Prognostic value and potential function of splicing events in prostate adenocarcinoma

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Abstract. Prostate adenocarcinoma (PRAD) is one of the most common types of malignancy in males and at present, effective prognostic indicators are limited. The development of PRAD has been associated with abnormalities in alternative splicing (AS), a requisite biological process of gene expression in eukaryotic cells; however, the prognostic value of AS products and splicing events remains to be elucidated. In the present study, the data of splicing events and the clinical information of PRAD patients were obtained from The Cancer Genome Atlas (TCGA)SpliceSeq and TCGA databases, respectively. A prognostic index (PI) was generated from disease-free survival-associated splicing events (DFS-SEs), which were identified by univariate/multivariate Cox regression analysis. A total of 6,909 DFS-SEs were identified in PRAD. The corresponding genes for the DFS-SEs were significantly enriched in mitochondria and their associated pathways according to Gene Ontology annotation and in the pathways of fatty acid metabolism, oxidative phosphorylation and Huntington's disease according to a Kyoto Encyclopedia of Genes and Genomes pathway analysis. The PI for mutually exclusive exons had the greatest ability to predict the probability of five-year disease-free survival of patients with PRAD, with an area under the time-dependent receiver-operating characteristic curve of 0.7606. Patients with PRAD, when divided into a ‘low’ and a ‘high’ group based on their median PI for exon skip values, exhibited a marked difference in disease-free survival (low vs. high, 3,588.45±250.51 vs. 1,531.08±136.50 days; P=7.43x10⁻⁹). A correlation network between DFS-SEs of splicing factors and non-splicing factors was constructed to determine the potential mechanisms in PRAD, which included the potential regulatory interaction between the splicing event of splicing factor RNA binding motif protein 5-alternate terminator (AT)-64957 and the splicing event of non-splicing factor heterochromatin protein 1 binding protein 3-AT-939. In conclusion, the PIs derived from DFS-SEs are valuable prognostic factors for patients with PRAD, and the function of splicing events in PRAD deserves further exploration.

Introduction

Prostate adenocarcinoma (PRAD) is one of the most common malignant tumor types of the male reproductive system and primarily occurs in elderly individuals (aged >65 years) (1,2). According to the most recent global statistical data, the number of newly diagnosed PRAD cases in 2012 has reached 1.1 million (3). The morbidity and mortality associated with PRAD are higher in developed than in developing countries. In the US, the predicted number of new cases of PRAD for 2018 is 164,690 and the number of associated mortalities is 29,430 (4). The mortality from PRAD is substantially reduced by early screening for prostate-specific antigen (PSA); However, the number of individuals who die from PRAD is second only to the number of patients that die from lung cancer, and PRAD therefore ranks second among cancer-associated deaths in males (4,5). According to cancer statistics for 2015, 60,300 individuals were diagnosed with PRAD and 26,600 patients succumbed to the disease in China. While these figures are lower than those for the US, the incidence of PRAD is rapidly increasing each year in China (6).

Only a small number of indicators are currently used for predicting the prognosis of PRAD patients, and each indicator has its own advantages and disadvantages. PSA is currently the most commonly used diagnostic screening and prognostic indicator for PRAD; however, its application has poor specificity (7,8). Circulating tumor cells provide excellent prognostic evaluation of PRAD, but their use is limited by high equipment requirements and costs (9). micro (mi)RNAs have also been demonstrated to have a high prognostic value for PRAD, but the technology for the extraction and identification of relevant miRNAs in limited samples remains to be fully developed and implemented (8,10). Therefore, novel and effective predictors for the prognosis of patients with PRAD are urgently required.

One strategy for identifying prognostic predictors for PRAD is the assessment of alternative splicing (AS), but this
field has remained largely unexplored. AS is an indispensable process in prokaryotic gene expression (11) and is able to turn mRNA precursors into different types of mature mRNAs through different processes to increase the diversity of mRNA types (12). Increasing evidence indicates that AS has an important role in the development of PRAD (13-16). The splice isoforms generated by AS are involved in different aspects of tumor physiology, including growth, apoptosis, infiltration, metastasis, angiogenesis and metabolism (14). The study of AS in tumors is helping to elucidate the mechanisms underlying tumor pathogenicity and may be a strategy in the search for reliable and effective diagnostic and prognostic targets. AS is classified into the following 7 types, based on the type of splicing: Alternate acceptor site (AA), alternate donor site (AD), alternate promoter (AP), alternate terminator (AT), exon skip (ES), mutually exclusive exons (ME) and retained intron (RI). The products of AS are termed splicing events (17,18). In the present study, the percent-spliced-in (PSI) value (ranging from 0 to 1) was used to perform an intuitive quantitative analysis and comparison of splicing events (18,19).

AS involves the use of splicing factors as executive proteins, so that changes in splicing factor expression directly cause an abnormal expression of splicing events (20). Numerous studies have confirmed that mutations in splicing factors are closely associated with the development and progression of tumors (21,22) and that the expression levels of splicing factors differ significantly between tumor cells and juxtacancerous tissues (20,23). For instance, serine- and arginine-rich splicing factor 1 was reported to be highly expressed in various tumor types (24), and pre-mRNA processing factor 6 was indicated to promote tumor cell growth (25). Therefore, exploring the potential regulatory association between splicing factors and splicing events may aid in identifying the pathogenic mechanisms underlying tumor development.

In recent years, the analysis of splicing events in tumor research has become possible with the introduction of deep sequencing techniques. The Cancer Genome Atlas (TCGA) database (https://portal.gdc.cancer.gov/) now provides RNA-sequencing data of various types of tumor and juxtacancerous tissues, while no normal tissues were included. Ryan et al (26) used the RNA-seq data provided by TCGA to establish a TCGASpliceSeq database (http://bioinformatics.mdanderson.org/TCGASpliceSeq/index.jsp), which offers a convenient way for researchers to study splicing events in tumors. The aim of the present study was to perform a re-calculation and in-depth analysis to determine the disease-free survival-associated splicing events (DFS-SEs) in patients with PRAD. The present study endeavored to construct a prognostic index (PI) for DFS-SEs and to evaluate the prognostic values of these indicators in PRAD using time-dependent receiver-operator characteristic (tROC) and Kaplan-Meier curve analyses. In addition, a correlation analysis was used to construct a potential regulatory network to explain the associations between splicing factors and splicing events.

Material and methods

Data acquisition and organization. The splicing event data for PRAD were downloaded from the TCGASpliceSeq database. In addition, the clinical data for 500 patients with PRAD were acquired from TCGA database.

In total, 91 patients with a ‘Person neoplasm cancer status’ of ‘Discrepancy’, ‘Not Available’ or ‘Unknown’, 3 patients with a ‘patient death reason’ of ‘Other, non-malignant disease’ or ‘Unknown’ and 9 patients with ‘<30’ in ‘days to last followup’ were excluded, leaving 397 patients. After matching the patients with their TCGASpliceSeq database entries according to their labeling numbers, 394 patients were included in the present study.

‘With tumor’ in ‘person neoplasm cancer status’ was regarded as the endpoint. The DFS time in ‘days to last followup’ was used to calculate the DFS of patients with PRAD.

Identification and analysis of DFS-SEs. Univariate Cox regression analysis was performed to analyze and identify DFS-SEs. Cytoscape software was used to plot the interactive association between the genes corresponding to DFS-SEs. The Database for Annotation, Visualization and Integrated Discovery (DAVID) Functional Annotation Result Summary (https://david.ncifcrf.gov/summary.jsp, version 6.8) was used for Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analysis of the genes corresponding to the top 500 most significant DFS-SEs in PRAD. The top five pathways identified by the GO and KEGG analyses were displayed.

Construction of the PI model for PRAD based on splicing events. To construct powerful PIs, multivariate Cox regression analysis was used to exclude irrelevant splicing events. Multivariate Cox regression analyses were performed for the top 10 DFS-SEs that had the highest prognostic values for each splicing type and the top 10 DFS-SEs that had the highest prognostic values for all splicing types. Splicing events with P<0.05 were selected to constitute PIs. The calculation formula was as follows:

\[ PI = \sum_{i=1}^{n} PSI_i \times \beta_i \]

where \( \beta \) is the regression coefficient.

Evaluation of the prognostic value of PIs. The efficacy of the PIs in predicting the cancer status after five years was evaluated by tROC analysis with the R package ‘survivalROC’ (27). Kaplan-Meier curve analysis was performed for the DFS time and cancer status of the patients with PRAD to evaluate the prognostic efficacy of the most significant top 10 DFS-SEs and PIs. Univariate Cox regression analysis was used to calculate the hazard ratio (HR) values of the PIs and other clinical parameters for PRAD recurrence.

Construction of a correlation network of DFS-SEs in PRAD. Information on the splicing factors was downloaded from the SpliceAid2 database (http://www.introni.it/splicing.html) (28) and included 66 splicing factors. All splicing events for the splicing factors were identified based on the results of the univariate Cox regression analysis. A total of 22 splicing events with P<0.01 were selected. In addition, the top 50 splicing events with the highest prognostic value according to the univariate Cox analysis were assessed, and events with
P<1x10^{-20} were inputted into Cytoscape 3.5.1 (The Cytoscape Consortium, New York, NY, USA) to construct the network.

Statistical analysis. UpSet was used to visualize the intersection between genes and the 7 splicing types (29). SPSS software version 22 (IBM Corp., Armonk, NY, USA) was used for the univariate and multivariate Cox regression, Kaplan-Meier curve and Pearson correlation analyses. The tROC analysis was performed with R (version 3.4.3). P<0.05 was considered to indicate a statistically significant difference. GraphPad 7.0 (GraphPad Inc., La Jolla, CA, USA) was used to plot the ROC and Kaplan-Meier curves.

Results

Comprehensive analysis of splicing events in the PRAD cohort from TCGA dataset. In the present study, a total of 44,070 splicing events were identified in the cohort of PRAD patients, including 3,524 of the AA type, 3,101 of the AD type, 9,035 of the AP type, 8,663 of the AT type, 16,772 of the ES type, 228 of the ME type and 2,747 of the RI type (Table I). UpSet, which is similar to a Venn diagram, was used to illustrate the splicing events in PRAD (Fig. 1).

Single splicing events were accurately described using the labelled name of each splicing event, which contained the gene name, splicing type and AS ID. For instance, for pleckstrin homology domain containing N1 (PLEKHN1)-ES-1, the gene name is PLEKHN1, the splicing type is ES and the AS ID is 1.

DFS-SEs in PRAD identified from TCGA. In total, 6,909 splicing events were identified as DFS-SEs by univariate Cox proportional hazards regression analysis; these included 373 of the AA type, 8,663 of the AD type, 1,474 of the AP type, 2,275 of the AT type, 1,956 of the ES type, 42 of the ME type and 425 of the RI type (Table I). The HRs of the top 10 splicing events for PRAD recurrence, ranked by P-value for each splicing type, are presented in Fig. 2. Kaplan-Meier curve analyses were performed for the 10 most significant splicing events among all the DFS-SEs. Patients with PRAD were divided into ‘high’ and ‘low’ groups based on the median PSI value of a certain splicing event. The patients with PRAD who had a high PSI value for elongin B-ES-33303, FK506 binding protein 2 (FKBP2)-AP-16603, NHL repeat containing 3-ES-25701 and thioredoxin-ES-87183 also had a relatively high chance of DFS, whereas patients with high PSI values for abhydrolase domain containing 17A-AD-46558, FKBP2-AP-16602, yippee like 3-AD-36074, high mobility group AT-hook 2-AT-22879 and prostaglandin D2 synthase-AT-88235 had a reduced probability of DFS (Fig. 3).

The UpSet was also used to quantitatively display the intersection of genes and splicing types of the DFS-SEs in PRAD (Fig. 4A). Numerous splicing events of certain genes, including inositol hexakisphosphate kinase 2, mammalian target of rapamycin-associated protein, LST8 homolog and autophagy related 16 like 2, were associated with DFS. Selection of the top 500 DFS-SEs and input of their corresponding 356 genes into Cytoscape for analysis provided the gene interaction network (Fig. 4B). Of note, a total of 356 counterpart genes of the 500 most significant DFS-SEs were linked to mitochondria and associated pathways according to GO annotation, including ‘mitochondrial electron transport, NADH to ubiquinone’, ‘mitochondrial translational elongation’ and ‘mitochondrial translational termination’ in the category Biological Process (BP) and ‘mitochondrion’, ‘mitochondrial inner membrane’ in the category Cellular Component (CC) (Fig. 5).
PI models based on the DFS-SEs in patients with PRAD. Multivariate Cox regression analyses were performed based on the top 10 significant DFS-SEs for each of the seven splicing types and for all splicing types. PIs constructed based on splicing events of the AA, AD, AP, AT, ES, ME, RI and all splicing types, and their associated data are presented in Figs. 6-13, respectively. For simplification, these eight PIs were abbreviated as PI-AA, PI-AD, PI-AP, PI-AT, PI-ES, PI-ME, PI-RI and PI-ALL, respectively. The calculation of PIs was performed according to the formula mentioned above.

Prognostic value of PIs in patients with PRAD. The tROC curve analyses indicated that PI-ME was the most effective PI at predicting the cancer status after five years, with an AUC value of 0.7606 (Fig. 13D). This was followed by PI-ALL and PI-ES, which exhibited AUC values of 0.7558 (Fig. 10E) and 0.7403 (Fig. 6E), respectively. The median values of the eight PIs were then used to classify the patients with PRAD into low- and high-level groups for the Kaplan-Meier curve analyses. The low and high groups into which the patients with PRAD were stratified based on the median value of PI-ES exhibited the most significant difference in DFS (low vs. high, 3,588.45±250.51 vs. 1,531.08±136.50 days; P=7.43x10^-9). After this, the most significant differences between low and high groups were those with stratification based on PI-ME and PI-AD (Table II).
The prognostic value of the PIs and other clinical parameters, including age, Gleason score, PSA value, pathologic T stage and pathologic N stage, were then analyzed using univariate Cox analysis. Except for age, the other clinical parameters and the eight PIs all had a high predictive value regarding the DFS of patients with PRAD (Table III). In addition, the multivariate analysis revealed that only the pathologic T stage, PI-AT, PI-ME and PI-ALL had an independent significant prognostic value for predicting the DFS of patients with PRAD (Table III).

Potential correlation network of DFS-SEs in PRAD. Splicing factors are key genes for the regulation of the development of splicing events (30). Splicing factors also have their own corresponding splicing events. A correlation analysis was used to examine the potential regulatory association between the splicing events of splicing factors and non-splicing factors and to construct a regulatory network.

A total of 66 splicing factors were obtained from the SpliceAid2 database. Based on the univariate Cox analysis, 22 splicing events of the 66 splicing factors with P<0.01 were selected, as well as the most significant top 50 splicing events of non-splicing factors with P<2.5x10^{-6}. The correlations between the PSI values of the 22 splicing events of the splicing factors and the PSI values of the top 50 splicing events of non-splicing factors were calculated. Splicing events with P<1x10^{-20} according to the correlation analysis were then introduced into Cytoscape to generate the network (Fig. 14A). The red nodes in the graph indicate the splicing events of the splicing factors (n=17) and the light blue nodes indicate the splicing events associated with poor prognosis. The triangular nodes indicate the splicing events associated with good prognosis and the oval nodes indicate the splicing events associated with poor prognosis. The blue lines indicate negative correlations between two splicing events and the green lines indicate positive correlations between two splicing events.

The correlation network revealed a complex association between the splicing events of splicing and non-splicing factors. For instance, the splicing event of splicing factor serine and arginine-rich splicing factor 2-RI-43660 was positively correlated with the splicing event of PI-AT-22879 and negatively correlated with the splicing event of PI-ME-25701. The red nodes in the graph indicate the splicing events of the splicing factors (n=17) and the light blue nodes indicate the splicing events associated with poor prognosis. The triangular nodes indicate the splicing events associated with good prognosis and the oval nodes indicate the splicing events associated with poor prognosis. The blue lines indicate negative correlations between two splicing events and the green lines indicate positive correlations between two splicing events.

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correlated with ectonucleoside triphosphate diphosphohydrolase 5 (ENTPD5)-AT-28353 and was negatively correlated with ENTPD5-AT-28354. Of note, the splicing events with good DFS were predominantly negatively correlated with the splicing events of the splicing factors, whereas the splicing events with poor DFS were predominantly positively correlated with the splicing events of the splicing factors (Fig. 14A).

Fig. 14B and C also present two pairs of splicing events of splicing and non-splicing factors with the most significant positive and negative correlations among all combinations and their Kaplan-Meier curve analyses. The PSI values of RNA binding motif protein 5 (RBM5)-AT-64957 and heterochromatin protein 1 binding protein 3 (HP1BP3)-AT-939 were negatively correlated (Fig. 14B), while the PSI values of RBM5-AT-64955 and HP1BP3-AT-939 were positively correlated (Fig. 14C). Patients with PRAD were then divided into low and high groups based on the median PSI value of certain splicing events. PRAD patients in the low group

Figure 4. UpSet plots of DSEs and gene interaction network in PRAD. (A) UpSet plots of DSEs in PRAD. (B) Gene interaction network based on counterpart genes of the top 500 DSEs in PRAD. DSEs, disease-free survival-associated splicing events; PRAD, prostate adenocarcinoma; AA, alternate acceptor; AD, alternate donor; AP, alternate promoter; AT, alternate terminator; ES, exon skip; ME, mutually exclusive exons; RI, retained intron.
of RBM5-AT-64957 had a significantly higher DFS than those in the high group (Fig. 14D), while PRAD patients in the high RBM5-AT-64955 or HP1BP3-AT-939 group had a markedly higher DFS than those in the respective low group (Fig. 14E and F).

Discussion

AS is an important regulatory mechanism of gene expression, and abnormalities of this mechanism may cause the development of various diseases (14), including the entirety of the
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processes of cancer development and progression. Certain products of AS may therefore be used as diagnostic and prognostic indicators, as well as therapeutic targets for cancer (14,20). A small number of studies have been published on the use AS in...
PRAD, but these have primarily focused on splicing variants of the androgen receptor (14,15,31-33), proprotein convertase subtilisin/kexin type 6 (34), CD44 (35) and staphylococcal nuclease and tudor domain containing 1 (36). However, studies...
Figure 11. Analysis of PI-ME. The five splicing events of the ME type from which the PI-ME was generated, were identified from DFS-associated splicing events by multivariate Cox regression analysis. (A) Illustration of ME site. (B) Scatter plot of time of DFS vs. cancer status of PRAD patients. (C) PI value curve for patients with PRAD. (D) Time-dependent receiver-operator characteristic curve of PI for predicting the tumor status after five years. (E) Kaplan-Meier curve of PI for evaluating the DFS time of PRAD patients stratified by median PI value into low and high group. PI-ME, prognostic index based on the Mutually Exclusive Exons type of splicing events; PRAD, prostate adenocarcinoma; AUC, area under curve; DFS, disease-free survival; TBRG1, transforming growth factor β regulator 1; GTF2H3, general transcription factor IIH subunit 3; CCT3, chaperonin containing TCP1 subunit 3; VSTM1, V-set and transmembrane domain containing 1; ANKRD10, ankyrin repeat domain 10.

Figure 12. Analysis of PI-RI. The two splicing events of the RI type from which the PI-RI was generated, were identified from DFS-associated splicing events by multivariate Cox regression analysis. (A) Illustration of RI site. (B) Scatter plot of time of DFS vs. cancer status of PRAD patients. (C) PI value curve for patients with PRAD. (D) Time-dependent receiver operating characteristic curve of PI for predicting the tumor status after five years. (E) Kaplan-Meier curve of PI for evaluating the DFS time of PRAD patients stratified by median PI value into low and high group. PI-RI, prognostic index based on the Retained Intron type of splicing events; PRAD, prostate adenocarcinoma; AUC, area under curve; DFS, disease-free survival; CCL14, C-C motif chemokine ligand 14; CSAD, cysteine sulfinic acid decarboxylase.
providing a comprehensive evaluation of splicing events in PRAD are scarce.

In the present study, splicing events in PRAD were comprehensively evaluated based on data from the TCGASpliceSeq database. In those analyses, the total number of splicing events greatly exceeded the number of genes, indicating that AS markedly increases the diversity of gene expression products in PRAD, as has been observed in other diseases (14,20,37). One of the highlights of the present study was the comprehensive and systematic presentation of all splicing events and DFS-SEs in PRAD, which may provide clues for studying the functions of AS and its products in PRAD.

The present study also analyzed the functions of genes corresponding to DFS-SEs in PRAD, indicating that these genes were mainly enriched in the fatty acid (FA) metabolism and oxidative phosphorylation pathways. An increasing number of cohort studies and experiments have confirmed a close association between FA metabolism and PRAD. For instance, a cohort study by Brasky et al (38) indicated an increased risk for the development of PRAD in males who had high blood concentrations of long-chain ω-3 PUFAs, particularly eicosapentaenoic acid, in ω-3 species to induce cancer cell death (42).

Table I. Overview of the splicing events in prostate adenocarcinoma (n).

| Type   | Splicing events | Genes | Splicing events | Genes |
|--------|----------------|-------|----------------|-------|
| AA     | 3524           | 2488  | 373            | 341   |
| AD     | 3101           | 2185  | 364            | 325   |
| AP     | 9035           | 3621  | 1474           | 826   |
| AT     | 8663           | 3781  | 2275           | 1171  |
| ES     | 16772          | 6577  | 1956           | 1466  |
| ME     | 228            | 221   | 42             | 42    |
| RI     | 2747           | 1849  | 425            | 375   |
| Total  | 44070          | 10380 | 6909           | 3645  |

DFS-SEs, disease-free survival-associated splicing events; AA, alternate acceptor; AD, alternate donor; AP, alternate promoter; AT, alternate terminator; ES, exon skip; ME, mutually exclusive exons; RI, retained intron.

Table II. Analysis of Kaplan-Meier curves of PI for evaluating the DFS time of patients with prostate adenocarcinoma stratified by the PI value (cutoff at median).

| Type/group | DFS time (days) | P-value |
|------------|-----------------|---------|
| PI-AA      |                 |         |
| Low (<-5.88) | 2583.742±174.316 | 4.13x10^{-4} |
| High (≥5.88) | 1778.112±200.024 |       |
| PI-AD      |                 |         |
| Low (<-8.67) | 3306.887±246.458 | 5.99x10^{-4} |
| High (≥8.67) | 1704.737±145.898 |       |
| PI-AP      |                 |         |
| Low (<-4.89) | 1721.248±135.488 | 1.95x10^{-2} |
| High (≥4.89) | 2741.437±305.572 |       |
| PI-ME      |                 |         |
| Low (<-1.56) | 3412.259±231.568 | 2.95x10^{-7} |
| High (≥1.56) | 1450.951±129.526 |       |
| PI-RI      |                 |         |
| Low (<-0.87) | 2395.445±160.811 | 4.59x10^{-7} |
| High (≥0.87) | 1784.987±247.797 |       |
| PI-ALL     |                 |         |
| Low (<-491.40) | 2751.999±171.559 | 1.45x10^{-7} |
| High (≥-491.40) | 1700.247±184.410 |       |

PI, prognostic index; DFS, disease-free survival; PI-ALL, PI based on all seven types of splicing events; PI-AA, PI based on AA type of splicing events; AA, alternate acceptor; AD, alternate donor; AP, alternate promoter; AT, alternate terminator; ES, exon skip; ME, mutually exclusive exons; RI, retained intron.
was 0.7403 for predicting the cancer status after five years. In addition, the HR of PI-ES for tumor recurrence was 4.097 (95% CI: 2.439-6.881, P=9.77x10^-8) according to univariate Cox analysis and 2.599 (95% CI: 1.329-5.084, P=0.005) according to multivariate Cox analysis, which were better than previously reported values focusing on only a single indicator, including programmed death-1 receptor methylation (44), pituitary homeobox 3 methylation (45) and cysteine dioxygenase 1 promoter methylation (46).

The functions of different transcripts of the same gene may differ. Therefore, an increasing number of studies have been assessing the functions of specific mature mRNAs or

Table III. Logistic regression analysis of the association of clinical parameters and the PIs with the risk of prostate adenocarcinoma recurrence.

| Clinical feature | Univariate Cox analysis | Multivariate Cox analysis |
|------------------|-------------------------|--------------------------|
|                  | HR (95% CI)             | P-value                  | HR (95% CI)             | P-value                  |
| Age (≥61 vs. <61 years) | 1.463 (0.952-2.246) | 8.20x10^-2              | 1.103 (0.663-1.834) | 7.06x10^-1              |
| Gleason score (≥8 vs. <8) | 4.515 (2.738-7.444) | 3.47x10^-9              | 1.429 (0.689-2.961) | 3.38x10^-1              |
| PSA value (≥0.1 vs. <0.1) | 5.587 (3.451-9.043) | 2.54x10^-12             | 2.946 (1.702-5.099) | 1.14x10^-4              |
| Pathologic T stage (T3/T4 vs. T1/T2) | 7.010 (3.233-15.199) | 8.17x10^-7              | 4.180 (1.403-12.450) | 1.00x10^-1              |
| Pathologic N stage (N1 vs. N0) | 2.635 (1.666-4.166) | 3.42x10^-5              | 1.024 (0.589-1.780) | 9.34x10^-1              |
| PI-AA (≥-5.88 vs. <5.88) | 2.531 (1.598-4.008) | 7.57x10^-8              | 1.588 (0.847-2.980) | 1.50x10^-1              |
| PI-AD (≥-8.67 vs. <8.67) | 2.479 (1.567-3.922) | 1.05x10^-4              | 0.942 (0.502-1.769) | 8.53x10^-1              |
| PI-AP (≥-4.89 vs. <4.89) | 0.578 (0.370-0.902) | 1.60x10^-2              | 1.080 (0.600-1.946) | 7.97x10^-1              |
| PI-AT (≥-34.84 vs. <34.84) | 2.604 (1.668-4.065) | 2.54x10^-5              | 1.173 (0.635-2.167) | 6.11x10^-1              |
| PI-ES (≥-648.78 vs. <648.78) | 4.097 (2.439-6.881) | 9.77x10^-8              | 2.599 (1.329-5.084) | 5.00x10^-3              |
| PI-ME (≥-1.56 vs. <1.56) | 3.152 (1.985-5.004) | 1.12x10^-6              | 1.775 (0.999-3.155) | 5.00x10^-2              |
| PI-RI (≥0.87 vs. <0.87) | 1.836 (1.199-2.811) | 5.00x10^-3              | 0.624 (0.337-1.157) | 1.34x10^-1              |
| PI-ALL (≥-491.40 vs. <491.40) | 3.512 (2.131-5.789) | 8.28x10^-7              | 1.273 (0.641-2.528) | 4.91x10^-1              |

HR, hazard ratio; CI, confidence interval; PSA, prostate-specific antigen; PI, prognostic index; PI-ALL, PI based on all seven types of splicing events; PI-AA, PI based on AA type of splicing events; AA, alternate acceptor; AD, alternate donor; AP, alternate promoter; AT, alternate terminator; ES, exon skip; ME, mutually exclusive exons; RI, retained intron.

Figure 13. Analysis of PI-ALL. The five splicing events from which the PI-ALL was generated were identified from DFS-associated splicing events by multivariate Cox regression analysis. (A) Scatter plot of time of DFS vs. cancer status of PRAD patients. (B) PI value curve for patients with PRAD. (C) Time-dependent receiver-operator characteristic curve of PI for predicting the tumor status after five years; (D) Kaplan-Meier curve of PI for evaluating the DFS time of PRAD patients stratified by median PI value into low and high group. PI-ALL, prognostic index based on all seven types of splicing events; PRAD, prostate adenocarcinoma; AUC, area under curve; DFS, disease-free survival; NHLRC3, NHL repeat containing 3; TXN, thioredoxin; FKBP2, FK506 binding protein 2; HMGA2, high-mobility group AT-hook 2; STXB2, syntaxin binding protein 2.
transcripts of genes generated by AS (47-50). In addition, as the major executors of AS, splicing factors have important roles in the production of the splicing variants of genes (51,52). Therefore, a regulatory network of splicing factors and the AS products of regulated genes was constructed in the present study to elucidate pathogenic mechanisms of PRAD. The

Figure 14. Correlation analysis between splicing events of splicing factors and DFS-associated splicing events. (A) Correlation network of splicing events in PRAD. Red nodes represent splicing events of splicing factors, light blue nodes represent splicing events of non-splicing factors, triangle nodes represent splicing events positively associated with DFS, oval nodes represent splicing events negatively associated with DFS, blue lines represent negative correlations and green lines represent positive correlations. (B) Correlation of the PSI value between RBM5-AT-64957 and HP1BP3-AT-939. RBM5-AT-64957 is one of the splicing events of the splicing factor RBM5. (C) Correlation of the PSI value between RBM5-AT-64955 and HP1BP3-AT-939. RBM5-AT-64955 is one of the splicing events of the splicing factor RBM5. (D-F) Kaplan-Meier curves for evaluating the DFS time of PRAD patients stratified by the median PSI value of (D) RBM5-AT-64957, (E) RBM5-AT-64955 and (F) HP1BP3-AT-939 (high/low). PRAD, prostate adenocarcinoma; DFS, disease-free survival; AT, Alternate Terminator type of splicing events; PSI, percent-spliced-in; RBM4, RNA binding motif protein 4; DAZAP1, DAZ-associated protein 1; QKI, QKI, KH domain containing RNA binding; HNRNPDL, heterogeneous nuclear ribonucleoprotein D like; SRSF2, serine and arginine rich splicing factor 2; TSTD1, thiosulfate sulfurtransferase like domain containing 1; HP1BP3, heterochromatin protein 1 binding protein 3; DDRGK1, DDRGK domain containing 1; GLI4, GLI family zinc finger 4; PPHLN1, periphilin 1; YPEL3, yippee like 3; NCAFP2, non-SMC condensin II complex subunit H2; LARP1B, La ribonucleoprotein domain family member 1B; TYK2, tyrosine kinase 2; ABHD17A, abhydrolase domain containing 17A; H2AFJ, H2A histone family member J; TPP3, tumor protein p53 inducible protein 11; RAD51C, RAD51 paralog C; KLK2, kallikrein related peptidase 2; PTGDS, prostaglandin D2 synthase; NHLRC3, NHL repeat containing 3; TCEB2, elongin B; FKBP2, FK506 binding protein 2; ENTPD5, ectonucleoside triphosphate diphosphohydrolase 5.
DFS-SEs identified by univariate Cox analysis were subjected to correlation analyses for the construction of a more accurate correlation network between splicing events of splicing factors and other genes. The splicing events of non-splicing factors associated with a relatively long DFS exhibited a significant negative correlation with the splicing events of the splicing factors, while the opposite trend was identified for splicing events of non-splicing factors associated with poor DFS. At present, no relevant literature is available to confirm the potential regulatory associations for the splicing events discovered in the present study.

The major data of the present study comprehensively assessed the prognostic value of splicing events in PRAD. Of note, the present study had certain limitations. The lack of PRAD-associated mortalities in the TCGA database meant that only ‘with tumor’ was available as the endpoint in the prognostic analysis, rather than ‘death’. Validation in a larger cohort in a future study is also required. The present study also identified numerous splicing events that may theoretically influence the biological behavior of PRAD, so these splicing events require verification in experimental biological studies.

In conclusion, the present study comprehensively assessed AS events in PRAD. PIs generated from DFS-SEs had a high prognostic value. In addition, a more accurate regulatory network for AS in PRAD was provided based on a correlation analysis.

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Availability of data and materials

The PSI value of splicing events and clinical information in PRAD analyzed in this study may be acquired from the TCGASpliceSeq and TCGA.

Author's contributions

ZNM provided the study design. ZGH and RQH performed the analyses and calculations. ZGH, RQH and ZNM all contributed to the writing of the manuscript.

Ethics approval and consent to participate

Not applicable.

Patient consent for publication

Not applicable.

Competing interests

The authors declare that they have no competing interests.

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