

Six mutator-derived IncRNA signature of genome instability for predicting the clinical outcome of colon cancer

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Background: Colon adenocarcinoma (COAD) is one of the most common malignancies worldwide. Genomic instability is one of the hallmarks of colon cancer and is associated with prognosis. Nevertheless, the impact of genome instability-associated long non-coding RNAs (lncRNAs) along with their clinical significance in cancers has remained mostly unexplored.

Methods: In this study, a mutator hypothesis-derived computational frame integrating the somatic mutation profiles and lncRNA expression profiles in a tumor genome was developed, which enabled the identification of 137 novel genomic instability-associated lncRNAs in colon cancer. Subsequently, a genome instability-derived lncRNA signature (GILncSig) segregated the patients into low- and high-risk groups with prominent differences in outcomes.

Results: Combined with the overall survival data, we established 6 six lncRNA-based signature to predict prognosis, which were LINC00896, AC007996.1, NKILA, AP003555.2, MIRLET7BHG, and AC009237.14. We found that the expression level of PD-L1 (CD274) and somatic mutations in the high-risk group were higher than those in the low-risk group. This suggests that high-risk patients may be sensitive to immunotherapy. We further found that the prognosis of patients in the high-risk group was significantly lower than that of patients in the low-risk group, and that patients' prognosis was likely to be worse as the patient's risk score increased.

Conclusions: In conclusion, this study explores the role of lncRNAs in genomic instability and cancer prognosis and provides a new idea for the prognostic prediction of colon cancer.

Keywords: Genome instability; mutator phenotype; long non-coding RNAs (lncRNAs)

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Introduction

Worldwide, despite advances in the diagnosis and treatment of colon adenocarcinoma (COAD), its morbidity and mortality rates remain high (1). The TNM staging system remains the world's gold standard for selecting cancer treatment or predicting prognosis. Nonetheless, despite identical clinical characteristics, the prognosis amongst patients can differ significantly due to the high levels of heterogeneity in colon cancer (2). From the progress, development, and response to treatment, colon cancer has shown high complexity in terms of clinical and molecular heterogeneity (3). Hence, to assess the clinical outcomes of patients with colon cancer more accurately, the urgent identification of novel biomarkers is imperative.

One of the hallmarks of cancer has been reported to be genomic instability. There are many forms of genomic instability. Most cancers have a form called chromosomal instability (CIN), which refers to the high rate of change in the structure and number of chromosomes in cancer cells over time compared to normal cells. Other forms of genomic instability have also been described, including microsatellite instability (MSI; also known as MIN), and forms of genomic instability characterized by an increased frequency of base pair mutations (4). Genomic instability has been linked with the progression and survival of colon cancer, and is an important prognostic factor (5). In colorectal cancer (CRC), the cancer was induced by mutated genes The evolution of CRC is caused by a variety of genetic and epigenetic abnormalities, including defective DNA mismatch repair (DMMR) and mutations in Kirsten-RAS (KRAS) and BRAF proto-oncogenes (6,7). This suggests the potential of molecular signatures of genomic instability to predict prognosis. For example, a 12-gene genomic instability signature was identified through an analysis of the gene expression profiles of breast cancer specimens by Habermann et al. (8). A meta-analysis of the expression of miRNAs showed their association with MSI in colon cancer tissues (9). Many studies have also shown that upregulated immune checkpoints (e.g., CTLA 4, PD-1, and/or PD-L1/CD274) have been found in highly mutated tumors with DMMR or high MSI (MSI-H), which may benefit from immunotherapy (10,11).

Transcripts that do not indicate any potential for protein coding and are longer than 200 nt have been broadly classified as long non-coding RNAs (lncRNAs) (12). Growing *in vivo* and *in vitro* evidence over the years have indicated the significant role played by lncRNAs across various biological processes. In particular, aberrant lncRNA expression has been shown to play a role in metastasis, tumor progression, and cell proliferation (13). Despite lncRNAs being abnormally expressed in numerous cancers, the functions of these lncRNAs have not been determined. In the maintenance of genomic instability, the crucial role of lncRNAs has been revealed from emerging evidence. A study showed that the non-coding RNA NORAD regulates genomic stability by chelating pumilio proteins (14). Another study showed that RNA exosomes control superenhancer activity by regulating lncRNA transcription (15). In colon cancer, it has been shown that lncRNA CCAT2 induces CIN through BOP1-AURKb signaling, leading to poor prognosis (16). However, so far, lncRNAs associated with genomic instability and their clinical significance have not been mapped and explored.

To evaluate the potential of the lncRNA signature being considered as an indicator of genomic stability in colon cancer, a mutator hypothesis-derived computational frame integrating lncRNA expression and somatic mutation profiles in a tumor genome was developed in this study. This study explored the role of lncRNAs in genomic instability and cancer prognosis in colon cancer which was the main difference from the article you mentioned. We further explored the expression of immune checkpoint between high-risk group and low-risk group in this article. We used the CIBERSORT algorithm to estimate the proportions of 22 immune cell types in colon cancer samples to further investigate the relationship between the genome instabilityderived lncRNA signature (GILncSig) and immune cell infiltration.

We present the following article in accordance with the TRIPOD reporting checklist (available at https://dx.doi. org/10.21037/jgo-21-494).

Methods

Data download

The Cancer Genome Atlas (TCGA) database (https:// portal.gdc.cancer.gov/) provided the information of the somatic mutations, the fragments per kilobase of exon model per million mapped fragments (FPKM) type of RNA-seq expression data, and the clinical features of patients with COAD. For further study, a total of 446 paired samples with common clinicopathological features, somatic mutation information, survival information, and mRNA and lncRNA expression profiles were obtained.

Covariates	Туре	Total	Testing	Training	P value
Age	≤65	183	93	90	0.786
	>65	263	129	134	
Gender	Female	212	104	108	0.8459
	Male	234	118	116	
Stage	Stage I–II	250	126	124	0.7071
	Stage III-IV	185	89	96	
	Unknown	11	7	4	
Т	T1–2	86	47	39	0.3878
	T3–4	359	175	184	
	Unknown	1	0	1	
Μ	M0	329	161	168	1
	M1	61	30	31	
	Unknown	56	31	25	
Ν	N0	265	134	131	0.7585
	N1-2	181	88	93	

Table 1 Clinicopathological information of the three COAD patient cohorts in this study

COAD, colon adenocarcinoma.

All the colon cancer patients were divided into two sets, namely the test and training sets. To identify the prognostic lncRNA signature and build a prognostic risk model, a total of 224 patients from TCGA were placed in the training set, while 222 patients were used to validate the performance of the prognostic risk model. Furthermore, from TCGA database, the corresponding lncRNA expression data and the somatic mutation information of patients were also downloaded. *Table 1* provides a brief summary of the pathological and clinical characteristics. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

Identification of genome instability-associated lncRNAs

We developed a mutator hypothesis-derived computational framework by integrating somatic mutation profiles and lncRNA expression profiles in tumor genomes to identify lncRNAs associated with genomic instability using the following steps: (I) the cumulative number of somatic mutations in each patient was calculated; (II) the cumulative number of somatic mutations among patients was ranked in decreasing order; (III) the bottom 25% of patients were defined as the genomically stable (GS)-like group, while the

top 25% were defined as the genomically unstable (GU)like group; (IV) the expression profiles of the lncRNAs between the GS and GU groups were compared; (V) the genome instability-associated lncRNAs were identified as differentially expressed lncRNAs [absolute value of the fold change greater than 2 and their false discovery rate (FDR) adjusted P value being less than 0.05].

Enrichment analysis

We identified mRNAs pairwise expressed with lncRNAs associated with genomic instability. The top 10 mRNAs associated with each lncRNA were selected. On this basis, a co-expression network was constructed. The 'clusterProfiler' R package and the 'ggplot2' R package were used for Gene Ontology (GO) enrichment analysis and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway analysis (17,18).

Estimation of immune cell infiltration

We used the CIBERSORT algorithm to estimate the proportions of 22 immune cell types in colon cancer samples from gene expression data to further investigate 2160

the relationship between the GILncSig and immune cell infiltration (19). We removed samples with P>0.05 and retained samples with P<0.05 for further analysis. We used the Wilcoxon rank-sum test to identify if there was a significant difference in the proportion of immune cells between the low- and high-risk groups. In addition, the Kaplan-Meier method was used to evaluate the relationship between immune cell infiltration and patient prognosis.

Statistical analysis

To evaluate the overall survival and the association with the expression levels of genome instability-associated lncRNAs, multivariate and univariate Cox proportional hazard regression analyses were conducted. A GILncSig for prognostic prediction was developed as described below, based on the coefficients from the multivariate regression analysis and the levels of expression of the prognostic genome instability-associated lncRNAs:

$$GILncSig(patient) = \sum_{i=1}^{n} coef(IncRNAi) \times expr(IncRNAi)[1]$$

where GILncSig (patient) represented the prognostic risk score for the colon cancer patient, lncRNAi indicated the ith prognostic lncRNA, and expr (lncRNAi) was the expression level of lncRNAi for the patient. The coef (lncRNAi) was representative of the contribution of the lncRNAi to the prognostic risk scores that were obtained from the multivariate Cox analysis regression coefficient. To segregate patients into the low-risk group with a low GILncSig or the high-risk group with a high GILncSig, the median scores of patients in the training set were used as the risk cutoff.

To assess the difference in survival between the lowand high-risk groups with a 5% significance level, the log-rank test was applied. To determine the survival rate, the Kaplan-Meier method was applied. For assessing the independence of the GILncSig from the other key clinical factors, stratified and multivariate Cox regression analyses were conducted. With the Cox analysis, the 95% confidence interval (CI) and hazard ratio (HR) were calculated. The time-dependent receiver operating characteristic (ROC) curve was used to evaluate the performance of the GILncSig. R version 4.0.3 was used to perform all statistical analyses.

Results

Identification of genomic instability-related lncRNAs in colon cancer patients

We calculated and sorted the cumulative number of somatic mutations per patient to identify the lncRNAs associated with genomic instability. Based on the results of somatic mutations in decreasing order, the top 25% (n=112) of patients were divided into the GU-like group and the bottom 25% (n=101) of patients were divided into the GSlike group. In order to identify the lncRNAs with significant differences, the lncRNA expression profiles of the 112 patients from the GU-like group were compared with the 101 patients in the GS-like group. A total of 137 lncRNAs were found to be significantly differentially expressed, with an absolute value of fold change greater than 2 and FDR-adjusted P value less than 0.05. In total, 81 lncRNAs were upregulated and 56 downregulated, as detailed in Table S1. Figure 1A shows the top 40 lncRNAs that were significantly different. Using the set of 137 differentially expressed lncRNAs, the samples from TCGA cohort were divided into GS-like and GU-like groups based on the expression levels of the differentially expressed lncRNAs (Figure 1B). Between the two groups, the somatic mutation pattern was found to be significantly different. The GU-like group contained higher cumulative somatic mutations compared to the GS-like group (P=0.014; Figure 1C). Further study showed higher PD-L1 (CD274) expression in the GU group, consistent with another previous study (P= 1.4×10^{-8} ; Figure 1D) (11).

Enrichment analysis

We measured the expression correlation between the 137 differentially expressed lncRNAs and protein coding genes, and constructed a lncRNA-mRNA co-expression network (*Figure 2A*). The mRNAs co-expressed with lncRNAs were enriched by GO and KEGG analysis. Based on GO enrichment analysis, the mRNAs in this co-expression network were mainly enriched in gland development, organelle subcompartment, and steroid binding (*Figure 2B*). According to KEGG pathway enrichment analysis, the main enrichment pathways were the chemokine signaling pathway, proteoglycans in cancer, and the NOD-like receptor signaling pathway (*Figure 2C*). These pathways are



Figure 1 Identification of genomic instability-related lncRNAs in patients with colon cancer. (A) Heatmap showing significant changes in the top 40 lncRNAs between the GU-like group and the GS-like group. (B) Based on the expression patterns of 137 candidate lncRNAs associated with genomic instability, 446 colon cancer patients were clustered. (C) Boxplots of somatic mutations in the GU-like group and GS-like group. (D) Boxplots of CD274 expression levels in the GU-like group and GS-like group. LncRNAs, long non-coding RNAs; GU, genomically unstable; GS, genomically stable.

closely related to the occurrence and progression of cancer.

Development of the GILncSig

Colon cancer patients from TCGA were randomly divided into the training set and test set to further investigate the prognostic roles of these candidate genome instabilityassociated lncRNAs, as detailed in *Table 1*. It was observed that 12 genome instability-associated lncRNAs were significantly associated with the prognosis of colon cancer patients in the training set (all P<0.05), and the prognostic forest plot of these 12 lncRNAs is shown in *Figure 3.* Furthermore, the multivariate Cox proportional hazards regression analysis was conducted between the 12 candidate lncRNAs to single out the lncRNAs with independent prognostic value. Finally, 6 of the 12 candidate lncRNAs (LINC00896, AC007996.1, NKILA, AP003555.2, MIRLET7BHG, and AC009237.14) were identified as independent prognostic lncRNAs as their P values in the multivariate Cox analysis were less than 0.05 (*Table 2*). Subsequently, based on the expression levels of the six independent prognostic genomic instability-associated lncRNAs and the coefficients of the multivariate Cox analysis, a GILncSig was developed to determine the



Figure 2 Co-expression network and enrichment analysis of differential lncRNAs. (A) Co-expression network of genomic instabilityrelated lncRNAs and mRNAs. The blue circles represent lncRNAs and red circles represent mRNAs. (B) GO enrichment analysis and (C) KEGG pathway analysis of mRNAs associated with lncRNAs. LncRNAs, long non-coding RNAs; GO, Gene Ontology; KEGG, Kyoto Encyclopedia of Genes and Genomes.

prognostic risk of patients with colon cancer. GILncSig score = $(0.244155 \times \text{expression} \text{ level of LINC00896}) + (0.626462 \times \text{expression} \text{ level of AC007996.1}) + (0.295227 \times \text{expression} \text{ level of NKILA}) + (0.209866 \times \text{expression} \text{ level of AP003555.2}) + (0.351340 \times \text{expression} \text{ level of MIRLET7BHG}) + (0.196452 \times \text{expression} \text{ level of AC009237.14})$. In the GILncSig score, the coefficients of all lncRNAs were positive, and their high expression levels were associated with poorer survival. The expression levels of lncRNAs in the GILncSig in the high- and low-risk groups are shown in *Figure 4*. Results showed that all 6 lncRNAs in the GILncSig were upregulated in the high-risk group, both in the training group (*Figure 4A*) and in TCGA patient groups (*Figure 4C*). However, only four lncRNAs (AC007996.1, NKILA, AP003555.2, and AC009237.14) were upregulated in the test group (*Figure 4B*).

Risk scores for patients in both the training set and the test set were obtained, and these patients were then classified into different prognostic groups using the median risk score as a threshold. We then performed Kaplan-Meier analyses of patients in the high- and low-risk groups, and the results showed that survival outcomes in the low-risk group were significantly better than those in the high-risk group in TCGA groups (training set, P<0.001, *Figure 5A*; test set, P=0.005, *Figure 5B*; TCGA sets, P<0.001, *Figure 5C*). Time-dependent ROC curve analysis of the GILncSig

	pvalue	Hazard ratio	1
LINC00896	0.024	1.416(1.048–1.915)	¦ ⊢ 4
AL138789.1	0.044	1.286(1.007–1.642)	▶ — ■ —— 1
AC020656.2	0.038	1.336(1.016–1.757)	→
AC007996.1	0.006	1.688(1.164–2.449)	· • • • • • • • • • • • • • • • • • • •
LINC00941	0.033	1.266(1.019–1.573)	→
NKILA	0.017	1.298(1.048–1.607)	! ⊢− ■−−−4
LINC00654	0.038	1.116(1.006–1.237)	H=-1
AC007991.2	0.046	1.060(1.001–1.122)	j e i
AP003555.2	0.002	1.310(1.109–1.549)	i 🛏 🖬 🖌
MIRLET7BHG	<0.001	1.386(1.159–1.658)	↓ ⊢− ■−−4
AC009237.14	0.018	1.163(1.027–1.318)	-=
PTPRD-AS1	0.037	0.624(0.401–0.972)	⊢ ∎(
			0.0 0.5 1.0 1.5 2.0
			Hazard ratio

Figure 3 The prognostic forest plot of 12 genome instability-associated lncRNAs in the training set. LncRNAs, long non-coding RNAs.

Table 2 LncRNAs associated with the prognosis of colon cancer patients obtained after multivariate Cox analysis

ID	Coef	HR	HR.95L	HR.95H	P value
LINC00896	0.2441547	1.276541	0.960602	1.696392	0.092392
AC007996.1	0.6264617	1.870978	1.234731	2.835079	0.003133
NKILA	0.2952265	1.343430	1.080575	1.670227	0.007871
AP003555.2	0.2098663	1.233513	1.002866	1.517206	0.046915
MIRLET7BHG	0.3513404	1.420971	1.164792	1.733492	0.000532
AC009237.14	0.1964520	1.217076	1.068190	1.386715	0.003169

LncRNAs, long non-coding RNAs; coef, coefficient; HR, hazard ratio.

showed that the areas under the curve (AUCs) of the training set, testing set and TCGA groups were 0.691, 0.661, and 0.675, respectively (*Figure 5D-5F*).

The number of somatic mutations and the expression level of PD-L1 in each group

We then explored the number of somatic mutations and the differences in PD-L1 expression levels between the highand low-risk patients. We first developed a set of risk figures for the three datasets, including a heat map of lncRNA expression and the distribution of patient risk scores. As is shown in *Figure 6A*, the expression levels of all lncRNAs in the training set increased with the increase in GILncSig score. These results were further verified in the test set and the TCGA set (*Figure 6B,6C*). As shown in *Figure 6D-6F*, the count of somatic mutations in patients in the high-risk group was higher than that of patients in the low-risk group (P=0.088, training set; P=0.066, test set; P=0.013, total set). The expression level of PD-L1 (CD274) was significantly higher in the high-risk group than in the low-risk group (P=0.01, training set; *Figure 6G*). However, in the test set, the expression of PD-L1 (CD274) was not significantly different among the high- and low-risk groups (P=0.15, Figure 6H). The expression level of PD-L1 (CD274) was significantly higher in the high-risk group than in the lowrisk group (TCGA set; Figure 61). We also did mutations of the molecule CD274 across different groups, but we found that it was very stable, with only two mutations in 400 patient samples (Table S2). Our results show that PD-L1 expression is higher in the high-risk group, suggesting that these patients may benefit from immunotherapy.



Figure 4 The expression levels of all six lncRNAs of the GILncSig in high- and low-risk groups from the training (A), test (B), and TCGA (C) sets. **, P<0.01; ***, P<0.001. LncRNAs, long non-coding RNAs; GILncSig, genome instability-derived lncRNA signature; TCGA, The Cancer Genome Atlas.

GILncSig and clinical factors

We used the chi-square test to investigate the relationship between risk and clinical characteristics. As shown in *Table 3*, the stages, T stage, and N stage were significantly correlated with the risk score in the training group and the total set. The higher a patient's risk score, the higher their stage and the higher their N stage. All the clinical characteristics showed no significant differences between the high- and low-risk groups in the test set.

The landscape of immune infiltration in COAD

To further investigate the relationship between risk score and immune cell infiltration, we used the CIBERSORT algorithm to estimate the proportions of 22 immune cell types in the COAD cohort from gene expression data. The Wilcoxon rank-sum test was used to explore whether there were differences between the 22 kinds of immune cells in different groups. As shown in *Figure 7A*, T follicular helper cells (P=0.018), resting NK cells (P=0.029), and M1



Figure 5 Kaplan-Meier curves and ROC curves of patients in the high- and low-risk groups. Kaplan-Meier curves of overall survival in low- and high-risk patients from the training (A), test (B), and TCGA (C) sets. Time-dependent ROC curve analysis of the GILncSig in the training (D), test (E), and TCGA (F) sets. ROC, receiver operating characteristic; GILncSig, genome instability-derived lncRNA signature; TCGA, The Cancer Genome Atlas.

macrophages (P=0.026) varied significantly between highand low-risk score patients. In addition, Kaplan-Meier analysis showed that a lower proportion of regulatory T cells (Tregs) were associated with better overall survival (P=0.018, *Figure 7B*). However, a higher proportion of resting mast cells were associated with better overall survival (P=0.008, *Figure 7C*).

Discussion

The initiation, development, and treatment of colon cancer have been the subject of investigation for the past several years (20-22). Patients continue to be classified according to different therapeutic groups based on their pathological features, while the most important prognostic factors, such as the traditional histopathological features of tumor size, grade, and stage, also continue to be used (23,24). Nonetheless, due to the various limitations of traditional clinicopathological features, the clinical outcome of patients with colon cancer remains highly heterogeneous (25). Genomic instability is not only a common feature of most cancers, but is also considered to be one of the factors affecting the prognosis of colon cancer (26). In colon cancer progression and recurrence, genomic instability has a crucial and dominating role, thereby suggesting the important diagnostic and prognostic implications indicated by the pattern and degree of genomic instability (27,28). Nevertheless, quantitative measures of the degree



Figure 6 The number of somatic mutations and the expression level of CD274 in each group. LncRNA expression patterns with increasing GILncSig score in the training (A), test (B), and TCGA (C) sets. The distribution of somatic mutations in the training (D), test (E), and TCGA (F) sets. The expression levels of CD274 in the high- and low-risk groups from the training (G), test (H), and TCGA (I) sets. LncRNAs, long non-coding RNAs; GILncSig, genome instability-derived lncRNA signature; TCGA, The Cancer Genome Atlas.

of genomic instability have remained a challenge. For predicting genomic instability, concerted efforts are being made to develop gene or miRNA signatures and identify miRNAs associated with genomic instability (29).

LncRNAs, a novel class of non-coding RNAs, have

recently gained significance as important components of tumors. The dysregulated expression of lncRNAs in cancer is related to disease progression, and lncRNAs have the potential to be used as prognostic markers for patients (30,31). With recent advances in the understanding of the

Table 3 Correlations between the risk scores of mutator-derived lncRNAs and the clinicopathological characteristics in the training set, test set, and total set

Parameters	Training set (n=224)			Testing set (n=222)			Total set (n=446)					
	HR	LR	χ^2	Р	HR	LR	χ^2	Р	HR	LR	χ^2	Р
Age (y)												
≤65	42	48	0.4643	0.4956	52	41	0.1127	0.7371	94	89	0.0178	0.8938
>65	70	64			68	61			138	125		
Gender												
Male	57	59			66	52			123	111		
Female	55	53	0.0179	0.8936	54	50	0.2145	0.6432	109	103	0.0218	0.8826
Stage												
Stage I–II	52	72	7.4875	0.0062	64	62	0.9353	0.3335	116	134	7.3455	0.0067
Stage III-IV	59	37			52	37			111	74		
Unknown	4	1							5	6		
т												
T1–2	18	21	0.1035	0.7476	19	28	3.79	0.0516	37	49	2.9457	0.0861
T3–4	93	91			101	74			194	165		
Unknown	1	0							1	0		
М												
MO	79	89	1.5986	0.2061	86	75	0.6441	0.4222	165	164	2.5732	0.1087
M1	19	12			19	11			38	23		
Unknown	14	11			15	16			29	27		
Ν												
N0	53	78	10.5905	0.0011	68	66	1.1722	0.2789	121	144	9.956	0.0016
N1-2	59	34			52	36			111	70		

LncRNAs, long non-coding RNAs; HR, high risk; LR, low risk.

functional mechanisms of lncRNAs, lncRNAs are also critical to genome stability (32). The systematic exploration of their clinical significance in cancers and the genome-wide identification of genome instability-associated lncRNAs remain in their infancy despite certain efforts being made. Hence, to identify genome instability-associated lncRNAs combining lncRNA expression and the tumor mutator phenotype, a computational frame was developed. The results showed that 137 novel genome instability-associated lncRNAs were identified after the lncRNA expression profiles were combined with the somatic mutation profiles of colon cancer. We then constructed a lncRNA-mRNA co-expression network analysis of 128 genes that coexpressed lncRNAs associated with genomic instability. GO enrichment analysis showed that the mRNAs in this co-expression network were mainly enriched in gland development, organelle subcompartment, and steroid binding. According to KEGG pathway enrichment analysis, the main enrichment pathways were the chemokine signaling pathway, proteoglycans in cancer, and the NODlike receptor signaling pathway, which are closely related to the occurrence and progression of cancer.

Colon cancer patients from TCGA were randomly divided into the training set and test set. Through multivariate Cox analysis, we observed that six genomic instability-related lncRNAs (LINC00896, AC007996.1, NKILA, AP003555.2, MIRLET7BHG, and AC009237.14) were significantly associated with the prognosis of patients in the training set.



Figure 7 The landscape of immune infiltration in COAD. (A) The comparison of the fractions of immune cells between the high- and low-risk groups. Kaplan-Meier survival analysis of overall survival between patients with high and low levels of infiltrating Tregs (B) and resting mast cells (C). COAD, colon adenocarcinoma; Tregs, regulatory T cells.

A GILncSig was developed to determine the prognostic risk of patients with colon cancer: GILncSig score = $(0.244155 \times 10^{-5})$ expression level of LINC00896) + (0.626462 × expression level of AC007996.1) + (0.295227 × expression level of NKILA) + (0.209866 × expression level of AP003555.2) + (0.351340 × expression level of MIRLET7BHG) + (0.196452 × expression level of AC009237.14). The results showed that all six lncRNAs were associated with poor prognosis, which has rarely been reported in the literature. NKILA is a lncRNA that interacts with NF-KB. A study showed that NKILA promotes tumor immune evasion by sensitizing T cells to activation-induced cell death in breast cancer (33). Another study also showed that lncRNA AP003555.2 was associated with poor prognosis in patients with colon cancer (34). Furthermore, one study suggested that MIRLET7BHG was associated with polycystic ovary syndrome in women (35). Another study showed that autophagy-related lncRNA AC009237.14 was associated with poor prognosis in colon cancer patients (36). We further investigated the association between the GILncSig and clinical features. The results showed that higher risk scores were associated with more advanced stage and N stage. In addition, there was an obvious survival difference between the high- and low-risk groups according to our model, that is, the high-risk group had a shorter survival time and worse prognosis, and the AUC in TCGA groups was greater than 0.65, indicating that the prediction results of our model were relatively accurate.

Maintaining genomic stability to a large extent, nucleotide excision repair can specifically prevent mutations induced by environmental carcinogens. Moreover, some patients with MSI-H or DMMR CRC appear to be prone to persistent clinical responses to checkpoint inhibitors,

providing a new treatment option for patients with advanced disease (37). We then examined the number of somatic mutations and the expression of PD-L1 (CD274) between high- and low-risk patients. The results showed that the number of somatic mutations in the high-risk group was higher than that in the low-risk group, and the expression level of PD-L1 (CD274) in the high-risk group was significantly higher than that in the low-risk group. This suggests that our GILncSig is beneficial in differentiating patients who are sensitive to immunotherapy.

We further investigated the relationship between risk score and immune cell infiltration, and showed that there were more T follicular helper cells, resting NK cells, and M1 macrophages in the high-risk group. Besides, a low proportion of Tregs is related to a better prognosis. In a previous study, Tregs from cancer patients inhibited the mechanism of traditional T cell migration and thus affected patient outcomes (38). Mast cells have been shown to promote the development of colon cancer and could be a potential therapeutic target (39). All of these studies confirm our conclusions.

Although our study provides important insights for better assessing genomic instability and prognosis in colon cancer patients, it still has some limitations that require further investigation. We need more independent datasets such as Gene Expression Omnibus (GEO) datasets to validate the GILncSig to ensure its robustness and repeatability. In addition, we need to conduct further experimental validation to understand the regulatory mechanisms of the GILncSig in maintaining genomic instability.

Conclusions

In conclusion, examining the role of lncRNAs in genome instability through a mutator hypothesis-derived computational frame for identifying genome instabilityassociated lncRNAs was proposed by this study, providing a critical approach and resource for further studies. The levels of somatic mutations and PD-L1 (CD274) expression in the high-risk group were higher than those in the lowrisk group, which indicated that high-risk groups may be sensitive to immunotherapy. The GILncSig can predict the outcome of patients with COAD and provides a new therapeutic direction.

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For contributors to the TRIPOD statement, see https://www.annals.org.

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Footnote

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