

trim only as described in (2) above. This will result in all trimmed reads having the desired RF orientation. Even when the adapter is located at or within a few bases of the 5' end of the read—as described in (1) above—trimming strategy (2) can still be employed. However, the resulting sequence will contain all or mostly adapter sequence and may not align. Thus trimming only as described in (2) will result in a slight decrease in data yield, but will ensure that aligned read pairs maintain the desired RF orientation.

An alternative trimming approach is to apply either (1) or (2) on a per-read basis such that longest possible fragment is preserved. This is achieved by trimming from the end of the read closest to the junction adapter, resulting in a set of read pairs with both RF and FR orientations. Reads with an FR orientation will have an insert size distribution with a mean less than 1 Kb. This subset should be segregated during adapter trimming and aligned separately unless an aligner is used that can handle reads pairs with differing orientations and insert size distributions.

As described above, the junction adapter may appear in one of three forms (Table 1, adapter elements 1–3). Therefore, two rounds of adapter trimming must be performed, first trimming the single adapter and then its reverse complement (Table 1, adapter elements 2–3 respectively). This process will also comprehensively trim the duplicate junction adapter sequence as well. If external adapter sequence is present within a read, then the trimming process described above is optional. This is because reads containing external adapter sequence have a very low probability of alignment or assembly. In most cases, it is therefore preferable to throw out read pairs that contain external adapter sequence, thus saving time during the alignment or assembly process.

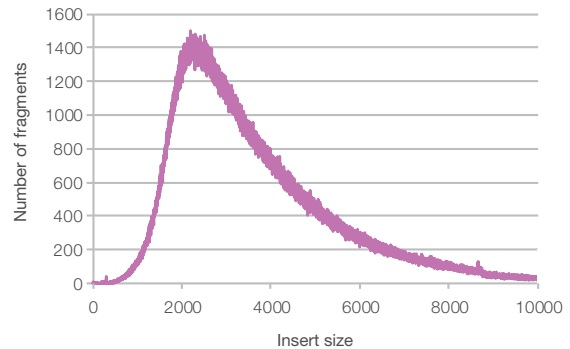
In order for trimming to be effective, the software tools used to perform read trimming must also be able to trim partial adapters. Examples of tools that can perform partial adapter trimming include Biopieces¹ and adapterremoval². Both tools will trim one or more adapters from either the 3' or 5' end of the read, though neither has the option of trimming from the end closest to the junction adapter. Ideally, the tool used for adapter trimming should also allow for potential sequencing errors in the adapter sequence. The degree to which mismatches are allowed will depend on the length of the adapter and on the desired sensitivity in identifying the adapter sequence. For example, a stringent mismatch setting (e.g., requiring zero mismatches) may cause the trimming algorithm to miss some of the adapter sequences; while a looser mismatch setting may cause the trimming algorithm to identify and trim sequences that do not originate from the adapter.

Secondary Analysis

Commercially-available secondary analysis tools such as Novoalign³ and CLC Genomics Workbench⁴ are able to analyze read pairs in an RF orientation. Certain open source tools such as BWA⁵, used for alignment, and Velvet⁶, used for *de novo* assembly, require the paired reads to be in an FR orientation. For these latter tools, read pairs in the RF orientation must be reverse complemented prior to alignment or assembly. If the sequence data contains a mixture of RF and FR pairs, then only the RF pairs should be reverse complemented.

If performing *de novo* assembly using Velvet and trimming from the end of the read closest to the internal adapter (i.e., per-read trimming to preserve the longest possible fragment), it is conventional to separate the reads pairs that are trimmed according to strategy

Figure 3: *E. coli* Mate Pair Library Insert Size Distribution



The insert size distribution of an *E. coli* K-12 MG1655 mate pair library sequenced on the MiSeq[®] platform. The insert size per read pair was calculated after aligning the data to the reference genome using BWA. Histogram summary statistics were calculated across all values. The median insert size was 3,125 bp.

(1) versus strategy (2) into separate fastq files. Each fastq file can then be passed separately to Velvet for assembly via command line arguments, '-shortPaired -fastq strategy1.fastq -shortPaired2 -fastq strategy2.fastq'. Velvet makes use of the insert size distribution of each fastq file during the assembly process. Since the FR reads defined by strategy (1) have a lower mean insert size than the RF reads defined by strategy (2), passing them simultaneously in a single fastq file would confound the algorithm and result in a suboptimal assembly. This topic is addressed in more detail in the Velvet user manual⁷.

Most *de novo* assemblers are based on the manipulation of de Bruijn graphs. Simply, this approach involves breaking reads into smaller pieces, known as k-mers (where k denotes its length in bases), in order to reduce the computational load of assembly. To achieve contiguous *de novo* assemblies, long k-mers are crucial. After adapter trimming, the average read length of a paired 100 bp run is approximately 90 bp. The maximum k-mer length is obviously bounded by the read length; for trimmed 90 bp reads, reasonable k-mers sizes are between 55 and 60 bp for a bacterial genome. The k-mer size is a configurable parameter in most assembly tools; so it is reasonable to try several k-mer sizes and choose the assembly that offers an optimal combination of scaffold sizes and assembly errors. Tools such as VelvetOptimiser⁸ automate the process of choosing the optimal k-mer size.

Example Applications of Mate Pair Data

The following data generated by *E. coli* and human samples demonstrate that Nextera mate pair libraries support both small and complex genomes.

De Novo Assembly of *E. coli* K-12 MG1655 Genome

Nextera mate pair library sequencing allows the user to tune the experimental insert size distribution of the reads, meaning that read pairs from the same experiment can offer both long-range and short-range sequence information. The ultimate effect of this additional information is improved *de novo* assemblies.

Table 2: Assembly Summary Statistics from an *E. coli* Mate Pair Library Run

Assembly Metric	Size (bp)
Genome Size	4,630,577
N50 (Scaffold)	4,617,839
Max Scaffold Size	4,617,839
N50 (Contig)	170,901
MaxContig Size	432,335

Assembly summary statistics from a 2 × 150 bp *E. coli* K-12 MG1655 mate pair library sequenced on the MiSeq platform.

Table 3: Assembly Summary Statistics from Combined Mate Pair and Paired-End Runs

Application	Mate Pair	Paired-End
Mean Coverage	280.2	1805
Number of Bases < 8× Coverage	4	20,964
% Mismatches	0.63%	0.20%
% Indels	0.02%	0.00%
% Aligned Bases > Q30	82.59%	85.61%

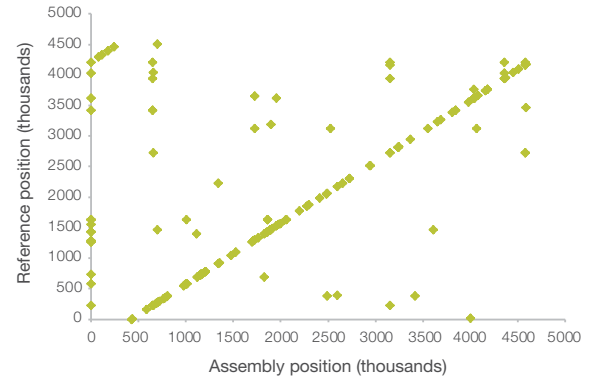
Assembly summary statistics from a 2 × 150 mate pair library and a 2 × 250 standard paired-end library of *E. coli* K-12 MG1655 sequenced on a MiSeq instrument. The additional information offered by long-range mate pairs allows much higher coverage of the reference genome. When assessing coverage, reads with mapping quality score (MAPQ) less than 1 were filtered from analysis.

To demonstrate the use of mate pair libraries for *de novo* assembly, the genome of *E. coli* K-12 MG1655—a laboratory strain of bacteria used to benchmark many NGS applications—was assembled. A MG1655 library was generated as described in the Nextera Mate Pair Sample Preparation User Guide (PN 15035209)⁹. Each mate pair read was sequenced on the MiSeq platform for 150 cycles resulting in a total of 300 bases of sequencing data per mate pair library. Total data yield was 1.8 Gb of raw data, the median insert size was 3,125 bp, and the insert size distribution is shown in Figure 3.

Prior to performing *de novo* assembly, the data were pre-processed as follows. In short, the adapters were trimmed and the RF reads were reverse complemented. Additionally, the data were randomly down-sampled to 500 Mb so as to limit the mean coverage to 50–100×, thereby reducing the computational requirements of the assembler. The VelvetOptimiser software was used for assembly. The optimal k-mer size was 99 bp. Summary statistics for the assembly are shown in Table 2.

To confirm the consistency of the scaffolded assembly with the reference genome, a dot plot was constructed comparing the alignment of the scaffold to the reference genome (Figure 4). Dot plots similar to the one shown in Figure 4 can be generated with tools such as MUMmer¹⁰. The plot shows that the assembled scaffold correlates perfectly with the reference genome. The observed discontinuity (i.e., two lines) between the position of the assembly and the reference is

Figure 4: Dot Plot of *E. coli* Assembly Aligned to the Reference Genome



Dot plot of an *E. coli* K-12 MG1655 mate pair assembly aligned to the reference genome. The assembled scaffold correlates perfectly with the reference, indicated by the diagonal line. The appearance of two distinct lines is due to the circular structure of the *E. coli* genome.

Table 4: Alignment Summary Statistics from a Human Mate Pair Sequencing Run

Metric	Mate Pair Result
Mean Coverage	15.70
% Coverage >8×	92.04%
% Mismatches	0.46%
% Indels	0.03%
% Aligned Bases >Q30	92.04%

Alignment summary statistics from a Human NA12877 mate pair library, run at 2 × 100 on a MiSeq instrument.

due to the circular genome structure of *E. coli*; that is, the first base in the assembled scaffold is unlikely to align with the first base of the reference genome because the first position in the assembly is arbitrarily selected.

To further assess run quality, *E. coli* read pairs were aligned to the reference genome using BWA. This alignment was then compared to the alignment generated from a standard 2 × 250 bp paired-end MiSeq run. The results are summarized in Table 3. To assess coverage, reads having a mapping quality score (MAPQ) less than 1 were filtered out. Due to the additional information offered by long-range mate pair reads, the percentage of the genome having unique coverage is significantly improved, as demonstrated by the 5000× reduction in the number of bases with coverage below 8×.

Sequencing of A Human NA12877 Sample

A Nextera mate pair library of a human male (NA12877) sample was sequenced on the HiSeq 2500 platform for 2 × 100 cycles and aligned using BWA. The raw data yield was 67 Gb, and the aligned yield was 45 Gb, corresponding to a mean coverage of approximately 16×. The loss in aligned yield is primarily due to adapter trimming. The median

