Comparative genomic analysis of oral versus laryngeal and pharyngeal cancer


DOI
10.1016/j.oraloncology.2018.04.006

Publication date
2018

Document Version
Other version

Published in
Oral Oncology

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Citation for published version (APA):

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Supplementary Methods

Sequencing and bioinformatics protocol of NKI dataset
Genomic DNA was isolated with the AllPrep DNA Mini Kit (Qiagen). Paired-end libraries were prepared and subsequently target enriched with the TruSeq DNA library preparation kit (Illumina) and a SureSelect custom-based bait library (Agilent) targeting 556 genes (Supplementary Table 1). HPV gene baits were included in order to determine the HPV status of tumor samples. The enriched libraries were sequenced on the Illumina Hiseq 2000 using a 2x75bp paired-end protocol. The reads were aligned to the NCBI build 37 (hg19) human reference genome using the Burrows-Wheeler Aligner backtrack algorithm [1]. Potential PCR duplicates were removed with Picard Tools (http://picard.sourceforge.net). The average read coverage was 352.

We called single nucleotide variants and indels with VarScan 2.3.9 [2] in conjunction with Samtools mpileup 0.1.19 [3]. We annotated these variants with Annovar version date 11-05-2016 [4]. The bash script that was used to call and annotate the variants is available at https://github.com/dvossen/OSCC-versus-LPSCC. Variants were classified as germline variants or SPMs using public SNP databases because a matched normal sample was unavailable for most tumor samples. Variants were classified as SPMs if they (1) were absent from the 1000 Genomes Project [5], Exome Sequencing Project (Exome Variant Server, NHLBI GO Exome Sequencing Project (ESP), Seattle, WA (URL: http://evs.gs.washington.edu/EVS/) (Accessed 22 December 2014)) and Exome Aggregation Consortium [6] databases, or (2) occurred in more than 20 samples of the Catalogue Of Somatic Mutations In Cancer (COSMIC), using an R script available at https://github.com/dvossen/OSCC-versus-LPSCC. All analyses were based on these SPMs; listed in Supplementary Table 2.

Copy number and ploidy were estimated from DNA sequencing data with PureCN [7], following the protocol in its manual. We used thresholds recommended by PureCN for calling homozygous deletions (copy number ≤ 0.5) and focal amplifications (copy number ≥ 7).

Mutational signatures
Different mutational processes (e.g. DNA repair defects or tobacco exposure) cause different patterns of somatic point mutations (SPMs) throughout the genome. Such characteristic patterns are often referred to as ‘mutational signatures’ and COSMIC currently lists 30 curated mutational signatures (http://cancer.sanger.ac.uk/cosmic/signatures, 18 Dec 2017). We used the R package deconstructSigs [8] to determine which of these mutational signatures contributed to the SPMs found in each sample. A signature was classified as contributing to a tumor sample’s SPMs when its weight was at least 0.06, following deconstructSigs’ recommendations [8]. Following these, we only included samples with at least 50 SPMs in this analysis and ran deconstructSigs with default normalization.

We then focused on signatures that made a contribution to at least 25% of OSCC or L/P-SCC samples (signatures 1, 2, 3, 4, 6, 7, 13, 15, 24). For each signature, we compared the
proportion of OSCC and L/P-SCC tumor samples to which the signature made a contribution, using Fisher’s exact test. We controlled the false discovery rate at 0.10 by correcting for multiple hypothesis testing with the Benjamini and Hochberg method. The corrected P-values are reported as Q-values.

References


