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Co-expression network analysis of long noncoding RNAs (IncRNAs) and cancer genes reveals *SFTA1P* and *CASC2* abnormalities in lung squamous cell carcinoma

Guang-Qing Huang^{a,*}, Zun-Ping Ke^{b,*}, Hai-Bo Hu^{c*}, and Biao Gu^d

^aDepartment of Intensive Care Unit, Renmin Hospital, Hubei University of Medicine, Shiyan, Hubei, P.R. China; ^bDepartment of Cardiology, The Fifth People's Hospital of Shanghai, Fudan University, Shanghai, China; ^cDepartment of Thoracic Surgery, The Affiliated Huai'an Hospital of Xuzhou Medical University and The Second People's Hospital of Huai'an, Huai'an, China; ^dDepartment of Thoracic Surgery, Huai'an First People's Hospital, Nanjing Medical University, Huai'an, Jiangsu, China

ABSTRACT

Lung squamous cell carcinoma(LSCC) is the most common and aggressive lung tumor with poor clinical outcome. Previously studies showed that deregulation of long noncoding RNAs (IncRNAs) were involved in LSCC. We intended to figure out the role of IncRNAs in the regulation process of cancer-related genes and pathways they are involved. Data of 552 samples, including 501 cancer samples and 51 normal ones, were extracted from The Cancer Genome Atlas (TCGA). Differentially expressed IncRNAs (DEIs) were screened out (FDR<0.05, |logFC|>1) and then followed by GO ontology and KEGG annotation analysis. Oncogenes from COSMIC data set and Tumor suppressor genes (TSGs) from TSGene data set were collected and analyzed by gene Set Enrichment Analysis (GSEA) . The differentially expressed oncogenes and tumor supressor gene (TSGs) were obtained and co-expression analysis was conducted to generate co-expression lncRNA-gene pairs, which can be helpful in figuring out the role of lncRNA in the regulation of oncogenes and tumor suppressor genes. A total of 31 IncRNAs with low expression levels and 37 IncRNAs with high expression levels were screened out and most of them were enriched in pathways such as meiosis, male gamete generation, defensins. Of note, SFTA1P and CASC2 were found to be related with most of the oncogenes and TSGs by co-expression analysis. We suggested SFTA1P and CASC2 played important role in the regulation of both oncogene and TSGs during the carcinogenesis of LSCC and have the potential to be applied in future diagnosis, prognostic process and target therapy of LSCC.

Introduction

Lung squamous cell carcinoma is one of the most common type of cancer, accounting for $40 \sim 50\%$ of all primary lung cancer.¹ Currently, the prognosis of NSCC patients is still very poor. There was not much research about the prognostic and predictive markers for LSCC other than TP53 mutations, which was identified as a causative somatic aberration.²

Long noncoding RNA(lncRNA), with the length of 200 nt and pervasively transcribed in the mammalian genome,³ is involved in a series of processes regulating tumor biology.^{4, 5} It was reported that lncRNAs are differentially expressed in many kinds of cancer, exerting important regulations on tumor biology via regulating oncogenes or tumor suppressors.⁶ Given the fact that the number and functions of all kinds of lncRNAs are not well known, differences in IncRNA levels from one specific type of lung cancer might help us better in identifying their potential role as biomarker of LSCC. Therefore, the subtypes of NSCLC and SCLC were not included in our study. We could expound role of the differentially expressed lncRNA in the occurrence and development of cancer more convincingly. Several reviews have emphasized that the role of ncRNAs in different cellular processes has been largely underestimated. It's reported that non-coding RNAs may be involved in the silence of tumor suppressor genes (TSGs) epigenetically, which can inhibit normal cellular growth in cancer.⁷ LncRNAs mainly played roles in regulation of transcriptional, posttranscriptional and epigenetic mechanisms in the process of tumorigenesis.⁸ Several studies confirmed the deregulation role of lncRNA in cancers, such as hepatocellular carcinoma, breast,⁹ osteosarcoma,¹⁰ colon and prostate cancers.^{11, 12}

Differentially expressed IncRNAs provide essential evidence to figure out the difference between cancer samples and normal specimens, which can help us in exploring the relevant pathways they were involved. For example, in previous studies, *NOTCH1* or *NOTCH2* were found to be 2 tumor suppressor specific to epithelial malignancies, thus coming to the conclusion that targeted inhibition of the Notch pathway may affect squamous epithelial malignancies.⁴

In spite of many studies on the potential important roles of lncRNAs in various kinds of cancer, there had not been much report on lncRNAs associated with LSCC except one study which

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CONTACT Hai-Bo Hu or dr_huhaibo@aliyun.com Department of Thoracic Surgery, The Affiliated Huai'an Hospital of Xuzhou Medical University and The Second People's Hospital of Huai'an, Huai'an, China; Biao Gu or prof_zh@sina.com Department of Thoracic Surgery, Huai'an First People's Hospital, Nanjing Medical University, 6 Beijing West Road, Huai'an, Jiangsu 223300, China. *Co-first authors.

indicated that lower expression of CASC2 was accompanied by poor prognosis of non small cell lung cancer(NSCLC).¹³ In this study, we attempted to provide large-scale survey of lncRNAs within LSCC, the subtype of non-small cell lung cancer, and to figure out the role of lncRNA in the regulation process of cancer-related genes of LSCC patients, which may help us in further analysis of identifying pathways they are involved, and thus to identify prognostic marker and novel therapy for LSCC patients.

Materials and methods

Source of data and data pre-processing

A total of 552 expression data (501 LSCC samples and 51 normal ones) were extracted from TCGA (The Cancer Genome Atlas), based on the platform of IlluminaHiSeq. A total of 175 lncRNAs were screened out by annotation profiles from ensemble data set and FRKM values were extracted out and used in the formation of data matrix.^{14, 15}

Identification of differentially expressed IncRNA

Differentially expressed lncRNAs (DEIs) were extracted out from the expression matrix by the algorithm method of limma package.¹⁶ The thresholds were FDR<0.05 and |logFC|>1.

GO annotation and KEGG pathway analysis

We applied lncRNA2 function ¹⁷ software which include functional annotation of 9625 human lncRNA genes to annotate the DEIs. All nodes in Gene Ontology (GO) and a total of 4380 function pathways were included in this functional pathway. GO functional annotation and KEGG pathway analysis of DEIs involved was analyzed by lncRNA2 function.

Collection of oncogenes and Tumor suppressor genes (TSGs)

Data of all the 634 oncogenes were from COSMIC data set,^{17, 18} which included information of somatic mutation associated



with all kinds of cancers. The whole 1452 tumor suppressor gene data were extracted from TSGene data set and TAG data set.¹⁹ There were 1217 TSGs collected from more than 9000 studies in TSGene 2.0 (1018 protein coding genes and 199 non-coding genes) while 265 TSGs were included in TAG data set.

Identification of differentially expressed oncogenes and TSGs

Cluster heatmap of the top 100 most significantly differentially expressed oncogenes and TSGswere painted by heatmap 2.0 of Limma package in R language with the thresholds of FC = 2 and FDR < 0.05.

Enrichment analysis of gene set

Differentially expressed oncogenes and TSGs were annotated by GSEA (Gene Set Enrichment Analysis).^{20, 21} GO, KEGG and cancer gene neighborhood were all included in GSEA, which can help us in obtaining the pathways and functions of differentially expressed cancer genes.

Co-expression analysis

Pearson correlation coefficient of each lncRNA and cancer gene, as well as confidence interval and p-value, was calculated out using the function of cor ()andcor.test() of R fuction. The formula were as follow,

$$r = \frac{1}{n=1} \sum_{i=1}^{n} \left(\frac{X_i - \overline{X}}{s_X} \right) \left(\frac{Y_i - \overline{Y}}{s_Y} \right)$$

Co-expression pair of lncRNA-gene was screened out with the threshold of |correlation coefficient|>0.5 and FDR<0.05 to analyze the regulation role of lncRNApalyed on oncogenes and TSGs.²²

Down_regulated	logFC	AveExpr	P.Value	adj.P.Val
C14orf132	-1463.05	491.5892	2.95E-99	2.58E-97
SFTA1P	-1061.46	143.3066	8.72E-153	1.53E-150
MALAT1	-831.21	2824.661	0.038047	0.097914
IPW	-154.206	261.4601	2.58E-06	1.46E-05
HCG11	-95.327	295.2437	0.00017	0.000675
C10orf95	-62.8257	22.00978	2.72E-42	7.94E-41
CASC2	-51.757	34.08992	4.99E-40	1.25E-38
HCG26	-41.9228	63.43647	5.44E-07	3.53E-06
MGC27382	-25.3669	4.139212	2.17E-62	9.49E-61
INE2	-9.15558	7.585014	6.68E-10	6.15E-09
up_regulated	logFC	AveExpr	P.Value	Adj.P.Val
TUG1	1672.372	4458.745	4.68E-09	3.90E-08
GAS5	1102.298	2215.222	1.12E-05	5.30E-05
SNHG1	940.4861	1150.743	1.07E-18	2.33E-17
MIAT	865.0085	1007.605	0.006695	0.020554
SNHG5	574.2911	1803.889	0.006891	0.020791
SNHG6	516.5869	1115.43	1.87E-06	1.13E-05
TP53TG1	168.8525	387.5029	5.51E-06	2.92E-05
PVT1	138.7953	159.0165	4.58E-16	8.90E-15
SNHG12	136.07	307.6558	1.55E-05	6.80E-05
SNHG11	100.5078	203.8896	3.76E-12	5.49E-11



Figure 1. The top 5 pathways of GO and KEGG enrichment analysis in differentially expressed IncRNA. The horizontal axis included the score of enrichment while the vertical axis represented the enriched pathways. GO annotation of upregulated IncRNA was displayed in (A) while the downregulated ones were in (B). (C) was the KEGG annotation of upregulated IncRNA while (D) was that of downregulated IncRNA. The higher the enrichment score was, the more the pathways were involved in cancer regulation.

Results

Identification of DEIs

DEIs were identified by Limma package with the thresholds of FDR<0.05 and |logFC|>0. As a result, a toal of 32 downregulated lncRNA and 37 upregulated lncRNA were screened out (Table 1). Multiple of the top 10 listed lncRNAs of all the up-and-downregulated genes was extremely significant, indicating expression of these lncRNAs were quite different between cancer samples and normal ones. Full information of DEIs were in Table S1 and results of TSG difference in detail were listed in Table S2 while the difference of oncogene in detail was listed in Table S3.

GO and KEGG annotation of IncRNAs

There were 45 biologic processes in mostly enriched GO annotation of upregulated lncRNA, along with 11 cellular_component, 2molecular_function and 2 KEGG annotations (Fig. 1). The most enriched GO terms were spermatogenesis while the most enriched KEGG wasdefensins. There were 58 biologic processes in the mostly enriched GO terms of lncRNA, as well

Table 2. The top 10 most enriched genes of TSGs.

Up_regulated	logFC	AveExpr	P.Value	adj.P.Val
KRT19	28845.27	33309.9	2.72E-14	1.05E-13
JUP	17128.21	20583.57	2.68E-21	1.49E-20
SFN	14604.06	15739.99	2.21E-12	7.32E-12
P4HB	14486.82	28042.46	3.14E-14	1.21E-13
S100A2	14480.6	13260.58	5.90E-07	1.34E-06
CLDN1	13579.75	13894.82	9.24E-07	2.07E-06
IGFBP3	11991.54	12730.01	5.97E-06	1.26E-05
S100A11	11224.27	18430.29	1.23E-13	4.55E-13
HSP90B1	10926.5	22550.73	5.72E-16	2.40E-15
TP63	10801.41	10001.16	3.78E-16	1.63E-15
Down_regulated	logFC	AveExpr	P.Value	adj.P.Val
B2M	-61028.7	57288.05	1.70E-24	1.07E-23
EPAS1	-43960.2	9017.274	9.54E-152	1.34E-149
AHNAK	-39444.6	25397.52	4.28E-54	6.11E-53
VIM	-34186.1	17043.41	2.63E-72	5.76E-71
TIMP3	-29060.2	10151.26	5.52E-58	8.41E-57
EMP2	-26385.1	6813.104	1.23E-170	2.88E-168
CAV1	-24996.5	5456.84	6.22E-147	7.93E-145
GPX3	-22076.3	4341.227	1.65E-134	1.44E-132
SPTBN1	-20717.6	8233.457	1.47E-161	2.28E-159
TXNIP	-20001.2	9212.048	3.83E-71	7.77E-70

Table 3. The top 10 up and downregulated oncogenes.

Up_regulated	logFC	AveExpr	P.Value	adj.P.Val
COL1A1	55834.14	74270.7	3.93E-05	7.82E-05
HSP90AB1	15214.92	32400.67	8.79E-14	3.66E-13
LAMC2	11714.45	14225.67	0.000332	0.000591
TFRC	11713.08	17040.72	1.66E-05	3.45E-05
HSP90AA1	10399.82	27118.23	3.11E-10	1.04E-09
CALR	8826.698	24686.2	7.66E-12	2.77E-11
GPX2	6511.432	5962.844	1.25E-09	3.85E-09
GNAS	6483.241	24242.39	2.23E-05	4.51E-05
HMGA1	5950.289	6397.132	3.91E-32	3.87E-31
ATP2A2	5798.927	10919.86	4.03E-20	2.28E-19
down_regulated	logFC	AveExpr	P.Value	adj.P.Val
CD74		49787.68	7.05E-62	1.47E-60
SLC34A2	-61364.2	11025.63	3.18E-137	6.17E-135
FOS	—19955.7	6364.295	1.14E-52	2.01E-51
MSN	-12400.6	13715.76	6.16E-37	7.33E-36
SDC4	-9841.3	7256.907	9.19E-26	6.96E-25
ALDH2	-9295.91	3688.583	2.70E-105	1.05E-103
HLA-A	-8937.56	26264.03	0.000358	0.000634
EZR	-7375.86	9158.441	9.26E-30	7.71E-29
MYH11	-6842.01	4572.208	1.61E-05	3.35E-05
LMNA	-5908.48	10217.69	7.28E-13	2.83E-12

as 20 cellular_component and one molecular_function. Of all the 8 KEGG annotations, the most significant GO term was spermatogenesis while the most significant KEGG term was APC-C-mediated degradation. The 2 terms were both GO annotation of spermatogenesis, indicating that these significant lncRNAs were quite different from those normal samples, thus affecting the regulation of reproduction and inheritance process. lncRNAs in these pathways were involved in regulation of cellular cycles and pathways of defensins.

Identification of differentially expressed oncogene and TSGs

A total of 518 downregulated TSGs and 503 upregulated TSGs were screened out by Limma package with the thresholds of

FDR<0.05 and |logFC|>1. The top 10 most genes enriched in TSGs were list in Table 2 Similarly, there were 177 upregulated oncogenes and 234 downregulated oncogenes (Table 3). Heatmap of the top 100 differentially expressed TSGs and oncogeneswas shown in Fig. 2 and 3, respectively, indicating that there was significant difference of cancer related genes between normal samples and tumor specimens.

GSEA analysis of differentially expressed oncogenes

No upregulated KEGG gene set but 3 downregulated ones were screened out with the threshold of NOM p-val <0.25. downregulated oncogene was mainly enriched in pathways as CYTOKINE_CYTOKINE_RECEPTOR_INTERACTION, KEGG_MAPK_SIGNALING_PATHWAY and KEGG_JAK_-



Figure 2. Heatmap of the top 100 differentially expressed TSGs between normal samples and tumor specimens. The horizontal axis were different samples, the gray ones were tumor samples while the yellow ones were normal specimens. The vertical axis was genes, which indicated that there was obvious difference among normal samples and tumor ones.



Figure 3. The top 100 differentially exprssed oncogenes between normal samples and tumor specimens. The horizontal axis was samples while the gray ones were tumor samples and the yellow ones were normal specimens. The vertical axis stands for different genes.



Figure 4. The most enriched data set of differentially expressed up and downregulated oncogenes. (A) represents the downregulated genes in KEGG_MAPK_SIGNA-LING_PATHWAY data set and (B) represented how many genes were upregulated in GO data set RESPONSE_TO_DNA_DAMAGE_STIMULUSj. (C) represents the downregulated genes in PLASMA_MEMBRANE data set and (D) represented how many genes were upregulated in MORF_AATF.

STAT_SIGNALING_PATHWAY.A total of 3 upregulated GO gene sets and 14 downregulated ones were enriched. The 3 upregulated Go gene sets were RESPONSE_TO_ DNA_DAMAGE_STIMULUS, DNA_METABOLIC_PROCESS, RESPONSE_TO_ENDOGENOUS_STIMULUS, while the downregulated ones were PLASMA_MEMBRANE, PLAS-MA_MEMBRANE_PART, MEMBRANE. There were 14 gene sets in upregulated cancer neighborhood gene while there was no gene set in downregulated genes. Genes sets of upregulated cancer neighborhood gene were MOR-F_AATF, MORF_DEK, MORF_UBE2I. On the other hand, upregulated gene of differentially expressed oncogene were-RESPONSE_TO_DNA_DAMAGE_STIMULUS and MOR-F AATF (Fig. 4). AATF was a kind of transcription factor against apoptosis process, indicating that the main cause in the occurrence of NCSC was DNA damage and cell obtained the characteristic of anti-apoptosis.

GSEA analysis of differentially expressed TSGs

Two upregulated KEGG gene sets and 5 downregulated KEGG gene sets were enriched. The upregulated gene sets were *KEGG_CELL_CYCLE* and *KEGG_P53_SIGNALING_PATH-WAY* while the downregulated ones were *KEGG_TIGHT_JUNCTION*, *KEGG_LEUKOCYTE_TRANSENDOTHELIAL_MIGRATION*, and *KEGG_MAPK_SIGNALING_PATHWAY*. A total of 17 upregulated GO terms and 25 downregulated GO terms were screened out. The upregulated terms were

RESPONSE_TO_DNA_DAMAGE_STIMULUS,RESPONSE_TO _ENDOGENOUS_STIMULUS, DNA_METABOLIC_PROCESS while the downregulated ones were PLASMA_MEMBRANE, INTRINSIC_TO_MEMBRANE, INTEGRAL_TO_MEMBRANE.

Cancer neighborhood gene sets

A total of 24 upregulated cancer neighborhood gene sets and 2 downregulated gene sets were figured out. The upregulated genes sets of cancer neighborhood gene were *MORF_BUB3*, *MORF_EIF3S2* and *MORF_XRCC5* while the downregulated ones were *GCM_MAP4K4* and *MORF_CDC2L5*.

The most significant terms in TSG were mainly enriched in *KEGG_CELL_CYCLE* and *RESPONSE_TO_DNA_DAMAGE_S-TIMULUS* (Fig. 5). Genes were downregulated in *KEGG_TIGHT_JUNCTION* and *PLASMA_MEMBRANE*.

Since cells in LSCC samples appeared with the characteristic of keratosis and intercellular bridge, including spindle cell carcinoma, pathways associated with the structure of cell membrane could become abnormal. Growth and division of cells could be constant and hereditary substance became variated after malignant transformation process, which could stimulate the overexpression of suppressor genes in cells.

Co-expression of DEIs and TSG or oncogenes

Co-efficient of up-or-downregulated lncRNA with TSGs or oncogene was calculated out respectively by R function. The



Figure 5. The most enriched data sets of tumor supressor gene. (A) represented upregulated genes in KEGG_CELL_CYCLE and (B) represented downregulated genes in KEGG_TIGHT_JUNCTION. (C) represented upregulated genes in RESPONSE_TO_DNA_DAMAGE_STIMULUS while (D) represented how many downregulated genes in PLASMA_MEMBRANE.

thresholds of 0.5 and -0.5 was used to screen out the co-expression pair of lncRNAs-genes.*SFTA1P* appeared to be the most co-expressed gene between downregulated lncRNAs and onco-genes (Table 4). There were 7 co-expression genes, that were*PTPRB*, *SLC34A2*, *FGR*,*ROS1*,*CSF3R*, *CXCL2*, *RND1*. *CACNA1D* was found to be the most co-expression oncogenes with lncRNAs, while all these co-expression oncogenes were downregulated.There were 2 co-expressed genes, *KCNH8* and *SOX2*, between upregulated lncRNAs and oncogenes and *MIAT* was the one with the most co-expression. *PABPC1L* was

found to be the oncogene co-expressing with more than one lncRNAs.

There were 3 genes with most co-expression between downregulated lncRNAs and TSGs, *SFTA1P* with the 16 co-expression genes, while *C10orf95* with 8 co-expression genes and *CASC2* with 7 co-expression genes (Table 5). All the TSGs, that is,*GADD45B*, *CSRNP1*, *RHOBTB2*, *DOK2* were all downregulated.Four co-expression genes in *DLEU1*, 5 co-expression genes in *GAS5* and 4 co-expression genes in *SNHG12*. The 3 TSGs, *RPL15*, *SELT*, *RBX1*, were all overexpressed.

Table 4. Co-expression pairs of IncRNA and oncogenes.

down_IncRNA	oncogene	cor_value	up_IncRNA	oncogene	cor_value
SFTA1P	PTPRB	0.606947516	MIAT	KCNH8	0.57164
SFTA1P	SLC34A2	0.585553018	MIAT	SOX2	0.501148
SFTA1P	PRAM1	0.621849534	HCG18	TRIM27	0.550524
SFTA1P	FGR	0.522879132	SNHG12	PABPC1L	0.529173
SFTA1P	ROS1	0.526725187	C9orf163	RALGDS	0.506619
SFTA1P	CSF3R	0.537765065	TUG1	PATZ1	0.536315
SFTA1P	CXCL2	0.723380194	SNHG3	PABPC1L	0.57944
SFTA1P	RND1	0.563852798	SNHG4	KIAA1598	0.503864
HCG26	HLA-A	0.506811902	SNHG11	PABPC1L	0.542648
HCG26	TNFRSF14	0.579783753			
HCG26	WAS	0.503636572			
CASC2	ECT2L	0.608696582			
RMST	ASPSCR1	0.527844789			
KIAA0087	CACNA1D	0.628973394			
DKFZp779M0652	CACNA1D	0.676408943			
C9orf106	CACNA1D	0.654643601			

Table 5.	Co-expression	pairs	between	IncRNA and	TSGs
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down_IncRNA	TSG	cor_value	up_IncRNA	TSG	cor_value
SFTA1P SFTA2P SFTA1P SFTA2P SF	SCGB3A1 LEFTY2 GADD45B CITED2 BCL6B CSRNP1 RHOBTB2 DOK2 COL4A3 RPS6KA2 ADAMT58 SUSD2 EDNRB SELENBP1 CDH5 GKN2 CASC1 PACRG KCNRG SNTN ZMYND10	0.557159 0.668367 0.682329 0.510908 0.544449 0.662952 0.565762 0.51381 0.655631 0.530706 0.750708 0.573774 0.581291 0.716833 0.67137 0.820088 0.682196 0.588571 0.603464 0.594486 0.645154	DLEU1 DLEU1 DLEU1 GAS5 GAS5 GAS5 GAS5 SNHG12 SNHG12 SNHG12 SNHG12	RPL15 SELT RBX1 NDUFA13 RPL15 EEF1A1 HINT1 TOMM20 DLEU1 RBM6 ARGLU1 RBM5 RASSF1	0.505824 0.572985 0.542194 0.508297 0.544716 0.505638 0.512073 0.508488 0.505154 0.593483 0.523387 0.559451 0.516762
CASC2 CASC2	C2orf40	0.590077			

Discussion

Long non-coding RNA (lncRNA), is a kind of RNAs whose length larger than 200 bases with biologic functions. LncRNAs can suppress translation by biding to their corresponding mRNAs and that's the main reason why they have an essential role in gene regulation. It's involved in the growth and pathological process by the way of chromatin reprogramming, cis regulation at enhancers or post-trtanscriptional regulation, which was similar with way of the encodinggenes. Since the close relation between gene expression patterns and tumor subtypes was essential in understanding the molecular basis of tumorigenesis,^{23, 24} the 2 differentially expressed IncRNAs between normal samples and patients with SCLC we identified in this study, SFTA1P and CASC2, may of much value in clinical management of the disease.

There were reports about abnormal expression of some IncRNA in the occurrence, invasion and development in many types of tumor tissues, indicating their association with the suppression of cancer. The expression degree of these IncRNAs may reflect the degree of prognosis process. Given their specialty in cancer tissues, these IncRNAs were regarded as prognostic biomarkers of cancer disease.

In this study, spermatogenesis was the most enriched GO terms in both up and downregulated IncRNAs. We speculated that there was much difference of differentially expressed IncRNAs between normal samples and cancer specimens, which thus affected the regulation of reproduction and inheritance process. Besides, the results of pathway analysis revealed that these IncRNAs were also involved in the pathways of cell cycle regulation and alexin.

Functional analysis of oncogenes revealed that they had large effect in signaling pathways associated with cytokine and DNA damage, indicating that there was large variation in genetic material of NSCC and activation of signaling pathways led to the deterioration of the cancer. Two IncRNA, SFTA1P and CASC2 were found to be associated with most of tumor supressor genes and oncogenes and had an great effect on the regulation of LSCC. It was reported that SFTA1P was downregulated in NSCC and its main function was lying on epidermal growth, cell attachment and response to DNA damage.²⁵ The other IncRNA, CASC2, was reported to be inhibited by miR-21 in gliomas and,^{26, 27} which could be the ceRNA of other oncogenes.

It was reported that lncRNAs could in cis regulate the expression of their neighboring genes,^{4, 28} we suspected the effect of these 2 genes may be related with this pathway, which needed more relevant research.

In summary, we identify 2 potential biomarkers, that is, SFTA1P and CASC2, associated with the regulation and development of lung squamous cell carcinoma, which could provide more specific and accurateprognostic and predictive indicators to clinical outcome of LSCC patients, implying their application in clinical diagnosis and treatment of this disease.

Conclusions

In summary, we identify 2 potential biomarkers, that is, SFTA1P and CASC2, associated with the regulation and development of lung squamous cell carcinoma, which could provide more specific and accurate prognostic and predictive indicators to clinical outcome of LSCC patients, implying their application in clinical diagnosis and treatment of this disease.

Disclosure of potential conflicts of interest

No potential conflicts of interest were disclosed.

Author contributions

GQ.H, ZP.K and B.G conceived and designed the experiments; GQ.H, ZP. K performed the experiments; ZP.K and B.G analyzed the data; Y.C contributed analysis tools; GQ.H, ZP.K, B.G and HB.H wrote the paper.

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