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# GENOMICS AND BIOINFORMATICS Article title

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# **ABSTRACT**

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#### INTRODUCTION

Cancer develops as the result of the accumulation of somatic mutations and clonal selection of cells with mutations that confer a selective advantage on the cell. Understanding the forces that shaped the evolutionary history of a tumor, the mutations that are responsible for its growth, the rate at which mutations are occurring, or how much genetic diversity is likely present in the tumor, requires accurate variant calling, particularly at low variant allele frequency (Bozic et al., 2016, Williams et al., 2016, 2018). Accurate variant identification is also critical in optimizing the treatment regime for an individual patients disease (Borad et al., 2014, Chen et al., 2013, Ding et al., 2012, Findlay et al., 2016, Mardis, 2012). Low frequency mutations present a significant problem for current mutation calling methods because their signature in the data is difficult to distinguish from the noise introduced by Next Generation Sequencing (NGS), and this problem increases as sequencing depth increases.

Methods for identifying true somatic mutations - i.e. variant calling - from NGS data are an active area of research in bioinformatics. The earliest widely used somatic variant callers aimed specifically at tumors, Mutect1 and Varscan2, used a combination of heuristic filtering and a model of sequencing errors to identify and score potential variants, setting a threshold for that score designed to balance sensitivity and specificity (Cibulskis et al., 2013, Koboldt et al., 2012). Subsequent research gave rise to a number of alternate variant calling strategies including haplotype based callers (Garrison and Marth, 2012), joint genotype analysis (SomaticSniper, JointSNVMix2, Seurat, and CaVEMan, MuClone) (Christoforides et al., 2013, Dorri et al., 2019, Jones et al., 2016, Larson et al., 2012, Roth et al., 2012), allele frequency based analysis (Strelka, MuTect, LoFreq, EBCall, deepSNV, LoLoPicker, and MuSE)(Carrot-Zhang and Majewski, 2017, Fan et al., 2016, Gerstung et al., 2012, Saunders et al., 2012, Shiraishi et al., 2013, Wilm et al., 2012), and a mixture of ensemble and deep learning methods (MutationSeq, SomaticSeq, SNooPer, and BAYSIC). All of these methods have varying levels of complexity, and some are focused on specific types of data. The one thing they all have in common is that they either implicitly or explicitly assume that the probability of a mutation occurring at a particular site is proportional to the overall mutation rate, and the same at every site in the genome.

Single nucleotide substitions, i.e. simple mutations, arise in tumors at a rate and at genomic locations driven by two main processes. The first is the spontaneous accumulation of mutations that occurs in all dividing tissues, and has a characteristic mutation signature that describes the probability of mutation in a given genomic context (Alexandrov et al., 2015, Lee-Six et al., 2018, Nik-Zainal et al., 2012). The second, and far more complex, process is the accumulation of mutations through exposure to mutagens or degradation - via mutation or deletion - of cellular machinery responsible for the identification and repair of damage or replication errors. Many mutagens and DNA repair mechanism defects also have highly specific mutation signatures, such that they can be identified by observing the mutations in the tumor (Alexandrov et al., 2016, 2013, Helleday et al., 2014, Kandoth et al., 2013, Nik-Zainal et al., 2016).

Here we present an algorithm f estimating the prior probability of mutation at a given site using the observed mutation spectrum of the tumor, and show that the addition of this prior to the MuTect variant calling model produces a superior variant classifier in both simulated and real tumor data. We then extend the method with an application of the local false discovery rate by computing the probability that a site is non-null under an assumption of clonal expansion with either early or small selective differences between clones. We provide a simple implementation in R that takes MuTect caller output as input, and returns the posterior probability that a site is variant for every site observed by MuTect.

# MATERIALS AND METHODS

#### Materials subsection one

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# 2 Nucleic Acids Research, 2009, Vol. 37, No. 12

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Materials subsubsection one. Text. T

$$LD^r = \frac{LD}{A_{iso}} = 1.5S \left(3\cos^2 \alpha_i - 1\right) \tag{1}$$

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## Materials subsection two

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$$\mathrm{LD}(t) = \sum_{i} a_{i} \exp\left(\frac{-t}{\tau_{i}}\right)$$

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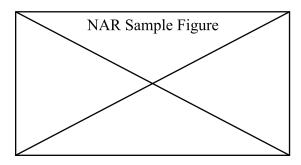


Figure 1. Caption for figure within column.

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#### **RESULTS**

### Results subsection one

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## **Results subsection two**

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Table 1. This is a table caption

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Row 2	Row 2	Row 2	Row 2	Row 2

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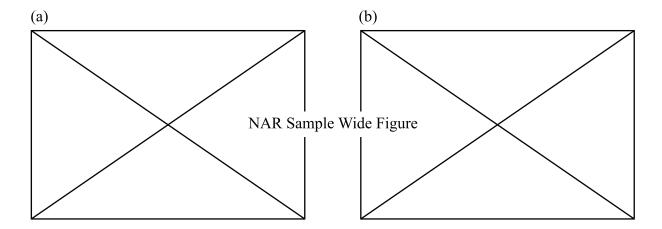


Figure 2. Caption for wide figure over two columns. (a) Left figure. (b) Right figure (see (a)).

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#### Results subsection three

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## DISCUSSION

# Discussion subsection one

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#### Discussion subsection two

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# Discussion subsection three

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#### 4 Nucleic Acids Research, 2009, Vol. 37, No. 12

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## **CONCLUSION**

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## **ACKNOWLEDGEMENTS**

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Conflict of interest statement. None declared.

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