>
<td>‘utg’ = assembly</td>
</tr>
<tr>
<td>obt</td>
<td></td>
</tr>
<tr>
<td>utg</td>
<td></td>
</tr>
<tr>
<td>cor</td>
<td>ovl</td>
</tr>
<tr>
<td></td>
<td>Overlap generation using the</td>
</tr>
<tr>
<td></td>
<td>‘overlapInCore’ algorithm for</td>
</tr>
<tr>
<td></td>
<td>‘cor’ = correction</td>
</tr>
<tr>
<td></td>
<td>‘obt’ = trimming</td>
</tr>
<tr>
<td></td>
<td>‘utg’ = assembly</td>
</tr>
<tr>
<td>obt</td>
<td></td>
</tr>
<tr>
<td>utg</td>
<td></td>
</tr>
<tr>
<td></td>
<td>ovb</td>
</tr>
<tr>
<td></td>
<td>Parallel overlap store bucketizing</td>
</tr>
<tr>
<td></td>
<td>ovs</td>
</tr>
<tr>
<td></td>
<td>Parallel overlap store bucket sorting</td>
</tr>
<tr>
<td></td>
<td>cor</td>
</tr>
<tr>
<td></td>
<td>Read correction</td>
</tr>
<tr>
<td></td>
<td>red</td>
</tr>
<tr>
<td></td>
<td>Error detection in reads</td>
</tr>
<tr>
<td></td>
<td>oea</td>
</tr>
<tr>
<td></td>
<td>Error adjustment in overlaps</td>
</tr>
<tr>
<td></td>
<td>bat</td>
</tr>
<tr>
<td></td>
<td>Unitig/contig construction</td>
</tr>
<tr>
<td></td>
<td>cns</td>
</tr>
<tr>
<td></td>
<td>Unitig/contig consensus</td>
</tr>
</tbody>
</table>

For example, ‘mhapMemory’ would set the memory limit for computing overlaps with the mhap algorithm; ‘cormhap-
Memory’ would set the memory limit only when mhap is used for generating overlaps used for correction.

The ‘minMemory’, ‘maxMemory’, ‘minThreads’ and ‘maxThreads’ options will apply to all jobs, and can be used
to artificially limit canu to a portion of the current machine. In the overlapper example above, setting maxThreads=4
would result in two concurrent jobs instead of four.
5.14 Overlap Error Adjustment

The Overlap Error Adjustment module adjusts the error rate claimed by each overlap to account for sequencing error and true polymorphism. A pair-wise multialignment is generated for all overlaps to a given read. Each multialignment column is examined to determine if the base in the given read is correct, is part of a true difference, or is a likely sequencing error. For the latter case, a base change is noted. Once all base changes in all columns of all reads have been determined, all overlaps are recomputed with said changes applied, and the new fraction error stored for each overlap.

Three parameters exist to change the behavior:

- **oeaErrorRate <real=unset>** Do not use overlaps that are above this fraction error for finding true differences and errors. Mostly useful only when oeaMaskTrivial is enabled. For PacBio and Nanopore data, the default is utgOvlErrorRate. For HiFi data, the default is 0.003.

- **oeaHaploConfirm <integer=5>** A set of this many reads with the same base in a column is declared to be a true haplotype difference and will increase the fraction error for the overlap. The default for all read types is 5.

- **oeaMaskTrivial <boolean=unset>** If true, errors that occur in low-complexity sequence are ignored when finding base changes and when recomputing overlap error rates. Unless set explicitly, it is true (enabled) for PacBio HiFi data and false otherwise.

This module consists of two steps: RED (read error detection), and OEA (overlap error adjustment). They have slightly different run-time requirements. RED can use multiple threads and is slightly more computationally expensive; OEA can not use multiple threads and is slightly more I/O intensive. The batch length and batch size parameters can tune the size of each job, however, the default values have worked well (so well that we don’t actually document what these values should be set to).

- **oeaBatchLength <unset>** Number of bases per overlap error correction batch
- **oeaBatchSize <unset>** Number of reads per overlap error correction batch
- **redBatchLength <unset>** Number of bases per fragment error detection batch
- **redBatchSize <unset>** Number of reads per fragment error detection batch

5.15 Unitigger

- **unitigger <string="bogart">** Which unitig construction algorithm to use. Only “bogart” is supported.
- **utgErrorRate <float=unset>** Stringency of overlaps used for constructing contigs. The bogart algorithm uses the distribution of overlap error rates to filter high error overlaps; bogart will never see overlaps with error higher than this parameter.
- **batOptions <unset>** Advanced options to bogart

5.16 Consensus Partitioning

STILL DONE BY UNITIGGER, NEED TO MOVE OUTSIDE

- **cnsConsensus** Which algorithm to use for computing consensus sequences. Only ‘utgcn5’ is supported.
- **cnsPartitions** Compute consense by splitting the tigs into N partitions.
- **cnsPartitionMin** Don’t make a partition with fewer than N reads
- **cnsMaxCoverage** Limit unitig consensus to at most this coverage.
**cnsErrorRate**  Inform the consensus generation algorithm of the amount of difference it should expect in a read-to-read alignment. Typically set to `utgOvlErrorRate`. If set too high, reads could be placed in an incorrect location, leading to errors in the consensus sequence. If set too low, reads could be omitted from the consensus graph (or multialignment, depending on algorithm), resulting in truncated consensus sequences.

### 5.17 Read Correction

The first step in Canu is to find high-error overlaps and generate corrected sequences for subsequent assembly. This is currently the fastest step in Canu. By default, only the longest 40X of data (based on the specified genome size) is used for correction. Typically, some reads are trimmed during correction due to being chimeric or having erroneous sequence, resulting in a loss of 20-25% (30X output). You can force correction to be non-lossy by setting `corMinCoverage=0`, in which case the corrected reads output will be the same length as the input data, keeping any high-error unsupported bases. Canu will trim these in downstream steps before assembly.

If you have a dataset with uneven coverage or small plasmids, correcting the longest 40X may not give you sufficient coverage of your genome/plasmid. In these cases, you can set `corOutCoverage=999`, or any value greater than your total input coverage which will correct and assemble all input data, at the expense of runtime.

**corErrorRate <integer=unset>**  Do not use overlaps with error rate higher than this when computing corrected reads.

In Canu v2.2, this parameter was changed from 0.50 (for -nanopore) and 0.30 (for -pacbio) to 0.30 and 0.25, respectively.

The tables below show a significant speedup for Nanopore reads without much loss in output quantity. There is indication of a slight improvement in corrected read quality at lower corErrorRate, however, read quality was not directly evaluated.

For PacBio reads, with a smaller change in corErrorRate, the speedup is about 10%.

<table>
<thead>
<tr>
<th>CHM13 Chromosome X, nanopore, 105x input coverage</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>corErrorRate</strong></td>
</tr>
<tr>
<td>------------------</td>
</tr>
<tr>
<td>5</td>
</tr>
<tr>
<td>10</td>
</tr>
<tr>
<td>15</td>
</tr>
<tr>
<td>20</td>
</tr>
<tr>
<td>25</td>
</tr>
<tr>
<td>30</td>
</tr>
<tr>
<td>40</td>
</tr>
<tr>
<td>50</td>
</tr>
</tbody>
</table>
HG002 Chromosome X, nanopore, 20x input coverage

<table>
<thead>
<tr>
<th>corErrorRate</th>
<th>Corrected Coverage</th>
<th>Trimmed Coverage</th>
<th>Bogart Error Rate</th>
<th>CPU Time (hours)</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>–.-x</td>
<td>–.-x</td>
<td>-%</td>
<td>31</td>
</tr>
<tr>
<td>10</td>
<td>3.9x</td>
<td>–.-x</td>
<td>-%</td>
<td>66</td>
</tr>
<tr>
<td>15</td>
<td>9.6x</td>
<td>–.-x</td>
<td>-%</td>
<td>105</td>
</tr>
<tr>
<td>20</td>
<td>11.4x</td>
<td>11.2x</td>
<td>1.71%</td>
<td>134</td>
</tr>
<tr>
<td>25</td>
<td>11.9x</td>
<td>11.6x</td>
<td>1.79%</td>
<td>154</td>
</tr>
<tr>
<td>30</td>
<td>12.0x</td>
<td>11.8x</td>
<td>1.83%</td>
<td>169</td>
</tr>
<tr>
<td>40</td>
<td>12.2x</td>
<td>12.0x</td>
<td>1.94%</td>
<td>221</td>
</tr>
<tr>
<td>50</td>
<td>12.6x</td>
<td>12.3x</td>
<td>2.29%</td>
<td>709</td>
</tr>
</tbody>
</table>

corConsensus <string="falconpipe"> Which algorithm to use for computing read consensus sequences. Only ‘falcon’ and ‘falconpipe’ are supported.

corPartitions <integer=128> Partition read correction into N jobs

corPartitionMin <integer=25000> Don’t make a read correction partition with fewer than N reads

corMinEvidenceLength <integer=unset> Limit read correction to only overlaps longer than this; default: unlimited

corMinCoverage <integer=4> Limit read correction to regions with at least this minimum coverage. Split reads when coverage drops below threshold.

corMaxEvidenceErate <integer=unset> Limit read correction to only overlaps at or below this fraction error; default: unlimited

corMaxEvidenceCoverageGlobal <string="1.0x"> Limit reads used for correction to supporting at most this coverage; default: 1.0 * estimated coverage

corMaxEvidenceCoverageLocal <string="2.0x"> Limit reads being corrected to at most this much evidence coverage; default: 10 * estimated coverage

corOutCoverage <integer=40> Only correct the longest reads up to this coverage; default 40

corFilter <string="expensive"> Method to filter short reads from correction; ‘quick’ or ‘expensive’ or ‘none’

5.18 Output Filtering

contigFilter <minReads, integer=2> <minLength, integer=0> <singleReadSpan, float=1.0> <lowCovSpan, float=0.5> <lowCovDepth, integer=5>

A contig that needs any of the following conditions is flagged as ‘unassembled’ and removed from further consideration:

- fewer than minReads reads (default 2)
- shorter than minLength bases (default 0)
- a single read covers more than singleReadSpan fraction of the contig (default 1.0)
- more than lowCovSpan fraction of the contig is at coverage below lowCovDepth (defaults 0.5, 5)

This filtering is done immediately after initial contigs are formed, before potentially incorrectly spanned repeats are detected. Initial contigs that incorrectly span a repeat can be split into multiple contigs; none of these new contigs will be flagged as ‘unassembled’, even if they are a single read.
Canu is derived from Celera Assembler, which is no longer maintained.

Celera Assembler [Myers 2000] was designed to reconstruct mammalian chromosomal DNA sequences from the short fragments of a whole genome shotgun sequencing project. Celera Assembler was used to produce reconstructions of several large genomes, namely those of Homo sapiens [Venter 2001], Mus musculus [Mural 2002], Rattus norvegicus [unpublished data], Canis familiaris [Kirkness 2003], Drosophila melanogaster [Adams 2000], and Anopheles gambiae [Holt 2001]. Celera Assembler was shown to be very accurate when its reconstruction of the human genome was compared to independent reconstructions completed later [Istrail 2004]. It was used to reconstructing one of the first large-scale metagenomic projects [Venter 2004, Rusch 2007] and a diploid human reference [Levy 2007, Denisov 2008]. It was adapted to 454 Pyrosequencing [Miller 2008] and PacBio sequencing [Koren 2012], demonstrating finished bacterial genomes [Koren 2013] and efficient algorithms for eukaryotic assembly [Berlin 2015].

Celera Assembler was released under the GNU General Public License, version 2 as a supplement to [Istrail 2004].

Canu [Koren and Walenz 2017] was branched from Celera Assembler in 2015, and specialized for single-molecule high-noise sequences. For the most recent license information on Canu, see README.licences.

6.1 References

• Levy et al. (2007) The Diploid Genome Sequence of an Individual Human. PLoS Biology 0050254
• Koren et al. (2013) Reducing assembly complexity of microbial genomes with single-molecule sequencing. Genome Biology 14:R101.

Canu is a fork of the Celera Assembler designed for high-noise single-molecule sequencing (such as the PacBio RSII or Oxford Nanopore MinION).


The easiest way to get started is to download a release. If you encounter any issues, please report them using the github issues page.

Alternatively, you can also build the latest unreleased from github:

```
git clone https://github.com/marbl/canu.git
cd canu/src
make -j <number of threads>
```
CHAPTER 9

Learn

- **Quick Start** - no experience or data required, download and assemble *Escherichia coli* today!
- **FAQ** - Frequently asked questions
- **Canu tutorial** - a gentle introduction to the complexities of canu.
- **Canu pipeline** - what, exactly, is canu doing, anyway?
- **Canu Parameter Reference** - all the parameters, grouped by function.
- **Canu History** - the history of the Canu project.