Supplementary material for

Homeobox B13 G84E mutation and prostate cancer risk

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Systematic review and meta-analysis

Methods

We carried out a systematic review and meta-analysis with the aim to synthesise the available evidence on the association between the *HOXB13* G84E mutation and prostate cancer (PCa) risk. Specifically, we aimed to derive a pooled estimate which reflects the strength of association for unselected men in the general population.

On 1st February 2018, we searched Pubmed, MEDLINE and Embase for original publications using the search query

("HOXB13" OR "Homeobox B13" OR "rs138213197" OR "G84E" OR "p.Gly84Glu" OR "c.251G>A") AND "Prostate"

with no time period or language restrictions.

From the publications, we extracted RR estimates (odds ratio, risk ratio or hazard ratio), and approximated standard errors from the associated confidence interval widths. When RR and/or CI had not been reported but sufficient information was otherwise available in the publication (e.g. absolute frequencies in case-control studies), we calculated missing measures. Whenever available, we used adjusted estimates as reported in the publications.

For articles that reported on multiple studies, we extracted results from the individual studies rather than results based on pooled datasets whenever possible. We contacted the authors of three publications to ask for clarifications.

We grouped the RR estimates according to study design, and for case-control studies according to case and control selection as well as other study covariates, and tested for differences by fitting models with covariate moderators. To assess heterogeneity between published estimates, we used the DerSimonian-Laird heterogeneity of effects chi-square test and corresponding *l*² statistic. We assessed potential publication bias by producing funnel plots and tested for funnel plot asymmetry using the rank correlation test. We derived pooled meta-analysis estimates according to both fixed effects and random effects models. Finally, we performed a leave-one-out sensitivity analysis to assess the impact of individual studies on the results, by omitting one of the included studies at a time and refitting the models.

The meta-analysis was done using R software (version 3.4.0), with the meta and metafor packages.

Results

We identified 121 original research articles, of which 20 presented an estimate of the RR of prostate cancer for *HOXB13* G84E mutation carriers compared to the general population or non-carriers based on original data (Supplementary Fig. 1; (Akbari *et al.*, 2012; Breyer *et al.*, 2012; Ewing *et al.*, 2012; Gudmundsson *et al.*, 2012; Kluźniak *et al.*, 2013; Laitinen *et al.*, 2013; MacInnis *et al.*, 2013; Stott-Miller *et al.*, 2013; Witte *et al.*, 2013; Xu *et al.*, 2013; Chen *et al.*, 2013, 2018; Karlsson *et al.*, 2014; Albitar *et al.*, 2015; Kote-Jarai *et al.*, 2015; Beebe-Dimmer *et al.*, 2015; Hoffmann *et al.*, 2015; Storebjerg *et al.*, 2016; Karyadi *et al.*, 2017; FitzGerald *et al.*, 2017)). Supplementary Table 1 provides a summary of these publications.

A total of 17 case-control studies from 13 publications reported estimates based on unselected cases (Akbari *et al.*, 2012; Breyer *et al.*, 2012; Gudmundsson *et al.*, 2012; Laitinen *et al.*, 2013; Kluźniak *et al.*, 2013; Karlsson *et al.*, 2014; Albitar *et al.*, 2015; Kote-Jarai *et al.*, 2015; Beebe-Dimmer *et al.*, 2015; Hoffmann *et al.*, 2015; Storebjerg *et al.*, 2016; Karyadi *et al.*, 2017; Chen *et al.*, 2018), eight studies from six publications reported estimates based on cases selected on the basis of a young age at PCa diagnosis and/or family history (Breyer *et al.*, 2012; Ewing *et al.*, 2012; Gudmundsson *et al.*, 2012; Laitinen *et al.*, 2013; Stott-Miller *et al.*, 2013; Kote-Jarai *et al.*, 2012; Gudmundsson *et al.*, 2012; Laitinen *et al.*, 2013; Stott-Miller *et al.*, 2013; Kote-Jarai *et al.*, 2012; Gudmundsson *et al.*, 2012; Laitinen *et al.*, 2013; Stott-Miller *et al.*, 2013; Kote-Jarai *et al.*, 2015; and three studies from three publications reported estimates based on case-family study designs (Witte *et al.*, 2013; Xu *et al.*, 2013; FitzGerald *et al.*, 2017). Among the unselected case-control studies, 15 reported RR estimates that were significantly different from 1. All RR estimates from case-control studies with other case selections showed statistically significant deviation from 1. In addition, two independent prospective cohort studies where initially PCa-negative were followed for PCa onset (Chen *et al.*, 2013; Laitinen *et al.*, 2013), and a kin-cohort study based on modified segregation analysis (MacInnis *et al.*, 2013) reported statistically significant RR estimates.

Meta-analysis

Case-control studies

Fig. 1 shows a forest plot of published RRs. Among case-control studies, the RR estimates showed moderate heterogeneity (l^2 =52%, p<0.001), and differed by case selection (unselected, young age, familial, or both young age and familial cases; test for subgroup differences, p=0.007). In particular, estimates from case-control studies based on cases selected for family history were on average higher than estimates from studies on unselected cases (p=0.005).

We restricted the meta-analysis to case-control studies that compared unselected cases (with respect to age at diagnosis or family history) to unrelated controls. In these unselected case-control studies, RR estimates still showed moderate heterogeneity (l^2 =42%, p=0.036), and the corresponding funnel plot showed a tendency for asymmetry with many smaller studies reporting estimates higher than average (Supplementary Fig. 2; test for asymmetry, p=0.12). Overall, based on all unselected case-control studies, we derived a fixed effects model estimate of RR=3.23 (95% CI 2.80-3.74), and a random effects model estimate of RR=3.43 (95% CI 2.78-4.23).

Within the 17 unselected case-control studies, we found no statistically significant differences by control selection (population-based, screening-negative, or other disease controls; test for subgroup differences, p=0.7), nor by population setting (Europe or North America; test for subgroup differences, p=0.8). Two of these studies relied on cases from PSA screening trials (Laitinen et al., 2013; Karyadi et al., 2017), while the remaining 15 studies relied on observational case series. Restriction to the studies with observational case series did not substantially change the estimated RRs (fixed effects model RR=3.32, 95% CI 2.85-3.88; random effects model RR=3.57, 95% Cl 2.84-4.49) nor the estimated heterogeneity ($I^2=44\%$, p=0.036). Two of the studies used age-matched controls and four adjusted for age and/or other potential confounders. Five did no age adjustment but relied on screening-negative controls of similar age to the cases, or controls that were otherwise of similar ages. In the remaining six studies no adjustments were done. There was no significant difference by age adjustment scheme (matching/covariate adjustment, no adjustment but cases and controls were of similar ages, or no adjustment but cases and controls were of different ages; test for subgroup differences, p=0.9). Regressing on the average (mean or median) age of cases in the unselected case-control studies¹ revealed indications of a decrease in RRs with higher average age of cases, but this trend was not statistically significant and would only explain a minor proportion of the heterogeneity between estimates (test for continuous moderator, p=0.12; residual $l^2=39\%$, p=0.058).

In the leave-one-out sensitivity analysis, the fixed effects RR estimates varied between 3.09 and 3.52, and the random effects RR estimates between 3.18 and 3.60, and two studies were revealed to have outlier RRs whose omission had a substantial effect on the estimated heterogeneity (Supplementary Table 2; after omission, l^2 =25%, p=0.2 (Beebe-Dimmer *et al.*, 2015); after omission, l^2 =26%, p=0.2 (Gudmundsson *et al.*, 2012)). One of these two studies reported an RR estimate of 1.99 (95% CI 1.37-2.90), and relied on self-reported personal cancer history through a questionnaire to classify cases and controls (Beebe-Dimmer *et al.*, 2015). Misclassification of cases and controls is therefore a possible explanation for this outlying low RR estimate, as nondifferential misclassification may result in bias towards the null. We could not identify any likely methodological explanation for the outlying RR estimate of 7.51 (95% CI 3.99-14.11) from the other study (Gudmundsson *et al.*, 2012). After omission of the Beebe-Dimmer *et al* (2015) estimate, the estimated heterogeneity between the remaining 16 unselected case-control studies was low (l^2 =25%, p=0.2), the fixed effects model RR estimate was 3.52 (95% CI 3.01-4.13), and the random effects model RR estimate was 3.60 (95% CI 2.97-4.38). The corresponding funnel plot however showed similar indications of asymmetry as before the omission, with several smaller studies reporting higher than average RRs (Supplementary Fig. 3; test for asymmetry, p=0.11).

¹ For the purposes of this analysis, we used the mean age of the cases when available, but otherwise the median age if only the median had been reported. One study only reported mean ages by case subgroups (Gudmundsson *et al.*, 2012); we estimated the global mean for cases using the corresponding subgroup sizes. One study only reported the cases' ages in age groups (Karyadi *et al.*, 2017); we approximated the mean age based on the frequencies and midpoint age of each age group. Three studies did not report the age of cases, and we used the reported mean age of cases and controls combined (Beebe-Dimmer *et al.*, 2015; Hoffmann *et al.*, 2015), or the corresponding mean age of case-probands from the current analysis of the UKGPCS dataset (Kote-Jarai *et al.*, 2015).

Other study designs

Two prospective cohort studies have been performed, among biopsy-confirmed initially PCa-negative men with moderately heightened PSA (Chen *et al.*, 2013) and among biopsy-confirmed initially PCa-negative men with benign prostate hyperplasia (Laitinen *et al.*, 2013). While both studies relied on inclusion criteria with strong selection, we are not aware of any evidence to imply that PSA or BPH would act as effect modifiers. Finally, one previous kin-cohort study recruited cases based on young age, but adjusted for this ascertainment and analysed PCa patterns in family members at any age (MacInnis *et al.*, 2013). When these three studies were added to the previous 17 unselected case-control studies there was no substantial change in the estimated heterogeneity between estimates (I^2 =40%, p=0.032). The fixed effects model RR estimate was 3.21 (95% CI 2.80-3.69), and the random effects model estimate was 3.42 (95% CI 2.81-4.17). When one outlier unselected case-control study may somitted, the corresponding fixed effects model RR estimate was 3.46 (95% CI 2.98-4.02 and the random effects model RR estimate was 3.57 (95% CI 2.96-4.30).

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Supplementary Fig. 1 – Systematic review and meta-analysis: Flowchart detailing the identification of original research articles on the relative risk of prostate cancer for *HOXB13* G84E mutation carriers



^aNo mutation carriers in cases and/or controls: n=2. Pooled analysis of previously published studies: n=1. ^bNo prostate cancers observed in mutation carriers: n=1.

Supplementary Table 1 – Systematic review and meta-analysis: Summary of published studies that estimate the relative risk of prostate cancer for HOXB13 G84E mutation carriers

Case-control studies

					Cases		Controls		Cases	Controls					
			Case	Control	Mutation-	Cases	Mutation-	Controls	Average	Average				Age-	
Publication	Population	Dataset	selection	selection	carriers	Total	carriers	Total	Age	Age	RR	95% CI	Adjustments	adjustment	Comment
Ewing et al.,	2012														
		University of	Young age												
		Michigan	(≤55 years)												
		Prostate	and/or											No, but	
		Cancer	familial (≥1						Mean:					used	
	USA,	Genetics	affected	Concenting					52.2,					screening-	
	European	Project (UM-	FDR or	Screening-					Median:	Not				negative	
	ancestry	PCGP)	SDR)	negative	26	1130	1	1401	52	stated	33.0		No	controls	
			Mixed,												
			unselected											No, but	
			and						Mean:					used	
	USA,		familial (≥2	Concenting					58.5,					screening-	
	European	John Hopkins	affected	Screening-					Median:	Not				negative	
	ancestry	University	FDR)	negative	46	3953	1	1401	59	stated	16.5		No	controls	
Akbari <i>et al.</i> ,	2012														
														No, but	
														used	
	Canada,													screening-	
	European			Screening-					Mean:	Mean:				negative	
	ancestry	Sunnybrook	Unselected	negative	10	1525	2	1759	65.6	63.0	5.8	1.3-26.5	No	controls	
Breyer et al.,	2012														
			Familial												
			(≥1												
	USA,	Familial	affected												
	European	Prostate	FDR or	Screening-					Mean:	Mean:				Covariate	
	ancestry	Cancer Study	SDR)	negative ^a	16	858	2	825	60.1	60.1	7.9	1.8-34.5	Age	adjustment	
	USA,			C											
	European	Singleton		Screening-					Mean:	Mean:				Covariate	
	ancestry	cases	Unselected	negative ^a	4	268	2	825	58.4	60.1	5.6	0.9-33.9	Age	adjustment	
Gudmundsso	on et al., 2012														
															Per allele OR;
															Relied on
															imputed
				Other					Mean:	Mean:					<i>HOXB13</i> G84E
1	Iceland		Unselected	disease		4537	1	54444	71	84	3.67	1.71-7.90	No	No	status

									Mean:						
				Deve lation					60,	Net					
			Unsolocted	Population-		1092		1260	so	NOT	147	2 50 60 14	No	No	Por allala OP
	USA		Unselected	Daseu		1902		1200	Mean [.]	Stateu	14.7	3.33-00.14	NO	NO	Fel allele OK
									69.						
				Other					Median:	Not					
	Spain ^b		Unselected	disease		716		1692	70	stated	~		No	No	Per allele OR
										Not					
										stated,					
										however					
										age-					
	The			Population-						matched					
	Netherlands		Unselected	based		1520		1916	64.3	to cases	7.51	3.99-14.11	No	Matching	Per allele OR
									Mean:	Mean:					
				Others					70,	60,					
	Pomania ^C		Uncolocted	Other		772		957	iviedian:	iviedian:	1 10	0~~	No	No	Por allala OP
	KUIIIdilid		Unselected	uisease		122		657	/1	02	1.19	0-00	NO	No but	Per allele OK
														used	
			Young age											screening-	
			(50-69	Screening-					Mean:	Mean:				negative	
	UK	ProtecT	years)	negative		511		1825	63.0	62.5	14.44	4.74-44.03	No	controls	Per allele OR
Kluźniak at a	2012	•		-											•
KIUZIIIAK EL UI	., 2013		Ι	Dopulation					Mooni	Moon:					
	Poland		Unselected	based	20	3515	3	2604	68.8	61.8	5	1.5-16.7	No	No	
Laitinon et al	2012 ^d														
Latinenetur	, 2015	Unselected													
		cases vs													
		Population		Population-					Mean:	Not					
	Finland	controls	Unselected	based ^a		3197		923	68.6	stated	3.6	2.2-5.7	No	No	
		ERSPC													
		screening												No, but	
		cases vs												used	
		ERSPC												screening-	
		screening		Screening-					Mean:	Not				negative	
	Finland	controls	Unselected	negative		1184		4544	67.0	stated	2.1	1.2-3.6	No	controls	
		Familial	Familial												
		e	(≥1												
		cases ⁻ vs	affected	Population-						Nut					
	Finland	Population	FDR or	hacad ^a		114		022	Mean:	NOt	0.7	4 2 16 0	No	No	
	Finland	controls	SDK)	Daseu		114		923	٥2.8	stated	ð.2	4.3-16.0	NO	INO	
Stott-Miller e	et al., 2013														

									Mean:	Mean:					
									~60.3;	~60.0;					
									Not	Not					
									stated,	stated,					
									but	but					
			(two data						approxi-	approxi-					
		Fred	collections:						mated	mated					
		Hutchinson	40-64						from	from					
		Cancor	40-04						age-	age-				Matching	
		Bosoarch	first 2E 74						group	group				nuc	
	Europoon	Contor	voars in	Population					frequen-	frequen-				pius	
	ancostry		years in	Population-	17	1210	5	1250	cies ^f	cies ^f	22	1 21 8 06	٨٥٥	adjustment	
	ancestry	(FICKC)	second	Daseu	17	1510	5	1259	CIE3	CIE3	5.5	1.21-0.90	Age	aujustment	
Witte et al., 2	013														
			Mixed,												
			familial (all												
			affected												
			brothers												
			from												
			families												
			with ≥1												
			brother												
			diagnosed												
			at age ≤73	Mixed,											
			years), or	family-											
			aggressive	based											
			disease	(healthy											
			(Gleason	brothers)											
			score ≥ 7 .	and ,											
			stage >T2c	unrelated									Age, race		
	USA, mixed		or PSA	(screening-					Mean:	Mean:			and	Covariate	
	ancestry		>10 ng/ml)	negative)	20	1645	3	1019	63.2	64.3	4.79		institution	adjustment	
		I	1 20 118/ 1111/		20	1013		1015	03.2	01.5	1.75		institution	adjustment	
Xu et al., 201	3			·			r	r	1	r	1				
			Familial	Family-											
			(affected	based											
			men in	(unaffected											
			multiple	men in											
			case	multiple											
			families	case											
		International	with at	families											
		Consortium	least one	with at											
		for Prostate	HOXB13	least one											
		Cancer	G84E	HOXB13											
		Genetics	mutation	G84E					Mean:	Not					
	International	(ICPCG) ^g	carrier;	mutation	154	326	36	117	64.4	stated	4.3	2.32-7.96	No	No	

			varying definitions)	carrier; varying definitions)											
Karlsson et al	., 2014		1											1	
	Sweden	Cancer of the Prostate Sweden (CAPS)	Unselected	Population- based	130	2805	24	1709	Median: 64.9	Median: 67.2	3.4	2.2-5.4	No	Matching	
	Sweden	Stockholm-1	Unselected	Screening- negative	91	2098	37	2880	Median: 65.8	Median: 64.0	3.5	2.4-5.2	No	No, but used screening- negative controls	
Albitar et al.,	2015	T	I	I	I			I	I		1		T	I	1
	USA, ancestry not specified		Unselected	Other disease	2	232	1	110	Median: 68	Median: 65			No	No, but controls were of similar ages	
Beebe-Dimm	er <i>et al.,</i> 2015														
	USA, mixed ancestry	Mayo Clinic Biobank	Unselected	Population- based	19	1362	23	5898	Not stated	Not stated	1.99	1.37-2.90	Age	Covariate adjustment	
Hoffmann et	al., 2015 ^h														
	USA, European ancestry	Kaiser Permanente Research Program on Genes, Environment and Health Genetic Epidemiology Research on Adult Health and Aging (RPGEH GERA)	Unselected	Population- based		3976		29517	Not stated	Not stated	3.63	2.48-5.85	Age and genetic ancestry	Covariate adjustment	Relied on imputed HOXB13 G84E status
Kote-Jarai et	al., 2015		1		1									1	1
	UK	PRM probands vs controls	Unselected	Screening- negative ^a	38	3494	28	5252	Not stated	Mean: 55	2.12	1.30-3.46	No	No	

			Familial												
			(men from												
			families												
			with >2												
			with 22												
			FDR, SDR												
			or IDR,												
			one of												
			whom												
			diagnosed												
			at age ≤65												
			years; or,												
			≥3 FDR,												
		PRS	SDR or TDR												
		probands vs	diagnosed	Screening-					Not	Mean:					
	UК	controls	at any age)	negative ^a	25	786	28	5252	stated	55	6.36	3.67-11.01	No	No	
	0	DRV	acany age,		20		20	0202	stated	00	0.00	0.07 11.01			
		nrobands vs	Voung ago	Screening-					Not	Moon					
		probatius vs	(CEO years)	nogativo ^a	71	4272	20	5252	stated		2 10	2 02 E 01	No	No	
	UK	controis		negative	/1	4373	20	JZJZ	Stateu	33	3.19	2.03-3.01	NO	NO	
Storebjerg et	al., 2016		•	•					•	1				•	
														No, but	
		Aarhus							Mean:	Mean:				used	
		Prostate							62.7,	61.2,				screening-	
		Cancer		Screening-					Median:	Median:				negative	
	Denmark	Project	Unselected	negative	25	995	8	1622	63	62	5.12	2.26-13.38	No	controls	
Karvadi <i>et al</i>	2017 ⁱ														
nar yaar ee ang	, 2027		I	I					Mean:	Mean:				I	
									~69.3	~68.3					
									Not	Not					
									stated	stated					
									but	but					
									approvi	approvi					
									approxi-	approxi-					
									from	from					
									from	from					
									age-	age-					
									group	group					
				Screening-					trequen-	trequen-				Covariate	
	USA	PLCO	Unselected	negative	47	4222	9	2899	cies [†]	cies [†]	3.78	1.94-8.28	Age	adjustment	
FitzGerald et	al., 2017														
	, _0_,		Mixed.	Mixed.										1	
			familial	family-											
			(affected	based					Not	Not					
	Australia		men from	(unaffected		744		805	stated	stated	6.59 ^j		No	No	
		1		,					0.0.00	0.0.00		1			1

			families with ≥5 affected relatives) and young age (<75 years)	men from families with ≥5 affected relatives) and population-										
				based										
Chen <i>et al.,</i> 2	.018													
		Norwegian												
		Orological		Dopulation					Not	Madian				
		Cancer		Population-	10		4.2	1640	NOT	wedian:				
	Norway	Group	Unselected	based	18	/79	12	1643	stated	64	3.8	NO	NO	

Other study designs

				Follow-up				
Publication	Population	Dataset	Inclusion criteria	onset	Observed rates	RR	95% CI	Adjustments
Prospective cohort stud	dies							
		REDUCE	Aged 50-75 years, PSA of 2.5-10 ng/ml if <60 years old or 3-10 ng/ml if ≥60 years old, prostate volume ≤80 cm ³ , negative initial biopsy, no high-grade prostatic intraepithelial neoplasia or atypical small acinar		No. PCa/Total No. (%):			
Chen <i>et al.,</i> 2013	International	trial	proliferation	4 years	701/3186 (22%) Non-carriers	2.45	1.48-4.07	No
Laitinen <i>et al.,</i> 2013	Finland	врн	Diagnosed with benign prostate hyperplasia (BPH)	8-14 years	254 PCa (2.6% mutation carriers) 516 without PCa (0.6% mutation carriers)	4.6	1.3-16.2	No
Kin-cohort study								
MacInnis et al., 2013	Australia		Family members of men with PCa at age <55 years, with additional inclusion of		9 relatives with PCa, in 19 families with a mutation carrier proband	16.4	2.5-107.2	Ascertainment

		men diagnosed at ages 55-59 years			
		some study years			

^a Used as control group to multiple case series in same publication.

^b Omitted from meta-analysis: no mutation carriers in controls.

^c Omitted from meta-analysis: infinite confidence interval.

^d Includes three case datasets (Unselected, Familial, and ERSPC) and two control datasets (Population, and EPSPC), and all combinations of case and control datasets are compared in the publication. For this meta-analysis, we selected the comparisons listed in the table so as to not count the same study populations multiple times.

^e Restricted to familial cases not previously reported on in Xu *et al.* (2013).

^f The mean age was approximated based on the frequencies and midpoint age of each age group.

^g Excludes families previously reported on in Ewing *et al.* (2012), but includes 5 *HOXB13* G84E carrier families from the UKGPCS familial PRS cohort.

^h Also analysed same data as a retrospective cohort and presented hazard ratios. This is not included in the present meta-analysis, to not count the same study populations multiple times.

ⁱ Karyadi *et al.* (2017) also reports results from the FHCRC study, which were previously published in Stott-Miller *et al.* (2013). Hence, we do not include the FHCRC results from Karyadi *et al.* (2017) in the present meta-analysis.

^j Calculated by a quasi-likelihood based method which accounts for relatedness between subjects but does not produce confidence intervals.

Supplementary Fig. 2 – Systematic review and meta-analysis: Funnel plot of published relative risk estimates from unselected case-control studies



Supplementary Table 2 – Systematic review and meta-analysis: Leave-one-out sensitivity analysis for the unselected case-control studies

Omitted study	Fixed effects RR (95% CI)	Random effects RR (95% CI)	<i>I</i> ² , p-value
After omitting: Akbari2012	3.22 (2.78-3.72)	3.40 (2.75-4.21)	44%, p=0.029
After omitting: Breyer2012, Singleton cases	3.22 (2.79-3.73)	3.42 (2.76-4.23)	45%, p=0.027
After omitting: Gudmundsson2012, Iceland	3.22 (2.78-3.73)	3.43 (2.75-4.28)	45%, p=0.025
After omitting: Gudmundsson2012, USA	3.18 (2.75-3.68)	3.30 (2.72-4.02)	35%, p=0.083
After omitting: Gudmundsson2012, The Netherlands	3.09 (2.66-3.58)	3.18 (2.64-3.83)	26%, p=0.2
After omitting: Kluzniak2013	3.21 (2.78-3.72)	3.40 (2.74-4.22)	45%, p=0.028
After omitting: Laitinen2013, Unselected cases vs Population controls	3.20 (2.75-3.73)	3.44 (2.73-4.32)	45%, p=0.026
After omitting: Laitinen2013, ERSPC screening cases vs ERSPC screening controls	3.34 (2.87-3.88)	3.56 (2.88-4.42)	40%, p=0.050
After omitting: Karlsson2014, CAPS	3.22 (2.76-3.75)	3.46 (2.75-4.36)	45%, p=0.025
After omitting: Karlsson2014, Stockholm-1	3.19 (2.73-3.73)	3.45 (2.73-4.36)	45%, p=0.026
After omitting: Albitar2015	3.25 (2.81-3.76)	3.46 (2.81-4.27)	43%, p=0.033
After omitting: Beebe-Dimmer2015	3.52 (3.01-4.13)	3.60 (2.97-4.38)	25%, p=0.2
After omitting: Hoffmann2015	3.19 (2.73-3.72)	3.44 (2.73-4.33)	45%, p=0.027
After omitting: Kote-Jarai2015, PRM	3.37 (2.89-3.92)	3.57 (2.89-4.43)	39%, p=0.058
After omitting: Storebjerg2016	3.19 (2.76-3.70)	3.37 (2.72-4.18)	43%, p=0.033
After omitting: Karyadi2017, PLCO	3.21 (2.77-3.73)	3.42 (2.74-4.27)	45%, p=0.0 <mark>26</mark>
After omitting: Chen2018	3.21 (2.77-3.73)	3.42 (2.74-4.27)	45%, p=0.026



Supplementary Fig. 3 – Systematic review and meta-analysis: Funnel plot of published relative risk estimates from unselected case-control studies, after omission of Beebe-Dimer *et al.* (2015)

Model parameterisation

The PCa incidence was assumed to follow a model of the form

$$\lambda(t_i, k_i) = \lambda_0(t_i, k_i) \times \exp(G_i(t_i, k_i) + P_i)$$

where $\lambda_i(t_i, k_i)$ is the incidence for the *i*th man at age t_i in birth cohort k_i ; $\lambda_0(t_i, k_i)$ is the baseline incidence; G_i is a variable dependent on the man's *HOXB13* G84E genotype and represents the RRs for heterozygous and/or homozygous mutation carriers (depending on the model of inheritance); and P_i is the man's polygenotype.

We constrained the average PCa incidence across all genotypes to agree with calendar-period- and cohortspecific PCa incidences for England and Wales (1,2). We split individuals into birth cohorts according to years of birth \leq 1909, 1910-1919, ..., \geq 1960, and assumed that the PCa incidence in each birth cohort followed that of those born in the mid-point years (e.g. 1915 for the second birth cohort, and using 1905 and 1965 for the earliest and latest cohort, respectively). To avoid large variations in incidences between successive years, we smoothed population incidences using LOWESS regression (3).

We parameterised the model in terms of the relative risk (RR) at the average polygenic load, i.e. averaged over the levels of the polygenic component in the population. Let $\overline{RR}_g(t, k)$ denote the RR at average polygenic load for men with *g* copies of *HOXB13* G84E compared to non-carriers. In this general form, \overline{RR}_g may depend on age *t* and/or birth cohort *k*.

Assuming independence between *HOXB13* G84E and the polygenotype, and that the polygenotype acts multiplicatively with *HOXB13* G84E, the incidence at time t in birth cohort k for a man with *HOXB13* G84E genotype g and polygenotype P is

$$\lambda_{g,P}(t,k) = \lambda_{g,0}(t,k)e^P$$

where $\lambda_{g,0}(t,k)$ is the corresponding incidence when P = 0. The average incidence over the polygenotypes in the population is

$$\bar{\lambda}_{g}(t,k) = \frac{\sum_{p} \phi_{p} f_{g,p}(t,k)}{\sum_{p} \phi_{p} S_{g,p}(t-1,k)} = \frac{\sum_{p} \phi_{p} S_{g,p}(t-1,k) e^{p}}{\sum_{p} \phi_{p} S_{g,p}(t-1,k)} \lambda_{g,0}(t,k)$$

where ϕ_p is the population frequency of polygenotype p, $S_{g,p}(t-1,k)$ is the survival function for men with genotype g and polygenotype p, and $f_{g,p}(t,k) = \lambda_{g,p}(t,k)S_{g,p}(t-1,k)$ the corresponding failure density function.

Hence, the RR for mutation carriers with *g* copies of *HOXB13* G84E compared to non-carriers at average polygenic effect in the population is

$$\overline{RR}_g(t,k) = \frac{\overline{\lambda}_g(t,k)}{\overline{\lambda}_0(t,k)} = C_g(t,k)RR_{g,0}(t,k)$$

where $RR_{g,0}(t,k)$ is the RR at polygenotype P = 0 and

$$C_g(t,k) = \left(\frac{\sum_p \phi_p S_{g,p}(t-1,k)e^p}{\sum_p \phi_p S_{g,p}(t-1,k)}\right) / \left(\frac{\sum_p \phi_p S_{0,p}(t-1,k)e^p}{\sum_p \phi_p S_{0,p}(t-1,k)}\right).$$

Explicitly, the model given at the start of this appendix of the incidence for the i^{th} individual with g_i copies of the *HOXB13* G84E mutation and polygenotype P_i , is thus specified with

$$G_{i}(t_{i},k_{i}) = \begin{cases} \ln[C_{g}^{-1}(t_{i},k_{i})\overline{RR}_{g\geq1}(t_{i},k_{i})] \times 1_{\{g_{i}\geq1\}}, \text{ dominant model} \\ \ln[C_{g}^{-1}(t_{i},k_{i})\overline{RR}_{g=2}(t_{i},k_{i})] \times 1_{\{g_{i}=2\}}, \text{ recessive model} \\ \sum_{\gamma=1}^{2} \ln[C_{g}^{-1}(t,k_{i})\overline{RR}_{\gamma}(t_{i},k_{i})] \times 1_{\{g_{i}=\gamma\}}, \text{ general model} \\ \ln[C_{g}^{-1}(t_{i},k_{i})\overline{RR}_{\text{per-allele}}(t_{i},k_{i})] \times g_{i}, \text{ multiplicative model} \end{cases}$$

where $\mathbf{1}_{\{-\}}$ is an indicator function taking values 0 or 1, and

 $P_i \sim N(0, \sigma_P^2)$

is normally distributed with standard deviation σ_P .

In the main effects models, we assumed constant $\overline{RR}_g(t, k) \equiv \overline{RR}_g$. In addition, we assessed modification of the effect of G84E by allowing the RR to vary by age and birth cohort.

When exploring modification in the effect of *HOXB13* G84E between birth cohort groups, we divided men into N_c birth cohort groups and estimated separate RR as

$$\ln \overline{RR}_g(t,k) = \sum_{c=1}^{N_C} \beta_c \times \mathbb{1}_{\{k \in \text{group } c\}}$$

where β_c is the ln*RR* in the c^{th} birth cohort group.

Similarly, we allowed the RR to vary between age groups by an analogue parameterisation. We also allowed the RR to vary with age according to a log-linear model as

$$\ln \overline{RR}_{a}(t,k) = \alpha_{0} + \alpha_{1} \times (t-70)$$

where α_0 corresponds to the estimated lnRR at age 70 and α_1 the change in lnRR per year of age.

Finally, we fit a model which allowed for modification by both age and birth cohort, by fitting separate birth cohort group specific intercepts in the above log-linear model as

$$\ln \overline{RR}_g(t,k) = \sum_{c=1}^{N_C} \gamma_c \times \mathbb{1}_{\{k \in \text{group } c\}} + \alpha \times (t-70)$$

where γ_c is the lnRR at age 70 in the c^{th} birth cohort group, and α the change in lnRR per year of age.

References

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All models

Supplementary Table 3 shows the main effects models for the *HOXB13* G84E mutation under different assumptions for the mode of inheritance. We found that a multiplicative model for the *HOXB13* G84E mutation fits the data best as measured by the Akaike information criterion (AIC), both for single gene models (AIC=44312.8) and models which allowed for a polygenic component to capture residual familial effects (AIC=40616.1). However, multiplicative, dominant and general models of inheritance showed similar model fit for both single gene models (multiplicative: AIC=44312.8; dominant: AIC=44314.9; general: AIC=44314.7) and polygenic models (polygenic multiplicative: AIC=40616.1; polygenic dominant: AIC=40619.4; polygenic general: AIC=40617.8).

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Models which allowed for a polygenic component improved the AIC as compared to the corresponding single gene model for every assumed mode of inheritance for *HOXB13* G84E (polygenic multiplicative: AIC=40616.1, versus multiplicative: AIC=44312.8; polygenic dominant: AIC=40619.4, versus dominant: AIC=44314.9; polygenic recessive: AIC=40635.8, versus recessive: AIC=44368.2; polygenic general: AIC=40617.8, versus general: AIC=44314.7).

As the polygenic multiplicative model had the overall lowest AIC among the considered models, we chose the polygenic multiplicative model as the main model for all subsequent analyses.

Supplementary Table 3	– All models: N	Main effects mo	odels by assi	umed inheritar	nce						
				Likelihood				Minor		Polygenic	
	Log-	No.		ratio test				allele		standard	
Model	likelihood	parameters	AICa	p-value	<i>HOXB13</i> G84E	RR ^b	(95% CI ^c)	frequency	(95% CI)	deviation	(95% CI)
Sporadic	-22182.9	1	44367.8	-		1.00 ^d		0.77%	(0.67%-0.89%)	0.00 ^d	
Dominant	-22155.5	2	44314.9	<0.001 ^e	Het/Hom ^f	3.84	(3.02-4.89)	0.17%	(0.13%-0.24%)	0.00 ^d	
Recessive	-22182.1	2	44368.2	0.2 ^e	Hom	2.81	(0.61-12.9)	0.76%	(0.66%-0.88%)	0.00 ^d	
General	-22154.4	3	44314.7	<0.001 ^e	Het	3.81	(2.99-4.86)	0.17%	(0.13%-0.24%)	0.00 ^d	
					Hom	12.9	(3.40-49.0)				
Multiplicative	-22154.4	2	44312.8	<0.001 ^e	Per-allele	3.79	(3.00-4.80)	0.18%	(0.13%-0.24%)	0.00 ^d	
Polygenic	-20315.6	2	40635.2	<0.001 ^e				0.77%	(0.67%-0.89%)	2.73	(2.65-2.81)
Polygenic dominant	-20306.7	3	40619.4	<0.001 ^g	Het/Hom	3.52	(1.78-6.97)	0.22%	(0.11%-0.44%)	2.72	(2.64-2.80)

Polygenic recessive	-20314.9	3	40635.8	0.2 ^g	Hom	2.70	(1.43-5.08)	0.76%	(0.66%-0.88%)	2.73	(2.65-2.81)
Polygenic general	-20304.9	4	40617.8	<0.001 ^g	Het	3.56	(1.72-7.34)	0.22%	(0.11%-0.45%)	2.72	(2.64-2.80)
					Hom	19.5	(3.24-117.6)				
Polygenic multiplicative	-20305.0	3	40616.1	<0.001 ^g	Per-allele	3.86	(2.16-6.88)	0.20%	(0.11%-0.36%)	2.72	(2.64-2.80)

^a AIC: Akaike information criterion.

^b RR: Relative risk.

^c CI: Confidence interval.

^d Constrained to constant value.

^e Compared to sporadic model.

^f Het: heterozygous carriers; Hom: homozygous carriers.

^g Compared to polygenic model.

Supplementary Table 4 shows the polygenic multiplicative model refitted with alternative assumptions to those used for the main model.

The main polygenic multiplicative model assumes calendar-period- and birth-cohort-specific population incidences.

• When we instead assumed that the latest available (2015) population incidence applied to all men, the model fit was considerably worse (AIC=83526.5, versus AIC=40616.1 for the main model).

The main polygenic multiplicative model assumes the polygenic component to act multiplicatively on both mutation carriers and non-carriers with equal polygenic standard deviation.

- When we assumed that the polygenic component does not act on *HOXB13* G84E mutation carriers (i.e. that the polygenic standard deviation is equal to zero for mutation carriers), the model fit was worse (AIC=40666.4, versus AIC=40616.1 for the main model).
- When we allowed for the polygenic component to act differently on *HOXB13* G84E mutation carriers and non-carriers (i.e. estimating separate polygenic standard deviations for mutation carriers and non-carriers), the AIC did not improve (AIC=40616.8, versus AIC=40616.1 for the main model), and there was no statistically significant difference between this model and the main model (p=0.3). There was a statistically significant difference between the model which allowed a polygenic standard deviation for mutation carriers, and the model where the polygenic component did not act on mutation carriers (p<0.001).

As these results support the initial assumptions of calendar-period- and birth-cohort-specific population incidences, as well as of a polygenic component that acts multiplicatively on both mutation carriers and non-carriers, we chose to retain the main polygenic multiplicative model.

Supplementary Table 4 –	Supplementary Table 4 – All models: Polygenic multiplicative model, alternative fits														
				Likelihood				Minor		Polygenic					
	Log-	No.		ratio test		Per-allele		allele		standard					
Model	likelihood	parameters	AIC ^a	p-value	Group	RR ^b	(95% CI°)	frequency	(95% CI)	deviation	(95% CI)				
Assuming latest available population incidence (2015) for all birth cohorts	-41760.2	3	83526.5	-		4.20	(3.49-5.04)	0.14%	(0.10%-0.18%)	3.26	(3.23-3.30)				
Assuming polygenic component does not act on mutation carriers	-20330.2	3	40666.4	-	Non-carriers G84E carriers	9.73	(7.74-12.2)	0.09%	(0.07%-0.11%)	2.72 0.00 ^d	(2.64-2.80)				
Assuming separate polygenic standard	-20304.4	4	40616.8	0.3 ^e ; <0.001 ^f		4.98	(2.74-9.05)	0.16%	(0.09%-0.29%)						

deviation for mutation	Non-carriers	2.72	(2.64-2.80)
carriers	G84E carriers	3.28	(2.63-4.09)

^a AIC: Akaike information criterion.

^b RR: Relative risk.

^c CI: Confidence interval.

^d Constrained to constant value.

^e Compared to polygenic multiplicative model.

^f Compared to polygenic multiplicative model where the polygenic component does not act on mutation carriers.

Supplementary Table 5 shows the considered polygenic multiplicative models that allowed for separate RRs for subgroups of *HOXB13* G84E mutation carriers split by age and/or birth cohort.

As compared to the main polygenic multiplicative model, a model which allowed the RR for *HOXB13* G84E mutation carriers to vary by ten-year age groups did not improve the fit (AIC=40621.0, versus AIC=40616.1 for the main model; p=0.5), but point estimates indicated higher risks at younger ages. Most age-specific RR models with broader age groups similarly showed no improvement in model fit (four age groups: AIC=40619.4; three age groups: AIC=40619.3). Only one of the considered models with dichotomous age groups showed a minor improvement in AIC (ages 35-54 and 55-84: AIC=40615.6, versus AIC=40616.1 for the main model; p=0.12). The best fitting age-specific model was one which allowed the log-RR to vary linearly with age (AIC=40614.8, versus AIC=40616.1 for the main model; p=0.068).

A model with separate RRs for *HOXB13* G84E mutation carriers for each of the seven birth cohorts showed improved fit over the main polygenic multiplicative model (AIC=40612.1, versus AIC=40616.1 for the main model; p=0.014). When we created broader birth cohort groups by collapsing adjacent birth cohorts, results were similar (four birth cohort groups: AIC=40613.1; three birth cohort groups: AIC=40613.4). The most parsimonious cohort-specific model as measured by AIC, was a model with two birth cohort groups split by years-of-birth \leq 1929 and \geq 1930 (AIC=40609.6, versus AIC=40616.1 for the main model; p=0.004).

The best fitting model with birth-cohort-specific RRs fit the data better than the best fitting model with age-specific RRs (AIC=40609.6 for the model with two birth-cohort-specific RRs, versus AIC=40614.8 for the model where the RR varied log-linearly with age). A model that allowed the RR for *HOXB13* G84E mutation carriers to vary by both age and birth cohort, according to the parametrisations used in the best fitting age- and cohort-specific models, did not improve the model fit (AIC=40611.3, versus AIC=40609.6 for the most parsimonious model with birth-cohort-specific RRs; p=0.6). In this model with age- and birth-cohort-specific RRs, the point estimates for the birth-cohort-specific RRs were similar to those of the birth-cohort-specific model (\leq 1929: RR=3.09 versus RR=3.13; \geq 1930: RR=5.96 versus RR=5.71), but the age-specific effect was attenuated as compared to the age-specific model (per-year-of-age RR=1.00, versus per-year-of-age RR=0.98).

We thus chose the model with the two birth cohort groups \leq 1929 and \geq 1930 as the most parsimonious model out of all considered.

				Likelihood				Minor		Polygenic	
	Log-	No.		ratio test		Per-allele		allele		standard	
Model	likelihood	parameters	AIC ^a	p-value	Group	RR ^b	(95% CI°)	frequency	(95% CI)	deviation	(95% CI)
Age-specific RR: five	-20303.5	7	40621.0	0.5 ^d				0.19%	(0.12%-0.31%)	2.72	(2.64-2.80)
age groups					Age 35-44	7.06	(2.12-23.5)				
					Age 45-54	4.83	(2.67-8.73)				
					Age 55-64	3.91	(2.37-6.43)				
					Age 65-74	3.73	(2.31-6.02)				

	_				Age 75-84	3.46	(1.94-6.16)				
Age-specific RR: four	-20303.7	6	40619.4	0.4 ^d				0.19%	(0.11%-0.33%)	2.72	(2.64-2.80)
age groups					Age 35-54	4.99	(2.81-8.85)				
					Age 55-64	3.95	(2.25-6.92)				
					Age 65-74	3.77	(2.23-6.35)				
					Age 75-84	3.49	(1.57-7.77)				
Age-specific RR: three	-20304.6	5	40619.3	0.7 ^d				0.19%	(0.12%-0.30%)	2.72	(2.64-2.80)
age groups					Age 35-64	4.22	(2.63-6.77)				
					Age 65-74	3.76	(2.38-5.93)				
					Age 75-84	3.49	(1.88-6.50)				
Age-specific RR:	-20304.6	4	40617.1	0.3 ^d				0.20%	(0.12%-0.34%)	2.72	(2.64-2.80)
dichotomous age					Age 35-44	6.73	(2.62-17.3)				
groups					Age 45-84	3.79	(2.26-6.35)				
	-20303.8	4	40615.6	0.12 ^d				0.20%	(0.12%-0.32%)	2.72	(2.64-2.80)
					Age 35-54	4.87	(2.90-8.18)				
					Age 55-84	3.77	(2.33-6.10)				
	-20304.7	4	40617.3	0.4 ^d				0.19%	(0.11%-0.33%)	2.72	(2.64-2.80)
					Age 35-64	4.22	(2.39-7.45)				
					Age 65-84	3.70	(2.16-6.35)				
	-20304.9	4	40617.8	0.6 ^d				0.20%	(0.11%-0.36%)	2.72	(2.64-2.80)
					Age 35-74	3.92	(2.20-7.01)				
					Age 75-84	3.41	(1.70-6.84)				
	-20303.4	4	40614.8	0.068 ^d				0.18%	(0.11%-0.31%)	2.72	(2.64-2.80)
					Baseline (Age 70)	3.70	(2.23-6.14)				

Age-specific RR: per year of age (log-linear	-				Per year of age	0.98	(0.97-1.00)				
model) ^e											
Birth-cohort-specific	-20297.1	9	40612.1	0.014 ^d				0.13%	(0.08%-0.20%)	2.72	(2.64-2.80)
RR: seven birth cohorts					Born ≤1909	1.59	(0.65-3.89)				
					Born 1910-19	4.64	(1.74-12.4)				
					Born 1920-29	3.64	(1.83-7.24)				
					Born 1930-39	6.32	(4.03-9.91)				
					Born 1940-49	6.08	(3.78-9.79)				
					Born 1950-59	6.37	(3.73-10.9)				
					Born ≥1960	11.8	(5.99-23.1)				
Birth-cohort-specific	-20300.6	6	40613.1	0.030 ^d				0.13%	(0.09%-0.21%)	2.72	(2.65-2.80)
RR: four birth cohort					Born ≤1919	2.74	(1.54-4.87)				
P. 0419					Born 1920-29	3.59	(1.94-6.65)				
					Born 1930-39	6.10	(3.89-9.58)				
					Born ≥1940	6.18	(3.86-9.89)				
Birth-cohort-specific	-20301.7	5	40613.4	0.035 ^d				0.12%	(0.08%-0.19%)	2.72	(2.65-2.80)
RR: three birth cohort					Born ≤1919	2.80	(1.72-4.56)				
P. 0419					Born 1920-39	5.69	(3.87-8.37)				
					Born ≥1940	6.62	(4.33-10.1)				
Birth-cohort-specific	-20301.5	4	40611.0	0.008 ^d				0.13%	(0.09%-0.19%)	2.72	(2.65-2.80)
RR: dichotomous birth					Born ≤1909	1.78	(0.77-4.11)				
conort groups					Born ≥1910	5.92	(4.11-8.52)				
	-20302.1	4	40612.3	0.016 ^d				0.13%	(0.09%-0.19%)	2.72	(2.65-2.80)
					Born ≤1919	2.75	(1.63-4.63)				

					Born ≥1920	5.97	(4.17-8.54)				
	-20300.8	4	40609.6	0.004 ^d				0.14%	(0.09%-0.21%)	2.72	(2.65-2.80)
					Born ≤1929	3.09	(2.03-4.71)				
					Born ≥1930	5.96	(4.01-8.88)				
	-20303.9	4	40615.8	0.13 ^d				0.17%	(0.10%-0.31%)	2.72	(2.64-2.80)
					Born ≤1939	3.76	(2.19-6.45)				
					Born ≥1940	4.82	(2.61-8.91)				
	-20304.2	4	40616.5	0.2 ^d				0.19%	(0.11%-0.32%)	2.72	(2.64-2.80)
					Born ≤1949	3.93	(2.36-6.52)				
					Born ≥1950	4.76	(2.60-8.72)				
	-20302.4	4	40612.7	0.021 ^d				0.19%	(0.12%-0.33%)	2.72	(2.64-2.80)
					Born ≤1959	3.87	(2.32-6.45)				
					Born ≥1960	7.69	(3.75-15.8)				
Age- and birth-cohort- specific RR ^f	-20300.7	5	40611.3	0.020 ^g ; 0.6 ^h				0.14%	(0.09%-0.21%)	2.72	(2.65-2.80)
					Born ≤1929 (Age 70)	3.13	(1.96-5.01)				
					Born ≥1930 (Age 70)	5.71	(3.71-8.78)				
					Per year of age	1.00	(0.98-1.01)				

^a AIC: Akaike information criterion.

^b RR: Relative risk.

^c CI: Confidence interval.

^d Compared to polygenic multiplicative model.

^e The model with log-linear age-specific RR was specified as $\ln \overline{RR}(t) = \alpha_0 + \alpha_1 \times (t - 70)$ at age *t*, where α_0 corresponds to the estimated RR at age 70 and α_1 the change in RR per year of age; see the Supplementary material (model parameterisation).

^f The model with age- and birth-cohort-specific RR was specified as $\ln \overline{RR}(t,k) = \gamma_{\leq 1929} \times 1_{\{k \in \text{birth cohort group} \leq 1929\}} + \gamma_{\geq 1930} \times 1_{\{k \in \text{birth cohort group} \geq 1930\}} + \alpha \times (t-70)$ for birth cohort k at age t, where $\gamma_{\leq 1929}$ and $\gamma_{\geq 1930}$ corresponds to the estimated RR at age 70 for men born ≤ 1929 and ≥ 1930 , respectively, and α the change in RR per year of age; see the Supplementary material (model parameterisation).

^g Compared to polygenic multiplicative model with log-linear age-specific RR. ^h Compared to polygenic multiplicative model RR specific to birth cohorts ≤1929/≥1930. **Supplementary Fig. 4** – Predicted cumulative prostate cancer risks for a 35 year old man carrying a single copy of the *HOXB13* G84E mutation with average (unknown) prostate cancer family history, as estimated by the most parsimonious model, by birth cohort



Age

Sensitivity analyses

Supplementary Table 6 shows the results of the sensitivity analyses by refitting the most parsimonious model (with birth-cohort-specific RRs) to various subgroups of the data.

To assess the potential impact of self-reported family history on second-degree relatives, we refitted the most parsimonious model using family data on first-degree relatives only. The RR and minor allele frequency point estimates of the resulting model were similar to those of the model fitted to the entire dataset.

To assess the potential impact of imperfect ascertainment adjustments, we refitted the model to the population ascertained PRM and young-onset ascertained PRY arms separately. Refitting to the family ascertained PRS arm is not possible, because the residual familial polygenic component is not identifiable in this dataset due to that the ascertainment adjustment involves conditioning on all phenotypes in the relatives. The PRM arm yielded lower RR point estimates than the PRY arm (\leq 1929: RR=3.18 for PRM versus RR=4.01 for PRY; \geq 1930: RR=4.72 for PRM versus RR=10.4 for PRY), but the confidence intervals were overlapping.

	Log-	No.		Likelihood ratio test				Minor allele		Polygenic standard	
Subgroup	likelihood	parameters	AIC ^a	p-value	Group	\mathbf{Per} -allele \mathbf{RR}^{b}	(95% CI°)	frequency	(95% CI)	deviation	(95% CI)
Subgroup: First degree-	-12708.0	4	25424.0	0.021 ^e				0.13%	(0.08%-0.20%)	2.92	(2.85-3.00)
relatives					Born ≤1929	3.54	(1.99-6.32)				
					Born ≥1930	6.31	(4.12-9.66)				
Subgroup: PRM	-5000.1	4	10008.3	0.3 ^f				0.14%	(0.05%-0.37%)	2.35	(2.16-2.54)
(population					Born ≤1929	3.18	(1.18-8.55)				
ascertaineu					Born ≥1930	4.72	(1.59-14.0)				
Subgroup: PRY	-15164.6	4	30337.3	0.037 ^g				0.08%	(0.04%-0.17%)	2.80	(2.72-2.89)
(ascertained by young					Born ≤1929	4.01	(2.10-7.64)				
age at diagriosis)					Born ≥1930	10.4	(5.03-21.5)				

Supplementary Table 6 – Sensitivity analyses: Most parsimonious polygenic multiplicative model with birth-cohort-specific RRs, refitted to subgroups

^a AIC: Akaike information criterion.

^b RR: Relative risk.

^d First degree relatives for PRM and PRY cohorts, but first- and second-degree relatives for the PRS cohort where families were ascertained on the basis of family history.

^e Compared to Polygenic multiplicative model, fit using first-degree relatives (not shown).

^g Compared to Polygenic multiplicative model, fit using PRY families (not shown).

^c CI: Confidence interval.

^f Compared to Polygenic multiplicative model, fit using PRM families (not shown).

Supplementary Table 7 shows the results for the sensitivity analyses by refitting the most parsimonious model (with birth-cohort-specific RRs), to datasets where alternative imputation or censoring schemes had been used to handle missing data on ages-at-diagnosis. The primary dataset used in the analysis was based on sampling from the observed age-at-diagnosis distribution for men in the same birth cohort as those in the population-based PRM arm.

- To assess the potential impact of relying on the observed distribution in the population-based arm, we performed an alternative imputation which was based on external population-based data. We sampled ages-at-diagnosis from historical calendar-period- and birth-cohort-specific population prostate cancer incidences, adjusted for year-of-birth-specific mortality. The resulting point estimates were very similar to those from the primary dataset.
- To allow for the fact that ages-at-diagnosis in relatives of men with familial or young-age prostate cancer may differ from population distributions, we performed an alternative imputation based on the mean observed age-at-diagnoses within each ascertainment group and birth cohort. The resulting point estimates were very similar to those from the primary dataset.
- To assess the robustness of the results to extreme values, we performed three constant value imputations: assigning ages-at-diagnosis 60, 70 or 80 to all with missing ages-at-diagnosis. RR estimates were higher when age 60 was assumed and lower when age 80 was assumed, but all three imputations resulted in models which showed statistically significant differences in the RRs between the two birth cohort groups.
- To assess the impact of not performing imputations we instead censored all men with missing ages-at-diagnosis at age zero, which is conceptually equivalent to excluding all with missing ages-at-diagnosis. The resulting model showed overall lower RR point estimates, but the differences in RRs between the two birth cohort groups were statistically significant.
- To assess the impact if all men with missing ages-at-diagnoses had been diagnosed with prostate cancer at a high age, we assumed all with missing data to be censored at age 85. The resulting model showed lower RR point estimates, but the differences in RRs between the two birth cohort groups were statistically significant.

In summary, all alternative imputation and censoring schemes resulted in statistically significant differences in RRs between the two birth cohort groups based on years-of-birth ≤1929 and ≥1930.

				Likelihood				Minor		Polygenic		
Imputation or	Log-	No.		ratio test				allele		standard		
censoring scheme	likelihood	parameters	AIC ^a	p-value	Group	Per-allele RR ^b	(95% CI°)	frequency	(95% CI)	deviation	(95% CI)	
Alternative imputation:	-20316.4	4	40640.9	0.004 ^e				0.14%	(0.09%-0.22%)	2.72	(2.65-2.80)	
Sampled from historical					Davia	2.01						
population prostate					Born	3.01	(1.95-4.64)					
cancer incidences ^d					≤1929							
					Born	5.80	(3.75-8.97)					
					≥1930							
	-20393.4	4	40794.8	0.005 ^e				0.14%	(0.09%-0.20%)	2.76	(2.69-2.84)	

Supplementary Table 7 - Sensitivity analyses: Most parsimonious polygenic multiplicative model with birth-cohort-specific RRs, alternative imputation schemes

Alternative imputation: Mean observed age in					Born ≤1929	3.15	(2.07-4.81)				
each ascertainment group and birth cohort					Born ≥1930	6.06	(4.11-8.93)				
Alternative imputation:	-24147.8	4	48303.6	0.046 ^e				0.13%	(0.08%-0.20%)	2.95	(2.88-3.01)
Assuming age 60 for all with unknown age at diagnosis					Born ≤1929	3.92	(2.22-6.93)				
diagnosis					Born ≥1930	6.28	(4.06-9.72)				
Alternative imputation:	-21356.7	4	42721.4	0.015 ^e				0.12%	(0.08%-0.20%)	2.81	(2.74-2.88)
Assuming age 70 for all with unknown age at diagnosis					Born ≤1929	4.02	(1.93-8.37)				
U					Born ≥1930	6.70	(4.12-10.9)				
Alternative imputation:	-20375.7	4	40759.3	0.017 ^e				0.20%	(0.14%-0.30%)	2.59	(2.51-2.69)
Assuming age 80 for all with unknown age at diagnosis					Born ≤1929	2.45	(1.70-3.53)				
					Born ≥1930	4.04	(2.75-5.93)				
Censoring all with	-11885.4	4	23778.8	0.013 ^e				0.20%	(0.12%-0.33%)	1.70	(1.60-1.80)
unknown age at diagnosis at age 0					Born ≤1929	2.18	(1.32-3.60)				
					Born ≥1930	4.08	(2.48-6.70)				
Censoring all with	-12258.0	4	24524.0	0.014 ^e				0.23%	(0.14%-0.38%)	1.30	(1.21-1.39)
unknown age at diagnosis at age 85					Born ≤1929	1.88	(1.17-3.02)				

(assuming diagnoses at	Born	3.47	(2.15-5.59)	
ages above 85)	≥1930			

^e Compared to the corresponding Polygenic multiplicative model, based on the data using the same alternative imputation or censoring scheme (not shown).

^a AIC: Akaike information criterion.

^b RR: Relative risk.

^c CI: Confidence interval.

^d Based on population prostate cancer incidences for England and Wales (1960-2015), adjusted for population mortality.