

Prognosis model of colorectal cancer patients based on *NOTCH3*, *KMT2C*, and *CREBBP* mutations

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Background: Colorectal cancer (CRC) is one of the most common cancers. The aim of our study was to explore its related mutations, identify novel mutation markers, and construct predictive models for postoperative CRC patients, so as to provide evidence for the diagnosis, treatment, and prognosis of CRC.

Methods: A total 50 CRC patients were collected, and the mutations in tissue samples were detected through next-generation sequencing (NGS). Meanwhile, 246 CRC cases with complete mutation data were downloaded from The Cancer Genome Atlas (TCGA) database. Afterwards, the co-mutations in both clinical and TCGA cohorts were identified, and the high-frequency mutation genes were selected. Subsequently, functional enrichment analysis was performed, and overall survival (OS) and progression-free survival (PFS) predictive models were constructed.

Results: In all, 18 out of 238 co-mutation genes mutated in at least 20% of the samples and were selected out as common high-frequency mutation genes. They were significantly enriched in 460 Gene Ontology (GO) terms and 87 Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways (P<0.05), which were closely related to the occurrence and development of CRC. Among the 18 genes, *NOTCH3*, histone lysine methyltransferase 2C (*KMT2C*), and cAMP-response element binding protein-BP (*CREBBP*) were respectively associated with tumor position, stage, and PFS (P<0.05), and could be considered as potential biomarkers of CRC. Finally, OS and PFS predictive models were constructed and verified using the 50 clinical cases, with both models demonstrating high fitting degrees useful for predicting the OS and PFS of CRC patients.

Conclusions: *NOTCH3*, *KMT2C*, and *CREBBP* were found to be prospective biomarkers for the diagnosis and prognosis of CRC. The prognosis prediction models had high sensitivity and could be used to predict the OS and PFS of CRC patients.

Keywords: Colorectal cancer (CRC); mutation; prognosis; predictive model

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Introduction

Colorectal cancer (CRC) is one of the most common cancers in the world. According to the latest global cancer epidemiological statistics, new cases of CRC account for 10.2% of all malignant tumors, ranking third among all cancers, and the total number of deaths account for 9.2%, ranking second, with the proportion of CRC continuing to rise (1,2). In 2020, a projected 150,000 new CRC cases and more than 50,000 CRC-related deaths will occur in the United States (3). According to the current data published in China, in 2015, the number of new cases of CRC reached

388,000, with 225,000 of these cases being men. At present, the goal of CRC screening in China involves improving the screening and detection rates of early CRC and important precancerous lesions (4). Despite the progress in the diagnosis and treatment of the disease, the prognosis of CRC patients is still poor due to the late stage of the initial diagnosis and the high frequency of metastasis and recurrence. Therefore, it is necessary to develop an effective method to improve the diagnosis rate, predict metastasis and recurrence, and monitor the curative effect in real time, so as to improve the overall cure level (5). A thorough understanding of the molecular genetic characteristics of CRC is the key to solving this problem.

CRC develops through a series of differentially expressed or mutated genes which affect the homeostasis of oncogenes or tumor suppressors (6). In recent years, the identification of CRC tumor markers has seen rapid progress. Changes in non-coding RNA (ncRNA) have been confirmed as a key factor in the development of CRC (7). A variety of ncRNAs, including long non-coding RNAs (lncRNAs) and microRNAs (miRNAs), have recently been found to have functional features in the development of CRC (8-11). With the swift emergence of sequencing technology, the somatic mutations of a large number of genes, including TP53, KRAS, PIK3CA, APC, and RNF43, etc., have been proven as drivers in the development of CRC (12). Hence, the current study combined clinical samples and data from The Cancer Genome Atlas (TCGA) database to identify somatic mutations in postoperative CRC patients, and analyzed their correlation with clinical parameters. Finally, a prognostic model was constructed by regression analysis, and the clinical cohort was used as a verification group to further evaluate the prognostic ability of the model in patients with CRC. We present the following article in accordance with the MDAR reporting checklist (available at http://dx.doi.org/10.21037/jgo-21-28).

Methods

Patient enrolment and sample collection

From January 2017 to October 2019, 50 CRC patients were enrolled in this study from Tianjin Medical University Cancer Institute and Hospital, including 23 colon cancer cases and 27 rectal cancer cases. Subsequently, their tumor and paracancerous tissues were collected and sequenced through next-generation sequencing (NGS), and their clinical information was also recorded. This study was approved by the ethics committee of Tianjin Medical University Cancer Institute and Hospital (LLSP2019-016). All participants voluntarily signed informed consent to participate in the study. All procedures performed in this study involving human participants were in accordance with the Declaration of Helsinki (as revised in 2013).

TCGA data screening

Mutation data of CRC patients were downloaded from TCGA database (https://tcga-data.nci.nih.gov/docs/ publications/tcga/). The screening criteria were the following: (I) diagnosed as CRC with the pathological type of adenocarcinoma; (II) colon and rectal tumor sites, with a ratio of 23:27; (III) a tissue sample type with complete mutation data; and (IV) complete and detailed clinical information. Finally, a total of 246 cases were enrolled, comprising 110 colon cancer and 136 rectal cancer samples.

Mutation detection

Single-nucleotide variations (SNVs) and insertions/ deletions (InDels) were identified with VarScan version 2.4.3, MuTect version 1.1.4, and Genome Analysis Toolkit (GATK) version 2.3.9. CONTRA version 2.0.4 was used for copy number variations (CNVs) detection. An independently developed fusion program was used to detect gene fusion.

The mutation frequency of each genes in 50 clinical cases and 246 TCGA cases was counted by a self-developed Python script, and the common mutation genes were screened. Finally, the genes with a mutation frequency greater than 20% were selected as common high-frequency mutation genes.

Functional enrichment analysis

Gene Ontology (GO) terms and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analysis were performed using clusterProfiler package (v. 3.14.3) in R v.3.6.3 (R Foundation for Statistical Computing, Vienna, Austria), and the database of human reference genome sequence hg19 in org.Hs.eg.db package (v. 3.10.0) was used as the reference data. The P value was corrected by the Benjamin–Hochberg (BH) method, with a P value <0.05 and a q value <0.05 being the cut-off criteria.

Establishment of predictive models and statistical analysis

According to the mutations, overall survival (OS) and progression-free survival (PFS) predictive models were constructed using a multiple linear regression function Im() in R (v. 3.6.3). Statistical analyses were performed using Chi-square test or Fisher's exact test for categorical variables, and by *t*-test or Mann–Whitney U test for continuous variables through IBM SPSS Statistics v.21 software (IBM Corp., Armonk, NY, USA). A P value <0.05 was considered statistically significant.

Results

Clinical characteristics

Clinical characteristics of 50 clinical CRC patients are listed in *Table 1*. In TCGA cases, there were 130 (52.85%) males and 116 (47.15%) females, ranging in age from 31 years to 90 years (65.85±12.58). Furthermore, 129 cases were in stages I–II, 108 were in stages III–IV, and 9 had no recorded stage. In all, 200 survived and 46 died.

Screening of bigb-frequency mutation genes

In TCGA cohort, 42,514 mutations were found in 16,378 genes. In clinical cases, a total of 1465 mutations with a frequency $\ge 0.5\%$ in 255 genes were detected, in which TP53 p.Arg342* (22.09%), FBXW7 p.Arg658* (19.71%), BRAF p.Asp594Gly (17.33%), PIK3CA p.His1047Arg (16.67%), and NRAS p.Gly12Val (15.52%) had higher mutation frequencies. A total of 238 co-mutation genes were found in both clinical and TCGA cases. Afterwards, 18 genes with mutation frequency ≥20% were selected as highfrequency mutation genes; among them, TP53, ARID1A, and APC had a mutation frequency over 50%. Figure 1 shows the mutation distribution of 18 genes in all clinical samples, including 191 missense (67.97%), 38 nonsense (13.52%), 32 coding sequence (CDS) InDels (11.39%), 13 frameshift deletions (4.63%), and 7 frameshift insertions (2.49%).

Functional and pathway enrichment of high-frequency mutation genes

After enrichment analysis, the 18 genes were found to be enriched in 460 GO terms and 87 KEGG pathways. The GO terms included 419 biological processes (BPs), 10 cellular components (CCs), and 31 molecular functions (MFs). BPs were mainly enriched in histone modification, covalent chromatin modification, regulation of neuron apoptotic process, and regulation of neuron apoptotic process. CCs were enriched in mixed-lineage leukemia protein 3 (MLL3)/MLL4 complex, nuclear chromatin, lamellipodium, extrinsic component of membrane, etc. MFs were enriched in 1-phosphatidylinositol-3kinase activity, phosphatidylinositol 3-kinase activity, histone methyltransferase activity (H3-K4 specific), and phosphatidylinositol kinase activity. Figure 2A lists the top 10 BPs, MFs, and all CC terms according to P value. KEGG pathways of 18 high-frequency mutation genes were significantly enriched in pancreatic cancer, breast cancer, CRC, and FoxO signaling pathway, thyroid hormone signaling pathway, cell cycle, Wnt signaling pathway, and others (Figure 2B).

Clinical significance of the mutation genes

Next, we analyzed the correlation between high-frequency mutation genes and clinical characteristics, including tumor position, stage, recurrence, metastasis, OS, and PFS. Among the 18 genes, *NOTCH3* was significantly correlated with the tumor positions of CRC (P=0.021, *Figure 3A*), histone lysine methyltransferase 2C (*KMT2C*) with stage (P=0.042, *Figure 3B*), and cAMP-response element binding protein-BP (*CREBBP*) with PFS (P=0.015, *Figure 3C*). The mutations of *NOTCH3*, *KMT2C*, and *CREBBP* might play a potential role to identify the tumor location, cancer stage and PFS

Predictive models of prognostic index

According to the 18 mutated genes, OS and PFS prediction models were constructed by multiple regression analysis, and the significance test of regression equations were examined by F test. The P values of OS and PFS predictive models were 0.006 and 0.013 respectively, indicating that the test of regression equations was significant. Comparing the OS and PFS of clinical patients respectively with the predictive models, both of the predictive models had high fitting degrees (*Figure 4A*,*B*), and could be used to predict the OS and PFS of CRC patients.

Discussion

The ability to predict the prognosis of patients has great clinical value, as it can prolong the survival time and

Table 1 Clinical characteristics of 50 colorectal cancer patients

Table I Chincar characteristics of 50 cor	orectar cancer patients
Parameters	N [%]
Sex	
Male	30 [60]
Female	20 [40]
Age	
<60 years	27 [54]
≥60 years	23 [46]
Tumor stage	
Ι	1 [2]
II	12 [24]
III	23 [46]
IV	1 [2]
None	13 [26]
Pathological type	
Adenocarcinoma	50 [100]
Smoking	
Yes	16 [32]
No	34 [68]
Drinking	
Yes	24 [48]
No	26 [52]
Tumor size (cm)	
<5	15 [30]
≥5	33 [66]
None	2 [4]
Tumor volume (cm ²)	
<100	20 [40]
≥100	28 [56]
None	2 [4]
Tumor differentiation	
Moderate	33 [66]
Low	6 [12]
Middle-low	6 [12]
None	5 [10]

 Table 1 (continued)

Table 1 (continued)	
Parameters	N [%]
Hypertension	
Yes	11 [22]
No	39 [78]
Diabetes	
Yes	1 [2]
No	49 [98]
Therapeutic method	
Surgery	28 [56]
Chemotherapy	2 [4]
Surgery + chemotherapy	20 [40]
Recurrence	
Yes	5 [10]
No	45 [90]
Metastasis	
Yes	12 [24]
No	38 [76]
Overall survival (OS)	
<6 months	19 [38]
≥6 months	31 [62]
Progression-free survival (PFS)	
<3 months	13 [26]
≥3 months	2 [4]
None	35 [70]

improve the quality of life of patients, and can be used to adjust follow-up management. In CRC, the combined application of multiple markers can improve the accuracy of CRC screening and diagnosis, thus improving the diagnostic efficiency of gene detection (13). In this study, we found 18 high-frequency mutation genes (mutation frequency $\geq 20\%$) in clinical CRC patients and TCGA patients, including two common cancer related genes, *TP53* and *KRAS*. Combined with clinical characteristics, we further screened out *NOTCH3*, *KMT2C*, and *CREBBP* as candidate markers for the diagnosis and prognosis of



Figure 1 Mutation landscape of 18 common high frequency mutation genes in The Cancer Genome Atlas (TCGA) and clinical cases. The upper side shows the details of clinical information of each patient. The lower panel illustrates the genetic alterations, with the different colors representing different mutation types. The histogram on the right shows the frequency of mutation genes in 50 clinical patients.

CRC patients. Meanwhile, the predictive models based on the mutation genes were proven, through comparison with the OS and PFS of clinical cases, to have high reliability in predicting the OS and PFS of CRC patients.

NOTCH3 is one of the important members of the NOTCH family, and is involved in the occurrence and development of various cancers by regulating tumor microenvironment, and promoting tumor formation, progression, angiogenesis, migration, and invasion (14,15). Its overexpression activates Notch signaling pathway, and promotes tumor cell growth and migration (16). The abnormal activation of Notch pathway has been detected in acute lymphoblastic leukemia, gliomas, CRC and other tumors, and has been significantly correlated with prognosis (16,17). Although few studies on NOTCH3 mutation in cancers have been published, 199 mutations of NOTCH3 gene are reported to be present in malignant tumors of the lung, breast, gastric system, prostate, and lymphoma in the COSMIC database (https://cancer.sanger.ac.uk/cell_lines/ search?q=NOTCH3#muts). Despite not applied in clinical by now, Notch3 targeting has been proved as an effective way against cancer, such as anti-NOTCH3 and targeted miRNAs (18-20). Therefore, the mutations of NOTCH3 gene might play a role in the occurrence and development of these tumors.

Histone lysine methyltransferase 2C (KMT2C), also known as myeloid/lymphoid or mixed-lineage leukemia protein 3 (MLL3), encodes a nuclear protein with histone methylation activity and participates in transcriptional synergistic activation. As an important regulator of epigenetics, KMT2C participates in the methylation of various histone amino acid sites, changing the structure of chromatin and affecting the transcription process of target genes, and is thus an attractive drug target for cancer treatment (21-23). KMT2C mutates in a variety of human cancers and is considered to be crucial to the occurrence and development of cancers. However, the research on its mutation function is still limited, which may be related to the lack of a mutation hotspot and mutation domain in KMT2C (22). Interestingly, KMT2C has been previously reported to have a well-established hotspot mutation, S338L (31%), in CRC, which may lead to new epigenetic insights into the carcinogenesis of CRC (24,25). In this study, although no KMT2C hotspot mutation was found in CRC samples, nine KMT2C mutation sites were screened out with a frequency more than 1%, and the KMT2C gene

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Figure 2 Enrichment analysis of high-frequency mutation genes. Top 10 enriched BPs, CCs, and MFs in GO terms (A), and the top 30 KEGG pathways (B) of 18 common high-frequency mutation genes among TCGA and clinical cases. The colors of the circles indicate *P* values of terms and pathways, and the size represents the number of genes enriched in each term and pathway. BP, biological process; CC, cellular component; MF, molecular function; GO, Gene Ontology; KEGG, Kyoto Encyclopedia of Genes and Genomes.



Figure 3 Correlation analysis between gene mutations and clinical characteristics. *NOTCH3* (A), *KMT2C* (B), and *CREBBP* (C) mutations were respectively correlated with the tumor position, stage, and progression-free survival (PFS) of colorectal cancer (CRC) patients.

was enriched in the GO term of histone modification. More in-depth study is necessary to understand the mechanism and function of these KMT2C mutations in CRC. The cAMP-response element binding protein-BP (CREBBP) gene encodes the CREBBP protein that binds to the cAMP response element (CRE). CREBBP acts as a transcription factor and plays a role in transcription by participating in chromatin remodeling and helping RNA aggregation, which is known to underlie general cancer pathogenesis (26,27). Recent research has found that mutations in CREBBP are associated with poor prognosis in head and neck squamous cell carcinoma (HNSCC), and synthetic cytotoxicity has been identified in CREBBP mutant tumors (28). Kim et al. (29) found a frameshift mutation of CREBBP in microsatellite instability-high (MSI-H) gastric cancer, which could lead to the premature stop of amino acid synthesis in CREBBP protein; however, they also found that the mutation rate of CREBBP was very low (1.4%) in gastric cancer and CRC patients with MSI-H. In our study, CREBBP mutation occurred in 43.14% of clinical patients, which was very different from that of TCGA patients (6.91%) and of previous research. These differences can perhaps be attributed to a few factors. (I) The above study by Kim et al. only analyzed the gene mutation frequency in microsatellite instability-high (MSI-H) patients, while our study included all CRC patients and ignored MSI status, and thus might have found a higher mutation frequency. Additionally, (II) most of the TCGA data are from Caucasians and African Americans, while our study included Asians, and thus the variability in ethnicity, environment, and lifestyle might also have resulted in a difference of mutation frequency. Therefore, this issue needs to clarified by further research.

Finally, we developed predictive OS and a PFS models based on the mutated genes, and the predictive ability of the models were validated in the validation cohort, with a good predictive accuracy for OS and PFS in CRC patients. The predictive models represent a significant advance in the prognostic monitoring of CRC. Indeed, another recent study has reported that the prediction of CRC using genetic markers is feasible (30). Although there have not been appropriate drugs targeted to *NOTCH3*, *KMT2C*, and *CREBBP*, the three-gene model still become a useful tool to predict CRC patients prognosis.

Figure 4 The predictive ability of 18-gene prognostic models for colorectal cancer (CRC). Predictive models of the overall survival (OS) (A) and progression-free survival (PFS) (B) of CRC patients were constructed based on the 18 high-frequency mutation genes. The X-axis represents the predictive days, and the Y-axis represents the real days.

This study has some limitations which should be addressed. First, our current research was retrospective in nature. Although the OS and PFS models were verified in the validation cohort, the bias inherent in this study could not be completely eliminated. Second, the sample size used for this analysis was relatively small (fewer than 100 cases), which might have led to deviation. Finally, this study included only CRC patients in our hospital, whose baseline characteristics may differ from those in western countries. Therefore, it is not clear whether our current prognosis models are directly applicable to populations with different ethnic composition. In the further study, we will focus on the Asian populations and collect more multi-center CRC cases to improve the predictive model.

Conclusions

Our study identified 18 common high frequency mutations in clinical and TCGA CRC cases, with *NOTCH3*, *KMT2C*, and *CREBBP* constituting a potential novel signature for the diagnosis and prognosis of CRC. Based on this signature, we constructed predictive OS and PFS models which were preliminary proven to be highly reliable. However, the findings of our study and the related mechanisms need to be thoroughly validated and explored in a larger cohort study.

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Footnote

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Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was approved by Tianjin Medical University Cancer Institute and Hospital (LLSP2019-016). All participants voluntarily signed the informed consent before inclusion into this study. All procedures performed in this study involving human participants were in accordance with the Declaration of Helsinki (as revised in 2013).

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