cfSNV: a software package for sensitive detection of somatic mutations from cell-free DNA

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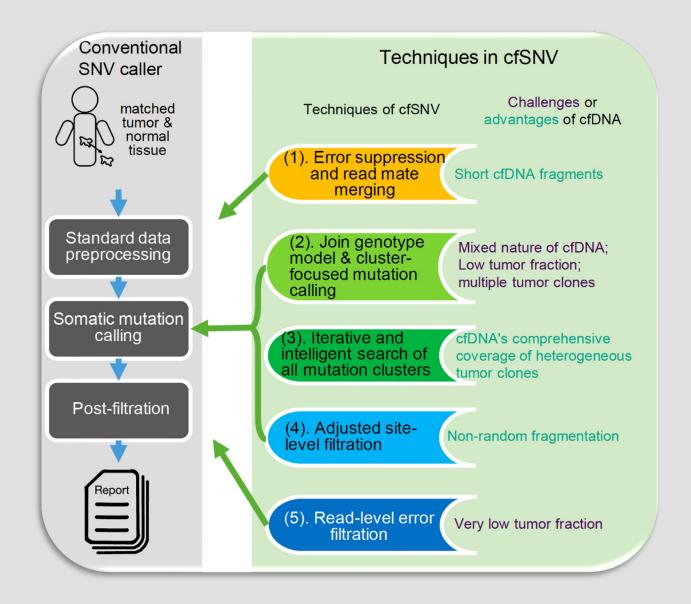
Disclosure

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■ W.L. and X.J.Z. are co-founders of EarlyDiagnostics Inc.

Challenges in single nucleotide variant (SNV) detection from cell-free DNA (cfDNA)

- Low tumor cfDNA fraction
- High tumor heterogeneity
- Existing methods cannot deal with low prevalence SNVs
 - Recall: lack of modelling for tumor content and clonal hierarchy
 - Precision: insufficient site-level statistics for error control



CfSNV

- ✓ cfSNV is an ultra-sensitive and accurate somatic SNV caller designed for cfDNA sequencing.
- ✓ Provide hierarchical mutation profiling and multi-layer error suppression
- ✓ Statistical model and machine learning approach

- o Li S, et al. Sensitive detection of tumor mutations from blood and its application to immunotherapy prognosis. Nature Communication. 2021 Jul 7;12(1):4172.
- o Li S, et al. cfSNV: a software tool for the sensitive detection of somatic mutations from cell-free DNA. Nature Protocol. 2022 Conditionally Accepted.

cfSNV workflow

Raw Data





DNA

WBC

Technique 1

Error suppression and bias correction by read mate merging Addressing cfDNA features:
Short cfDNA fragments

(Step 1) Meticulous preprocessing of cfDNA sequencing data

Merging of overlapping read pairs

Standard data preprocessing

Standard data preprocessing

Technique 2

Joint modeling of genotypes & cluster-focused mutation calling Addressing cfDNA features: Low tumor fraction and high heterogeneity

(Step 2) Iterative screening of mutation candidates following the clonal hierarchy

(2a) Estimating the mutation cluster frequency

(2b) Prediction of mutations via joint modeling genotypes in cfDNA and WBC (2d) Remove detected mutations and search for the next cluster

Technique 3

Iterative detection of all mutation clusters

Addressing cfDNA features: High heterogeneity

Technique 4

Adjusted site-level filtration Addressing cfDNA features: Non-random fragmentation (2c) Site-level filtration

(Step 3) Read-level error filtration by machine learning

Read-level filtration by a random forest model

Technique 5

Read-level error filtration

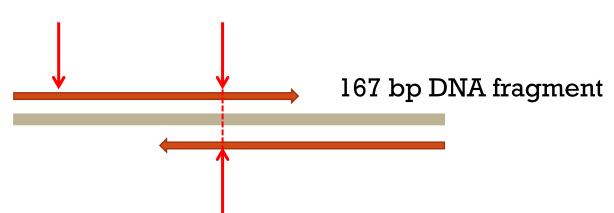
Addressing cfDNA features:
Low tumor fraction

somatic mutations

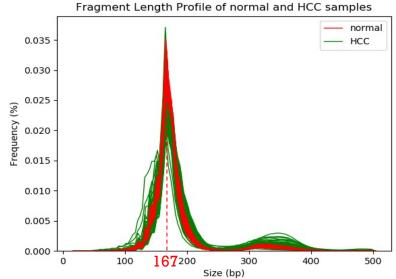
tumor fraction

Utilize overlapping read pairs for error suppression

- 1. Short fragment size
- 2. Non-random fragmentation



2 x 100 bp paired-end sequencing

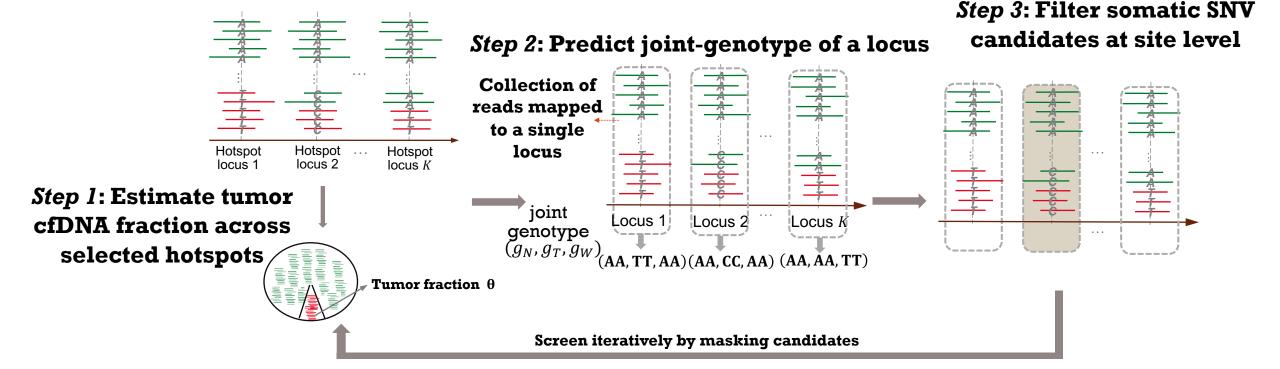


nucleosome + linker histone: 167 bp



Joint genotype modeling and iterative SNV calling

Mixed nature of cfDNA \rightarrow Joint genotype model cfDNA's coverage of all clones \rightarrow iterative SNV calling

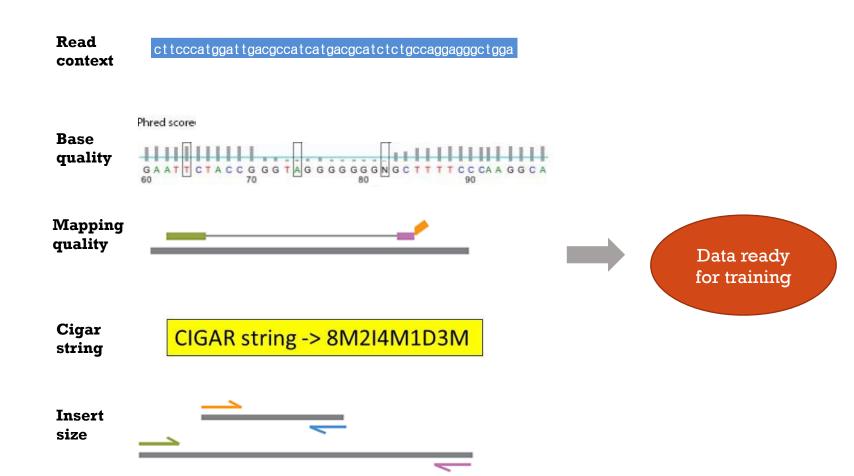


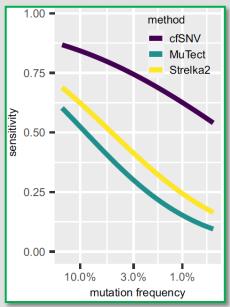
Read-level machine learning to distinguish true variants from sequencing errors for each read

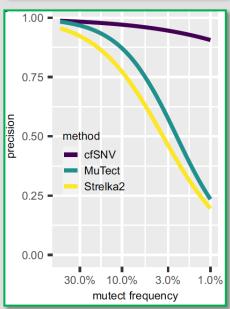
Reads with variant or error from selected loci



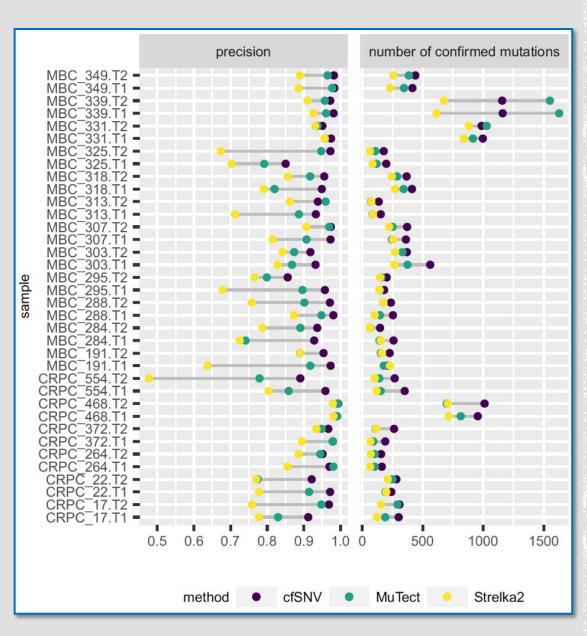
Extract features







Simulation data



Real data from cancer patients

CISNV

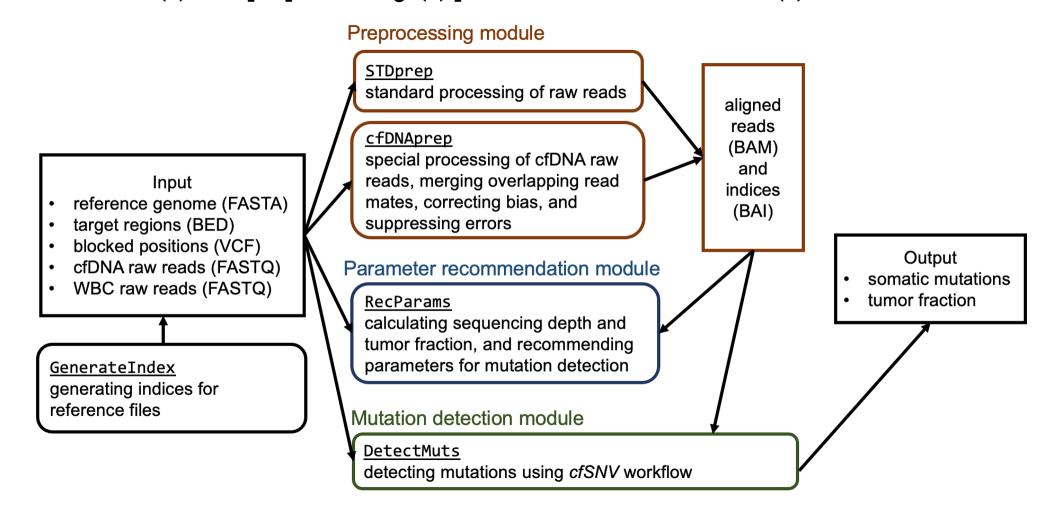
- ✓ Outperform existing tools in sensitivity while maintaining high precision.
- ✓ Improve mutation detection performance in medium-depth sequencing data, such as Whole-Exome Sequencing
- ✓ Wide-spectrum applications:
 - Cancer detection
 - Cancer monitoring
 - Therapy response prediction

Efficient implementation of cfSNV

- □ We implemented cfSNV by C++ and python, wrapped into an R package
- We built a Docker image, which is designed to enable researchers and clinicians with a limited computational background to easily carry out analyses on both high-performance computing platforms and local computers.
- Mutation calling from a standard preprocessed WES dataset (~250x and ~70 M target size) can be carried out in 3 hours on an Amazon Web Services cloud server with 8 vCPUs and 32 GB of RAM.
- ☐ It can automatically detect the statistics of the users' input data and recommend parameter settings that are tailored to the specific experimental protocol, sequencing coverage, and the tumor fraction of the dataset.
- □ cfSNV R package: https://github.com/jasminezhoulab/cfSNV
- □ cfSNV Docker image: https://github.com/jasminezhoulab/cfSNV docker

Modules of cfSNV package

Three modules: (1) data preprocessing, (2) parameter recommendation, (3) mutation detection.



Limitations and future work

- Does not support detection of somatic indels
- □ Does not consider haplotypes of somatic mutations

 Due to short size of cfDNA fragments, it might be difficult to resolve haplotypes from cfDNA
- □ The module, "error suppression in the overlapping read mates", does not support single-end sequencing data
- Does not provide data preprocessing of UMI-tagged sequencing data, due to the often customized UMI design