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Mutational load in carotid body tumor



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Abstract

Background: Carotid body tumor (CBT) is a rare neoplasm arising from paraganglion located near the bifurcation of the carotid artery. There is great intra-tumor heterogeneity, and CBT development could be associated with both germline and somatic allelic variants. Studies on the molecular genetics of CBT are limited, and the molecular mechanisms of its pathogenesis are not fully understood. This work is focused on the estimation of mutational load (ML) in CBT.

Methods: Using the NextSeq 500 platform, we performed exome sequencing of tumors with matched lymph node tissues and peripheral blood obtained from six patients with CBT. To obtain reliable results in tumors with low ML, we developed and successfully applied a complex approach for the analysis of sequencing data. ML was evaluated as the number of somatic variants per megabase (Mb) of the target regions covered by the Illumina TruSeq Exome Library Prep Kit.

Results: The ML in CBT varied in the range of 0.09–0.28/Mb. Additionally, we identified several pathogenic/likely pathogenic somatic and germline allelic variants across six patients studied (including TP53 variants).

Conclusions: Using the developed approach, we estimated the ML in CBT, which is much lower than in common malignant tumors. Identified variants in known paraganglioma/pheochromocytoma-causative genes and novel genes could be associated with the pathogenesis of CBT. The obtained results expand our knowledge of the mutation process in CBT as well as the biology of tumor development.

Keywords: Carotid body tumor, Mutational load, Somatic variants, Germline variants, Exome, High-throughput sequencing

Background

Carotid body tumor (CBT) is the most frequent paraganglioma of the head and neck that arises from carotid glomus [1]. This tumor is highly vascularized and commonly involves carotid artery and cranial nerves. Surgery is a main method for CBT treatment, since radiation therapy and chemotherapy are not very effective.

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Tumor development is closely associated with the accumulation of somatic mutations, which may be due to various processes such as endogenous and exogenous DNA damage, defective mechanisms of DNA replication, modification, and repair [2, 3]. These cause the changes in expression profiles of many genes, including activation of oncogenes and inactivation of tumor suppressor genes that lead to alterations in signaling pathways, cellular metabolism, and proliferation [4–14]. Distinct combinations of mutation types ("mutational signatures") depend on different mutation processes; multiple mutation processes generate jumbled composite signatures. In the study of

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Alexandrov et al. (2013), more than 21 mutational signatures for 30 cancers were identified, and it has been shown that the prevalence of somatic mutations across human cancer types are different, ranging from approximately 0.04 to 20 somatic mutations per megabase (Mb) [15, 16].

In recent years, immunotherapy has been successfully used in cancer treatments. Notably, the blockade of immunosuppressive checkpoints, such as T-lymphocyte-associated antigen 4 (CTLA4), programmed cell death 1 (PD1) protein, and programmed cell death-ligand 1 (PD-L1), has demonstrated objective clinical responses in various cancers and other malignant neoplasms [17, 18]. Inhibition of both CTLA4 and PD1/PD-L1 reactivates lymphocytes against tumor-expressing neoantigens. They result from different mutations in tumor cells, and the potential of neoantigen formation is correlated with mutational load (ML) [19]. Melanoma and lung cancer, which are characterized by high MLs, were demonstrated to have clinical benefits from the immunotherapy with antibodies targeting CTLA4 and PD1 [20, 21]. Responses to immune checkpoint blockade therapy have been described in patients with colorectal cancer characterized by microsatellite instability (MSI), which is accompanied in most cases with high ML, resulting from defects in mismatch-repair pathways [22]. High ML, MSI (surrogate marker of high ML), and neoantigen production have been demonstrated to be promising markers of sensitivity to immune checkpoint blockade for several tumors [23-25]. However, these criteria also exhibited inconsistent patterns in patients with ovarian and urothelial cancer as well as those with glioblastoma; therefore, their potential use as prognostic factors requires further studies [26–28]. Additionally, intestinal microbiota, expression of PD-L1 in tumor cells, and tumor-infiltrating lymphocytes (TILs) were also determined to have a predictive role in immunotherapeutic responses [29-33].

In this study, we estimated the ML in CBT. We performed exome sequencing of tumors with matched lymph node tissues and peripheral blood derived from six patients with CBT. Additionally, a number of pathogenic/likely pathogenic somatic and germline variants were identified.

Methods

Patients and samples

Formalin-fixed paraffin-embedded (FFPE) tumor and lymph node tissues as well as peripheral blood from six patients with CBT were collected from Vishnevsky Institute of Surgery, Ministry of Health of the Russian Federation for exome sequencing. We also used a collection of 52 CBTs (exome sequencing data, available in the NCBI Sequence Read Archive [SRA] under accession number PRJNA411769) from a previous study [34]. All patients provided written informed consent. Clinicopathologic characteristics of the patients with CBT are presented in Table 1. The study was approved by the ethics committee from Vishnevsky Institute of Surgery and performed according to the Declaration of Helsinki (1964).

Exome sequencing

DNA was extracted from blood cells using a MagNA Pure Compact Nucleic Acid Isolation Kit I (Roche, Switzerland) on a MagNA Pure Compact Instrument (Roche); DNA from tumor and lymph node tissues was isolated with High Pure FFPET DNA Isolation Kit (Roche). DNA (100 ng per sample) was sheared to 150 bp using Covaris S 220 System (Thermo Fisher Scientific, USA) and was then subjected to library preparation with TruSeq Exome Library Prep Kit (Illumina, USA) according to manufacturer's instructions. The exome sequencing was performed on a NextSeq 500 System (Illumina) with paired-end reads. Read length was 76 bp for tumor and lymph node tissues and 151 bp - for blood. The obtained coverage was at least 300×. Raw sequencing data have been deposited at the NCBI SRA under accession number PRJNA476932.

Raw reads were trimmed and adapter sequences were removed with Trimmomatic [35]. We aligned 100 K randomly selected reads to bacterial genomes (NCBI, all bacterial genomes submitted up to 2014) with BWA [36] in order to evaluate contamination levels. All the samples demonstrated no greater than 0.1% bacterial DNA ratios. Next, reads were mapped to the reference human genome GRCh37.75 (Ensembl) with BWA. The derived BAM files

Table 1 Clinicopathologic characteristics of the patients with CBT

Patient	Gender	Age (years)	Family history of paragangliomas	Metastasis (lymph node/distance)	Multifocal growth	Comments
Pat100	Female	35	N/A	No	No	_
Pat101	Female	46	N/A	No	No	Tumor recurrence was diagnosed in a year after surgery*
Pat102	Female	31	N/A	No	No	_
Pat103	Male	57	N/A	No	No	_
Pat104	Female	58	N/A	No	Yes	Carotid body tumor and vagal paraganglioma were diagnosed*
Pat105	Female	67	N/A	No	No	_

* - In the study, only primary CBT from Pat101 and CBT (not vagal paraganglioma) from Pat104 were analyzed

First, we applied VarScan to reveal somatic variants in paired ('tumor *versus* normal') mode. We merged BAMs for blood and lymph nodes per each patient and submit pileups (bcftools) from these BAMs to VarScan. Reads with mapping quality lower than 20 and the bases with base calling quality lower than 20 were filtered out. Only regions with 20x or higher coverage (for both tumor and 'merged' norm) were included in the analysis.

Second, we used Mutect2 (from GATK 4.0.8.1) to identify somatic variants, SNVs and indels. Before calling somatic variants, we ran Mutect2 in 'tumor-only' mode with all the 12 normal samples (lymph nodes and blood) to create a panel of norms (PoN). Next, we merged BAMs for blood and lymph nodes per each patient and used these BAMs along with PoN to call somatic variants with Mutect2 in 'tumor versus normal' mode. The derived VCFs were analyzed with GATK FilterMutect-Calls, and only passed somatic variants were included in the further analysis. We have decided to also include in the analysis clustered events, which are filtered out with FilterMutectCalls by default. Additionally, we called variants in artificial comparisons 'lymph node (FFPE) versus blood', 'tumor (FFPE) versus blood', 'blood versus tumor (FFPE)', etc.

The derived list of somatic variants was annotated using Annovar [39]. We included allele population frequency databases (gnomAD, 1000 Genomes Project, Kaviar, ESP 6500, and ExAC), public variant databases (dbSNP, ClinVar, and COSMIC), phastCons containing conservation data for vertebrates, primates, and placental mammals [40], and InterPro to analyze the localizations of variants in protein domains [41]. Additionally, prediction tools such as SIFT [42], PolyPhen2 [43], MutationTaster [44], LRT [45], InterVar [46], PRO-VEAN [47], M-CAP [48], MetaSVM, and MetaLR [49] were used to assess the pathogenicity of the variants. Variants were considered to be likely pathogenic if they were predicted as deleterious by at least three algorithms. However, in most cases the majority of algorithms gave consistent results.

We excluded variants with population frequency greater than 1%. Worth noting, the overall number of such variants comprised only 0-7% of all the exonic somatic variants (variants in gene coding regions) passed after analysis with FilterMutectCalls. Additionally, the list of somatic variants was filtered according to the minimal read coverage threshold (min 20 reads for a merged norm and min 10 reads for the tumor sample).

Results

Mutational load in CBT

First, we should mention once more that we used blood and FFPE samples taken from tumor and lymph node tissues. The accurate detection of variants in FFPE samples is often problematic because of DNA fragmentation and the occurrence of sequence artifacts resulted from fixation of tissues in formaldehyde [50]. To evaluate the effect of FFPE artifacts on the results, we compared 'lymph node (FFPE) versus blood' and revealed a great number of variants (hundreds) that was almost equal to the number of somatic variants found in the 'tumor (FFPE) versus blood' comparison (including variant with low alternative allele (AF) frequency). In contrast, when comparing either 'blood versus tumor (FFPE)' or 'blood versus lymph node (FFPE), a very moderate number of variants (dozens) was revealed. This pronounced trend was observed for all six patients. It suggests that the most of variants identified in FFPE samples may be formalin-induced DNA artifacts. However, there were a variety of SNVs (A > T, C >A, G > A, A > G, etc.); and only a moderate bias towards the typical FFPE-induced transition, C > T, was observed. About 25% of all SNVs were represented with C > * substitutions, and approximately 40-60% of them were C > T transitions (10–15% of all SNVs).

When 'tumor *versus* lymph node' was compared, e.g. two FFPE samples, we derived about 1.3–1.5-fold lower amount of somatic variants relatively 'tumor (FFPE) *versus* blood' comparison, because read coverage for FFPE lymph node samples was 2-fold greater (on the average) than the coverage for blood samples. Additionally, this may suggest slight co-occurrence of formalin-induced variants, which are partially 'subtracted' when comparing two FFPE samples.

Among Mutect results, there are several false-positive somatic variants with low AF (5-10%) but high coverage (10-30 reads for alternative allele; the total coverage was at least 300x) observed in tumor, lymph node and blood samples. These variants obviously are neither germinal (too low AF), nor somatic (they are present in norms and tumors), nor FFPE artifacts (they are present in blood samples). Remarkably, 20-40% of reads that support the alternative alleles (AA) have low base quality score (before and after BQSR). This strongly suggests the presence of context-dependent sequencing errors that are not eliminated with BQSR procedure and are not sufficiently addressed by a variant calling algorithm (either by Mutect2 or VarScan). Indeed, we observed many of these variants in polyN-tracts (especially polyG), e.g. GGGT>GGGG, CCCCG>CCCCC, susceptible for Illumina NextSeq-specific sequencing errors.

To eliminate such artifacts, we applied three approaches. First, we used strand bias filter (StrandOddsRatio annotation provided with GATK Mutect2 or

HaplotypeCaller). The presence of reads bearing a variant on only one strand indicates a false-positive. However, this option alone does not allow eliminating artifacts when they come from both strands (e.g. GGGTGGG>GGGG GGG). Second, we filtered out variants with abnormal distribution of base quality scores across the reads (GATK BaseQualityRankSumTest annotation). In other words, we excluded variants that were abnormally supported with too many reads with low base quality score at a current position, even if there were also many reads with high base quality score at this position. Third, we manually filtered out variants that were observed after/before four or more identical nucleotides (mainly, C/G). In most cases, when we are calling somatic variants on various types of cancer, their number may significantly exceed the ratio of the described false-positives. In contrast, when analyzing tumors with a low ML, this issue becomes especially important.

Totally, we found 70–130 potentially somatic variants in each patient comparing 'tumor (FFPE) *versus* matched lymph node (FFPE) and blood + PoN'. Most of these variants have low AF (Fig. 1). Only 5–20% of variants passed threshold of AF > 15%.

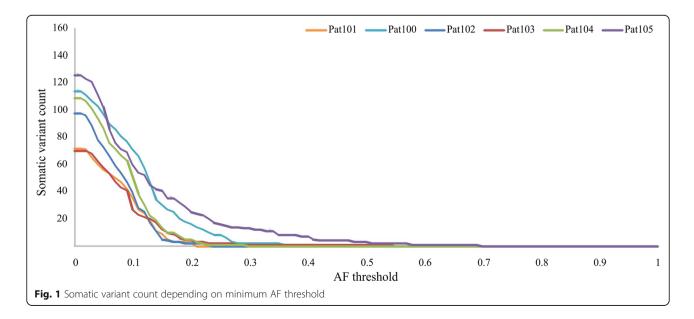
When calculating the total mutational load, it is incorrect to consider a somatic variant that is observed only in a fraction of the tumor cells (such variants are featured with low AF) as a "whole one". Otherwise, the higher sequencing coverage, the more we can find variants with very low AF values, and the higher calculated mutational load will be. The weight of such somatic variants should be adjusted for AF. Hence, we re-estimated the number of variants "in terms of heterozygous ones" as the sum of all AFs multiplied by 2 (Fig. 2, weighted somatic variants count). As can be seen from Fig. 2, two of six patients demonstrated an elevated number of potentially somatic variants. These differences are much more pronounced when an elevated AF threshold is set (AF > 0.2, e.g. > 40% of cells are for heterozygous variant). Nevertheless, it is worth noting that the initial number of tumor cells in the tumor samples did not reach 100%, and it could vary from 70 to 90%.

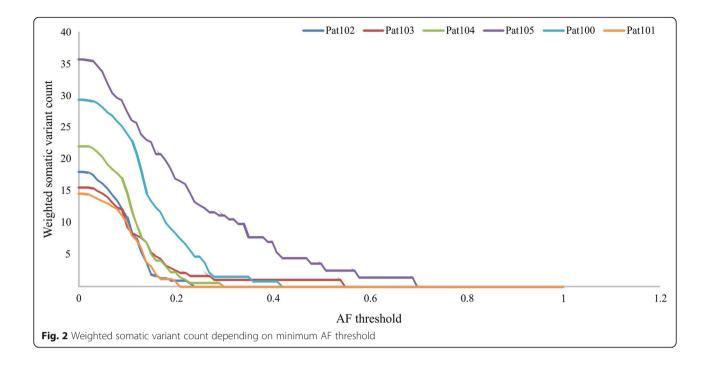
To finally assess the weighted mutational load (wML) in CBT, we should reasonably pick up the AF threshold. For most patients, a significant reduce in the number of variants occurs in the region of AF = 10-15%. Therefore, wML may be estimated as 4-12 variants per genome or 0.09-0.28 variants per megabase taking into account the fact that we have used TruSeq Exome Library Prep Kit (the total length of target regions is 43 Mb). It should be noted that the evaluation of ML in such cases is close to the limit of sensitivity/specificity of the method.

Considering the structure of a list of potential somatic variants, 25–73 variants (SNVs and indels) are located either in coding regions or splice sites and are supported with at least three reads corresponding to the alternative allele (Fig. 3). Remarkably, 2–18% of these ones have already been annotated in COSMIC databases, and only 2–7% of variants have maximal population frequency (across multiple databases) greater than 1% (before filtering). The total number of all variants (including UTRs, intronic, and intergenic) was 2-fold greater on the average than the number of variants in the coding regions.

Pathogenic and likely pathogenic somatic variants

Across six patients, we revealed 50 likely pathogenic variants, and among them several potential driver variants were observed (Fig. 4). In two patients (Pat103

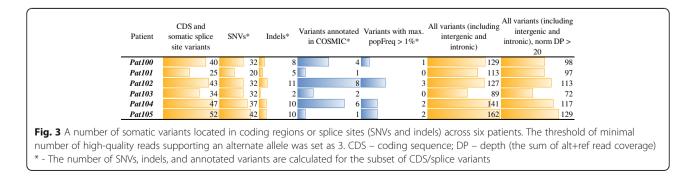




and Pat104) we found two co-occurred variants in *TP53* gene, NM_000546.5: c.842A > T, p.Asp281Val (chr17: 7,577,096, rs587781525) and NM_000546.5: c.A170A > G, p. Asp57Gly (chr17: 7,579,517). The first variant was described in dbSNP as both germline and somatic one, and has a pathogenic clinical significance according to the ClinVar database. The germline variant was associated with hereditary cancer-predisposing syndrome; the somatic variant has been found in many neoplasms, including neuroendocrine tumors (neuroblastoma and glioblastoma). Variant NM_000546.5: c.A170A > G, p. Asp57Gly (chr17: 7579517) was not previously described in databases and in the literature. In our study, this variant was found with high AF (0.55 and 0.23 for Pat103 and Pat104, respectively) and was classified to be likely pathogenic by most prediction tools. These two variants can be associated with biallelic inactivation of TP53 gene and are involved in the pathogenesis of CBT.

Potential driver variant NM_003002: c.A1A > T, p. Met1Leu (chr11: 111,957,632, rs104894307) in *SDHD* gene was revealed for Pat103. It was annotated in dbSNP and ClinVar databases as germline pathogenic variant associated with hereditary cancer-predisposing syndrome and pheochromocytomas/paragangliomas. This somatic variant was also described as pathogenic in sporadic paraganglioma [51].

We also observed somatic variants in several known cancer-associated genes (for example, *JAG1, PRDM2, PRDM8, SETD2, ASPM, ZIC, GRIK1,* etc.), which may be important for cell growth and proliferation. They did not overlap between the patients. These variants have low AF values (5–10%) occurring in low fraction of cells (10–20%), but they may represent driver events. This demonstrates genetic intra-tumor heterogeneity of CBT. Apparently, they occurred after tumor onset and initial progression. Having received these variants, such cells could gain an advantage in their clonal expansion.



						1	Гumo	r	Me	rged	norm		Ef	fect	on p	orote	in fu	metio	on				
Chr	Position	Ref.	Alt.	Gene	dbSNP	Alt. cov	Ref. cov	Alt. ratio	Alt. cov	Ref. cov	Alt. ratio	max Pop Freq	SIFT	Polyphen2 HDIV	Polyphen2 HVAR	PROVEAN	MetaSVM	MetaLR	M-CAP	phastCons 100-way vertebrate*	phastCons 20-way mammalian*	ClinVar	COSMIC
Pat100													_										
10	118687318		Α	SHTN1		3		0.13	1		0.02	0		D	D	D	Т		D	1	0.785		
14	54417376		G	BMP4		5		0.03	0		0.00	0		D	D	D	D		D	1	1		
2	131520986		G	AMER3	2 (2 (2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	8		0.15	1		0.01	0		D	D	N	Т			0.312	0.289		0000 (100010)
20	36769704		A	TGM2	rs368689999	5		0.06	0		0.00	0		D	D	D	D		D	1	0.556		<< COSM1229101
3	50339586		T	HYALI	rs104893743	5		0.03	0		0.00	0		D	D	D	D	_	D	1		<< Deficiency of	large intestine
4	38799482		A	TLR1	rs143619350	4		0.04	0		0.00	0		P	B	D	Т		T	1		iyaluronoglu-	<< COSM1055140
4	166021831	Α	Т	TMEM192		3		0.04	0		0.00	0		D	Р	D	Т		Т	1	L	osaminidase	large intestine,
6	39283003	Α	С	KCNK16		18	135	0.12	0	245	0.00	0	D	Р	В	Ν	Т	Т	Т	0.088	0.107		endometrium
Pat101				1										_					_	0	0		
1	114945479		G	TRIM33		3		0.04	0		0.00	0	_	D	D	D	D		D	1	1		
11	116744223	G	Α	SIK3		4		0.11	0	108		0		D	Р	Ν	Т		D	1	1		
12	110656018	G	Α	IFT81	rs756693673	4	17	0.19	0	36	0.00	0	D	D	Р	Ν	Т		Т	1	0.661		
19	35993779	G	Α	DMKN		3	20	0.13	0	76	0.00	0	D	D	Р	D	Т	Т	D	0.362	0.17		
2	25964912	G	Т	ASXL2		3	15	0.17	0	35	0.00	0	D	D	D	Ν	D	D	D	1	0.735		
Pat102	2																			0	0		
21	31062171		Т	GRIK1		8		0.05	0	250	0.00	0	D	Р	Р	Ν	D	D	D	1	0.785		
3	52559100	С	Т	NT5DC2		4	28	0.13	0	70	0.00	0	D	D	D	D	Т	Т	D	1	0.482		
Pat103	3																			0	0		
1	152185767	С	Т	HRNR		3	43	0.07	0	107	0.00	0	D	D	Р	Ν	Т	Т	Т	0.088	0.018		<< COSM2156983
11	111957632	Α	Т	SDHD	rs104894307	20	189	0.10	0	337	0.00	0	D	D	D	Ν	D	D	D	0.538		< Hereditary cancer-	central nervous
12	103237494	Α	С	PAH		6	101	0.06	0	255	0.00	0	D	D	D	D	D	D	D	1	0.252 P	redisposing syndrome	
13	49075950	G	Α	RCBTB2	rs376690025	4	71	0.05	0	127	0.00	0	D	В	В	D	Т	Т	D	1	0.343		r
17	7577096		Α	TP53	rs587781525	52	219	0.19	0	333	0.00	0	D	D	D	D	D	D	D	1		< Neuroblastoma;	<< COSM562341,
17	7579517	Т	С	TP53		71	59	0.55	0	123	0.00	0	D	Р	Р	Ν	D	D	D	0.088		Chronic lymphocytic	COSM45729, COSM1649342; lung,
17	39881161	G	Α	HAP1		4	59	0.06	0	168	0.00	0	D	Р	Р	D	Т	Т	Т	0.362	0.0651	eukemia; Renal cell	large intestine, breast
2	128459433	С	Α	SFT2D3		5	158	0.03	0	283	0.00	0	D	D	D	Ν	Т	Т	D	1		arcinoma, papillary; ancreatic	- ·
3	8667975	G	Α	SSUH2		6	32	0.16	0	79	0.00	0	Т	D	Р	D	Т	Т	Т	1	0 245	denocarcinoma; Skin	
Pat104													_	_	_				_	0		quamous cell carcinoma;	
1	200956292		Α	KIF21B		4		0.06	0		0.00	0		D	D	Ν	Т		D	1		(total 19 tumors)	<< COSM902032;
11	94731996		Т	KDM4D		2		0.67	0	22		0		Р	В	D	Т			0.192	0.048		endometrium
13	74420426		Т	KLF12		3		0.02	0	252		0		Р	Р	Ν	Т			0.487	0.434		
17	7577096		Α	TP53	rs587781525	22		0.08	1	454		0		D	D	D	D		D	1		< Neuroblastoma;	<< COSM562341,
17	7579517		С	TP53		16		0.23	1	169	0.01	0		Р	Р	Ν	D			0.088	0.212	Chronic lymphocytic eukemia; Renal cell	COSM45729,
2	46605215		Т	EPAS1		4		0.05	1	165		0		D	Р	Ν	Т			0.538		arcinoma, papillary;	COSM1649342;
2	230652330		С	TRIP12		10		0.12	0	134		0		Р	Р	D	Т		D	1	1 P	ancreatic	lung, large intestin
20	10639239		G	JAG1		11		0.06	0		0.00	0		D	D	D	D		D	1	0.845 a	denocarcinoma; Skin	breast
20	31017790		Т	ASXL1	rs564841799	4		0.05	0		0.00	0		D	D	D	Т		Т	1	0.785 s	quamous cell carcinoma;	
6	70410745	T	C	LMBRD1		8		0.15	0		0.00	0		D	D	D	Т		D	1	1.	(total 19 tumors)	
6	160232781	Т	С	PNLDC1		13	- 96	0.12	0	173	0.00	0	D	D	Р	D	Т	Т	Т	1	0.488		
at105			C	DDD1/2	75/0051-0	50	00	0.10	0	100	0.00	~	-		-	-	T	T	-	0	0		
1	14106977		G	PRDM2	rs756985448	50		0.69	0		0.00	0		D	D	D	Т		Т	1	0.845		
1	157491079	G	A	FCRL5	rs142943534	4		0.08	0	64		0	-	P	B	D	Т		D	1	0.222		
1	167849838	A	T	ADCY10		10		0.03	0		0.00	0		D	D	D	Т		D	1	1		
1	197097759		T	ASPM	B/00-101	64		0.32	0		0.00	0		D	D	D	Т		D	1	1		
1	201832596		T	IPO9	rs763876238	10		0.26	0		0.00	0		D	D	D	D		D	1	0.785		
14	102469248		G	DYNCIHI		38		0.39	0		0.00			D	D	D		T		1	0.20		
2	43986107		A		rs372835795	8		0.07	0		0.00		D	D			D	D	D	0.502	0.39		
20	31473396		A	EFCAB8		3		0.04	0		0.00	0		D	D	D	D	D	D	0.593	0.488		
20	43108887		A	TTPAL	rs368005756	6		0.05	1		0.01		D	D	D	D			D	1	1		
3	47061257		T	SETD2		3		0.05			0.00		T	B	B	N			D	1	1		
3	47061259		T	SETD2		3		0.05			0.00		D	D	D	D	D		D	1	0.525		000000000000000000000000000000000000000
3	142542410		T	PCOLCE2		34		0.35	0		0.00			D	D	D	Т	T		1	0.525		COSM1693554; sk
4	94376840		C	GRID2		4		0.13	1		0.01		D			D	Т	Т		1	1		
9	74970993		С	ZFAND5	#0791440672	39		0.40			0.00			D				T		1	0.521		
9	95263292	- C	Т	ECM2	rs781440673	14	63	0.18	0	107	0.00	- 0	D	D	D	IN	1	Т	1	1	0.521		

Pathogenic and likely pathogenic germline variants in causative genes

We analyzed germline variants in 42 genes (VHL, SDHA, SDHB, SDHC, SDHD, NF1, RET, HRAS, KRAS, EPAS1 (HIF2A), ATRX, CSDE1, BRAF, FGFR1, FGFR2, FGFR3, FGFR4, FGFRL1, SETD2, ARNT, TP53, TP53BP1, TP53BP2, TP53I13, KMT2D, BAP1, IDH1, IDH2, SDHAF1, SDHAP2,

FH, EGLN1, MDH2, TMEM127, MAX, KIF1B, MEN1, GDNF, GNAS, CDKN2A, BRCA1, and BRCA2) reported previously to be involved in the development of paragangliomas/pheochromocytomas [34, 52]. Three pathogenic and two likely pathogenic germline variants were found across six patients with CBT according to the predicted algorithms and public databases (Table 2, Fig. 5). These

variants were characterized by high conservation scores (PhastCons) and had an allele frequency less than 0.01% in 1000 Genomes Project, ESP 6500, and ExAC databases.

Discussion

The frequency of somatic mutations and neoantigen production correlates with responses to immunotherapy. In high mutated cancer, the ML was considered as a prognostic factor of prolonged survival with immune checkpoint inhibitors [20, 21]. However, this association has not been observed for tumors with low ML. In the present work, we estimated the ML in CBT. CBT belongs to rare neoplasms, and it is difficult to collect a representative set of CBT samples. Moreover, matched morphological normal tissues ("conventional norm") are unavailable due to the tumor localization. We collected and performed exome sequencing of tumor samples with matched lymph node tissues and peripheral blood from six patients with CBT. This revealed actual somatic variants and allowed estimating the ML in the six samples studied. CBT carried low ML (0.09-0.28/Mb) compared to other common cancers [15]. Even though high ML is a factor for immunotherapy, it also generally indicates high aggressiveness of the tumors and correlates with increased genetic instability and poor prognosis [53-56]. CBT is usually a slow-growing tumor, and in 10% of cases, it can become aggressive and metastatic [57]. Thus, low ML in CBT is possibly associated with its non-aggressive behavior and probably indicates inefficiency of immune checkpoint blockade therapy.

Paragangliomas/pheochromocytomas can be caused by germline and somatic variants in at least 42 known genes [52]. Potential driver somatic variants were found in *SDHD* (Pat103) and TP53 (Pat103 and Pat104) genes. Likely pathogenic variants were revealed in EPAS1 (HIF2A) (Pat104) and SETD2 (Pat105) genes, and no somatic variants were observed in other known genes. However, several pathogenic and likely pathogenic germline variants in SDHB, SDHC, SDHD, and RET genes were detected. A majority of them were identified in SDHx genes that are often associated with hereditary pheochromocytomas/paragangliomas [58]. Two variants were observed in *SDHB* gene in different patients (Pat101 and Pat102): a novel missense variant NM_003000.2: c.463C > A, p.Pro155Thr (chr1: 17,354,321) and a nonsense high-impact variant NM_003000.2: c.136C > T, p.Arg46* (chr1: 17,371,320, rs74315370). The latter was described in the dbSNP and ClinVar databases as a pathogenic germline variant associated with hereditary cancer predisposition syndrome, paragangliomas/pheochromocytomas, and gastrointestinal stromal tumor [59-62]. Notably, this variant has been reported in patients with aggressive extra-adrenal paraganglioma in the chest and CBT and has been considered as a high-risk factor for malignancy or recurrence of paragangliomas/pheochromocytomas [63-66]. Indeed, the patient tested (Pat101) was characterized by the tumor recurrence that is one of the features indicating aggressive phenotype of CBT. The variant NM_003001.3: c.149G > A, p.Arg50His (chr1: 161,298,257, rs769177037) in SDHC was also found in Pat102. It was described in dbSNP as a variant of uncertain clinical significance. Germline variants in SDHC are more rarely associated with the development of paragangliomas/pheochromocytomas than variants in SDHB or SDHD [58]. It should be noted that in Pat102 we observed two somatic likely pathogenic variants in GRIK1 and NT5DC2 genes. Therefore, according to our previous data, the formation of CBT can be probably caused by the cumulative effect of several highly or not highly pathogenic variants [34]. In this particular case, it seems that the main driver is the pathogenic germline variant in SDHB gene.

The germline variant NM_003002.3: c.305A > G, p.His102Arg (chr11: 111,959,726, rs104894302) in *SDHD* was identified in two patients – Pat100 and Pat104. This variant is found in dbSNP and ClinVar databases as a pathogenic germline variant associated with hereditary cancer-predisposing syndrome, paragangliomas/pheochromocytomas, gastric stromal sarcoma, and Cowden syndrome 3 [67, 68]. This variant has been detected in malignant CBT [69]. Data on the aggressive behavior of the tumor in tested patients have not been reported; one patient (Pat104) was characterized by multiple tumors (vagal paraganglioma and CBT) with multifocal growth. In this patient (Pat104), we also found pathogenic and likely pathogenic somatic variants in *TP53* gene.

One patient (Pat103) carried the germline variant NM_020975.4: c.1946C > T, p.Ser649Leu (chr10: 43,609,994, rs148935214) in the proto-oncogene RET. It was deposited to dbSNP from the gnomAD database as a germline variant. In ClinVar, another allele was reported with conflicting interpretations of pathogenicity found in hereditary cancer-predisposing syndrome (uncertain significance) and multiple endocrine neoplasia (MEN) type 2 (uncertain significance/likely benign) characterized by medullary thyroid carcinoma, pheochromocytomas, and hyperparathyroidism [70-72]. In this patient, we also identified pathogenic and likely pathogenic somatic variants in TP53 gene (NM_000546.5: c.842A > T, p.Asp281Val (chr17: 7,577,096, rs587781525) and NM_000546.5: c.A170A > G, p. Asp57Gly (chr17: 7,579,517) that are the same in Pat104. These variants can be potential driver ones.

Interestingly, we did not reveal any pathogenic germline variants in known paraganglioma/pheochromocytoma-causative genes in the patient Pat105, which is characterized with the greatest number of somatic variants and the highest ML. However, this patient was characterized by at least two somatic variants in CBT in one known CBT-causative gene – *SETD2*. This gene

Gene	Gene dbSNP	GenBank	Coordinate	Nucleotide	Nucleotide Amino acid Genotype Predictions	Genotype	Predictions				Clinical significance
				change	change		SIFT	PolyPhen2	MutationTaster	LRT	(dbSNP/ClinVar)
SDHB	I	NM_003000.2	NM_003000.2 chr1: 17,354,321	c.463C > A	c.463C > A p.Pro155Thr Het	Het	Deleterious (0)	Deleterious (0) Probably damaging (1)	Disease-causing (<i>p</i> -values = 1)	Neutral (0) N/A	N/A
SDHB	rs74315370	NM_003000.2	SDHB rs74315370 NM_003000.2 chr1: 17,371,320	c.136C > T p.Arg46*	p.Arg46*	Het	Deleterious (0) Benign (0)	Benign (0)	Disease-causing (<i>p</i> -values = 1)	Neutral (0) Pathogenic	Pathogenic
SDHD	rs104894302	NM_003002.3	SDHD rs104894302 NM_003002.3 chr11: 111,959,726 c.305A > G p.His102Arg Het	c.305A > G	p.His102Arg	Het	Deleterious (0)	Deleterious (0) Probably damaging (0.99) Disease-causing (<i>p</i> -values = 1)	Disease-causing (<i>p</i> -values = 1)	Neutral (0) Pathogenic	Pathogenic
RET	rs148935214	NM_020975.4	rs148935214 NM_020975.4 chr10:43,609,994	c.1946C > T	c.1946C > T p.Ser649Leu Het	Het	Deleterious (0)	Deleterious (0) Probably damaging (1)	Disease-causing (<i>p</i> -values = 1)	Neutral (0) Pathogenic	Pathogenic
SDHC	rs769177037	NM_003001.3	SDHC rs769177037 NM_003001.3 chr1: 161,298,257	c.149G > A	c.149G > A p.Arg50His	Het	Deleterious (0)	Deleterious (0) Probably damaging (1)	Not disease-causing (<i>p</i> -values = 0)	Neutral (0)	Not disease-causing Neutral (0) Uncertain significance (<i>p</i> -values = 0)

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Pat100	Pat101	Pat102	Pat103	Pat104	Pat105	dbSNP ID	Class	Gene	Amino acid/nucleotide change	Impact
							missense	SDHB	p.Pro155Thr/c.463C>A	Moderate
						rs74315370	nonsense	SDHB	p.Arg46*/c.136C>T	High
						rs104894302	missense	SDHD	p.His102Arg/c.305A>G	Moderate
						rs148935214	missense	RET	p.Ser649Leu/c.1946C>T	Moderate
						rs769177037	missense	SDHC	p.Arg50His/c.149G>A	Moderate
ia. 5 Die			0	e mutation		ermline variants a	cross six carotic	l body tumor	samples	

encodes for histone methyltransferase, an epigenetic modifier with tumor suppressor functionality [73]. Mutations in *SETD2* are found in many tumors, including neoplasms of the central nervous system [74].

In the previous work, we performed exome sequencing of 52 archival FFPE samples of CBT [34]. Peripheral blood or other normal tissues were unavailable; therefore, germline variants were excluded with strong filtering using the 1000 Genomes Project and ExAC databases. We estimated the ML as the number of potentially somatic deleterious variants per megabase of coding regions. However, we derived an obviously elevated ML: the average ML was 6-8 variants per Mb. Therefore, such approach does not allow efficient elimination of germline variants. Indeed, this method excludes 96-98% or more germline variants, but the remaining 2–3% of the germline variants may significantly outnumber somatic ones. In this study, we filtered the pool of somatic variant candidates from a previous work (52 patients) using exome sequencing data on blood and lymph nodes derived in the present work (6 patients). This resulted in at least 2-fold reduction of the estimated ML, but this value was still excessively high. Thus, the use of matched normal tissues is necessary to be able to accurately estimate ML.

Recently, Roche (Switzerland) announced the AVE-NIO ctDNA Analysis Kits for personalized oncology assays [75, 76]. The AVENIO ctDNA Surveillance Kit targets frequently mutated regions across 197 genes and has been optimized for monitoring of ML in lung and colorectal cancers. This kit contains the main genes that are associated with lung, colorectal, breast, gastric, prostate, ovarian, thyroid, and pancreatic cancers, as well as glioma and melanoma according to the U.S. National Comprehensive Cancer Network (NCCN) Guidelines (https://www.nccn.org/). The kit did not include genes that have been shown to be involved in the pathogenesis of paragangliomas and pheochromocytomas, except TP53, BRCA1, and BRCA2. A panel of genes accurately reflecting the ML in CBT is also unknown. Moreover, the kit and the appropriate analysis software focus on quantitating ML basing on ctDNA sequencing, and this approach is more acceptable for malignant tumors with a high frequency of metastases, while CBT is primarily a slow-growing tumor with indeterminate potential of malignancy. Thus, whole exome sequencing, which was used in the study, is currently the only method for estimating the ML in CBT.

Conclusion

The ML varied in the range of 0.09–0.28/Mb in the analyzed cohort of patients with CBT (six individuals). Several pathogenic/likely pathogenic somatic and germline allelic variants in both known paraganglioma/pheochromocytoma-causative genes and novel ones were identified. These results improve the understanding of CBT pathogenesis.

Abbreviations

AA: Alternative allele; AF: Allele frequency; CBT: Carotid body tumor; ML: Mutational load

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Availability of data and materials

All data generated or analyzed in this study are included in the published article. The sequence data are available in the NCBI Sequence Read Archive (SRA) under the accession numbers PRJNA411769 and PRJNA476932.

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Authors' contributions

AVK, AVS, and ENL conceived and designed the work; MSF, AVK, EAP, SLK, DVK, ARZ, ALG, NVM, AVP, AAK, and AVS performed the experiments; ENL, GSK, APP, MVK, VSP, and AAD analyzed the data; AVS, AVK, BYA, ADK, and ENL wrote the manuscript. All authors read and approved the final manuscript.

Ethics approval and consent to participate

The study was approved by the ethics committee from Vishnevsky Institute of Surgery and performed according to the Declaration of Helsinki (1964).

Consent for publication

Not applicable.

Competing interests

The authors declare that they have no competing interests.

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