Analysis of over 140,000 European descendants identifies genetically-predicted blood protein biomarkers associated with prostate cancer risk
Running title: Genetically predicted protein biomarkers for prostate cancer
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### 38 Abstract

39 Several blood protein biomarkers have been associated with prostate cancer (PrCa) risk. 40 However, most studies assessed only a small number of biomarkers and/or included a small 41 sample size. To identify novel protein biomarkers of PrCa risk, we studied 79,194 cases and 42 61,112 controls of European ancestry, included in the PRACTICAL/ELLIPSE consortia, using 43 genetic instruments of protein quantitative trait loci (pQTLs) for 1,478 plasma proteins. 31 44 proteins were associated with PrCa risk including proteins encoded by GSTP1, whose 45 methylation level was shown previously to be associated with PrCa risk, and MSMB, SPINT2, 46 *IGF2R*, and *CTSS*, which were previously implicated as potential target genes of PrCa risk 47 variants identified in genome-wide association studies. 18 proteins inversely correlated and 13 48 positively correlated with PrCa risk. For 28 of the identified proteins, gene somatic changes of 49 short indels, splice site, nonsense, or missense mutations were detected in PrCa patients in The 50 Cancer Genome Atlas. Pathway enrichment analysis showed that relevant genes were 51 significantly enriched in cancer related pathways. In conclusion, this study identifies 31 52 candidates of protein biomarkers for PrCa risk and provides new insights into the biology and 53 genetics of prostate tumorigenesis. 54

#### **Statement of Significance** 55

56 Integration of genomics and proteomics data identifies biomarkers associated with prostate

57 cancer risk

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# 62 Introduction

63	Prostate cancer (PrCa) is the second most frequently diagnosed malignancy and the fifth leading
64	cause of cancer mortality among males worldwide(1). In the United States, there were 164,690
65	estimated new PrCa cases and 29,430 estimated deaths due to PrCa in 2018, making it a
66	malignancy with the highest incidence and second highest mortality in males(2). The survival
67	rate is higher when cancer is diagnosed at a localized stage while it drops substantially when
68	PrCa is diagnosed at a metastatic stage(3). Biomarkers are needed for screening and the early
69	detection of PrCa. Prostate-specific antigen (PSA) has been used widely for PrCa screening(4,5);
70	however, there are controversies in using PSA screening due to the lack of a clear cutoff point
71	for high sensitivity and specificity(6-8), unclear benefit in reducing mortality in some
72	populations (9-11), and overdiagnosis of PrCa(12). Thus, there is a critical need to identify
73	additional screening biomarkers aiming to reduce the mortality of PrCa.
74	
75	Several other protein biomarkers measured in blood have been reported to be potentially
76	associated with PrCa risk, such as IGF-1, IGFBP1/2, and IL-6(13-16). However, findings have
77	been inconsistent from previous studies. Most existing studies have assessed only a small

78 number of candidates. With the recent development of proteomics technology, there have been

respectively several studies searching the whole proteome to identify novel biomarkers for PrCa early

80 detection and diagnosis(17-20). These studies have generated some promising findings.

81 However, these have only included a relatively small number of subjects as it is expensive to

82 profile the proteome in a large population-based study. More importantly, there are multiple

83 limitations that are commonly encountered in conventional epidemiologic studies, including

84 selection bias, potential confounding, and reverse causation. These limitations may explain some

85 of the inconsistent results from previous studies.

86

87	To reduce these biases, we used genetic variants associated with blood protein levels as the
88	instruments to assess the associations between genetically predicted protein levels and PrCa risk.
89	Because of the random assortment of alleles transferred from parents to offspring during gamete
90	formation, this approach should be less susceptible to selection bias, reverse causation, and
91	confounding effects. Over the past few years, genome-wide association studies (GWAS) have
92	identified hundreds of protein quantitative loci (pQTL)(21,22). With a large sample size, many
93	of these genetic variants can serve as strong instrumental variables for evaluating the
94	associations of genetically predicted protein levels with PrCa risk. Herein, we report results from
95	the first large study investigating the associations between genetically predicted blood protein
96	levels and PrCa risk using genetic instruments. We used the data from 79,194 cases and 61,112
97	controls of European descent included in GWAS consortia PRACTICAL, CRUK, CAPS, BPC3
98	and PEGASUS, as described previously(23).

99

# 100 Methods

101 A literature search was performed to identify the GWAS that uncovered genetic variants that 102 were significantly associated with protein levels. After careful evaluation, the study conducted 103 by Sun et al represents the largest and most comprehensive study to date(24). By using the data 104 from two sub-cohorts of 2,731 and 831 healthy European-ancestry participants from the 105 INTERVAL study, Sun et al identified 1,927 genetic associations with 1,478 proteins at a 106 stringent significance level(24). The detailed information of this study has been described 107 elsewhere(24). In brief, an aptamer-based multiplex protein assay (SOMAscan) was used to 108 quantify 3,620 plasma proteins. The robustness of the protein measurements was verified using

109	several methods(24). Genotypes were measured using the Affymetrix Axiom UK Biobank array,
110	which were further imputed using a combined reference panel from 1000 Genomes and UK10K.
111	pQTL analyses were performed within each subcohort, with adjustments for age, sex, duration
112	between blood draw and processing, and the first three principal components. After combining
113	the association results from the two subcohorts via fixed-effects inverse-variance meta-analysis
114	using METAL, the genetic associations between 1,927 variants and 1,478 proteins showed a
115	meta-analysis of $P < 1.5 \times 10^{-11}$ , and a consistent direction of effect and nominal significance
116	(P < 0.05). These pQTLs were used to construct the instrumental variables for assessing
117	associations between protein levels and the risk of developing prostate cancer. When two or
118	more variants located at the same chromosome were identified to be associated with a particular
119	protein, we assessed the correlations of the SNPs using the Pairwise LD function of SNiPA
120	( <u>http://snipa.helmholtz-muenchen.de/snipa/index.php?task=pairwise_ld</u> ). For each protein, only
121	SNPs independent of each other, as defined by $r^2 < 0.1$ (based on 1000 Genomes Project Phase 3
122	version 5 data focusing on European populations), were used to construct the instruments.
123	
124	We used the summary statistics data for the association of genetic variants with PrCa risk that
125	were generated from 79,194 PrCa cases and 61,112 controls of European ancestry in the
126	consortia PRACTICAL, CRUK, CAPS, BPC3 and PEGASUS(23,25). In brief, 46,939 PrCa
127	cases and 27,910 controls were genotyped using OncoArray, which included 570,000 SNPs
128	(http://epi.grants.cancer.gov/oncoarray/). Also included were data from several previous PrCa
129	GWAS of European ancestry: UK stage 1 and stage 2; CaPS 1 and CaPS 2; BPC3; NCI
130	PEGASUS; and iCOGS. These genotype data were imputed using the June 2014 release of the

- 131 1000 Genomes Project data as a reference. Logistic regression summary statistics were then
  132 meta-analyzed using an inverse variance fixed effect approach.
- 133

134 For estimating the association between genetically predicted circulating protein levels and PrCa 135 risk, the inverse variance weighted (IVW) method, using summary statistics results, was 136 used(26). The beta coefficient of the association between genetically predicted protein levels and PrCa risk was estimated using  $\sum_i \beta_{i,GX} * \beta_{i,GY} * \sigma_{i,GY}^{-2} / (\sum_i \beta_{i,GX}^2 * \sigma_{i,GY}^{-2})$ , and the corresponding 137 standard error was estimated using  $1/(\sum_{i} \beta_{i,GX}^2 * \sigma_{i,GY}^{-2})^{0.5}$ . Here,  $\beta_{i,GX}$  represents the beta 138 139 coefficient of the association between *i* th SNP and the protein of interest generated from the 140 pQTL study by Sun et al;  $\beta_{i,GY}$  and  $\sigma_{i,GY}$  represent the beta coefficient and standard error, 141 respectively, for the association between *i* th SNP and PrCa risk in the PrCa GWAS. The 142 association odds ratio (OR), confidence interval (CI), and P value were then estimated based on 143 the calculated beta coefficient and standard error. A Benjamini-Hochberg false discovery rate 144 (FDR) of < 0.05 was used to adjust for multiple comparisons. Furthermore, to evaluate whether 145 the identified associations between genetically predicted circulating protein levels and PrCa risk 146 were independent of association signals identified in GWAS, we performed conditional analyses, 147 adjusting for the closest risk SNPs identified in previous GWAS or fine-mapping studies. For 148 this analysis, we performed GCTA-COJO analyses(27-30) (version 1.26.0) to calculate 149 associations of SNPs with PrCa risk, after adjusting for the risk SNP of interest. We then re-ran 150 the IVW analyses using the association estimates generated from conditional analyses. 151 152 For each of the genes encoding the proteins that are identified in our study in association with

153 PrCa risk, we evaluated genetic variants/mutations/indels in prostate tumor tissues from PrCa

154	patients included in TCGA. The somatic level genetic changes were analyzed using MuTect(31)
155	and deposited to the TCGA data portal. Data were retrieved in April, 2016, through the data
156	portal. The proportion of assessed genes containing such somatic level genetic events tended to
157	be enriched, when compared with the proportion of all protein-coding genes across the genome.
158	Analysis was performed using MedCalc online software.
159	
160	To further assess whether our identified PrCa associated proteins are enriched in specific
161	pathways, molecular and cellular functions, and networks, we performed an enrichment analysis
162	of the genes encoding identified proteins using Ingenuity Pathway Analysis (IPA) software(32).
163	The detailed methodology of this tool has been described elsewhere(32). In brief, an
164	'enrichment' score [Fisher's exact test (FET) P-value] that measures overlap of observed and
165	predicted regulated gene sets was generated for each of the tested gene sets. The most significant
166	pathways and functions with an enrichment <i>P</i> -value less than 0.05 were reported.
167	
168	Results
169	Of the pQTLs for 1,478 proteins assessed in this study, association results for PrCA risk were
170	available for pQTLs of 1,469 proteins in the PrCa GWAS. For 1,106 of these proteins, only a
171	single pQTL was identified. Two pQTLs were identified for 302 proteins and three or more
172	pQTLs were identified for 71 proteins. Using the inverse variance weighted (IVW) method, we
173	identified 31 proteins for which their genetically predicted levels were associated with PrCa risk
174	at a false discovery rate of < 0.05 (Tables 1 and 2), including 22 encoded by genes located more
175	than 500 Kb away from any reported PrCa risk variants identified in GWAS or fine-mapping
176	studies (Table 1). The other nine associated proteins are encoded by genes locate at previously

177	reported PrCa risk loci (Table 2), including MSMB, SPINT2, IGF2R, and CTSS, which were
178	previously implicated as candidate target genes of PrCa risk variants identified in GWAS(33-35).
179	Interestingly, we also observed a significant association for glutathione S-transferase Pi, encoded
180	by GSTP1 (Table 2), whose methylation has been identified as a potential biomarker for PrCa
181	(36). In our study, an inverse association between protein level and PrCa risk was detected for
182	PSP-94, DcR3, IGF-II receptor, KDEL2, Cathepsin S, ZHX3, ZN175, GPC6, RM33, PIM1,
183	WISP-3, NCF-2, ATF6A, Laminin, Glutathione S-transferase Pi, GNMT, LRRN1, and SNAB
184	(ORs ranging from 0.69 to 0.97). Conversely, an association between a higher protein level and
185	increased PrCa risk was identified for TACT, GRIA4, PDE4D, TIP39, SPINT2, MICB, IL-21,
186	ARFP2, RF1ML, TPST1, KLRF1, TM149, and NKp46 (ORs ranging from 1.11 to 1.23).
187	
188	To determine whether the identified significant associations between genetically predicted
189	protein levels and PrCa risk were independent of GWAS-identified association signals, we
190	performed conditional analyses adjusting for the GWAS-identified risk SNPs closest to the genes
191	encoding our identified proteins (Tables 1 and 2)(27). For proteins listed in Table 1, the
192	analysis could not be performed for three proteins due to lack of data, and for all other proteins,
193	the associations remained essentially unchanged in the conditional analysis, suggesting these
194	associations may be independent of GWAS-identified association signals. On the other hand, for
195	proteins whose encoding genes locate at known PrCa risk loci, except for IGF2R, all other
196	associations were no longer statistically significant when conditioning on GWAS-identified risk
197	SNPs, suggesting these associations may be influenced by GWAS-identified association signals
198	( <b>Table 2</b> ).

199

200	By analyzing exome-sequencing data of prostate tumor-adjacent normal tissue and tumor tissue
201	obtained from 498 PrCa patients of The Cancer Genome Atlas (TCGA), we observed somatic
202	level changes of indels, nonsense mutations, splice site variations, or missense mutations in at
203	least one patient for 28 of the 31 genes encoding identified associated proteins (enrichment
204	p < 0.0001 compared with the proportion of all protein-coding genes across the genome)
205	(Supplementary Table 1). In addition to the somatic missense mutations detected in 24 genes,
206	indels were detected in four genes (ARFIP2, LRRN1, ZNF175, and PDE4DIP), splice site
207	variations were detected in four genes (IGF2R, IL21, MICB, and PTH2R), and a nonsense
208	mutation was detected in KLRF1 (Supplementary Table 1). Although the majority of these
209	somatic changes occurred in only one patient, a missense mutation in PTH2 occurred in nine
210	patients (1.8%) (Supplementary Table 1).
211	
212	Based on the IPA analysis, several cancer-related functions were enriched for the genes encoding
213	the associated proteins identified in this study (Supplementary Table 2). The top canonical

214 pathways identified included STAT3 Pathway ( $p=4.54 \times 10^{-3}$ ), Glutathione Redox Reactions I

215 (p=0.027), Glutathione-mediated Detoxification (p=0.030), Endoplasmic Reticulum Stress

216 Pathway (*p*=0.031), and tRNA Splicing (*p*=0.044).

217

# 218 **Discussion**

This is the first large-scale study to evaluate the associations of genetically predicted protein levels with PrCa risk using GWAS-identified pQTLs as instruments. We identified 31 proteins that demonstrated a statistically significant association with PrCa risk after FDR correction, including 22 whose encoding genes were located more than 500 Kb away from any reported

PrCa risk variants. Our study provides novel information to improve the understanding of
genetics and etiology for PrCa, and generates a list of promising proteins as potential biomarkers
for early detection of PrCa, the most common malignancy among men in most countries around
the world.

228	In the current work, we used data from large genome-wide association studies (GWAS)
229	involving 79,194 PrCa cases and 61,112 controls. The purpose and approach of the current
230	analysis are different from those of the study of Schumacher et al(23). In the GWAS study,
231	investigators evaluated each genetic variant across the genome one at a time, aiming to identify
232	novel susceptibility variants showing an association with PrCa risk(23). The current work aimed
233	to use genetically predicted protein expression levels as the testing unit to identify PrCa
234	associated proteins. We used a protein-based approach that aggregates the effects of several
235	SNPs into one testing unit whenever possible. The analysis unit for our study is proteins, while
236	the analysis unit in GWAS by Schumacher et al(23) is genetic variants.
237	
238	Previous research suggests that PSA, IGF-1, IGFBP1/2, and IL-6 measured in blood may be
239	associated with PrCa risk. For PSA, IGFBP1/2, and IL-6, there was no corresponding pQTL
240	identified in the study conducted by Sun et al(24), thus they were not investigated in the current
241	study. For IGF-1, by using its pQTL rs74480769 as instrument, we did not observe a significant
242	association with PrCa risk (OR=0.98, 95% CI: 0.90-1.07; P=0.70). The inconsistent finding of
243	IGF-1 with previous studies could be due to either a weak instrument used in the current study or
244	potential confounded estimates of associations in previous studies using a conventional
245	epidemiological design. Indeed, the significant positive association of IGF-1 was observed in the

Health Professionals Follow-up Study(15), but not in the Prostate Cancer Prevention Trial(14).
Further research would be needed to better understand the relationship between these proteins
and PrCa.

249

250 In this large study, we identified 22 associated proteins of which the encoding genes are located 251 at genomic loci not mapped by any of the previous GWAS. The statistical power in our study is 252 larger than GWAS because 1) the number of comparisons is smaller in our study than GWAS and thus we could use a less stringent statistical significance threshold rather than  $5 \times 10^{-8}$  in 253 254 GWAS and 2) the predicted protein levels are continuous variables, which improves statistical 255 power. It is worth noting that nine of the proteins identified in this study are encoded by genes 256 locating at the GWAS-identified loci. For many of the identified proteins, the genetic instrument 257 includes trans pQTL(s) beyond only cis pQTL(s) (**Tables 1-2**), thus explaining why the 258 corresponding protein-coding genes are not always at known susceptibility loci. In vitro/in vivo 259 studies and human studies have suggested that some of these novel genes may play an important 260 role in prostate tumorigenesis. For example, an inter-chromosomal interaction between a known 261 PrCa risk locus, 8q24, and CD96 was observed by the use of a chromosome conformation 262 capture-based multi-target sequencing technology(37). GPC6 was found to be recurrently altered 263 across tumors of advanced and lethal PrCa patients(38). PDE4D was shown to function as a 264 proliferation-promoting factor in PrCa and was overexpressed in human prostate carcinoma(39); 265 its inhibition had been shown to decrease PrCa cell growth (40). ATF6, which is related to the 266 unfolded protein response, was observed to be down-regulated in high-grade prostatic 267 intraepithelial neoplasia compared with normal prostate samples(41).

268

269	Of the nine associated proteins of which the encoding genes are located at GWAS-identified
270	PrCa risk loci, several have also been found to potentially play functional roles in PrCa
271	development. For example, the decreased GSTP1 expression was observed to accompany human
272	prostatic carcinogenesis(42). It is highly expressed in benign prostate glands while tends to not
273	express in prostate cancer glands(43). MSMB encodes MSP for prostatic secretory protein of 94
274	amino acids, which is secreted by the prostate and functions as a suppressor of tumor growth and
275	metastasis(44). Besides the study of Sun et al (24), several other studies also support the
276	potential of MSP as a serum marker for the early detection of high-grade PrCa(45,46). The
277	decreased expression of $IGF2R$ was thought to be partly responsible for the increased growth of
278	LNCaP human prostate cancer cells(47). In a mouse model, the mRNA of <i>IGF2R</i> was
279	significantly decreased in metastatic prostate lesions and androgen-independent PrCa(48). By
280	analyzing patient samples, it was identified that the loss of the heterozygosity of IGF2R was an
281	early event in the development of PrCa(49). In in vivo and human studies, it was suggested that
282	the shedding of MICB might contribute to the impairment of NK cell antitumor immunity in
283	PrCa formation(50,51). These previous studies provide support for a potential role of these genes
284	in prostate carcinogenesis.

285

The sample size for the main association analysis of our study was large, providing high statistical power to detect the protein-PrCa associations. Also, the design of using genetic instruments reduces biases, such as selection bias and potential confounding, and eliminates potential influence due to reverse causation. On the other hand, there are several potential limitations of our study. The possibility of pleiotropy effect cannot be excluded. For example, rs28929474, which was the instrument for proteins ZN175, ARFP2, GPC6, RM33, PIM1, and

292	WISP-3, as well as one of the two variants constituting an instrument for NCF-2, was also
293	reported to be associated with several other traits, including glycoprotein acetyls(52-54).
294	Similarly, rs429358, which was included in the instruments of LRRN1 and SNAB, was
295	associated with cerebral amyloid deposition and red cell distribution width(55,56); rs62143206,
296	which was included in the instrument of Glutathione S-transferase Pi, was also associated with
297	the monocyte percentage of white cells and the granulocyte percentage of myeloid white
298	cells(55). Further studies will be needed to validate our identified protein-PrCa associations.
299	Secondly, our analysis was constrained by the pQTLs identified in previous GWAS of
300	circulating protein levels, and thus we were unable to evaluate some important protein
301	biomarkers for PrCa as discussed previously. We anticipated that additional protein biomarkers
302	could be identified using newly identified pQTLs in the future. Furthermore, the current work
303	generates a list of promising protein candidates that show an association with PrCa, which can be
304	investigated further in future studies that directly measure levels of these proteins. Identification
305	of circulating protein biomarkers should be useful for PrCa risk assessment.
306	
307	In conclusion, in a large-scale study assessing associations between genetically predicted
308	circulating protein levels and PrCa, we identified multiple novel proteins showing a significant
309	association. Further investigation of these proteins will provide additional insight into the
310	biology and genetics of PrCa and facilitate the development of appropriate biomarker panels for
311	the early detection of PrCa.

312

313 Data availability

314	The OncoArray genotype data and relevant covariate information (i.e. ethnicity, country,
315	principal components, etc.) for prostate cancer study are available in dbGAP (Accession #:
316	phs001391.v1.p1). In total, 47 of the 52 OncoArray studies, encompassing nearly 90% of the
317	individual samples, are available. The previous meta-analysis summary results and genotype data
318	are currently available in dbGAP (Accession #: phs001081.v1.p1).
319	
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332	

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**Table 1.** Twenty-two novel protein-prostate cancer associations for proteins whose encoding genes are located at genomic loci at least500kb away from any GWAS-identified prostate cancer risk variants

					Distance of gene							P value after
		Protein-		Index	to the	Instrument	Type of				FDR P	adjusting for risk
Protein	Protein full name	gene	Region	$SNP(s)^{a}$	SNP (kb)	variants	pQTL	<b>OR</b> <sup>b</sup>	<b>95% CI</b> <sup>b</sup>	P value	value <sup>c</sup>	SNP <sup>d</sup>
	Cyclic AMP-	0	0		, , , ,							
	dependent											
	transcription factor					rs8111,	trans,			4	2	4
ATF6A	ATF-6 alpha	ATF6	1q23.3	rs4845695	6,824	rs61738953	trans	0.90	0.86-0.95	$1.31 \times 10^{-4}$	$9.18 \times 10^{-3}$	$1.31 \times 10^{-4}$
	Neutrophil cytosol					rs4632248,	trans,			0.00	<b>- - - - - 3</b>	*
NCF-2	factor 2	NCF2	1q25.3	rs199774366	20,932	rs28929474	trans	0.95	0.92-0.97	$9.93 \times 10^{-5}$	$7.29 \times 10^{-3}$	NA
<b>.</b>	<b>.</b>		1 05 0	100774266	01.077	rs62199218,	trans,	0.02	0.00.0.07	4.1.6	0.02	<b>NT 4</b> *
Laminin	Laminin	LAMCI	1q25.3	rs1997/4366	21,377	rs4129858	C1S	0.93	0.89-0.97	$4.16 \times 10^{-1}$	0.03	NA
DM22	39S ribosomal protein	MDDI 22	2-22.2	ma12295101	7 100		t	0.02		$0.61 \times 10^{-6}$	$7.42 \times 10^{-4}$	$0.42 \times 10^{-6}$
KIM33	Louine rich report	MKPL33	2p23.2	1813383191	7,100	rs28929474	trans	0.93	0.90-0.96	9.01 × 10	7.43 × 10	9.42 × 10
I PPN1	neuronal protein 1	IRRNI	3n26.2	rs2660753	83 221	18429558, rs6801780	ualis,	0.07	0.05.0.00	$7.21 \times 10^{-4}$	0.04	$7.21 \times 10^{-4}$
LIXINI	T-cell surface protein	LININI	3g13 13-	182000755	05,221	130001709	C15	0.97	0.95-0.99	7.21 ~ 10	0.04	7.21 ~ 10
ТАСТ	tactile	CD96	3a13.2	rs7611694	1 891	rs3132451	trans	1 22	1 16-1 29	$1.02 \times 10^{-12}$	$3.75 \times 10^{-10}$	$1.02 \times 10^{-12}$
mer	taethe	0000	5915.2	137011074	1,071	rs12368181	trans	1.22	1.10 1.27	1.02 ~ 10	5.75 ** 10	1.02 ** 10
IL-21	Interleukin-21	IL21	4a27	rs34480284	17.469	rs3129897	trans,	1.11	1.06-1.16	$7.77 \times 10^{-6}$	$7.43 \times 10^{-4}$	$NA^*$
	cAMP-specific 3 5-			1001100201	17,105	100129097	uuno		1100 1110		////0 10	
	cyclic		5q11.2-									
PDE4D	phosphodiesterase 4D	PDE4D	5q12.1	rs1482679	13,879	rs3132451	trans	1.17	1.12-1.22	$1.02 \times 10^{-12}$	$3.75 \times 10^{-10}$	$1.02 \times 10^{-12}$
	Glycine N-											
GNMT	methyltransferase	GNMT	6p21.1	rs4711748	763	rs57736976	cis	0.93	0.89-0.97	$6.80 \times 10^{-4}$	0.04	$2.78  imes 10^{-4}$
	Serine/threonine-											
PIM1	protein kinase pim-1	PIM1	6p21.2	rs9469899	2,345	rs28929474	trans	0.88	0.83-0.93	$9.61 \times 10^{-6}$	$7.43 \times 10^{-4}$	$9.42 \times 10^{-6}$
	WNT1-inducible-											
	signaling pathway									6		6
WISP-3	protein 3	WISP3	6q21	rs2273669	3,090	rs28929474	trans	0.83	0.77-0.90	$9.61 \times 10^{-6}$	$7.43 \times 10^{-4}$	$9.42 \times 10^{-6}$
	Protein-tyrosine									<b>z z z</b> 10-1		
TPST1	sulfotransferase 1	TPST1	7q11.21	rs56232506	18,233	rs313829	cis	1.14	1.06-1.22	$5.23 \times 10^{-4}$	0.03	$5.43 \times 10^{-4}$
ARFP2	Artaptin-2	ARFIP2	11p15.4	rs61890184	1,045	rs28929474	trans	1.23	1.12-1.35	$9.61 \times 10^{-0}$	$7.43 \times 10^{-4}$	$9.42 \times 10^{-6}$
GRIA4	Glutamate receptor 4	GRIA4	11q22.3	rs1800057	2,291	rs3132451	trans	1.17	1.12-1.22	$1.02 \times 10^{-12}$	$3.75 \times 10^{-10}$	$1.02 \times 10^{-12}$
KLRF1	Killer cell lectin-like	KLRF1	12p13.31	rs2066827	2,873	rs11708955,	trans,	1.13	1.05-1.20	$5.74 \times 10^{-4}$	0.03	$5.74 \times 10^{-4}$

	receptor subfamily F					rs62143194	trans					
	member 1											
			13q31.3-									
GPC6	Glypican-6	GPC6	13q32.1	rs9600079	20,151	rs28929474	trans	0.81	0.73-0.89	$9.61 \times 10^{-6}$	$7.43 \times 10^{-4}$	$9.42 \times 10^{-6}$
	IGF-like family											
TM149	receptor 1	IGFLR1	19q13.12	rs8102476	2,502	rs12459634	cis	1.06	1.02-1.09	$7.31 \times 10^{-4}$	0.04	$4.68 \times 10^{-3}$
	Tuberoinfundibular											
TIP39	peptide of 39 residues	PTH2	19q13.33	rs2659124	1,428	rs375375234	trans	1.22	1.13-1.32	$3.06 \times 10^{-7}$	$4.99 \times 10^{-5}$	$2.96 \times 10^{-7}$
	Zinc finger protein											
ZN175	175	ZNF175	19q13.41	rs2735839	710	rs28929474	trans	0.91	0.87-0.95	$9.61 \times 10^{-6}$	$7.43 \times 10^{-4}$	$9.42 \times 10^{-6}$
	Natural cytotoxicity											
NKp46	triggering receptor 1	NCR1	19q13.42	rs103294	620	rs2278428	cis	1.16	1.06-1.26	$9.91 \times 10^{-4}$	0.05	$9.65 \times 10^{-4}$
	Beta-soluble NSF					rs429358,	trans,					
SNAB	attachment protein	NAPB	20p11.21	rs11480453	7,945	rs7658970	trans	0.91	0.86-0.96	$9.77 \times 10^{-4}$	0.05	$9.77 \times 10^{-4}$
	Zinc fingers and											
ZHX3	homeoboxes protein 3	ZHX3	20q12	rs11480453	8,460	rs1694123	trans	0.79	0.71-0.88	$9.38 \times 10^{-6}$	$7.43 \times 10^{-4}$	$9.67 \times 10^{-6}$

<sup>a</sup> Closest risk variant identified in previous GWAS or fine-mapping studies for prostate cancer risk.

<sup>b</sup> OR (odds ratio) and CI (confidence interval) per one standard deviation increase in genetically predicted protein

<sup>c</sup> FDR *P* value: false discovery rate (FDR) adjusted *p* value; associations with a FDR  $p \le 0.05$  considered statistically significant

<sup>d</sup> using COJO method(27)

NA\*: the adjacent risk variant is not available in the 1000 Genomes Project data

**Table 2**. Nine novel protein-prostate cancer associations for proteins whose encoding genes are located at genomic loci within 500kb of previous GWAS-identified prostate cancer risk variants

Protein	Protein name	Protein- encoding	Region	Index SNP(s) <sup>a</sup>	Distance of gene to the index SNP (kb)	Instrument variants	Type of	OR <sup>b</sup>	95% CI <sup>b</sup>	<i>P</i> value	FDR <i>P</i> value <sup>c</sup>	P value after adjusting for risk SNPs <sup>d</sup>
Cathepsin S	Cathepsin S	CTSS	1021.3	rs17599629	44	rs41271951	cis	0.91	0.88-0.95	$2.73 \times 10^{-7}$	$4.99 \times 10^{-5}$	0.16
MICD	MHC class I polypeptide-related	MICD	C 21 22		122			1.00	1.05.1.12	2.07 × 10 <sup>-6</sup>	$2.76 \times 10^{-4}$	0.02
MICB	Sequence B	MICB	6p21.33	rs2596546	133	rs3134900	C1S	1.09	1.05-1.12	$2.07 \times 10^{-5}$	2.76×10	0.03
RF1ML	factor 1-like_ mitochondrial	MTRF1L	6q25.2	rs3968480	109	rs503366	cis	1.18	1.08-1.29	$4.67 \times 10^{-4}$	0.03	0.21
IGF-II receptor	Cation-independent mannose-6-phosphate receptor	IGF2R	6q25.3	rs651164	47	rs629849	cis	0.92	0.90-0.94	$3.98 \times 10^{-10}$	9.73 × 10 <sup>-8</sup>	$9.95 \times 10^{-11}$
PSP-94	Beta- microseminoprotein	MSMB	10q11.22	rs10993994	0.002	rs541781976, rs10993994	trans, cis	0.81	0.80-0.82	$3.60 \times 10^{-155}$	5.29 × 10 <sup>-152</sup>	NA <sup>*</sup>
Glutathione S-												
transferase	Glutathione S-					rs1695,	cis,			4		2
Pi	transferase P	GSTP1	11q13.2	rs12785905	399	rs62143206	trans	0.94	0.91-0.97	$5.91 \times 10^{-4}$	0.03	$3.12 \times 10^{-3}$
KDEL2	KDEL motif- containing protein 2	KDELC2	11q22.3	rs1800057	199	rs74911261	cis	0.89	0.86-0.93	$1.83 \times 10^{-8}$	$3.85 \times 10^{-6}$	0.42
SPINT2	Kunitz-type protease inhibitor 2	SPINT2	19q13.2	rs8102476 rs12610267	0	rs71354995	cis	1.05	1.03-1.06	$1.31 \times 10^{-6}$	$1.92 \times 10^{-4}$	0.07
DcR3	Tumor necrosis factor receptor superfamily member 6B	TNFRSF6B	20q13.33	rs6062509	33	rs62217798	cis	0.69	0.62-0.77	$1.98 \times 10^{-11}$	5.81 × 10 <sup>-9</sup>	0.05

<sup>a</sup> Closest risk variant(s) identified in previous GWAS or fine-mapping studies for prostate cancer risk

<sup>b</sup> OR (odds ratio) and CI (confidence interval) per one standard deviation increase in genetically predicted protein

<sup>c</sup> FDR *P* value: false discovery rate (FDR) adjusted *p* value; associations with a FDR  $p \le 0.05$  considered statistically significant

<sup>d</sup> using COJO method(27)

NA\*: the adjacent risk variant is the corresponding pQTL





# Analysis of over 140,000 European descendants identifies genetically-predicted blood protein biomarkers associated with prostate cancer risk

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