

# Identification of potential key genes and high-frequency mutant genes in prostate cancer by using RNA-Seq data

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**Abstract.** The aim of the present study was to identify potential key genes and single nucleotide variations (SNVs) in prostate cancer. RNA sequencing (RNA-seq) data, GSE22260, were downloaded from the Gene Expression Omnibus database, including 4 prostate cancer samples and 4 normal tissues samples. RNA-Seq reads were processed using Tophat and differentially-expressed genes (DEGs) were identified using the Cufflinks package. Gene Ontology enrichment analysis of DEGs was performed. Subsequently, Seqpos was used to identify the potential upstream regulatory elements of DEGs. SNV was analyzed using Genome Analysis Toolkit. In addition, the frequency and risk-level of mutant genes were calculated using VarioWatch. A total of 150 upregulated and 211 downregulated DEGs were selected and 25 upregulated and 17 downregulated potential upstream regulatory elements were identified, respectively. The SNV annotations of somatic mutations revealed that 65% were base transition and 35% were base transversion. At frequencies  $\geq 2$ , a total of 17 mutation sites were identified. The mutation site with the highest frequency was located in the folate hydrolase 1B (*FOLH1B*) gene. Furthermore, 20 high-risk mutant genes with high frequency were identified using VarioWatch, including ribosomal protein S4 Y-linked 2 (*RPS4Y2*), polycystin 1 transient receptor potential channel interacting (*PKD1*) and *FOLH1B*. In addition, kallikrein 1 (*KLK1*) and *PKD1* are known tumor suppressor genes. The potential regulatory elements and high-frequency mutant genes (*RPS4Y2*, *KLK1*, *PKD1* and

*FOLH1B*) may have key functions in prostate cancer. The results of the present study may provide novel information for the understanding of prostate cancer development.

## Introduction

Prostate cancer is the sixth leading cause of cancer-associated mortality in males worldwide, and is the second leading cause among males in the United States (1). The morbidity rates of prostate cancer vary widely across the world and are less frequent in South and East Asia, compared with that in Europe and United States (2). In spite of the high incidence, the etiology of prostate cancer remains unknown (3). Furthermore, early prostate cancer is typically asymptomatic, with ~66% of patients diagnosed with prostate cancer exhibiting no symptoms, preventing the early control of prostate cancer (4). The identification of novel therapeutic targets is required to diagnose early prostate cancer.

Established risk factors of prostate cancer include age, race and exhibiting a family history of prostate cancer (5). Genetic factors, which may increase the risk of developing prostate cancer, are associated with ethnicity and family history of the disease (6). A number of different genes have been implicated in the development of prostate cancer (7,8). A previous study validated that prostate cancer is prone to recurrent gene fusions of androgen-regulated genes, including transmembrane protease serine 2 (*TMPRSS2*), solute carrier family 45 member 3 and N-myc downstream regulated 1, and E26 transformation-specific (ETS) transcription factors, including ERG ETS transcription factor (*ERG*) and ETS variant 1 (9). In addition, loss of tumor suppressor genes [including Kruppel-like factor 6 (10), BTG anti-proliferation factor 3 (11) and phosphatase and tensin homolog (12)] and activation of oncogenes [including nuclear receptor coactivator 2 (*NCOA2*) and MYC proto-oncogene bHLH transcription factor] in the carcinogenesis of prostate cancer have been identified (13). Although previous studies have revealed a number of candidate genes, the causes of prostate cancer remain unclear.

Next-generation RNA sequencing (RNA-seq) enables the identification of genes that may be susceptible to prostate cancer (14). On the basis of RNA-seq data, the transcriptome profiles of primary prostate cancer are identified, including gene fusions, long non-coding RNAs, alternative splicing and somatic

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mutations (9). Ren *et al* (15) used RNA-seq to profile genetic aberrations in Chinese patients with prostate cancer and identify three recurrent gene fusions, including *TMPRSS2-ERG*, ubiquitin specific peptidase 9 Y-linked-testis-specific transcripts, Y-linked 15 on chromosome Y and the interchromosomal translocation of CTAGE family member 5 ER export factor-KH RNA binding domain containing signal transduction associated 3. Kannan *et al* (16) investigated chimeric RNAs expressed in human prostate cancer and obtained 1.3 billion sequence reads, which enabled the identification of 2,369 chimeric RNA candidates for distinguishing of prostate cancer. On the basis of this, Xu *et al* (17) identified 116 disruptive mutations in 92 genes with high confidence, including a frameshift insertion/deletion in the coding region of the TNF superfamily member 10 gene associated with apoptosis. However, differentially-expressed genes (DEGs) between prostate cancer samples and normal control samples were not investigated, and a number of somatic mutations remain unknown.

In the present study, raw RNA-seq data from the study by Kannan *et al* (16) was downloaded from the National Center for Biotechnology Information database and analyzed by a number of bioinformatics methods. First, DEGs between prostate cancer samples and normal control samples were identified. Subsequently, functional enrichment analysis of DEGs was performed, to understand the underlying molecular mechanisms of prostate cancer. Furthermore, the upstream regulatory elements of DEGs were identified and analyzed. In addition, single nucleotide variations (SNVs) were determined and the somatic mutation sites were annotated. The risk level of SNVs was assessed for selection of SNVs with high risk and high frequency. These candidate genes may contribute to further the understanding of prostate cancer.

## Materials and methods

**Raw RNA-seq data.** The RNA-seq data GSE22260 (16) were downloaded from the Gene Expression Omnibus ([www.ncbi.nlm.nih.gov](http://www.ncbi.nlm.nih.gov)) database, including 4 prostate cancer samples (GSM554078, GSM554082, GSM554086 and GSM554088) and 4 matched normal tissues samples (GSM554118, GSM554120, GSM554122 and GSM554124), respectively. The Gleason score of the four cancer samples were 7, 7, 7 and 6 (18). The sequencing platform was Illumina Genome Analyzer Ix (Illumina, Inc., San Diego, CA, USA). Reads were generated via a paired-end approach.

**Read alignment.** All RNA-Seq reads were mapped to the reference human genome (hg19) of the University of California Santa Cruz (Santa Cruz, CA, USA) by using Tophat software (19). Only the reads that mapped to specific genome locations were retained. A maximum of two mismatches in each read were permitted. Other parameters were set up according to the default settings of Tophat.

**DEG analysis.** On the basis of Refseq gene annotation (20), transcripts were assembled using Cufflinks (version 0.9.3) (21). Subsequently, the gene expression levels of transcripts were calculated using Cuffdiff (part of the Cufflinks package), on the basis of the fragments/kilobase/million reads method (22). The differentially expressed transcripts were identified by

Table I. Comparison between identified DEGs from prostate cancer samples and normal control samples.

Expression	DEGs	TF counts	TF genes
Downregulated	211	4	<i>IFI16, NEUROG3, RARG, SIM1</i>
Upregulated	150	4	<i>DMBX1, NCOA2, ONECUT2, ZNF83</i>

DEGs, differentially-expressed genes; TF, transcription factor.

calculating the fold change in the transcript and Student's t-tests in Cufflinks were performed to assess the difference. Transcripts with a log<sub>2</sub> fold change >2 and P<0.05 were considered DEGs.

**Function enrichment analysis of DEGs.** Gene Ontology (GO) (23) enrichment analysis was performed for the aforementioned DEGs, on the basis of the Database for Annotation, Visualization and Integrated Discovery database ([www.david.niaid.nih.gov](http://www.david.niaid.nih.gov)) (24). GO categories were classified into biological process (BP), molecular function and cellular component (CC) GO-terms. P<0.05 was set as the threshold criterion.

DEGs which demonstrated regulatory function were selected, labeled and inputted into the tumor suppressor gene (25) database ([bioinfo.mc.vanderbilt.edu/TSGene](http://bioinfo.mc.vanderbilt.edu/TSGene)) and the tumor-associated gene (26) database ([www.binfo.ncku.edu.tw/TAG/GeneDoc.php](http://www.binfo.ncku.edu.tw/TAG/GeneDoc.php)) to select cancer-related genes (tumor suppressor genes or oncogenes) for additional analysis.

**Upstream regulatory elements of DEGs.** In the present study, the upstream region (1.5 kb) of the transcription start site was defined as the promoter region. For the promoter region of up- and downregulated DEGs, motif identification was performed using Seqpos (27) to identify transcription factors. P<0.00001 and the frequency of motif targeted sequences >50% of up- and downregulated DEGs were set as the threshold criteria.

**Identification of SNVs.** The Genome Analysis Toolkit (28) software (available at [www.broadinstitute.org/gatk](http://www.broadinstitute.org/gatk)) was used to identify the SNVs. The coverage of each credible SNV was >5x. The minimum quality score of reliability SNV was 30 and SNVs with a quality score >50 were defined as high reliability SNVs. Based on the single nucleotide polymorphism (SNP) sites documented in dbSNP137 and 1,000 genome databases, the already known SNV callings in tumor tissues were removed. To remove the interference of RNA editing in the transcriptome, SNV calling was optimized by combining with RNA-seq data of normal controls.

**Somatic mutation sites annotation and high-risk mutation sites assessment.** The VarioWatch (29) software ([genepipe.ncgm.sinica.edu.tw/variowatch/main.do](http://genepipe.ncgm.sinica.edu.tw/variowatch/main.do)) was used to annotate the SNVs in coding sequence and assess the functional impact of gene products, on the basis of the risk

Table II. Ten most significant GO terms of up- and downregulated DEGs in prostate cancer.

A, Upregulated DEGs			
Category	Term	Count, n	P-value
CC	GO:0000786; nucleosome	5	0.0006
BP	GO:0006333; chromatin assembly or disassembly	6	0.0016
CC	GO:0032993; protein-DNA complex	5	0.0020
BP	GO:0006334; nucleosome assembly	5	0.0024
BP	GO:0031497; chromatin assembly	5	0.0027
BP	GO:0065004; protein-DNA complex assembly	5	0.0032
BP	GO:0034728; nucleosome organization	5	0.0034
BP	GO:0051223; regulation of protein transport	5	0.0070
BP	GO:0006323; DNA packaging	5	0.0077
CC	GO:0000785; chromatin	6	0.0083

B, Downregulated DEGs			
Category	Term	Count, n	P-value
CC	GO:0005902; microvillus	4	0.0033
CC	GO:0031226; intrinsic to plasma membrane	20	0.0055
BP	GO:0009100; glycoprotein metabolic process	7	0.0082
BP	GO:0009101; glycoprotein biosynthetic process	6	0.0120
CC	GO:0044459; plasma membrane part	29	0.0129
CC	GO:0046658; anchored to plasma membrane	3	0.0144
BP	GO:0006955; immune response	13	0.0162
CC	GO:0005887; integral to plasma membrane	18	0.0196
CC	GO:0005886; plasma membrane	43	0.0196
BP	GO:0043413; biopolymer glycosylation	5	0.0250

DEGs, differentially-expressed genes; GO, Gene Ontology; CC, cellular component; BP, biological process.

Table III. Potential upstream regulatory elements of DEGs in prostate cancer.

Expression	Count, n	Candidate TFs
Upregulated	25	<i>BARHL1, CD200, CEBPA, CUX1, EN1, ESX1, HNF4A, HOXA5, IRF8, LHX6, MEOX1, NOX1, NR1H4, PAX6, PBX1, PHOX2A, PITX2, PRKRA, RAX, RORA, SIX4, TCF7L2, VAX2, VSX2, ZEB1</i>
Downregulated	17	<i>CDX1, EPAS1, FOXC1, HMX1, HMX3, HOXA10, HOXA3, HOXC12, HSF1, HSF2, IKZF2, IRF1, IRF2, MSX1, NR3C1, POU2F1, ZEB1</i>

DEGs, differentially-expressed genes; TF, transcription factor.

assessment of the software. The SNVs with high-risk level of abnormal protein function were selected.

## Results

*Differential expression analysis and functional enrichment analysis.* According to the differential expression analysis of paired RNA-seq data between prostate cancer samples and normal control samples, a total of 150 upregulated and 211 downregulated DEGs were selected (Table I). On the basis of the annotation information of transcription factors, 8 differentially-expressed transcription factors were identified, 4 of which were upregulated [interferon- $\gamma$  inducible protein 16 (*IFI16*), neurogenin 3 (*NEUROG3*), retinoic acid receptor- $\gamma$  (*RARG*) and single-minded family bHLH transcription factor 1 (*SIMI1*)] and 4 were downregulated [diencephalon/mesencephalon homeobox 1 (*DMBX1*), nuclear receptor coactivator 2 (*NCOA2*), one cut homeobox 2 (*ONECUT2*) and zing finger protein 83 (*ZNF83*)] (Table I).

The most significantly enriched BP GO-terms of upregulated DEGs were chromatin assembly or disassembly

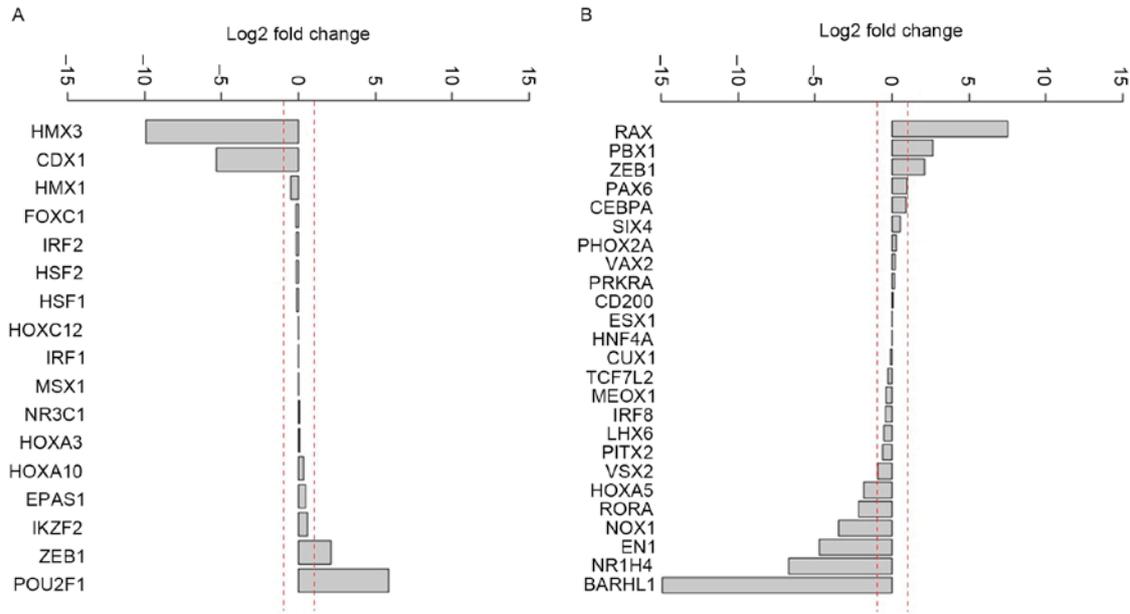


Figure 1. Differential expression of the potential upstream regulatory elements of DEGs in prostate cancer. (A) The expression of the upstream regulatory elements of downregulated DEGs. (B) The expression of the upstream regulatory elements of upregulated DEGs. The red dashed line represents log<sub>2</sub> fold change of 1 or -1. DEG, differentially-expressed gene.

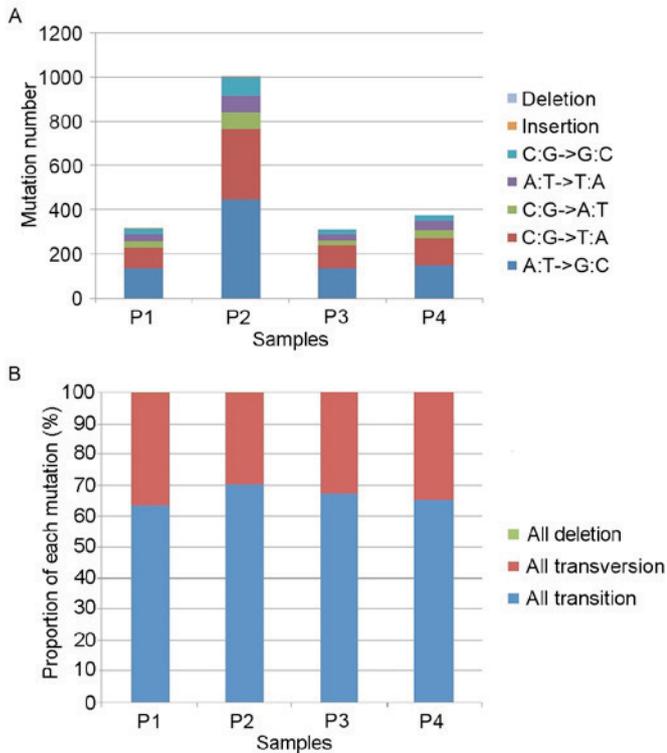


Figure 2. Statistical analysis of somatic mutations in 4 prostate cancer samples. (A) The number of each mutation type in 4 prostate cancer samples. (B) The proportion of each mutation type in 4 cancer samples. P1, P2, P3 and P4 represent the 4 prostate cancer samples.

( $P=0.0016$ ) and nucleosome assembly ( $P=0.0024$ ) (Table II). In addition, the most significant CC GO-terms identified were nucleosome ( $P=0.0006$ ) and protein-DNA complex ( $P=0.0020$ ). The results of the present study suggested that

Table IV. High-frequency mutation sites in prostate cancer.

Chr	Position	Base		Frequency, n
		Reference	Mutated	
Chr11	89431695	T	A	3
ChrY	22941450	A	C	2
ChrY	22921933	A	G	2
ChrX	51807602	G	A	2
ChrX	43634502	T	C	2
Chr9	96214432	G	C	2
Chr7	73245681	A	G	2
Chr6	29692069	C	A	2
Chr6	10749928	A	G	2
Chr19	51325077	G	T	2
Chr18	3253963	C	G	2
Chr18	21057145	C	G	2
Chr17	60631088	A	G	2
Chr16	2149965	G	A	2
Chr15	28421681	A	G	2
Chr1	89474720	T	C	2
Chr1	147580839	A	G	2

Chr, chromosome.

genes associated with chromatin structure stability may be highly expressed in prostate cancer.

Of the downregulated DEGs, BP GO terms were significantly enriched in glycoprotein metabolic process ( $P=0.0082$ ), glycoprotein biosynthetic process ( $P=0.0120$ ) and immune response ( $P=0.0162$ ); whereas, CC GO terms were

Table V. Frequency of high-risk genes in prostate cancer.

High-risk mutated gene	Gene name	Frequency	Expression level	Tumor-associated gene
<i>PPP4R2</i>	Protein phosphatase 4, regulatory subunit 2	2	-	-
<i>CLDN4</i>	Claudin 4	2	Upregulated <sup>a</sup>	Tumor-associated <sup>a</sup>
<i>FOLH1B</i>	Folate hydrolase 1B	3	Downregulated <sup>a</sup>	Tumor-associated <sup>a</sup>
<i>KIF27</i>	Kinesin family member 27	2	-	-
<i>MAOB</i>	Monoamine oxidase B	2	Upregulated <sup>a</sup>	Tumor-associated <sup>a</sup>
<i>DUXA</i>	Double homeobox A	2	-	-
<i>KLK1</i>	Kallikrein 1	2	Upregulated	Tumor suppressor gene
<i>MYL12A</i>	Myosin, light chain 12A, regulatory, non-sarcomeric	2	-	-
<i>HERC2</i>	Hect domain and RLD 2	2	-	-
<i>HLA-E</i>	Major histocompatibility complex, class I, E	2	Downregulated <sup>a</sup>	Tumor suppressor gene <sup>a</sup>
<i>HLA-G</i>	Major histocompatibility complex, class I, G	2	Downregulated <sup>a</sup>	Tumor-associated <sup>a</sup>
<i>RIOK3</i>	RIO kinase 3	2	-	-
<i>RPS4Y2</i>	Ribosomal protein S4, Y-linked 2	3	-	-
<i>PKD1</i>	Polycystic kidney disease 1 (autosomal dominant)	3	Upregulated <sup>a</sup>	Tumor suppressor gene
<i>USP10</i>	Ubiquitin specific peptidase 10	2	-	-
<i>TLK2</i>	Tousled-like kinase 2	2	-	-
<i>TMEM14B</i>	Transmembrane protein 14D; transmembrane protein 14B	2	-	-
<i>TUBA1A</i>	Tubulin, $\alpha$ 1a	2	-	-
<i>GBP3</i>	Guanylate binding protein 3	2	-	-
<i>MAGED4B</i>	Melanoma antigen family D, 4B; melanoma antigen family D, 4	2	-	Tumor-associated gene

<sup>a</sup>Data provided by the OncoSearch database (<http://oncosearch.biopathway.org/#/search>).

significantly enriched in the microvillus ( $P=0.0033$ ), intrinsic to plasma membrane ( $P=0.0055$ ) and plasma membrane part ( $P=0.0129$ ) (Table II).

*Identification of the upstream regulatory elements of DEGs.* Motif scanning of the upstream regulatory elements of up-regulated DEGs revealed 25 transcription factors that may regulate these DEGs (Table III). Additionally, 17 transcription factors were identified to regulate downregulated DEGs (Table III).

Differential expression analysis of the potential upstream regulatory elements demonstrated that the expression of H6 family homeobox 3 and caudal type homeobox 1 were markedly decreased (Fig. 1A). In addition, the following genes were activated in different degrees: Retina and anterior neural fold homeobox, PBX homeobox 1, zinc finger E-box binding homeobox 1, paired box 6 and CCAAT/enhancer binding protein- $\alpha$  (Fig. 1B).

*Detection of SNVs in prostate cancer.* Following the removal of polymerase chain reaction duplicates and data processing of RNA-seq in 4 prostate cancer samples, 317, 1,004, 310 and 373 somatic mutation sites were identified in 4 prostate cancer samples particularly (Fig. 2). In each sample, base transversion (C:G  $\rightarrow$  A:T; A:T  $\rightarrow$  T:A; C:G  $\rightarrow$  G:C) and base transition (A:T  $\rightarrow$  G:C; C:G  $\rightarrow$  T:A) were the most common types of base mutation (Fig. 2A). Only one insertion site was identified in the 4 prostate cancer samples. The proportions of each mutation type in the 4 prostate cancer samples were similar, suggesting that 65% of mutations were base transition and 35% were base transversion (Fig. 2B).

Mutation sites that occurred in  $>2$  prostate cancer samples were defined as a high-frequency site (frequency  $\geq 2$ ). A total of 17 mutation sites were identified as high-frequency sites (Table IV). The T  $\rightarrow$  A base transversion of chromosome 1189431695 was identified in 3 cancer samples and its loci was located in the folate hydrolase 1B (*FOLH1B*) gene. An

additional 16 mutation sites were detected in 2 cancer samples and loci were all located in the coding sequence (CDS) region of genes.

The frequency of high-risk mutant genes was calculated using VarioWatch (Table V). A total of 20 high-risk mutant genes were identified, of which ribosomal protein S4 Y-linked 2 (*RPS4Y2*), polycystin 1 transient receptor potential channel interacting (*PKDI*) and *FOLHIB* were identified in 3 cancer samples. Furthermore, kallikrein 1 (*KLK1*) and *PKDI* were known tumor suppressor genes (30,31). Although the expression level of the *KLK1* gene was increased in prostate cancer, the high frequency of mutation located in the CDS region may lead to dysfunction of the tumor suppressor. However, the expression levels of other 19 high-risk genes were not markedly different in prostate cancer, indicating that the abnormal function of these genes, induced by base mutation, may be associated with the development of prostate cancer.

## Discussion

The present study provided a survey of DEGs and mutant genes in human prostate cancer. RNA-seq data between 4 prostate cancer samples and 4 matched normal samples were analyzed. A total 150 upregulated and 211 downregulated DEGs were identified, 4 of which were upregulated transcription factors (*DMBX1*, *NCOA2*, *ONECUT2* and *ZNF83*), and 4 were downregulated transcription factors (*IFI16*, *NEUROG3*, *RARG* and *SIMI*).

*NCOA2* has been suggested as an oncogene in primary tumors by increasing androgen receptor signaling, which is known to have a function in early and late-stage prostate cancer (13). Furthermore, the upregulated expression of transcription factor *ONECUT2* was revealed in breast and prostate cancer cell lines (32), and increased *IFI16* protein in normal human prostate epithelial cells was associated with cellular senescence-associated cell growth arrest (33). *NEUROG3* was identified to be expressed in metastatic neuroendocrine prostate cancer cells (34). However, mechanism of abnormal regulation of *DMBX1*, *ZNF83*, *RARG* and *SIMI* in prostate cancer remains unknown. For example, the abnormal regulation of *ZNF83* has been identified in hepatocellular carcinoma (35) and colorectal cancer (36), but not in prostate cancer. *DMBX1* is a paired-class homeodomain transcription factor and may be determined in the brain, stomach and testis in adult normal tissues (37). A previous study focused on the expression pattern of *DMBX1* in the development of the neural network (38). Therefore, these aforementioned transcription factors may have important functions in the progress of prostate cancer.

Determining the SNVs in prostate cancer identified 17 mutation sites with a frequency  $\geq 2$ . The mutation site with the highest frequency was located in *FOLHIB*. Furthermore, 20 high-risk mutant genes with highest frequency were identified using VarioWatch and included *RPS4Y2*, *PKDI* and *FOLHIB*. *FOLHIB* originates from the duplication of folate hydrolase 1 (*FOLH1*) (39), which is an established biomarker for prostate cancer (40). *FOLH1*, also known as prostate-specific membrane antigen 1, is used as a diagnostic and prognostic indicator for prostate cancer, and is associated with aggressiveness and metastasis of prostate cancer (41). Of

the identified high-risk mutant genes exhibiting the highest frequency, *KLK1* and *PKDI* are known tumor suppressor genes and MAGE family member D4B is a tumor-associated gene (30,31). *PKDI* has been identified to be downregulated in advanced-stage prostate cancer and was present as a protein complex, combined with the androgen receptor, in prostate cancer cells (42). The kallikrein-related peptidases have been identified in a number of types of cancer, including prostate and ovarian, and a combination of the dysregulation of *KLK1*, 5 and 13 was associated with poorer disease-free survival for prostate cancer (43). The high frequency of mutation located in *KLK1*, *PKDI* and *D4B* may lead to dysfunction of the tumor suppressor functions and consequently contribute to the progress of prostate cancer.

There were a number of limitations in the present study; for example, the sample size used for the analysis was small. Furthermore, the results of the present study require validation using other RNA-seq data or microarray data of prostate cancer.

Combined with bioinformatics methods, the RNA-seq data were analyzed to determine candidate genes for diagnosing and/or treating of prostate cancer. The identified DEGs (*DMBX1*, *ZNF83*, *RARG* and *SIMI*) and mutant genes (*RPS4Y2*, *KLK1* and *FOLHIB*) may have important functions in the progression of prostate cancer. The results of the present study may enable an improved understanding of the molecular mechanisms that underlie prostate cancer pathogenesis.

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