Research Paper

Transcriptomic features of primary prostate cancer and their prognostic relevance to castration-resistant prostate cancer

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ABSTRACT

Although various mechanisms of castration-resistant prostate cancer (CRPC) have been discovered, reliable biomarkers for monitoring CRPC progression are lacking. We sought to identify molecules that predict the progression of advanced prostate cancer (AdvPC) into CRPC. The study used primary-site samples (N=45 for next-generation sequencing (NGS); N=243 for real-time polymerase chain reaction) from patients with prostate cancer (PC). Five public databases containing microarray data of AdvPC and CRPC samples were analyzed. The NGS data showed that each progression step in PC associated with distinct gene expression profiles. *Androgen receptor* (AR) associated with tumorigenesis, advanced progression, and progression into CRPC. Analysis of the paired and unpaired AdvPC and CRPC samples in the NGS cohort showed that 15 genes associated with progression into CRPC. This was validated by cohort-1 and public database analyses. Analysis of the third cohort with

AdvPC showed that higher serine peptidase inhibitor, Kazal type 1 (SPINK1) and lower Sp8 transcription factor (SP8) expression associated with progression into CRPC (log-rank test, both P < 0.05). Multivariate regression analysis showed that higher SPINK1 (Hazard Ratio (HR)=4.506, 95% confidence intervals (CI)=1.175–17.29, P=0.028) and lower SP8 (HR=0.199, 95% CI=0.063–0.632, P=0.006) expression independently predicted progression into CRPC. Gene network analysis showed that CRPC progression may be mediated through the AR-SPINK1 pathway by a HNF1Abased gene network. Taken together, our results suggest thatSPINK1 and SP8 may be useful for classifying patients with AdvPC who have a higher risk of progressing to CRPC.

INTRODUCTION

Prostate cancer (PC) is currently the most common malignancy in males [1]. Unfortunately, about 20% of all PC patients will present with or develop disease progression and metastasis [2, 3]. For patients with such advanced PC (AdvPC), the therapeutic options include prostatectomy, radiation therapy, and/or androgendeprivation therapy (ADT) [4]. However, despite these therapies, the prognosis of AdvPC remains poor: for example, the 5-year survival rate of patients with AdvPC who are treated with ADT is 23-26% [3]. This is because ADT can only induce short 2-3 year remissions since most PCs eventually develop into castration-resistant prostate cancer (CRPC) [3]. CRPC associates with a particularly poor prognosis. The mechanism by which PC cells become castration-resistant remains unclear. After PC cells acquire castration resistance, the treatment options are limited to second-line ADT (abiraterone or enzalutamide), new generation cytotoxic chemotherapy, or secondary symptomatic relief of bone metastases [5]. Thus, the management of AdvPC and CRPC remains difficult. There is a great need for robust methods that can identify AdvPC patients who have a higher risk of progressing to CRPC.

Recent advances in high-throughput next-generation sequencing (NGS) technologies have greatly improved our understanding of the genomics and epigenomics of PC [6–9]. In addition, integrative genomic approaches have revealed various novel molecular characteristics of CRPC [7, 9]. However, these studies have only identified a handful of reliable and practical criteria that can adequately predict progression of AdvPC. Moreover, the vast majority of these studies were conducted on metastasized CRPC discovered at distant sites. Since these sites are not the primary tumor site and metastatic sites have completely different microenvironments from the primary site [10], these studies do not reflect the precise molecular or clinical characteristics of progression of AdvPC into CRPC.

Here, we used NGS to explore the transcriptomes of primary-site samples from patients with PC at various stages of progression, namely, benign prostate hyperplasia (BPH), localized PC, AdvPC, and CRPC. The findings were validated by analyses of public databases containing microarray data of AdvPC and CRPC samples, and by subjecting primary-site samples from an independent cohort of patients with various PC stages to reverse transcriptase-polymerase chain reaction (RT-PCR) analysis. A third independent cohort of patients with AdvPC who underwent surgery and then prolonged follow-up was used to assess the ability of selected genes to predict progression of AdvPC into CRPC.

RESULTS

PC progression steps associate with distinct gene-expression profiles

All patients' characteristics were described in Table 1. The transcriptomes of 45 prostate samples from the NGS cohort were obtained by RNA-Seq. Unsupervised hierarchical clustering analysis yielded three main sample clusters that consisted of all BPH (normal) samples, all localized PC (T2 or T3 N0 M0) samples, and all AdvPC (T4 or any N1/M1) plus CRPC samples (Supplementary Figure 1). The AdvPC and CRPC samples were not clearly distinguishable from one another. Thus, as AdvPC progresses into CRPC, only a few genes change their expression significantly. By contrast, larger transcriptome changes are observed when BPH compares to localized PC and when localized PC compares to AdvPC.

The differentially expressed genes in the BPH, localized PC, AdvPC, and CRPC samples were determined by using the GLM likelihood ratio test. This yielded three gene lists that showed the genes whose expression changed when BPH compared to localized PC (list A), localized PC compared to AdvPC (list B), and AdvPC compared to CRPC (list C) (P<0.001, Supplementary Figure 2). Similar to the hierarchical cluster analysis, the transcriptomes of AdvPC and CRPC only differed by 90 genes. Greater differences between transcriptomes were observed in the other comparisons. A Venn diagram was then used to compare the three gene lists (Figure 1A). Several different patterns were observed: A only (BPH vs. localized PC: 2,445 genes), B only (localized PC vs. AdvPC: 403 genes), C only (AdvPC vs. CRPC: 50 genes), $A \cap B$ (95 genes), $B\cap C$ (12 genes), $C\cap A$ (27 genes), and $A\cap B\cap C$ (one gene) (Figure 1B). Thus, the genes in the A and B only categories had expression patterns that associated with tumorigenesis and advanced progression, respectively;

Characteristic	NGS cohort				Validation cohort				Prognostic cohort
	BPH controls	Localized PC	AdvPC	CRPC	BPH controls	Locally advanced or AdvPC	Hormone- suppressed PC	CRPC	Locally advanced or AdvPC
No.	8	16	9*	12*	58	62	14	15	94
Age, years, mean (range)	65.1 (54–70)	68.1 (54–75)	74.9 (69–82)	73.7 (59–82)	71.5 (54–90)	71.4 (48–85)	72.2 (52–78)	69.6 (53–85)	69.9 (52–86)
PSA at operation, ng/ml, mean±SD	1.8±0.4	15.1±7.9	332.8±276.0	62.3±56.0	1.7±1.3	387.9±1239.3	9.8±15.7	193.6±387.3	277.8±959.5
Operation, n (%)									
TUR-P	8	0	1	12	58	44 (71.0)	14 (100)	15 (100)	59 (62.8)
Radical prostatectomy	0	16	8	0		18 (29.0)	0 (0.0)	0 (0.0)	35 (37.2)
Gleason score, n (%)									
6 or less		2	0	0		5 (8.1)	0 (0.0)	0 (0.0)	0 (0.0)
7		11	3	2		20 (32.3)	2 (16.3)	3 (20.0)	42 (44.7)
8		0	0	2		14 (22.6)	3 (21.4)	1 (6.7)	17 (18.1)
9		3	6	7		21 (33.9)	8 (57.1)	8 (53.3)	32 (34.0)
10		0	0	1		2 (3.2)	1 (7.1)	3 (20.0)	3 (3.2)
TNM stage, n (%)									
T2 or T3, N0, M0		16	0	0		17 (27.4)	1 (7.1)	0 (0.0)	47 (50.0)
T4 or metastatic		0	9	12		45 (72.6)	13 (92.9)	15 (100)	47 (50.0)

Table 1: Clinical characteristics of the patients and controls in the NGS cohort, the validation cohort, and the prognostic cohort

*In the NGS cohort, four patients had biopsies of both their AdvPC and the CRPC that arose from it later.

Abbreviations: BPH, benign prostate hyperplasia; NGS, next-generation sequencing; PC, prostate cancer; AdvPC, advanced PC; PSA, prostate-specific antigen; SD, standard deviation; TUR-P, transurethral resection of the prostate.

the genes in the C only category had expression patterns that associated with progression into CRPC; the genes in the A \cap B category were common to tumorigenesis and advanced progression; the genes in the B \cap C category were common to advanced progression and CRPC development; and the genes in the C \cap A category were common to tumorigenesis and CRPC development. The only gene that associated with tumorigenesis, advanced progression, and CRPC development was *AR* (Figure 1): when tumorigenesis transformed BPH into localized PC, *AR* was significantly down-regulated. By contrast, it was up-regulated when localized PC progressed to AdvPC and further upregulated when AdvPC progressed into CRPC (Supplementary Figure 2). Thus, *AR* could be a crucial mediator of PC progression.

Progression of AdvPC into CRPC associates with expression changes in 12 genes

The NGS cohort samples included four pairs of AdvPC and CRPC samples from the same patient. These eight paired samples were assessed for genes that changed their expression when AdvPC progressed into CRPC: 309

genes were significantly differentially expressed. A similar analysis of the 13 unpaired AdvPC and CRPC samples in the NGS cohort showed that 182 genes were significantly differentially expressed (*P*<0.001 by the GLM likelihood ratio test). Comparison of these two gene lists showed 15 genes in common (Supplementary Figure 3). Of these 15, two (*AR* and *SPINK1*) were up-regulated in CRPC compared to in AdvPC and ten (*ALOX15B, ANPEP, CBLN2, CEACAM20, CEACAM22P, SERPINB11, SNCA, SP8, TRPM8*, and *WNT11*) were down-regulated in CRPC compared to in AdvPC (Supplementary Table 3). The remaining genes (*KLK1, PGC,* and *PPFIA2*) showed different expression changes in the paired and unpaired samples and were not considered further.

Confirmation of the 12 CRPC developmentassociated genes

Public gene-expression datasets were used to verify the expression patterns of the 12 genes whose expression changed significantly as AdvPC progressed into CRPC. The available probes that matched with the 12 genes were used only. In the GSE28403 dataset, comparison of the four AdvPC and nine CRPC samples in terms of gene expression showed that apart from a few exceptions, most gene expression changes were consistent with those seen in our cohort. However, the gene expression differences seen in GSE28403 only achieved statistical significance a few times (one was AR; Supplementary Figure 4). This may be due to the small sample size of GSE28403 (N=13) and the fact that all AdvPC and two CRPC samples were from metastatic sites.

A second verification was performed with the GSE32269 dataset, which included 21 primary-site localized PC and 29 metastatic CRPC samples. Again, most expression changes were consistent with those seen in our cohort. Several changes achieved statistical significance (one was AR; Supplementary Figure 5).



Figure 1: Differentially expressed genes in the prostate tissue samples from the next-generation sequencing (NGS) cohort. The NGS cohort consisted of patients with benign prostate hyperplasia (BPH), localized prostate cancer (PC), advanced PC (AdvPC), or castration-resistant PC (CRPC). (A) Venn diagram of the genes that were differentially expressed in the BPH, localized PC, AdvPC, and CRPC patients. The genes were selected by using the GLM likelihood ratio test with EdgeR software. Genes whose differentially expressed between BPH and localized PC. Genes in the orange circle (list B) indicate those that are differentially expressed between localized PC and AdvPC. Genes in the black circle (list C) indicate those that are differentially expressed between AdvPC and CRPC. (B) Expression patterns of selected genes in the Venn diagram. The data are presented in matrix format where rows indicate individual genes and columns show the indicated tissue. Red and green colors indicate high and low expression, respectively.

In a third dataset, GSE35988, which included 59 primary-site localized PC and 35 metastatic CRPC samples, all but one gene exhibited expression changes that were consistent with those seen in our cohort. There were also more statistically significant changes in this cohort than in the other cohorts (one was AR; Supplementary Figure 6).

To further confirm that the expression of these genes changed as AdvPC progressed into CRPC, we subjected primary-site PC samples from a second independent cohort to RT-PCR analysis. These samples consisted of BPH, locally advanced or AdvPC, hormone-suppressed PC, and CRPC samples. We only assessed 7 of the 12 genes because previous studies have already shown that ALOX15B, ANPEP, SNCA, TRPM8, and WNT11 associate strongly with AdvPC [11-16]. Of these seven genes, two (AR and SPINK1) were up-regulated by progression into CRPC and five (SERPINB11, SP8, CBLN2, CEACAM20, and CEACAM22P) were down-regulated. BPH only differed significantly from the PC samples in terms of SERPINB11, CEACAM20, and CEACAM22P expression (Figure 2). By contrast, the three PC stages differed significantly from each other in terms of AR, SPINK1, SP8, CEACAM20, and CEACAM22P (but not SERPINB11 and CBLN2) expression (Figure 2). Thus, these data corresponded largely with the RNA-Seq data from the NGS cohort.

Two genes can predict the risk of progression into CRPC

To determine the prognostic usefulness of the five genes that associated significantly with progression to CRPC in our validation cohort analysis, we employed our third independent cohort. All patients in this cohort underwent surgery for AdvPC and were placed on ADT and followed on average for 32.7 months. Of the 94 patients, 24 developed CRPC during follow-up. We analyzed the 94 primary-site AdvPC samples that were taken at surgery for the expression of the five candidate genes and then divided the patients into two groups on the basis on the expression of each gene by median values. The progression-free survival duration of the two groups (*i.e.*, duration between starting ADT and CRPC) was then analyzed (Figure 3). Only SPINK1 and SP8 were significantly predictive of time to CRPC. Patients with higher SPINK1 expression progressed to CRPC significantly more frequently than patients with lower SPINK1 expression (log-rank test, P=0.005). Similarly, patients with lower SP8 expression progressed to CRPC significantly more frequently than patients with higher SP8 expression (log-rank test, P=0.002).

To verify the prognostic usefulness of *SPINK1* and *SP8*, we subjected the third cohort data to multivariate Cox regression analysis to determine the independent



Figure 2: The seven genes that were differentially expressed between primary-site advanced prostate cancer (PC) and castration-resistant PC (CRPC) samples in the NGS cohort were validated by analyzing the validation cohort by using real-time PCR. The cohort consisted of patients with benign prostate hyperplasia (BPH), locally advanced or advanced PC (AdvPC), hormone-suppressed PC, or CRPC. The patterns of *AR*, *SPINK1*, *SP8*, *CEACAM20*, and *CEACAM22P* expression were consistent with those observed in the NGS cohort.

association between *SPINK1* and *SP8* with known clinicopathological risk factors of PC. Even after applying a variable selection procedure, type of operation, lymph node or distant metastasis, higher *SPINK1* expression (HR=4.506, 95% CI=1.175–17.29, P=0.028), and lower *SP8* expression (HR=0.199, 95% CI=0.063–0.632, P=0.006) were independent risk factors for CRPC progression (Table 2).

Biological insights into the gene profile of progression into CRPC

To identify signaling pathways that promote progression to CRPC, a gene-to-gene network analysis of the 90 genes that were differentially expressed between AdvPC and CRPC sample groups (Figure 1) was performed by using the IPA tool. Analysis of literaturebased gene networks revealed functional connectivity between *AR* and *SPINK1*, which indirectly regulated by *SP8*, and a putative gene network hub mediated by *HNF1A* (Figure 4). Analysis of publicly available datasets from five independent cohorts of patients with AdvPC or CRPC (including the three described above) confirmed that expression changes of *AR*, *SPINK1*, *HNF1A*, and several other network members were well agreed with those in the NGS cohort. (Supplementary Figures 8–12). Details are available in the Supplementary Materials.

DISCUSSION

PC is a heterogeneous disease with diverse progression steps that involves the activities of various molecules. Identification of the genes that associate specifically with the progression of AdvPC into CRPC will help identify the patients at risk of such progression and will aid the development of therapeutic strategies that block this progression step. Such therapies could greatly improve the overall quality of life of PC patients and prolong their survival. With this in mind, we subjected 45 primary-site samples from patients with BPH, localized PC, AdvPC, or CRPC to transcriptome analysis to identify the genes that associate with each PC progression step. Transcriptome profiling showed that 90 genes significantly associated with progression from AdvPC to CPRC. In



Figure 3: Prognostic usefulness of five genes that associated with progression from locally advanced or advanced prostate cancer (PC) to castration-resistant PC (CRPC). The prognostic cohort consisted of 94 patients with locally advanced or advanced PC (AdvPC) who underwent surgery followed by androgen-deprivation therapy (ADT) and were followed up on average for 32.7 months. The patients were divided into two groups depending on whether they had higher or lower *AR*, *SPINK1*, *SP8*, *CEACAM20*, or *CEACAM22P* expression. The time to CRPC in the two subgroups for each gene was analyzed. Higher *SPINK1* and lower *SP8* expression associated with progression to CRPC.

Table 2: Multivariate Cox regression modeling to identify risk factors for the progression of AdvPC into CRPC

Variable	Univariate analy	sis	Multivariate analysis		
	HR (95% CI)	Р	HR (95% CI)	Р	
Type of operation (RP vs. TUR–P)	8.02 (3.277–19.631)	< 0.001	3.572 (1.074–11.882)	0.038	
PSA (<20 vs. ≥20 ng/mL)	4.718 (1.407–15.82)	0.012			
Lymph node or distant metastasis (no vs. yes)	8.322 (2.831–24.462)	< 0.001	7.076 (1.435–34.886)	0.016	
Gleason score (7 vs. 8–10)	3.892 (1.451–10.439)	0.007			
SPINK1 expression (lower vs. higher)*	3.84 (1.404–10.503)	0.009	4.506 (1.175–17.29)	0.028	
SP8 expression (lower vs. higher)*	0.256 (0.101-0.646)	0.004	0.199 (0.063-0.632)	0.006	

Abbreviations: AdvPC, advanced prostate cancer; CRPC, castration-resistant prostate cancer; CI, confidence interval; HR, hazard ratio; PSA, prostate-specific antigen; RP, radical prostatectomy; TUR–P, transurethral resection of the prostate. *Expressions of gene were divided by medial value.

particular, one gene, AR, was down-regulated when BPH compared to localized PC, up-regulated when localized PC compared to AdvPC, and further up-regulated when AdvPC compared to CRPC. Thus, AR may participate in all PC progression steps. Bioinformatic- and RT-PCRbased experiments with four independent patient cohorts confirmed the significant role of five additional genes in progression to CRPC. An analysis of an independent cohort of patients with AdvPC who underwent surgery, ADT, and extensive follow-up then showed that the expression patterns of two of these five genes, SPINK1 and SP8, in the primary site predicted progression into CRPC: specifically, higher SPINK1 and lower SP8 expression associated with shorter times to CRPC development. Multivariate regression analysis confirmed that these genes are independent predictors of the risk of progression to CRPC, even after considering clinicopathological indicators. Gene network analysis also revealed signaling pathways that may be responsible for CRPC development.

SPINK1 encodes a trypsin inhibitor that is secreted by pancreatic acinar cells. Mutations in SPINK1 associate with hereditary pancreatitis and tropical calcific pancreatitis [17]. SPINK1 also participates in several important cellular activities, including abnormal morphology and proliferation. Moreover, it is up-regulated in many cancers, including PC [18-23]. There are several other biological links between SPINK1 and PC. First, Tomlins et al. showed that SPINK1 is highly up-regulated in a molecular subtype of ETS-rearrangement-negative PC and that SPINK1 was an independent predictor of biochemical recurrence of these PCs [23]. However, Flavin et al. showed that SPINK1encoded protein may not predict biochemical recurrence or lethal PC after radical prostatectomy [22]. This disparity may reflect the different composition of the patient cohorts in these two studies [23]. Second, SPINK1 is directly regulated by AR in prostate tissue and has been reported to be a potential therapeutic target in CRPC [3, 24]. However, we are the first to show that SPINK1 is of prognostic relevance in CRPC.

We also showed for the first time that *SP8* may be an independent indicator of CRPC progression as well. *SP8* is a SP family transcription factor that plays an essential role in proper limb development. There is also an association between *SP8* and the development of other urological diseases such as hypospadias [25]. However, there are very few reports on its role in the cell or its association with cancer.

We used public datasets to validate our finding that progression of PC into CRPC associated with expression changes in several genes. While the public datasets did generally confirm our findings, there were some discrepancies. First, while SPINK1 did tend to be up-regulated in CRPC in the public datasets relative to advanced or localized PC, this change never achieved statistical significance (Supplementary Figures 4-6 and 8-12). By contrast, it was strongly up-regulated in the CRPCs in validation cohort-1 (Figure 2). Second, although SP8 was down-regulated in CRPC in the NGS cohort relative to AdvPC, it was up-regulated in the CRPCs in the GSE28403 dataset relative to AdvPC (Supplementary Figure 4). This may be due to the small sample size of that public dataset and the fact that all AdvPCs and two of the nine CRPCs were from metastatic sites. Indeed, our analysis of the validation cohort confirmed that SP8 is down-regulated in CRPCs (Figure 2). These results underscore the importance of having a sufficient sample size and maintaining a consistent sampling method when obtaining tissue samples for genomic comparisons.

Our gene-to-gene network analysis of the 90 genes that were differentially expressed between AdvPC and CRPC revealed a putative *HNF1A*-mediated gene network and functional connectivity between *AR* and *SPINK1* (Figure 4). The genes that are regulated by *HNF1A* include UDP glucuronosyltransferase (UGT) family members, namely, *UGT1A1*, *UGT1A3*, and *UGT2B15*. Like *HNF1A*, these proteins participate in cancer, cell proliferation, and tumor morphology. Metabolomic profiling of PC has also shown that CRPCs differ significantly from androgen-dependent PC in terms of UGT activity-associated metabolites [26]. Thus, UGT activity may be important for modulating the activity of androgens [27]. The pathway that is regulated by UGT activity may also be a potential source of targets for drug discovery in CRPC [3]. Notably, when we explored the role of hormone interactions in our gene-to-gene networks, we found that many hormones mediated the signaling of *AR* and the downstream effectors of the *HNF1A* network, including the UGT family members (Supplementary Figure 13). This suggests that the *AR*- and *HNF1A*-mediated gene networks participate in the development of CRPC. However, the precise role that *HNF1A* and these hormone interactions play in CRPC development remains to be determined experimentally, because these results were derived from only computational analysis.

There were several limitations in our study. First, our patient cohorts only contained 27 CRPC samples. This is insufficient for rigorously determining the role of genes in progression to CRPC. Second, we analyzed the transcriptomics datasets without considering other biological events such as mutations, copy number alteration, or epigenetic changes. This may have hampered our mechanistic insights into CRPC development. Third, our analyses only provided indirect evidence for the role of *HNF1A* and its network member genes in CRPC progression; these activities must be validated by biological

assays. Forth, we used small part of whole cancer tissues, and some of harvested tissues could not explain the all cancer clones because of tumor heterogeneity of PC. Lastly, BPH tissues could not represent the normal prostate tissue, because the most of cancer and BPH are arose from the different zone of the prostate.

Despite these limitations, however, this investigation has several key advantages over similar studies. First, AdvPC and CRPC sample pairs were obtained from four patients and all methods were handled by a single institute, including extraction of patient clinical information, RNA extraction, and data processing; this is likely to have limited heterogeneity between patient samples. Second, the gene-expression profiles that associated with three PC progression steps were determined. Third, to ensure that we obtained a gene profile that associated specifically with the development of CRPC, we only tested primary CRPCs that were obtained from homogenous biopsy sites; none of our PC samples were from distant tissue sites.

In conclusion, during our analysis of the transcriptomic changes that occur during various multiple PC progression steps, two genes, namely, *SPINK1* and *SP8*, were found to be reliable prognostic indicators of progression to CRPC; these markers are independent of the classical pathological prognostic variables.





MATERIALS AND METHODS

Patients and tissue samples

Our cohort study was based on 288 primary-site samples from three independent cohorts of PC patients. All samples were obtained from patients treated at the Chungbuk National University Hospital. All tumor samples were acquired after patient consent for tissue sample donation and examination was received. The patients who received radiation therapy were excluded for the study. The study protocol was approved by the Institutional Review Board for Human Genetic and Genomic Research (IRB approval number: 2006-01-001 and GR2010-12-010). PC was defined as CRPC if the prostate-specific antigen (PSA) levels had risen three times in a row despite the patient having low serum testosterone levels (<50 ng/mL).

The first cohort consisted of 45 prostate tissue samples (eight BPHs, 16 localized PCs, nine AdvPCs, and 12 CRPCs) from 41 patients (Table 1). Four patients provided both AdvPC and CRPC samples at different time points (Supplementary Table 1). All samples were subjected to NGS to explore their transcriptomes. It should be emphasized that all samples in this cohort were collected from the primary cancer site in the prostate: none were metastasized samples that were discovered at distant tissue sites.

The second cohort was a validation cohort. It consisted of 58 non-cancer BPHs, 62 locally advanced or AdvPCs, 14 hormone-suppressed PCs, and 15 CRPCs (Table 1). Hormone-suppressed PC was defined as PC when the patient was receiving ADT and there was no indication of progression into CRPC.

The third cohort was used to determine the prognostic value of selected genes. It consisted of 94 patients with locally advanced or advanced disease who received ADT after surgery (Table 1). Of these patients, 24 developed CRPC within 32.7 months of follow up (6.1–140 months).

RNA-Seq experiments and data processing

RNA extraction and preparation for RNA-Seq was performed as described (Supplementary Materials).

Public gene-expression datasets

To validate the NGS data, five gene-expression datasets (GSE28403, GSE32269, GSE35988, GSE37199, and GSE70768) that were based on microarray analysis of samples of PC at various progression stages were obtained from the National Center for Biotechnology (NCBI) Gene Expression Omnibus (GEO) [9, 28–31].

These cohorts are described in the Supplementary Materials.

Real-time PCR

mRNA levels were measured by RT-PCR as described (Supplementary Materials).

Statistical analysis

Statistical analyses were performed by using R (ver. 3.2.5.) and SPSS (ver. 23). To determine the mRNAexpression profiles of the prostate samples, a hierarchical clustering algorithm with centered correlation coefficient for similarity measure and complete linkage clustering was used. For clustering, the fragments per kilobase of transcript per million fragments mapped (FPKM) of each sample were used to estimate each gene expression. The FPKM data was normalized by the quantile method, log2-transformed, and median-centered across genes and samples.

To compare sample subgroups in the NGS cohort in terms of gene expression, an EdgeR package that utilizes a negative binomial model was used to detect differentially expressed genes from count data [32]. The gene count dispersion was estimated by using a Cox-Reid profile-adjusted likelihood method. After model fitting and estimation of dispersion, differentially expressed genes were selected by using a generalized linear model (GLM) likelihood ratio test. To verify gene expression differences between sample subgroups in the validation cohorts, twosample *t*-tests were performed for each gene. Expression differences in genes were considered statistically significant if P was <0.001 and the fold difference in expression between two sample groups was ≥ 2 . The Kaplan-Meier method was used to calculate the time to CRPC progression and the difference in survival between two groups was assessed by using log-rank statistics. The association between potential risk factors and prognosis was assessed by using multivariate Cox regression models. A backward-forward step procedure was applied to generate the multivariate model with the most informative variables [33]. The results were expressed as Hazard Ratio (HR) and 95% confidence intervals (CI).

Gene-set enrichment analysis was performed to identify the most significant gene sets that associated with PC progression. The significance of over-represented gene sets was estimated by Fisher's exact test. To explore the relationships between the genes that associated with progression into CRPC, we searched for a connection between the genes by examining the previously reported literature and generated gene networks based on whether they had more connected genes than would be expected to occur by chance. Gene set enrichment and gene network analyses were performed by using the Ingenuity Pathway Analysis[™] (IPA) tool.

Author contributions

S.J. Yun and S.K. Kim designed studies, and tissue harvesting, *in vitro* experiments, and NGS analysis were performed by S.J. Yun, S.K. Kim, J. P. Jeong, H.W. Kang, Kim, S.Y. Kim, and W.J. Kim. Data analysis and interpretation were performed by S.J. Yun, S.K. Kim, Y.S. Ha, Y.H. Kim, H.W. Kang, S.H. Kim, J.L. Park, and J.H. Kim. The manuscript writing and revision were performed by S.J. Yun, S.K. Kim, J. Kim, E.J. Cha, P. Jeong, H.W. Kang, S.K. Moon, Y.H. Choi, S.Y. Kim, and W.J. Kim. Administrative, technical, and material support was assisted by E.J. Cha, J.S. Kim, S.J. Kim, and Y.K. Choi. All authors reviewed the manuscript prior to submission and provided comments.

CONFLICTS OF INTEREST

No potential conflicts of interest were disclosed.

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