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Additional SNPs improve risk stratification of a polygenic hazard score for prostate cancer

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Running Title

Adding SNPs improves prostate cancer polygenic score

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Abstract

Background: Polygenic hazard scores (PHS) can identify individuals with increased risk of prostate cancer. We estimated the benefit of additional SNPs on performance of a previously validated PHS (PHS46).

- 5 Materials and Method: 180 SNPs, shown to be previously associated with prostate cancer, were used to develop a PHS model in men with European ancestry. A machine-learning approach, LASSO-regularized Cox regression, was used to select SNPs and to estimate their coefficients in the training set (75,596 men). Performance of the resulting model was evaluated in the testing/validation set (6,411 men) with two metrics: (1) hazard ratios (HRs) and (2) positive
- 10 predictive value (PPV) of prostate-specific antigen (PSA) testing. HRs were estimated between individuals with PHS in the top 5% to those in the middle 40% (HR95/50), top 20% to bottom 20% (HR80/20), and bottom 20% to middle 40% (HR20/50). PPV was calculated for the top 20% (PPV80) and top 5% (PPV95) of PHS as the fraction of individuals with elevated PSA that were diagnosed with clinically significant prostate cancer on biopsy.
- 15 Results: 166 SNPs had non-zero coefficients in the Cox model (PHS166). All HR metrics showed significant improvements for PHS166 compared to PHS46: HR95/50 increased from 3.72 to 5.09, HR80/20 increased from 6.12 to 9.45, and HR20/50 decreased from 0.41 to 0.34. By contrast, no significant differences were observed in PPV of PSA testing for clinically significant prostate cancer.
- 20 Conclusion: Incorporating 120 additional SNPs (PHS166 vs PHS46) significantly improved HRs for prostate cancer, while PPV of PSA testing remained the same.

Introduction

Optimal prostate cancer screening strategies seek to strike a balance between identifying clinically significant and potentially lethal cases that require treatment, while minimizing overdiagnosis of indolent, lower-risk cases that do not need radical treatment¹⁻³.

5 Genetic risk models have emerged as potentially useful tools that identify individuals with greater risk for being diagnosed with prostate cancer^{4,5}, and so help inform if and when to initiate screening for an individual. A subset of these models called polygenic hazard scores (PHS) seeks to directly identify associations between common genetic variants and the age of diagnosis of prostate cancer by utilizing the framework of time-to-event analyses^{1,6}.

10 We have previously reported on a PHS model for prostate cancer, PHS46, that demonstrated excellent performance in an independent test set of men from varied genetic ancestries⁶. The model incorporates genetic data of 46 unique single nucleotide polymorphisms (SNPs), and was identified through a systematic search of European men genotyped on the iCOGS chipset (Illumina, San Diego, CA). With an ever-increasing list of loci associated with
15 prostate cancer in the literature⁷⁻⁹, we sought to determine what effect, if any, the incorporation of additional SNPs would have on the performance of PHS46.

To this end, we employed a machine-learning approach, LASSO-regularized Cox regression,^{10,11} to select SNPs from a list that included the 46 used in PHS46, as well as over 100 SNPs identified in previous analyses as having genome-wide significance for association
20 with prostate cancer⁷. LASSO-regularized regression is an established variable selection technique in datasets with a large number of predictors and has been previously implemented as a SNP selection tool for a breast cancer polygenic risk score¹². Performance metrics describing statistical model goodness-of-fit and clinically actionable screening utility of the LASSO-regularized PHS model for prostate cancer were compared with those achieved with
25 PHS46 to determine the potential benefit of incorporating additional SNPs in polygenic hazard models.

Material and Methods

Study dataset

We obtained genotype and phenotype data from the PRACTICAL¹³ consortium for this analysis. Genotyping was performed previously on either OncoArray¹³ or iCOGS⁹ chips, and these data were previously imputed using the 1000 Genomes reference panel¹⁴. Missing SNP calls were replaced with the mean of the genotyped data for that SNP in the training set^{1,15}. In total, data from 82,007 men with European genetic ancestry (Supplementary Table 1)^{13,16} were available for this analysis. A testing set consisting of 6,411 men (4,828 controls and 1,583 cases) enrolled in the ProtecT clinical trial was set aside for estimating the performance of the final PHS models. The data from ProtecT were chosen as the testing set because they are well characterized and were previously used for validation of PHS46¹, allowing us to directly benchmark the performance of the updated model against previous iterations. The ProtecT trial also included biopsies of participants with elevated prostate-specific antigen (PSA) level, which permits analysis of the positive predictive value of the current clinical standard for screening, PSA testing. The remaining 75,596 individuals (25,127 controls and 50,469 cases) were used for training of the model. This first analysis was limited to men of European descent because of much greater data availability in that population, but our previous work has shown that development in Europeans can inform careful future work to assess and improve performance in other ancestries¹⁷.

Model development using LASSO regularization

A list of published SNPs previously identified^{1,7} to be associated with prostate cancer was compiled. In total, 180 unique SNPs were considered for estimation within the PHS model framework. An initial screening was conducted to identify pairs of SNPs that were highly correlated ($R^2 > 0.95$). For each pair of highly correlated SNPs, a univariable Cox proportional hazards model using age of diagnosis of prostate cancer as the time to event was calculated for

each SNP in the pair, and the one with the larger p-value was discarded. The remaining SNPs were included as candidates for the new PHS model. The R (v.4.0.1) package 'glmnet' was used to estimate a LASSO-regularized Cox-proportional hazards model^{10,11} using age of diagnosis of prostate cancer as the time to event. The genetic data of candidate SNPs and first four European ancestry principal components were included as predictors. Controls were censored at age of last follow-up. The hyper-parameter of the LASSO-regularized model, lambda, was selected using 10-fold cross-validation^{10,11}. The final form of the LASSO model was estimated at the value of lambda that minimized the mean cross-validated error.

Characterization of LASSO-regularized PHS model

The PHS score for each of the individuals in the training and testing set was estimated as the weighted sum of the genetic counts of each of the SNPs in the PHS model, using the LASSO model coefficients as weights. Distributions of the new PHS score were compared qualitatively between training and testing groups to confirm that the model was appropriately calibrated for use in the testing set.

We also sought to assess how the LASSO-regularized PHS score compared to family history in explaining the variation in age at diagnosis of prostate cancer. A multivariable Cox proportional hazards model was estimated using the age at diagnosis of any prostate cancer as the time to event, and the PHS score and family history as predictors in both training and testing sets, separately. The family history variable was coded as a binary variable: "None" or "One or more affected first-degree relatives". Observations with missing family history values were removed from the analysis. The explained relative risk¹⁸ (ERR) of each of the covariables as well as the full model were estimated using the "clinfun" software package in R, and provided a quantifiable measure for the importance of each variable in the model. Empirical confidence intervals for ERR were estimated using 1000 bootstrapped iterations.

Performance comparison between PHS46 and LASSO-regularized PHS

Performance in the testing set was assessed using hazard ratios (HRs) and positive predictive value (PPV), as described below. In each case, performance metrics were generated for the newly developed LASSO PHS model and for PHS46. Model coefficients for PHS46 were
5 obtained from the literature¹⁷. For each performance metric, one thousand bootstrap samples of the testing set were used to generate empirical 95% confidence intervals for LASSO PHS and for PHS46. In addition, bootstrapped 95% confidence intervals were generated for the percentage change of each performance metric between the two models, using PHS46 as the reference. Percent changes were deemed statistically significant if the bootstrapped 95%
10 confidence interval did not include 0.

HR performance

Calibration Cox proportional hazards models were fit to the bootstrapped testing data using the PHS score as the sole predictor and the age-of-diagnosis of prostate cancer as the
15 dependent variable. The model coefficient of this Cox regression model is referred to as the calibration factor. Next, the hazard ratio between two PHS groups, such as those in the top 5% to the middle 40% (HR95/50), is estimated as the exponential of the product of the calibration factor and the difference in mean PHS scores of each group. Hazard ratios between the top 20% to the bottom 20% (HR80/20) and the bottom 20% to the middle 40% (HR20/50) were
20 similarly calculated. The PHS cutoffs used to define these groups were determined from the distribution of PHS in the training set controls under 70 years of age^{1,15}.

A similar strategy was used to estimate the HR performance for clinically significant prostate cancer. The criteria for clinical significance were any of: Gleason score ≥ 7 , stage T3-T4, PSA concentration $\geq 10\text{ng/mL}$, pelvic lymph nodal metastasis, or distant metastasis¹⁹. In
25 this analysis, controls and low-risk (i.e., not clinically significant) cancers were censored at age of last follow-up and age of diagnosis, respectively. HRs are reported after sample-weight

correction^{1,17,20} using the total number of cases and controls in the ProtecT trial to generate weighting factors.

Sample-weight corrected HR values were also generated using the age at diagnosis of non-clinically significant prostate cancer. Individuals with clinically significant prostate cancer were removed from this secondary analysis.

PPV performance

One indicator of clinical utility of a risk-stratification approach like PHS is whether it can be used to improve the PPV of the standard clinical screening test, prostate-specific antigen (PSA). As a population-based screening study, ProtecT provides biopsy results of both cases and controls with a positive PSA result (i.e., ≥ 3 ng/mL). PPV performance of each model was estimated by randomly sampling individuals within the testing set with positive PSA results, while maintaining the case to control ratio of the ProtecT study (1:2). PPV is calculated as the fraction of positive PSA individuals in the top 20% (PPV80) or top 5% (PPV95) of PHS scores that had clinically significant prostate cancer.

Cumulative incidence curves for LASSO-PHS in United Kingdom

To illustrate the utility of the LASSO PHS model in informing prostate cancer screening, cumulative incidence curves for various PHS risk groups were estimated, as described previously²¹. The age-specific general cumulative incidence curve for prostate cancer was estimated for the United Kingdom population, aged 40 to 70, using data from Cancer Research UK 2015-2017²². The proportion of clinically significant and non-clinically significant prostate cancer at each age was estimated using data from the Cluster Randomized Trial of PSA Testing for Prostate Cancer (CAP) trial²³. Disease-specific cumulative incidence curves for clinically significant and non-clinically-significant prostate cancer were estimated by multiplying the general cumulative incidence curve by their respective proportions. The risk-adjusted incidence

curves for individuals in the upper 5th percentile and upper 20th percentile were estimated by multiplying the disease-specific cumulative incidence curves by the mean value of HR95/50 and HR80/50 in the testing set, respectively. Hazard ratios were obtained using the age of diagnosis of clinically significant prostate cancer as the time-to-event and after sample-weight correction.

5

Results

SNP screening and PHS model training

Of the 180 SNPs originally considered for this study, 6 SNPs were discarded in the initial screening process of removing highly correlated SNPs. Of the 174 remaining candidate SNPs (Supplementary Table 2), 166 had non-zero LASSO model coefficients and were selected for the final PHS model (PHS166).

The majority of the 166 variants (ninety-seven, 53%) used in PHS166 were classified as intron variants (Supplementary Table 3). Of the genes associated with variants from PHS166, HNF1B on chromosome 17 was associated with the greatest number of variants (4). Additional genes that were associated with multiple variants included ITGA6(x2), LINC00506(x2), PDLIM5(x2), TERT(x2), CTD-2194D22.4(x2), RGS17(x2), LOC105375751(x2), and CASC8(x3). Two of the SNPs used in PHS166 (rs721048 and rs10993994) were designated as ‘pathogenic’ by ClinVar²⁴ and associated with hereditary prostate cancer.

20

PHS166 model characterization

Distributions of PHS166 score were visually consistent between training and testing sets (Supplementary Figure 1). The 20th, 30th, 70th, 80th, and 98th percentiles of the reference PHS risk scores (controls in training set) were estimated as -0.411, -0.307, 0.048, 0.154, and 0.557, respectively.

25

PHS166 contributed roughly 80 to 90 percent of the total explained relative risk (Supplementary Table 4) of a Cox proportional hazards model containing both family history and PHS166. Family history was not found to be statistically significantly associated with age at diagnosis of prostate cancer in the testing set¹.

5

Performance comparison – PHS46 vs. PHS166

All PHS166 HR-based performance metrics showed statistically significant improvements compared to PHS46 (Table 1), for both any and clinically significant prostate cancer. The mean HR95/50 and HR80/20 values for PHS166 were roughly 36 to 55% greater than those for PHS46. For example, HR80/20 for clinically significant prostate cancer increased from 6.12 to 9.45. Similarly, HR20/50 for PHS166 was, on average, 18% lower than that for PHS46. Similar trends were observed for non-clinically significant prostate cancer (Supplementary Table 5). No significant differences between models were observed in either of the PPV-based performance metrics (Table 2). Among individuals in the top 20% of risk scores with a positive PSA test, the estimated mean PPV for clinically significant prostate cancer was roughly 0.19 irrespective of the model used – indicating approximately 19% of positive PSA tests in this risk group yielded a diagnosis of clinically significant prostate cancer. By comparison, approximately 13% of all positive PSA tests resulted in a diagnosis of clinically significant prostate cancer.

20

Cumulative incidence curves for PHS166 in United Kingdom

Cumulative incidence curves for clinically significant and non-clinically significant prostate cancer for the upper 5th percentile (>95th percentile) and upper 20th percentile (>80th percentile) of PHS166 scores in the United Kingdom demonstrated expected stratification of prostate cancer risk (Figure 1).

25

Discussion

Using a machine-learning, LASSO-regularized Cox framework, we identified 166 SNPs to be included in a polygenic hazard model (PHS166) for association with age of diagnosis of prostate cancer in men of European genetic ancestry. Variants used in PHS166 were associated with several genes, including those encoding for hepatocyte nuclear factor-1 beta (HNF1B), cancer susceptibility 8 (CASC8), and telomerase (TERT). PHS166 also explained a much larger percentage of the total explained relative risk compared to family history, suggesting that the former is important for stratifying patients' risk. When compared to the original PHS, consisting of 46 SNPs, PHS166 demonstrated substantially improved HR performance. For example, the HR for clinically significant prostate cancer comparing the upper and lower quintiles of genetic risk increased by 56% when using PHS166. No significant improvements were found in the PPV of PSA testing when using PHS to stratify risk.

Increased separation in hazard rates between PHS risk groups may allow for more nuance in clinical decision making in certain scenarios. Accurate identification of low, intermediate, and high PHS risk groups in prostate cancer may help in decisions of when (or if) to initiate screening as well as possibly improving the interpretation of the disease screens²⁵. Targeting screening to men in the upper percentiles of polygenic risk as opposed to those in the lowest risk group may reduce the proportion of overdiagnosed indolent cancers from 43% to 19%^{26,27}. Risk stratification achieved here by PHS166 is similar or better than commonly used clinical tools for diseases such as breast cancer, diabetes, and cardiovascular disease^{25,28–30}. Clinically meaningful risk stratification is illustrated by the estimated cumulative incidence curves in Figure 1. This effect is particularly pronounced for clinically significant disease because of the increased proportion of clinically significant cases observed at older ages^{2,21,23}.

The lack of improvement in PPV in this study may suggest a “performance plateau” when using PHS to define broad risk categories for certain clinical applications. A similar effect has been previously described for prostate cancer polygenic models, in the context of using risk

scores to discriminate prostate biopsy outcomes³¹. Some of the precision in a score may also be diluted in broad clinical applications. The PPV analysis here is applied to participants in the ProtecT trial, which enrolled men aged 50 to 69 years, and screening in the trial was offered irrespective of underlying genetic risk². Further investigation is needed to learn whether timing
5 screening according to genetic risk might better leverage the superior HR performance of PHS166 risk score to improve the PPV of PSA testing.

LASSO frameworks have been used to identify SNPs for polygenic risk scores of several phenotypes, including fracture risk³², type 2 diabetes³³, and breast cancer¹². In this work, we have extended the application of LASSO to select SNPs in a polygenic hazard model of
10 prostate cancer from a list of candidates previously identified through logistic and time-to-event analysis. Simulation studies¹¹ have suggested that LASSO provides more robust estimates than stepwise selection in cases with both a few large effects, as well as many small effects. As new prostate cancer associated variants are discovered, this framework can be easily implemented to develop updated polygenic hazard models.

15 One limitation of PHS166 is that it was entirely developed and tested in European men. However, a well-vetted, well-tested PHS model for men of European genetic ancestry can be used as a starting block for developing models for other genetic ancestries, where large-scale databases are often more scarce, as has been shown for PHS46^{17,34}. Furthermore, some of the SNPs selected for incorporation into PHS166 were originally discovered in analyses that
20 included men from the ProtecT testing set. Therefore, the improvements in HRs observed for PHS166 may be somewhat overestimated. However, this bias is likely small, given that the testing set was only a small fraction (less than 5%) of the data used in prior discovery analyses, and the ProtecT data were not used to calculate SNP weights in PHS166. The LASSO-regularized Cox framework was also used to minimize any potential for over-fitting³⁵ by
25 introducing penalties for large effect sizes. In addition, this study uses age of diagnosis as the time-to-event variable, and any preceding period of undiagnosed disease is unknown.

Hypothetical perfect measurement of age of onset would likely further improve performance of the PHS model.

In conclusion, we applied a machine-learning, LASSO-regularized Cox regression framework to develop a larger PHS that includes 166 previously discovered SNPs. When
5 comparing the performance of PHS166 to the original model, PHS46, we found that incorporating 120 more SNPs significantly improved HRs for clinically significant prostate cancer. However, incorporating more SNPs did not improve on the ability of PHS46 to inform the PPV of PSA testing in the ProtecT dataset, perhaps illustrating a plateau effect and/or dilution of risk stratification in a broad clinical application.

Ethics Statement

All contributing studies were approved by the relevant ethics committees and performed in accordance with the Declaration of Helsinki; written informed consent was obtained from the study participants. The present analyses used de-identified data from the PRACTICAL

5 consortium and have been approved by the review board at the corresponding authors' institution.

Conflict of Interest:

All authors declare no personal or financial conflicts of interest for the submitted work except as follows. CCF is a scientific consultant for CorTechs Labs, Inc. RE reports honorarium as a speaker for GU-ASCO meeting in San Francisco Jan 2016, support from Janssen, and
5 honorarium as speaker for RMH-FR meeting Nov 2017. She reports honorarium as a speaker at the University of Chicago invited talk May 2018, and an educational honorarium by Bayer & Ipsen to attend GU Connect “Treatment sequencing for mCRPC patients within the changing landscape of mHSPC” at ESMO Barcelona, Sep 2019. She reports member of external Expert Committee on the Prostate Dx Advisory Panel. OAA received speaker’s honorarium from
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Data Availability Statement

The data used in this work were obtained from the Prostate Cancer Association Group to Investigate Cancer Associated Alterations in the Genome (PRACTICAL) consortium.

Readers who are interested in accessing the data must first submit a proposal to the Data

- 5 Access Committee. If the reader is not a member of the consortium, their concept form must be sponsored by a principal investigator (PI) of one of the PRACTICAL consortium member studies. If approved by the Data Access Committee, PIs within the consortium, each of whom retains ownership of their data submitted to the consortium, can then choose to participate in the specific proposal. In addition, portions of the data are available for request from dbGaP
10 (database of Genotypes and Phenotypes) which is maintained by the National Center for Biotechnology Information (NCBI):

<https://www.ncbi.nlm.nih.gov/gap/?term=lcogs+prostate>
<https://www.ncbi.nlm.nih.gov/gap/?term=lcogs+prostate>

Anyone can apply to join the consortium. The eligibility requirements are listed here:

- 15 http://practical.icr.ac.uk/blog/?page_id=9. Joining the consortium would not guarantee access, as a proposal for access would still be submitted to the Data Access Committee, but there would be no need for a separate member sponsor. Readers may find information about application by using the contact information below:

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Figure Legends

Figure 1. Cumulative incidence curves for PHS166. Risk-adjusted cumulative incidence curves for the upper 5th percentile (>95th percentile) and upper 20th percentile (>80th percentile) of PHS166 scores for clinically significant and non-clinically-significant prostate cancer.

- 5 Reference curves representing the population average cumulative incidence (i.e., unadjusted for genetic risk).

Table 1. HR performance in testing set. Sample-weight-corrected hazard ratios are estimated for PHS166 and PHS46 in the testing set, using age-of-onset of any or clinically significant prostate cancer. The percent change for each metric is calculated using the value of PHS46 as the reference. Mean values and 95% confidence intervals are reported.

Type of cancer	HR	PHS46	PHS166	Change (%)
Any	HR95/50	3.29 [2.73,3.77]	4.45 [3.68,5.06]	36 [18,53]
	HR80/20	5.15 [3.92,6.18]	7.85 [6.04,9.33]	53 [25,78]
	HR20/50	0.44 [0.40,0.49]	0.37 [0.33,0.40]	-18 [-25,-10]
Clinically Significant	HR95/50	3.72 [2.89,4.43]	5.09 [3.84,6.05]	37 [13,59]
	HR80/20	6.12 [4.18,7.67]	9.45 [6.17,11.79]	55 [17,88]
	HR20/50	0.41 [0.35,0.47]	0.34 [0.29,0.39]	-18 [-28,-9]

Table 2. PPV performance in testing set. Positive predictive value (PPV) of PSA testing for clinically significant prostate cancer using top 5% (PPV95) and top 20% (PPV80) cutoffs of PHS166 and PHS46 risk scores. The percent change for each metric is calculated using the value of PHS46 as the reference.

PPV	PHS46	PHS166	Change (%)
PPV95	0.227 [0.159,0.292]	0.239 [0.171,0.305]	6.3 [-25.5,32.1]
PPV80	0.192 [0.155,0.231]	0.187 [0.150,0.222]	-2.8 [-16.3,9.9]

5

Cumulative incidence (%)

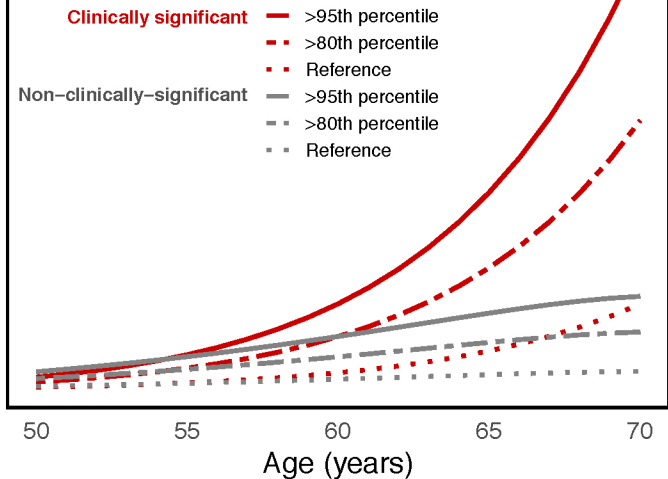


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Supplementary Table 1. Contributing Studies. Descriptions of contributing studies to training and testing sets.

Study Group Acronym	Study Group Name	cases	controls	Average age of cases
Training Set				
AHS	The Agricultural Health Study	491	1159	67.6
ATBC	Alpha-Tocopherol Beta-Carotene (BPC3)	1281	1913	72.2
Aarhus	Aarhus Prostate Cancer Study	1076	545	64.0
CCI	Cross Cancer Institute Prostate Brachytherapy Cohort	266	0	63.7
COH	City Of Hope	257	259	60.4
COSM	The Cohort Of Swedish Men	2298	1117	70.6
CPCS1	Copenhagen Prostate Cancer Study 1	536	258	68.2
CPCS2	Copenhagen Prostate Cancer Study 2	444	228	63.8
Canary PASS	Canary Prostate Active Surveillance Study (PASS)	364	0	62.1
CeRePP	French Prostate Case-Control Study	923	644	65.8
EPIC	European Prospective Investigation Into Cancer and Nutrition (BPC3)	635	693	66.7
ERSPC	European Randomised study of Screening for Prostate Cancer	71	65	71.2
ESTHER	Epidemiological investigations of the chances of preventing, recognizing early and optimally treating chronic diseases in an elderly population	324	315	64.8
FHCRC	Fred Hutchinson Prostate Cancer Studies	407	388	60.4
Gene-PARE	Genetic Predictors of Adverse Radiotherapy Response	242	0	66.2

HPFS	Health Professionals Follow-up Study (BPC3)	1168	1044	69.8
Hamburg-Zagreb		146	149	68.1
IMPACT	Identification of Men with a genetic predisposition to Prostate Cancer: Targeted screening in men at a higher genetic risk and controls	49	867	63.8
IPO-Porto	Portuguese Oncology Institute of Porto	374	180	56.3
KULEUVEN	Katholieke Universiteit Leuven	166	103	65.8
LAAPC	University of Southern California – Los Angeles Prostate Cancer Study	440	280	67.3
MCC-Spain	Multi Case Control Study- Spain	520	397	66.8
MCCS	Melbourne Collaborative Cohort Study	715	315	69.6
MDACC_AS	MD Anderson Cancer Center, Active surveillance trial	501	0	64.7
MEC	Multiethnic Cohort Study (BPC3)	598	642	69.9
MOFFITT	The Moffitt Group	403	203	64.7
Malaysia	Prostate cancer study in Malaysia	1	0	78.4
Oslo	COhort of NORway (CONOR)	1443	0	72.2
PCMUS	Prostate Cancer study Medical University Sofia	192	60	68.2
PHS	Physicians Health Study (BPC3)	622	257	68.7
PLCO	Prostate, Lung, Colorectal, and Ovarian Cancer Screening Trial (BPC3)	678	980	72.4
PRAGGA	PRostate cAncer Genetics in Galicia	129	100	68.1
PROCAP	PROgression in CAncer of the Prostate	659	236	64.3
PROFILE	Germline genetic profiling: correlation with targeted prostate cancer screening	13	21	59.4

	and treatment – The Pilot Profile Study			
PROGReSS	Prostate Cancer Group, Santiago, Spain	673	322	69.7
Poland	The Poland Group	484	317	69.1
ProMPT	Prostate cancer : Mechanisms of progression and Treatment	839	12	64.5
QLD	Prostate Cancer Supportive Care and Patient Outcomes Project (ProsCan), The QldMen and the Red Cross study	3282	1232	62.5
RAPPER	Radiogenomics: Assessment of Polymorphisms for Predicting the Effects of Radiotherapy	1914	0	70.2
SEARCH	Study of Epidemiology and Risk factors in Cancer Heredity	2511	1440	63.1
SFPCS	San Francisco Bay Area Prostate Cancer Study (former NC_CCPC)	279	205	64.8
SNP Prostate Ghent		316	135	65.3
SPAG	Serum Proteomic analysis for biomarkers of Aggressive prostate disease in the Guernsey population	39	169	65.3
STHM2	Stockholm 2	3019	1481	65.3
SWOG-PCPT	Prostate Cancer Prevention Trial	1072	1084	69.9
SWOG-SELECT	Selenium and Vitamin E Cancer Prevention Trial	1479	2070	67.4
TAMPERE	Finnish Genetic Predisposition to Prostate Cancer Study	2421	1183	67.2
TORONTO	Princess Margaret Biopsy Database	644	449	64.6
UKGPCS	U.K. Genetic Prostate Cancer Study and The Prostate Cancer Research Foundation Study	11939	1432	61.1

ULM	Familial Prostate Cancer Study Germany	457	178	64.3
WUGS	Washington University Genetics Study	669	0	61.2
Testing Set				
ProtecT	Prostate Testing for Cancer and Treatment	1553	1464	62.8
UKGPCS	U.K. Genetic Prostate Cancer Study and The Prostate Cancer Research Foundation Study	30	3364	61.0

Supplementary Table 2. SNP characteristics. RS-ID, Chromosome (Chr), Effect allele, Reference Allele (Ref), Base pair position (version 37), and LASSO-derived PHS coefficient for each of the 174 SNPs considered for this study. SNPs highlighted in gray (n=8) were not included in the final model – PHS166.

RS-ID	Chr	Effect	Ref	Position	PHS coefficient
rs56391074	1	AT	A	88210715	0.019
rs17599629	1	G	A	150658287	0.028
rs34579442	1	C	CT	153899900	0.033
rs1218582	1	G	A	154834183	0.023
rs4245739	1	C	A	204518842	-0.039
rs62106670	2	T	C	8597123	0.023
rs9287719	2	C	T	10710730	0.037
rs9306895	2	C	T	20878153	0.013
rs1465618	2	T	C	43553949	0.030
rs721048	2	A	G	63131731	0.002
rs6545977	2	A	G	63301164	-0.049
rs74702681	2	T	C	66652885	0.054
rs10187424	2	C	T	85794297	-0.035
rs11691517	2	G	T	111893096	-0.032
rs12621278	2	G	A	173311553	-0.116
rs16860513	2	A	T	173342367	-0.004
rs34925593	2	C	T	174234547	0.018
rs59308963	2	TATTCTGTC	T	202123479	-0.017
rs2292884	2	G	A	238443226	0.036
rs3771570	2	T	C	242382864	0.044
rs2660753	3	T	C	87110674	0.000
rs75219487	3	A	T	87147922	0.057
rs6788616	3	G	A	87205079	0.050
rs7611694	3	C	A	113275624	-0.041
rs4857841	3	A	G	128046643	0.048
rs6763931	3	A	G	141102833	0.016
rs182314334	3	C	T	152004202	-0.041
rs142436749	3	G	A	169093100	0.083
rs78416326	3	C	G	170074517	-0.091
rs10936632	3	C	A	170130102	-0.009
rs10009409	4	T	C	73855253	0.008
rs1894292	4	A	G	74349158	-0.026
rs12500426	4	A	C	95514609	0.000
rs6853490	4	G	A	95544718	0.020
rs17021918	4	T	C	95562877	-0.036
rs7679673	4	A	C	106061534	-0.073
rs2242652	5	A	G	1280028	-0.030

rs7725218	5	A	G	1282414	-0.040
rs2736108	5	T	C	1297488	0.039
rs10866527	5	T	C	1891800	0.026
rs12653946	5	T	C	1895829	0.003
rs2121875	5	C	A	44365545	0.009
rs10793821	5	C	T	133836209	-0.019
rs76551843	5	G	A	169172133	-0.138
rs4976790	5	T	G	177968915	0.027
rs4713266	6	T	C	11219030	-0.033
rs7767188	6	A	G	30073776	0.021
rs12665339	6	G	A	30601232	0.008
rs3096702	6	A	G	32192331	0.016
rs9296068	6	G	T	32988695	-0.018
rs9469899	6	A	G	34793124	0.034
rs1983891	6	T	C	41536427	0.034
rs4711748	6	T	C	43694598	0.019
rs9443189	6	G	A	76495882	-0.012
rs2273669	6	G	A	109285189	0.032
rs339331	6	C	T	117210052	-0.041
rs3910736	6	T	C	153412476	-0.020
rs1933488	6	G	A	153441079	-0.022
rs9364554	6	T	C	160833664	0.058
rs527510716	7	C	G	1944537	0.018
rs11452686	7	TA	T	20414110	0.000
rs12155172	7	A	G	20994491	0.048
rs10486567	7	A	G	27976563	-0.060
rs17621345	7	C	A	40875192	-0.029
rs56232506	7	A	G	47437244	0.023
rs6965016	7	C	A	97807882	0.051
rs2928679	8	A	G	23438975	0.028
rs11782388	8	C	T	23525358	0.047
rs11135910	8	T	C	25892142	0.043
rs9297746	8	C	T	127909361	-0.040
rs12543663	8	C	A	127924659	0.000
rs10086908	8	C	T	128011937	0.000
rs28556804	8	G	A	128014315	-0.061
rs77541621	8	A	G	128077146	0.152
rs1016343	8	T	C	128093297	0.081
rs183373024	8	G	A	128104117	0.361
rs16901979	8	A	C	128124916	0.022
rs60163266	8	A	G	128323157	0.025
rs620861	8	A	G	128335673	-0.051
rs6983267	8	T	G	128413305	-0.081

rs1447295	8	A	C	128485038	0.000
rs7812894	8	A	T	128520479	0.135
rs12549761	8	G	C	128540776	-0.083
rs1048169	9	C	T	19055965	0.021
rs17694493	9	G	C	22041998	0.022
rs10122495	9	T	A	34049779	0.004
rs1182	9	A	C	132576060	0.034
rs141536087	10	GCGCA	G	854691	0.058
rs76934034	10	C	T	46082985	-0.002
rs10993994	10	T	C	51549496	0.117
rs1935581	10	T	C	90195149	-0.029
rs3850699	10	G	A	104414221	-0.024
rs7094871	10	C	G	114712154	-0.008
rs4962416	10	C	T	126696872	0.037
rs1881502	11	T	C	1507512	0.008
rs72853963	11	A	G	2224664	0.022
rs7127900	11	A	G	2233574	0.069
rs61890184	11	A	G	7547587	0.032
rs547171081	11	CGG	C	47421962	0.014
rs2277283	11	C	T	61908440	0.033
rs12785905	11	C	G	66951965	0.033
rs12275055	11	G	A	68981359	0.076
rs7929962	11	C	T	68985583	-0.047
rs11290954	11	A	AC	76260543	-0.025
rs11568818	11	C	T	102401661	-0.031
rs1800057	11	G	C	108143456	0.052
rs11214775	11	A	G	113807181	-0.040
rs138466039	11	T	C	125054793	0.086
rs878987	11	G	A	134266372	0.025
rs2066827	12	G	T	12871099	-0.034
rs10845938	12	A	G	14416918	-0.034
rs80130819	12	C	A	48419618	-0.051
rs10875943	12	C	T	49676010	0.038
rs902774	12	A	G	53273904	0.008
rs55914512	12	T	G	53282274	0.067
rs7968403	12	C	T	65012824	-0.030
rs5799921	12	G	GA	90160530	-0.032
rs1270884	12	A	G	114685571	0.022
rs7295014	12	G	A	133067989	0.026
rs1004030	14	C	T	23305649	-0.014
rs11629412	14	G	C	37138294	-0.032
rs8008270	14	T	C	53372330	-0.042
rs4643253	14	C	T	69106108	0.000

rs7141529	14	T	C	69126744	-0.022
rs8014671	14	A	G	71092256	-0.021
rs4924487	15	G	C	40922915	-0.032
rs33984059	15	G	A	56385868	-0.074
rs112293876	15	C	CA	66764641	0.034
rs11863709	16	T	C	57654576	-0.061
rs201158093	16	TAA	TA	82178893	0.034
rs684232	17	C	T	618965	0.056
rs28441558	17	C	T	7803118	0.061
rs142444269	17	T	C	30098749	-0.027
rs11649743	17	A	G	36074979	-0.050
rs718961	17	A	G	36077099	-0.016
rs4430796	17	G	A	36098040	-0.011
rs11651052	17	A	G	36102381	-0.085
rs117576373	17	T	C	46820676	0.098
rs11650494	17	A	G	47345186	0.047
rs2680708	17	A	G	56456120	-0.024
rs1859962	17	G	T	69108753	0.083
rs8093601	18	C	G	51772473	0.018
rs28607662	18	C	T	53230859	0.026
rs12956892	18	T	G	56746315	0.013
rs533722308	18	CT	C	60961193	0.026
rs10460109	18	T	C	73036165	0.022
rs7241993	18	T	C	76773973	-0.048
rs11666569	19	T	C	17214073	-0.032
rs118005503	19	C	G	32167803	-0.024
rs8102476	19	T	C	38735613	-0.047
rs11672691	19	A	G	41985587	-0.049
rs61088131	19	C	T	42700947	-0.009
rs17632542	19	C	T	51361757	-0.197
rs2735839	19	A	G	51364623	-0.020
rs11480453	20	CA	C	31347512	-0.025
rs12480328	20	C	T	49527922	-0.039
rs6091758	20	G	A	52455205	0.045
rs2427345	20	T	C	61015611	-0.032
rs35897249	20	G	A	62233638	-0.017
rs6062509	20	G	T	62362563	-0.017
rs1041449	21	G	A	42901421	0.028
rs9625483	22	A	G	28888939	0.059
rs58133635	22	T	C	40471188	0.025
rs5759167	22	T	G	43500212	-0.064
rs73179053	22	C	T	43501620	-0.085
rs747745	22	C	T	43503547	-0.007

rs2405942	23	G	A	9814135	-0.021
rs17321482	23	T	C	11482634	-0.024
rs4907775	23	G	A	51263200	0.053
rs2807031	23	C	T	52896949	0.000
rs7888856	23	G	A	66751555	-0.034
rs5919432	23	C	T	67021550	-0.005
rs11795627	23	T	C	69957441	-0.020
rs6625711	23	A	T	70139850	0.004

Supplementary Table 3. Annotations for PHS166 variants. Annotations and nearby genes for 166 variants used in PHS166 were tabulated using publicly available data from dbSNP.

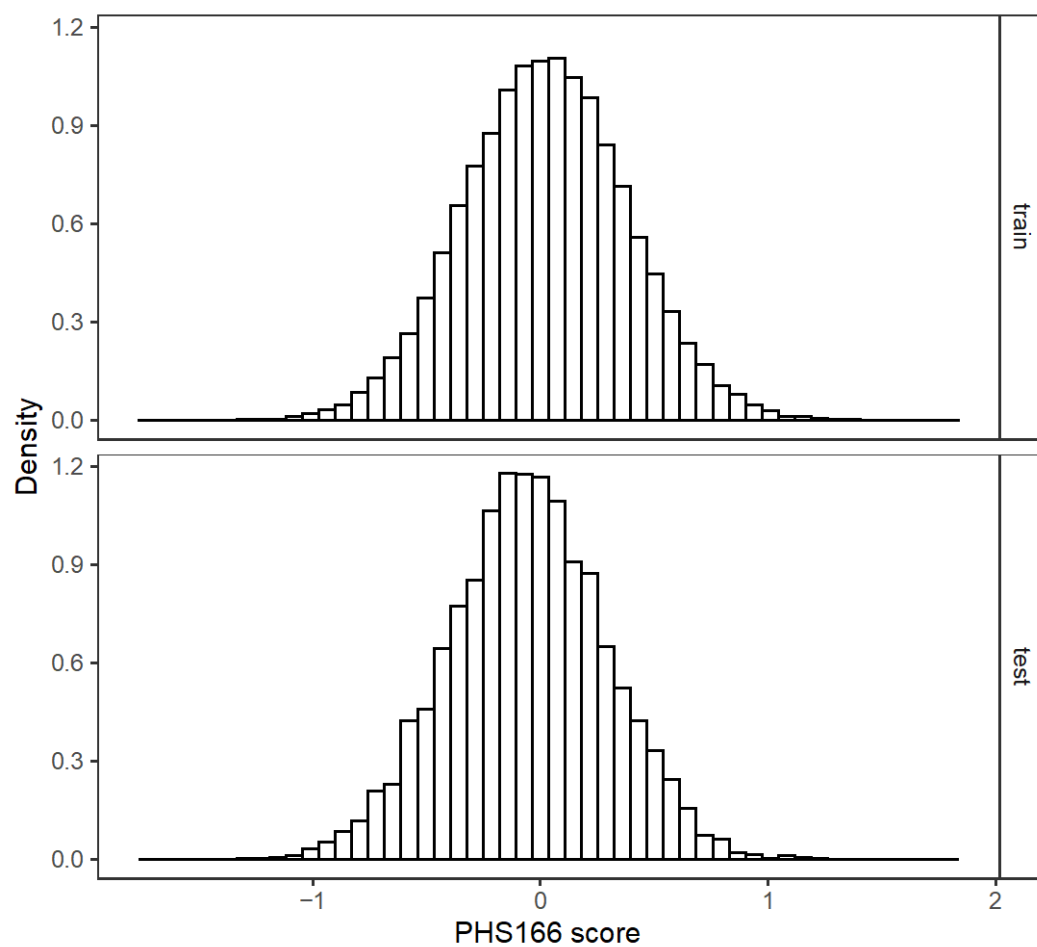
RS-ID	Gene annotations
rs56391074	None
rs17599629	GOLPH3L / Intron Variant
rs34579442	LOC101928059 / Intron Variant
rs1218582	KCNN3 / Intron Variant
rs4245739	MDM4 / Non Coding Transcript Variant
rs62106670	None
rs9287719	NOL10 / 500B Downstream Variant
rs9306895	GDF7 / 3 Prime UTR Variant
rs1465618	THADA / Intron Variant
rs721048	EHBP1 / Intron Variant
rs6545977	None
rs74702681	MEIS1-AS3 / Non Coding Transcript Variant
rs10187424	None
rs11691517	BCL2L11 / Intron Variant
rs12621278	ITGA6 / Intron Variant
rs16860513	ITGA6 / Intron Variant
rs34925593	None
rs59308963	CASP8 / Intron Variant
rs2292884	MLPH / Missense Variant
rs3771570	FARP2 / Intron Variant
rs75219487	LINC00506 / Intron Variant
rs6788616	LINC00506 / Intron Variant
rs7611694	SIDT1 / Intron Variant
rs4857841	EEFSEC / Intron Variant
rs6763931	ZBTB38 / Intron Variant
rs182314334	MBNL1 / Intron Variant
rs142436749	MECOM / Intron Variant
rs78416326	SKIL / 2KB Upstream Variant
rs10936632	None
rs10009409	LOC105377273 / Intron Variant
rs1894292	AFM / Intron Variant
rs6853490	PDLIM5 / Intron Variant
rs17021918	PDLIM5 / Intron Variant
rs7679673	None
rs2242652	TERT / Intron Variant
rs7725218	TERT / Intron Variant
rs2736108	None
rs10866527	CTD-2194D22.4 / Intron Variant
rs12653946	CTD-2194D22.4 / Intron Variant

rs2121875	FGF10 / Intron Variant
rs10793821	None
rs76551843	DOCK2 / Intron Variant
rs4976790	COL23A1 / Intron Variant
rs4713266	NEDD9 / Intron Variant
rs7767188	TRIM31 / Intron Variant
rs12665339	ATAT1 / Intron Variant
rs3096702	NOTCH4 / 2KB Upstream Variant
rs9296068	None
rs9469899	UHRF1BP1 / Intron Variant
rs1983891	FOXP4 / Intron Variant
rs4711748	None
rs9443189	MYO6 / Intron Variant
rs2273669	ARMC2 / Intron Variant
rs339331	RFX6 / Intron Variant
rs3910736	RGS17 / Intron Variant
rs1933488	RGS17 / Intron Variant
rs9364554	SLC22A3 / Intron Variant
rs527510716	MAD1L1 / Intron Variant
rs12155172	LINC01162 / Intron Variant
rs10486567	JAZF1 / Intron Variant
rs17621345	SUGCT / Intron Variant
rs56232506	TNS3 / Intron Variant
rs6965016	LMTK2 / Intron Variant
rs2928679	None
rs11782388	LOC107986930 / Intron Variant
rs11135910	EBF2 / Intron Variant
rs9297746	LOC105375751 / Intron Variant
rs28556804	LOC105375751 / Intron Variant
rs77541621	None
rs1016343	PCAT2 / Intron Variant
rs183373024	PRNCR1 / Non Coding Transcript Variant
rs16901979	None
rs60163266	CASC8 / Intron Variant
rs620861	CASC8 / Intron Variant
rs6983267	CASC8 / Intron Variant
rs7812894	None
rs12549761	None
rs1048169	HAUS6 / 3 Prime UTR Variant
rs17694493	CDKN2B-AS1 / Intron Variant
rs10122495	UBAP2 / 2KB Upstream Variant
rs1182	TOR1A / Non Coding Transcript Variant
rs141536087	LARP4B / 3 Prime UTR Variant

rs76934034	MARCHF8 / Intron Variant
rs10993994	MSMB / 2KB Upstream Variant
rs1935581	RNLS / Intron Variant
rs3850699	TRIM8 / Intron Variant
rs7094871	TCF7L2 / Intron Variant
rs4962416	CTBP2 / Intron Variant
rs1881502	MOB2 / Intron Variant
rs72853963	None
rs7127900	None
rs61890184	PPFIBP2 / Intron Variant
rs547171081	MIR4487 / 2KB Upstream Variant
rs2277283	INCENP / Missense Variant
rs12785905	KDM2A / Intron Variant
rs12275055	None
rs7929962	None
rs11290954	EMSY / Intron Variant
rs11568818	MMP7 / 2KB Upstream Variant
rs1800057	ATM / Missense Variant
rs11214775	HTR3B / Intron Variant
rs138466039	None
rs878987	B3GAT1 / Intron Variant
rs2066827	CDKN1B / Missense Variant
rs10845938	None
rs80130819	LOC105369750 / Intron Variant
rs10875943	None
rs902774	None
rs55914512	None
rs7968403	RASSF3 / Intron Variant
rs5799921	LOC107984543 / Intron Variant
rs1270884	None
rs7295014	FBRSL1 / Intron Variant
rs1004030	MMP14 / 2KB Upstream Variant
rs11629412	PAX9 / Intron Variant
rs8008270	FERMT2 / Intron Variant
rs7141529	RAD51B / Intron Variant
rs8014671	LOC101928075 / Intron Variant
rs4924487	KNL1 / Intron Variant
rs33984059	RFX7 / Missense Variant
rs112293876	None
rs11863709	ADGRG1 / Intron Variant
rs201158093	None

rs684232	VPS53 / 2KB Upstream Variant
rs28441558	CHD3 / Intron Variant
rs142444269	None
rs11649743	HNF1B / Intron Variant
rs718961	HNF1B / Intron Variant
rs4430796	HNF1B / Intron Variant
rs11651052	HNF1B / Intron Variant
rs117576373	LOC105371811 / Non Coding Transcript Variant
rs11650494	None
rs2680708	RNF43 / Intron Variant
rs1859962	CASC17 / Intron Variant
rs8093601	None
rs28607662	TCF4 / Intron Variant
rs12956892	None
rs533722308	BCL2 / Intron Variant
rs10460109	None
rs7241993	LOC105372225 / Intron Variant
rs11666569	MYO9B / Intron Variant
rs118005503	None
rs8102476	None
rs11672691	PCAT19 / Intron Variant
rs61088131	POU2F2 / Intron Variant
rs17632542	KLK3 / Missense Variant
rs2735839	None
rs11480453	None
rs12480328	ADNP / Intron Variant
rs6091758	None
rs2427345	LOC105372710 / Intron Variant
rs35897249	GMEB2 / Intron Variant
rs6062509	ZGPAT / Intron Variant
rs1041449	None
rs9625483	TTC28 / Intron Variant
rs58133635	TNRC6B / Intron Variant
rs5759167	None
rs73179053	None
rs747745	None
rs2405942	SHROOM2 / Intron Variant
rs17321482	ARHGAP6 / Intron Variant
rs4907775	None
rs7888856	None
rs5919432	None
rs11795627	TEX11 / Intron Variant
rs6625711	None

Supplementary Figure 1. Distribution of PHS166 score. Histograms of PHS166 scores for training and testing sets show similar patterns in distribution.



Supplementary Table 4. Explained relative risk (ERR) comparison. Values of ERR are tabulated for PHS166, Family History (None or ≥ 1 affected relative), and the full model (PHS166 + Family History). PHS166 contributed 82% of the overall ERR in the training set, and 98% of the overall ERR in the testing set. z

dataset	PHS166	Family History	PHS166 + Family History
training	0.116 [0.109,1.122]	0.016 [0.013, 0.018]	0.139 [0.132, 0.1460]
testing	0.147 [0.116,0.182]	0.001 [0, 0.006]	0.15 [0.12,0.185]

Supplementary Table 5. Testing performance for non-clinically significant prostate cancer. Sample-weight hazard ratios are estimated for PHS166 and PHS46 in the testing set using the age-of-onset of non-clinically significant prostate cancer.

HR	PHS46	PHS166	Change (%)
HR95/50	3.20 [2.59,3.78]	4.28 [3.49,4.96]	34 [16,51]
HR80/20	4.96 [3.61,6.14]	7.47 [5.49,9.07]	51 [23,77]
HR20/50	0.45 [0.39,0.50]	0.37 [0.33,0.42]	-17 [-25,-10]

Appendix 1. Additional members of the PRACTICAL Consortium

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CRUK and PRACTICAL consortium

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