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Detection of Copy Number Alterations Using Single Cell Sequencing

Kristin A. Knouse¹,²,³, Jie Wu⁴, Austin Hendricks⁵

¹Koch Institute for Integrative Cancer Research, Department of Biology, Massachusetts Institute of Technology
²Howard Hughes Medical Institute
³Division of Health Sciences and Technology, Harvard Medical School
⁴The Barbara K. Ostrom (1978) Bioinformatics and Computing Facility in the Swanson Biotechnology Center, Koch Institute for Integrative Cancer Research, Massachusetts Institute of Technology
⁵BioMicro Center, Department of Biology, Massachusetts Institute of Technology

Correspondence to: Kristin A. Knouse at kknouse@mit.edu

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Abstract

Detection of genomic changes at single cell resolution is important for characterizing genetic heterogeneity and evolution in normal tissues, cancers, and microbial populations. Traditional methods for assessing genetic heterogeneity have been limited by low resolution, low sensitivity, and/or low specificity. Single cell sequencing has emerged as a powerful tool for detecting genetic heterogeneity with high resolution, high sensitivity and, when appropriately analyzed, high specificity. Here we provide a protocol for the isolation, whole genome amplification, sequencing, and analysis of single cells. Our approach allows for the reliable identification of megabase-scale copy number variants in single cells. However, aspects of this protocol can also be applied to investigate other types of genetic alterations in single cells.

Introduction

Alterations in DNA copy number can range in size from several base pairs (copy number variants) (CNVs) to entire chromosomes (aneuploidy). Copy number alterations affecting large regions of the genome can have significant phenotypic consequences by altering the expression of up to thousands of genes¹,². CNVs that are present in all cells of a population can be detected by bulk sequencing or microarray-based methods³,⁴. However, populations can also be genetically heterogeneous, with CNVs existing in a subset of the population or even single cells. Genetic heterogeneity is common in cancer, driving tumor evolution, and also present in normal tissues, with unknown consequence⁵,⁶,⁷,⁸,⁹,¹⁰.

Traditionally, genetic heterogeneity was assessed either by cytologic approaches or bulk sequencing. Cytologic approaches, such as fluorescence in situ hybridization (FISH), chromosome spreads, and spectral karyotyping (SKY), have the benefit of identifying alterations present in individual cells but have high error rates due to artifacts of hybridization and spreading¹¹. These approaches are also limited in their resolution—only revealing copy number changes spanning several megabases. Sequencing or microarrays of bulk DNA, though higher in accuracy and resolution, is less sensitive. In order to detect genetic heterogeneity by population-based approaches, the variants must be present in a substantial fraction of cells in the population. The emergence of methods to amplify the genomic DNA from single cells has made it possible to sequence the genome of single cells. Single cell sequencing has the advantages of high resolution, high sensitivity, and, when appropriate quality control methods are applied, high accuracy¹².

Here, we describe a method for detecting megabase-scale copy number alterations in single cells. We isolate single cells by microaspiration, amplify genomic DNA using linker-adaptor PCR, prepare libraries for next-generation sequencing, and detect copy number variants by both hidden Markov model and circular binary segmentation.

Protocol

1. Isolating Single Cells

1. Prepare the microaspirator
   1. Remove the clear plastic end from the aspirator tube assembly and insert the narrow end into one end of a 1 foot-long PVC tubing with 3/16” inner diameter.
2. Whole Genome Amplification

1. To prevent contamination during whole genome amplification, add all reagents inside a tissue culture hood, use pipette tips with filters, and change the pipette tip in between wells.

2. Prepare a working lysis and fragmentation buffer solution from the whole genome amplification (WGA) kit. For each set of up to 32 cells, combine 32 µL of 10x single cell lysis and fragmentation buffer and 2 µL of proteinase K solution in a microcentrifuge tube. Vortex the tube to mix the solutions.

3. Add 1 µL of working lysis and fragmentation buffer solution to each well and pipette up and down to mix.

4. Cover the plate and seal all wells with a plastic film. Briefly centrifuge the plate in a mini plate spinner.

5. Thermal cycle as follows: 50 °C, 1 h and 99 °C, 4 min.

6. Cool the plate on ice and briefly centrifuge the plate in a mini plate spinner.

7. Prepare a working library preparation buffer solution from the whole genome amplification kit. For each cell, combine 2 µL of 1x single cell library preparation buffer and 1 µL of library stabilization solution in a microcentrifuge tube. Prepare the working solution for several cells in the same microcentrifuge tube.

8. Remove the plastic film and add 3 µL of working library preparation buffer solution to each well. Pipette the contents of the well up and down to mix. Replace the plastic film.

9. Briefly centrifuge the plate in a mini plate spinner and incubate at 95 °C for 2 min.

10. Cool the plate on ice and briefly centrifuge the plate in a mini plate spinner.

11. Remove the plastic film, add 1 µL library preparation enzyme to each well, and pipette the contents of the well up and down to mix. Keep the library preparation enzyme on ice or in a cold block throughout this step. Replace the plastic film.

12. Briefly centrifuge the plate in mini plate spinner and thermal cycle as follows: 16 °C, 20 min, 24 °C, 20 min, 37 °C, 20 min, 75 °C, 5 min, and hold at 4 °C.

13. Cool the plate on ice and briefly centrifuge the plate in a mini plate spinner.

14. Prepare a working amplification mix from the whole genome amplification kit. For each cell, combine 48.5 µL water, 7.5 µL 10x amplification master mix, and 5 µL WGA DNA polymerase in a microcentrifuge tube. Prepare the working mix for several cells in the same microcentrifuge tube. Keep the working mix on ice.

15. Remove the plastic film, add 61 µL working amplification mix to each well, and pipette contents of well up and down to mix. Keep the working amplification mix on ice or in a cold block throughout this step. Replace the plastic film.
16. Briefly centrifuge the plate in a mini plate spinner and thermal cycle as follows: 95 °C, 3 min, 25 cycles of 94 °C, 30 sec and 65 °C, 5 min, and hold at 4 °C.
17. Transfer 60 µL of each sample to a separate microcentrifuge tube and store at -20 °C for use in step 3.
18. Add DNA loading dye to the remaining volume of each sample (i.e. 3 µL of 6x DNA loading dye to 15 µL of sample) and run 5 µL from each reaction on a 1% agarose gel.

NOTE: Cells that successfully amplified will appear as a smear from 250 to 1,000 bp. Samples that do not show a smear or only show a faint smear are unlikely to yield useful sequencing data.

3. Sequencing

1. Purify samples with paramagnetic beads.
   1. Thaw DNA samples on ice. Vortex paramagnetic beads until the solution is homogenous and incubate the beads at room temperature for at least 30 min.
   2. Transfer 20 µL of each DNA sample into a clean microcentrifuge tube. The remainder of the sample can be stored at -20 °C.
   3. Add 30 µL (1.5x) of paramagnetic beads to each DNA sample and vortex to mix. Incubate the samples at room temperature for 10 min. Leave the stock solution of beads at room temperature for use in step 3.4.
   4. Place the tubes on a magnetic tube strip for 2 min, or until the beads form a pellet and the supernatant is clear.
   5. Using a P200 pipette, remove as much of the supernatant as possible without disturbing the beads.
   6. Add 180 µL of 80% ethanol to the DNA-bead mixture. Rotate the tubes several times relative to the magnet to *rinse* the beads through the ethanol solution.
   7. Using a P200 pipette, remove as much of the ethanol wash as possible without disturbing the beads.
   8. Repeat 3.1.6 and 3.1.7.
   9. Air dry the beads for approximately 10 min or until no more ethanol is visible. Proceed to the next step when the beads appear cracked or fall off the wall of the tube.
   10. Remove the samples from the magnetic strip. Add 40 µL of 10 mM Tris pH 8.0 to elute the DNA from the beads and vortex the samples to resuspend the beads. Incubate for 2 min at room temperature.
   11. Briefly centrifuge the samples to collect the liquid at the bottom of the tube and place the tubes on a magnetic tube strip for at least two min. Transfer the eluent into clean microcentrifuge tubes without disturbing the beads.

2. Sample normalization
   1. Quantify the concentration of each sample using a spectrophotometer. The concentrations should range from 10 to 30 ng/µL.
   2. Dilute each sample to 0.2 ng/µL using 10 mM Tris pH 8.0. The following steps will require a minimum input of 5 µL from this dilution.

3. Library preparation
   1. Prepare sequencing libraries by following the instructions of the library preparation kit.

4. Final cleanup
   1. Clean the samples according to step 3.1. For read lengths of 50 bp, 75 bp, or 150 bp, use a ratio of bead to sample volume of 1.5x, 1x, or 0.6x, respectively. Elute with 15 µL of 10 mM Tris pH 8.0.
   2. Run the samples on a fragment analyzer to check the size distribution of the library. The size distribution should be evenly spread from 150 to 900 bp.

5. Quantify samples using qPCR
   1. Using a library quantification kit, set up a qPCR reaction according to the kit instructions. Leave a column blank to add positive standards (DNA standards from the kit) and negative standards (water or other blank solution).
   2. Thermal cycle as follows: 95 °C, 5 min (ramp rate of 4.8 °C/sec) and 35 cycles of 95 °C, 30 sec (ramp rate of 4.8 °C/sec) and 60 °C, 45 sec (ramp rate of 2.5 °C/sec).

6. Pooling
   1. Determine how many lanes on the flowcell are needed and divide the samples into groups, with one group per lane.
   2. For each group of samples, choose the sample with the lowest concentration based on the qPCR data from step 3.5.2 and normalize all samples in that group to that concentration using 10 mM Tris pH 8.0.
   3. Pool the samples in each group together.

7. Sequencing
   1. Load pools on next-generation sequencing machine following standard procedures.

4. Data Analysis

NOTE: A Unix-based environment is required to run the programs and scripts in this section. Install the software mentioned in the protocol following their installation guides. All scripts can be found at [https://sourceforge.net/projects/singlecellseqcnv/](https://sourceforge.net/projects/singlecellseqcnv/).

1. Trim the reads to 40-nt using fastx_trimmer from FASTX-Toolkit version 0.0.13.
   fastx_trimmer -Q33 -i example.fastq -l 40 -o example.trim.fastq

2. Align the reads to the appropriate reference genome (mm9 for mouse, hg19 for human) using BWA version 0.6.1 with default options.
   bwa aln mm9.fa example.trim.fastq >example.sai
   bwa samse mm9.fa example.trim.fastq >example.sam

3. Remove alignments to chrM and random chromosomes, then sort and index the resulting BAM file using SAMTools version 0.1.19.
   grep -v -w chrM example.sam | grep -v random >example.filtered.sam
4. Run HMMcopy to detect CNVs.

1. Use gcCounter in HMMcopy to generate a GC percentage reference file for the genome. Use option "-w 500000" to specify window size. Use the same version of the fasta reference file used in step 4.2, but make sure chrM and random chromosomes are removed from the fasta file.
   gcCounter -w 500000 mm9.fa >mm9.gc.wig
2. Use generateMap.pl in HMMcopy to generate a wiggle file for mappability. Use option "-w 40" to specify read length. Use the same fasta reference file used in step 4.4.1.
   generateMap.pl -b mm9.fa
   generateMap.pl -w 40 -i mm9.fa mm9.fa -o mm9.bigwig
3. Use mapCounter in HMMcopy to generate a mappability reference file for the genome. Use option "-w 500000" to specify window size.
   mapCounter -w 500000 mm9.bigwig >mm9.map.wig
4. Use readCounter in HMMcopy to generate a wiggle file for each BAM file.
   readCounter -w 500000 example.filtered.bam >input.wig
5. Modify the paths to the reference files in the R scripts (run_hmmcopy.mm9.r or run_hmmcopy.hg19.r), use the files generated above in 4.4.1 (e.g., mm9.gc.wig) and 4.4.3 (e.g., mm9.map.wig) for the variables gfile and mfile, which refer to the GC percentage reference file and mappability reference file, respectively. Then run the provided R script "run_hmmcopy.mm9.r" or "run_hmmcopy.hg19.r".
   R CMD BATCH run_hmmcopy.mm9.r
   or
   R CMD BATCH run_hmmcopy.hg19.r
6. For batch processing of a set of BAM files, put the BAM files in a single folder and run the provided script "HMMpipe.pl" in the package to call segments and compute variability scores (VS). Follow the format:
   perl HMMpipe.pl Folder_to_BAMfiles mm9

5. Run DNAcopy to detect CNVs.

1. Use SAMtools version 0.1.19 to extract uniquely mapped reads from each BAM file.
   samtools view -h -F 0x0004 example.filtered.bam | egrep -i "^@|XT:A:U" | samtools view -Shu - >example.mapped.bam
2. Use MarkDuplicates in Picard version 1.94 to mark PCR duplicates in the BAM file.
   java -jar MarkDuplicates.jar INPUT=example.mapped.bam OUTPUT=example.nondup.bam METRICS_FILE=example.dup REMOVE_DUPLECTATES=true
3. Use coveragebed in bedtools version 2.17.0 to count mapped reads in each of the predefined dynamic 500kb mappable windows in the provided BED file "mm9.500k.dynamic.win.bed" or "hg19.500k.dynamic.win.bed".
   coverageBed -abam example.nondup.bam -b mm9.500k.dynamic.win.bed -counts | sort -k1,1 -k2,2n >example.nondup.counts
4. Use the provided Perl script "normalizeGC.pl" to normalize the read counts based on the provided GC content reference file "mm9.500k.dynamic.win.fa.gc.txt" or "hg19.500k.dynamic.win.fa.gc.txt".
   perl normalizeGC.pl mm9.500k.dynamic.win.fa.gc.txt example.nondup.counts >example.bam.norm.counts
5. Use the provided R script "dnacopy.r" to call the segments.
   R CMD BATCH dnacopy.r

6. Identify CNVs

1. Exclude cells for which the VS calculated in 4.4.6 exceeds 0.26.
2. Filter the segments called by HMMcopy in 4.4.6 to include only those for which the median log\(_2\) ratio is greater than 0.4 (putative gain) or less than -0.35 (putative loss).
3. Filter the segments called by DNAcopy in 4.5.5 to include only those for which the segment mean is greater than 1.32 or less than 0.6.
4. Overlap the segments from 4.6.2 and 4.6.3, calling CNVs only in regions where both HMMcopy and DNAcopy identify a putative gain or putative loss.

**Representative Results**

The assembled aspirator should appear similar to the one in Figure 1A. The needle should be drawn such that it is sufficiently wide to accommodate a single cell but not so wide that a large volume is drawn up with the single cell. Aspirating single cells is easiest when there are between one and five cells in a 10X field (Figure 1B).

If whole genome amplification is successful, the sample will appear as a smear on an agarose gel (Figure 2, lanes 1, 2, 4, 5, 6, and 7). A faint or absent smear indicates a failed amplification reaction and the sample should not be sequenced (Figure 2, lanes 3 and 8).

Following library preparation, the fragment size distribution of the samples should be assessed by capillary electrophoresis on a fragment analyzer. Successfully prepared libraries will have a rather even distribution of fragment sizes from 150 to 900 bp (Figure 3, A,B). Failed library preparation will result in a skewed fragment size distribution and such libraries should not be sequenced (Figure 3C).

Processing sequencing data through the hidden Markov model (HMMcopy) and circular binary segmentation (DNAcopy) will parse the genome of each cell into segments of estimated copy number. These segments can then be filtered to identify those with an estimated copy number consistent with gain or loss in a single cell (Table 1). These filtered segments from HMMcopy and DNAcopy should then be overlapped to identify high-confidence CNVs.
Figure 1. Single cell isolation. (A) An assembled microaspirator. (B) A 10X field showing dissociated cells (arrows) and microaspirator needle (lower right corner). Aspirating single cells is easiest when there are between one and five cells in a 10X field. Please click here to view a larger version of this figure.

Figure 2. Whole genome amplification. Agarose gel electrophoresis of 5 µl of whole genome amplification products. Samples that successfully amplify will appear as bright smears from 100 bp to 1 kb (lanes 1, 2, 4, 5, 6, and 7) and can be sequenced. Samples that do not successfully amplify will produce faint smears or no smear (lanes 3 and 8) and should not be sequenced. Please click here to view a larger version of this figure.
Figure 3. Library preparation. Representative results from a fragment analyzer. The graphs show fragment size (in bp) on X axis and relative fluorescence units (RFU) on Y axis. To the right of each graph is a simulated gel lane. (A) Results for an ideal sample, with an even distribution between 150 and 900 bp and without sharp peaks or bias toward one side. This sample is acceptable for sequencing. (B) Results from an okay sample, with a size distribution skewed toward lower fragment sizes. While this is not optimal, the sample can still be sequenced. (C) Results from a failed sample, with predominantly small fragment sizes. This is likely caused by prolonged incubation during the tagmentation step of library preparation. This sample should not be sequenced. Please click here to view a larger version of this figure.
Discussion

Traditionally, identifying CNVs and aneuploidy at the level of single cells required cytologic methods such as FISH and SKY. Now, single cell sequencing has emerged as an alternative approach for such questions. Single cell sequencing has advantages over FISH and SKY as it is both genome-wide and high resolution. Moreover, when appropriate quality control methods are applied, single cell sequencing can provide a more reliable assessment of CNVs and aneuploidy as it is not susceptible to the hybridization and spreading artifacts inherent to FISH and SKY. However, many of the recent applications of single cell sequencing have not been substantiated by thorough assessment of the sensitivity and specificity of the methods and analyses. Indeed, some of the analytic approaches used by other studies are associated with high frequencies (>50%) of false positive CNV calls. The approach we describe has been rigorously tested using cells of known CNV burden in order to determine true and false discovery rates and optimize the sensitivity and specificity of CNV detection. Using the quality control and analytic approaches described in this protocol, approximately 20% of 5 Mb gains, 75% of 5 Mb losses, and all CNVs exceeding 10 Mb can be detected. Though determining the false discovery rate of single cell sequencing is difficult, we have estimated it to be less than 25%. This protocol can be applied to cells from a variety of sources, the scripts can be modified to adjust the resolution of CNV detection, and the protocol can be adapted to identify other types of genomic alterations.

There are a variety of means of dissociating fresh tissues into single cells, and many publications describe procedures optimized for specific tissues, such as skin and brain. We prefer to isolate single cells by microaspiration as it allows for visual assessment of each cell to be sequenced. However, it is also possible to isolate single cells by fluorescence-activated cell sorting (FACS) and microfluidic devices. If single cell isolation and whole genome amplification is performed manually, it is reasonable to isolate and amplify up to forty cells in a single sitting. In order to obtain high quality single cell sequencing data, it is crucial that the amplification of single cell genomes is uniform and complete. We find that the quality of single cells isolated as well as the efficiency of lysis and amplification has a significant impact on the quality of sequencing data. As such, the cells should be harvested from their native environment just prior to isolation and whole genome amplification should begin immediately after cells are isolated. Moreover, the lysis and fragmentation step should be followed exactly as described in the steps 2.3-2.6.

The algorithms can be adjusted to change the resolution of CNV detection, with opposing effects on sensitivity and specificity. It is also possible to adjust the thresholds to detect whole-chromosome aneuploidy in the setting of tetraploidy. However, we find that our approach is limited to detecting CNVs greater than 5 Mb, as noise introduced during whole genome amplification complicates the detection of smaller variants. Future improvements in whole genome amplification approaches should ultimately enhance the resolution of CNV detection using single cell sequencing.

Single cell sequencing allows for investigation of not only copy number alterations but also single nucleotide variations and structural variation. Our protocol for single cell isolation can be applied to answer these other questions. However, the choice of whole genome amplification method depends on the specific application. The method described in this protocol, which is based on polymerase chain reaction, is best suited for detecting copy number alterations because it is associated with lower levels of amplification bias. For investigating other types of genomic alterations, such as single nucleotide polymorphisms, other methods of whole genome amplification are believed to be more suitable.

Disclosures

The authors have nothing to disclose.

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