**DNA methylation landscapes of prostate cancer brain metastasis are shaped by early driver genetic alterations**

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**Conflict of interest disclosure statement**

M.A.R. is listed as co-inventor on the patent US7718369B2 issued to the University of Michigan and the Brigham and Women's Hospital, on ETS gene fusions. He is also a co-inventor on SPOP mutations in prostate cancer diagnosis and therapy issued to Cornell University. S.G. plays a consulting or advisory role to Astellas Pharma (Inst), Curevac (Inst), Novartis (Inst), Active Biotech (Inst), Bristol-Myers Squibb (Inst), Ferring (Inst), MaxiVax, Advanced Accelerator Applications, Roche, Janssen (Inst), Innocrin Pharma (Inst), Sanofi, Bayer (Inst), Orion Pharma GmbH, Clovis Oncology (Inst), Menarini Silicon Biosystems (Inst), MSD (Inst). S.G. is a co-author of the patent Method for biomarker (WO 3752009138392 A1). S.G. is an honorary member of Janssen and also has ties to Nektar and ProteoMediX. M.A.R. is on the Scientific Advisory Board of NeoGenomics, inc., and Owkin, inc. C.K.Y.N. serves as a consultant for Repare Therapeutics USA. L.B. has served as an advisor for Janssen, Bayer, Roche, and Systems Oncology, has received honoraria from Janssen, and has sponsored research agreements with Novartis, Systems Oncology, and Thermo Fisher. The remaining authors declare no competing interests.

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**Abstract**

Metastases from primary prostate cancers to rare locations, such as the brain, are becoming more common due to longer life expectancy resulting from improved treatments. Epigenetic dysregulation is a feature of primary prostate cancer, and distinct DNA methylation profiles have been shown to be associated with the mutually exclusive *SPOP* mutant or *TMPRSS2-ERG* fusion genetic backgrounds. Using a cohort of prostate cancer brain metastases (PCBM) from 42 patients, with matched primary tumors for 17 patients, we carried out a DNA methylation analysis to examine the epigenetic distinction between primary prostate cancer and PCBM, the association between epigenetic alterations and mutational background, and particular epigenetic alterations that may be associated with PCBM. Multiregion sampling of PCBM revealed epigenetic stability within metastases. Aberrant methylation in PCBM was associated with mutational background and PRC2 complex activity, an effect that is particularly pronounced in *SPOP* mutant PCBM. While PCBM displayed a CpG island hypermethylator phenotype, hypomethylation at the promoters of genes involved in neuroactive ligand-receptor interaction and cell adhesion molecules such as GABRB3, CLDN8, and CLDN4 was also observed, suggesting that cells from primary tumors may require specific reprogramming to form brain metastasis. This study revealed the DNA methylation landscapes of PCBM and the potential mechanisms and effects of PCBM-associated aberrant DNA methylation.

**Significance**: DNA methylation analysis reveals the molecular characteristics of prostate cancer brain metastases and may serve as a starting point for efforts to identify and target susceptibilities of these rare metastases.

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**Introduction**

Clinical outcomes for prostate cancer are highly variable, ranging from indolent tumors requiring no intervention to highly aggressive, metastatic disease. Concerted efforts to characterize somatic features of prostate cancer have identified early and mutually exclusive events such as *TMPRSS2-ETS* gene fusions (most frequently *TMPRSS2-ERG*) and *SPOP* somatic mutations in therapy-naive localized tumors (1–3) and androgen receptor (AR) alterations in the metastatic setting under androgen deprivation therapy (ADT) (4–7), among other frequent molecular alterations.

Beyond their genomic profiles, numerous studies have compared gene expression and DNA-methylation profiles of primary prostate cancers and benign tissue (8–10), and the distinction between primary tumors harboring *SPOP* mutations or *TMPRSS2-ERG* fusions is detectable at the epigenetic level (11). Tumors with these two common genetic alterations show distinct methylomes (11), with the activity of EZH2 in non-TMPRSS2-ERG cases implicated in driving this distinction (12). A hypermethylator phenotype has been described in common sites of prostate cancer metastasis, such as bone, lymph node, and liver, along with regulation of oncogenes by intergenic DNA-methylation (13,14). More recently, epigenetic stability was reported between metastases from different sites within the same patient (15). However, while improvements in treatment for prostate cancer have led to increased survival, it has also brought an increase in metastasis to rare locations such as the brain (16), and a deep molecular understanding of these is lacking.

We, therefore, analyzed DNA methylation data from a cohort of 42 patients with metastasis from primary prostate cancer to the brain, with both primary and metastatic samples available for 17 of these. Multi-region sampling permitted the evaluation of intra-tumoural epigenetic heterogeneity at the primary and metastatic sites. Furthermore, we examined

whether the established DNA methylation features of prostate cancer are retained in this setting and sought insights into mechanisms driving metastasis to the brain.

## Materials and Methods

*Patient selection and tumor procurement*

Tumor samples were collected from Pathology Departments in University and Cantonal Hospitals across Switzerland (Institute of Pathology, Bern/ Institute of Neuropathology, Zurich/ Institute of Medical Genetics and Pathology, Basel/ Institute of Pathology, Aarau/ Institute of Pathology, Münsterlingen/ Institute of Pathology, Liestal/ Institute of Pathology, St. Gallen) and from the Department of Pathology and Laboratory Medicine, and Urology, Emory University School of Medicine, Atlanta, USA. Inclusion criteria were defined as patients having available formalin-fixed paraffin-embedded (FFPE) blocks from confirmed CNS or meningeal metastases of prostate carcinoma and, if available, from the matched primary tumor and normal tissue (**Supplementary Table S1**). All analyses were carried out following protocols approved by the Ethical Committee Bern (Project ID: 2019–00328). No participant compensation was applied for the current study.

*Study population*

Prostate cancer is a cancer that only involves men, thus all subjects are male. There was no exclusion of subjects due to age, although prostate cancer is most often diagnosed in the 6th or 7th decade of life. Our cohort included samples from 43 male patients, with PCBM tissue from 42. Only the primary tumor was available for patient P41 since the brain metastasis tissue did not pass quality controls for methylation analysis. Patients qualified for inclusion in this study if written consent or no documented refusal was available (Human Research Act, HRA, Swiss Confederation; Art. 34). We collected archived FFPE tissue from CNS (brain/spinal cord) and meningeal metastases. Most tumor samples corresponded to diagnostic biopsies (from prostate or CNS/dura), transurethral resections (TURP), or

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prostatectomy specimens. Primary tumors and metastases from patients P1, P32, P43, P44, P48, and P49 were taken from autopsy tissues. Additional diagnostic biopsies were available for patient P1. At least one metastatic sample was included from 42 patients (i.e., except P41). For patient P43, metastatic samples at several time points were collected. Additionally, from 17/42 patients with metastatic tissue, primary tumor tissue was also available, including five patients (P1, P6, P9, P29, and P44) with primary tumor samples at multiple time points. In total, we analyzed 155 areas, which included 57 from primary tumors and 95 from metastases, two from normal prostates, and one from normal brain. All 155 selected areas underwent DNA methylation analysis and were integrated with whole exome sequencing and targeted-RNA-seq data on the same samples, produced as previously described (17).

*DNA extraction and Illumina EPIC methylation array*

After deparaffinization, DNA was extracted from selected FFPE core biopsies (1 mm diameter) of matched tumor and normal tissue using the QIAamp DNA micro kit (Qiagen). Quality and quantity were determined by real-time PCR (Agilent NGS FFPE QC Kit, Catalog # G9700B) and Qubit Fluorometer (Invitrogen; Thermo Fisher Qubit fluorimeter, RRID:SCR_018095). Methylation profiling was performed using Infinium® MethylationEPIC BeadChips (850K; Catalog # 20087706) as previously described (18). The Epigenomic Services from Diagenode (RRID:SCR_014807) carried out this assay.

*EPIC methylation array processing*

After whole-genome amplification and enzymatic fragmentation, the samples were hybridized to the BeadChip, and scanning was conducted with Illumina iScan (RRID:SCR_016388). Idat files were exported and analyzed using the minfi package in R (RRID:SCR_012830; 19). Probes associated with SNPs or with a detection *P* value > 0.01 in any sample were removed before analysis. Data were normalized using the functional normalization algorithm from the minfi package (19). Probes were annotated using the IlluminaHumanMethylationEPICanno.ilm10b2.hg19 package in Bioconductor

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(RRID:SCR_006442). Estimating tumor purity was done using the R package InfiniumPurify v2.0, and beta values were purity-corrected using the R function adjustBetas (https://github.com/StaafLab/adjustBetas, latest commit 19a5bc9). Purity-corrected M-values were derived from the corrected beta values using the B2M function from the R package ENmix v.1.31.02. Principal component analysis was performed using the top 1% (8,038) most variable CpG sites in each set of samples being analyzed. Comparisons of the intragroup variation, defined as the within-group pairwise Euclidean distance based on their PCs, were performed using Wilcoxon's tests. Comparisons of the intergroup variation, as measured by pairwise Euclidean distance based on their PCs between samples of different groups, were performed using Wilcoxon's tests.

Probe-level differential methylation analysis was performed for 803,841 CpG sites using limma (RRID:SCR_010943; 20). Probes with |logFC| > 1.5 and FDR < 0.05 were considered differentially methylated (DM). Differentially methylated regions (DMRs) were called using DMRcate using the parameters 'lambda=500, C=5'(21). Regions with ≥ 5 CpGs and a mean change in B value > |20%| and FDR < 0.05 were considered differentially methylated. DMRs were annotated using the annotatePeak function from ChIPSeeker (RRID:SCR_021322) and the TxDb.Hsapiens.UCSC.hg19.knownGene packages from Bioconductor (22). Enrichment for DM at different genomic regions was performed using Fisher's exact test with Benjamini-Hochberg correction for multiple testing. Gene set enrichment analysis on the CpG level was performed using the ebGSEA algorithm implemented in the ChAMP package v.2.24.0 (RRID:SCR_012891; 21,23). Gene ontology analysis on promoter DMRs was performed using goregion from the missMethyl package v.1.28.0 (24).

Unsupervised hierarchical consensus clustering was performed using the ConsensusClusterPlus package (RRID:SCR_016954), 1000 subsamplings, and ward.D2 linkage. The optimal number of clusters was selected using the elbow method applied to the plot of change in area under the cumulative density function curve (25).

*Integration with published datasets*

Data from 100 non-brain prostate cancer metastases was kindly provided by Prof. Felix Feng (UCSF Helen Diller Family Comprehensive Cancer Center) (13) and analyzed using the bsseq package (RRID:SCR_001072). Data were filtered to sites covered to at least 10X and present in the post-filtering EPIC array. CpGs were then annotated using the IlluminaHumanMethylationEPICanno.ilm10b2.hg19 package to analyze CpG island methylation.

Illumina 450k methylation data for 502 primary prostate cancers from The Cancer Genome Atlas (TCGA) cohort (RRID:SCR_003193; 26) were downloaded as the TCGA-PRAD cohort using TCGABiolinks on 18/05/2022 (RRID:SCR_017683; 27,28).

Before the analysis of promoter DMR methylation levels, the three datasets were merged using the CpG IDs and batch-corrected using the removeBatchEffect function from the limma package. The batch-corrected data were then reduced to the CpGs included in the *GABRB3, CLDN8,* and *CLDN4* promoter DMRs using subsetByOverlaps from the GenomicRanges package (RRID:SCR_000025; 29). Levels of methylation at the CpG level, and after averaging the level of methylation at each DMR, were compared between datasets using the Wilcoxon test.

**Data availability**

The data generated in this study are publicly available in The European Genome-phenome Archive (**EGA**) (EGAD00010002370).

**Results**

*Inter and Intrapatient epigenetic heterogeneity in PCBM*

Using the Illumina Infinium EPIC array, we acquired DNA methylation data from 155 samples from 43 patients (17 patients with PCBM with matched primary tumor tissue, 25 with PCBM only, and 1 with primary tumor only). The 155 samples comprised 95 regions from brain metastases, 57 regions from primary tumors, two normal prostate, and one from normal brain tissue (**Supplementary Table S1**).

We performed principal component analysis (PCA) using the 1% of most variably methylated CpG sites (8,038 CpGs) across all primary and metastatic samples (**Supplementary Table S2**). No obvious separation of primary and metastatic samples was revealed by PC1 and PC2 (**Fig. 1A**), suggesting other factors contributing to the observed variance. Correlating the variation with the mutation status of prostate cancer driver genes, the presence of mutations in *TP53* and/or *FOXA1* and/or *TMPRSS2-ERG* fusions showed significant correlations with the calculated eigenvectors for PCs 1 and 2 (**Fig. 1B**). PC components 3-10 correlated with other features such as *SPOP* mutations, sample type (metastasis or primary tumor), and other genomic aberrations (**Fig. 1B**). To assess whether metastatic samples were more dissimilar to normal prostate tissues than primary samples, we calculated the pairwise distance between the tumor samples and the normal prostate reference tissues. Indeed, we found that the metastatic samples were slightly farther from the normal prostate samples than the primary samples were ($P$ = 0.0043, Wilcoxon test; **Fig. 1C**). Interestingly, the intra-patient variation between metastases was significantly lower compared to the variation between primary tumors or between metastases and primary tumors ($P < 2.22 \times 10^{-16}$, Wilcoxon test, **Fig. 1D**). The opposite was true for inter-patient variation, where metastases showed modestly higher variation compared to primary tumors ($P < 2.22 \times 10^{-16}$, Wilcoxon test, **Fig. 1E**).

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To obtain a focussed view of intra-patient heterogeneity, we performed unsupervised hierarchical clustering, using the top 1% most variably methylated CpGs, on a per-patient basis for the 17 patients for which primary and metastatic samples were available. This highlighted the separation between primary and metastatic samples within each patient, as these clustered separately in 12/16 cases with three or more samples (**Fig. 2, Supplementary Fig. S1**). We further observed that the top 1% of most variably methylated CpGs tended to be hypomethylated in the metastases compared to normal prostate. In some cases (such as P4 and P32), the metastatic samples harbored large blocks of hypomethylated CpG sites. However, the opposite (P9) or mixed patterns were observed in some cases (**Supplementary Fig. S1**). Interestingly, the only two samples annotated as small cell NE-carcinoma, in P1 and P49, clustered separately from the other samples from the same patient (**Fig. 2**), even though they are genetically related as they shared a large fraction of clonal somatic mutations (17). We further examined the top 1% most variably methylated CpG sites in the 18 cases with more than one metastatic sample and again observed the tendency of global hypomethylation among these samples (**Supplementary Fig. S2**). Some of these cases, e.g., P40, P45, and P47, had larger blocks of hypermethylated CpGs than the normal samples.

These data suggest that, while primary prostate cancer samples have epigenomes that distinguish them from normal prostate, metastatic samples acquire additional DNA methylation changes, further separating them epigenetically from normal tissue and primary cancer. The smaller distance between metastatic samples suggests reduced epigenetic heterogeneity within PCBM compared to primary prostate cancer.

*DNA methylation changes associated with early driver mutations distinguish PCBM*
It has previously been reported that prostate cancer with and without the *TMPRSS2-ERG* fusion show distinct methylomes (13), and *TMPRSS2-ERG* fusion-negative prostate cancers

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acquire DNA hypermethylation dependent on EZH2 (12). However, whether the epigenetic distinction resulting from these mutually exclusive driver events persists following metastasis has not been addressed. Using the 95 metastatic cases annotated with genotypic information in our previously published study (17), we performed unsupervised hierarchical consensus clustering using the top 1% of most variably methylated CpG sites across the metastatic samples. This separated most of the *TMPRSS2-ERG* fused samples from the non-fused samples, most of which harbored mutations in *SPOP, FOXA1,* or *TP53* **(Fig. 3A)**. Rather than separating by sample site, this separation by mutational background was also observed when clustering all metastatic and primary samples (**Supplementary Fig. S3**). We calculated the mean methylation change of the *SPOP* mutant, *TP53* mutant, or *TMPRSS2-ERG* fusion samples compared to normal prostate tissue. While the *TP53* mutants had overall more CpG sites that were significantly different from the normal tissue (either hypo- or hyper-methylated), we observed the previously reported hypermethylator phenotype in the *SPOP* mutant samples (13) (**Supplementary Fig. S4**). We observed that strongly hyper/hypomethylated CpG sites in tumors (compared to normal tissue) were more strongly associated with CpG islands than weakly hyper/hypomethylated CpG sites, suggesting the potential to regulate gene expression (**Fig. 3B**). Strongly hypomethylated CpGs were more frequent than strongly hypermethylated CpGs in all our tumor samples (**Supplementary Fig. S5)**, indicating global hypomethylation, which was also observed in other cancer types (30).

We used limma to call differentially methylated (DM) CpG sites ($|$log2FC$|$ > 1.5 and $q$ < 0.05) and DMRcate to call DM regions (DMR) ($|$mean difference in β value$|$ > 20%, $q$ < 0.05, ≥ 5 CpGs) in all primary prostate cancers against normal prostate tissue, and in metastases against normal prostate tissue. In contrast to the global hypomethylation of CpG sites, we detected 2,102 DMRs hypermethylated in the primaries compared to the normals and 180 hypomethylated DMRs (**Supplementary Table S3**). The metastases showed a greater acquisition of DMRs compared to normals (2,596 hypermethylated and 211 hypomethylated, **Supplementary Table S4**), but only a further 15 hypermethylated DMRs in comparison to

the primary prostate cancer tissue (**Fig. 3C** and **Supplementary Table S5**). Extending this analysis to look specifically at methylation changes occurring in the *SPOP* mutant or *TMPRSS2-ERG* fusion metastases, in comparison to normal prostate tissue, beyond those seen in metastases with neither alteration, reflected the observations from the global methylation analysis. While the *SPOP* mutant metastases showed hypermethylation at 542 regions and hypomethylation at 101 regions, the *TMPRSS2-ERG* fusion metastases had only 44 hypermethylated and 162 hypomethylated regions in comparison to metastases lacking either mutation (**Fig. 3D, Supplementary Tables S6** and **S7**).

Considering the distribution of the DM CpGs in relation to CpG islands, compared to the overall distribution of analyzed CpGs, we found enrichment for hypermethylated CpGs at CpG islands in the comparison of metastases to normal prostate samples (OR = 1.30, 95% CI = 1.26 - 1.33, $q < 2.22X10^{-16}$, Fisher's exact test; **Fig. 3E**). This enrichment of hypermethylated CpG islands was very similar when comparing *SPOP* mutant metastases and normal prostate (OR = 1.32, 95% CI = 1.30 - 1.34, $q < 2.22X10^{-16}$, Fisher's exact test), but was weaker in the *TMPRSS2-ERG* fusion metastases compared to normal prostate (OR = 1.16, 95% CI = 1.13 - 1.19, $q < 2.22X10^{-16}$, Fisher's exact test). However, the enrichment was more pronounced when examining the hypermethylated CpGs detected in *SPOP* mutant-specific metastases (i.e., metastases with *SPOP* mutations vs. metastases without *SPOP* mutations nor *TMPRSS2-ERG* fusions) (OR = 2.98, 95% CI = 2.89 - 3.08, $q < 2.22x10^{-16}$, Fisher's exact test). The same increase in enrichment for hypermethylation at CpG islands was not observed in the *TMPRSS2-ERG* fusion-specific methylation changes (OR = 0.59, 96% CI = 0.54 - 0.65, $q < 2.22X10^{-16}$, Fisher's exact test; **Fig. 3E**). Similarly, when compared to normal prostate, while we found a depletion of hypomethylated CpGs at CpG islands in the metastases (OR = 0.24, 95% CI = 0.22 - 0.26, $q < 2.22x10^{-16}$, Fisher's exact test; **Fig. 3E**), *SPOP* mutant metastases showed a stronger depletion of hypomethylated CpG island CpGs than *TMPRSS2-ERG* fusion metastases (*SPOP* OR =

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0.20, 95% CI = 0.18 - 0.21, $q$ < 2.22x10$^{-16}$, *TMPRSS2-ERG*, OR = 0.32, 95% CI = 0.30 - 0.34, $q$ < 2.22x10$^{-16}$, Fisher's exact tests).

Our results showed that early driver mutations in PCBM are associated with distinct methylation changes. In particular, the enrichment of hypermethylation and depletion of hypomethylation in CpGs at CpG islands in *SPOP* mutant metastases suggest a role in transcriptional regulation.

*Activity of the PRC2 complex drives PCBM-associated DNA methylation changes*
Examining the results of the DMR analysis, we found hypermethylation of the promoters of genes such as *SMARCA2* in the metastases, suggesting possible transcriptional downregulation (**Supplementary Fig. S6**), in keeping with reports of its role in the progression of metastatic castration-resistant prostate cancer (31,32). This hypermethylation was strongest in *SPOP* mutant metastases (**Supplementary Fig. S6**). We, therefore, examined genome-wide methylation occurring at and around CpG islands and in open sea regions (**Fig. 4A**). We found a significant reduction in global methylation at open sea regions in the primary samples, in keeping with the well-described phenomenon of global hypomethylation in cancer ($P$ = 0.038, Wilcoxon test) (33). However, this difference did not persist in the metastatic samples ($P$ = 0.14, Wilcoxon test), suggesting the level of methylation at open sea regions may have slightly increased in the PCBM. Indeed, when examining CpG island-associated CpGs, we observed a significant increase in overall methylation in the primary and PCBM samples compared to the normal prostate tissues ($P$ = 0.018 and $P$ = 0.016, respectively, Wilcoxon tests). However, there was also a marked increase in CpG island methylation in the PCBM compared to the primary samples ($P$ = 0.0049, Wilcoxon test). Interestingly, at CpG island shores and shelves (2Kb and 4Kb from the center of a CpG island, respectively), no significant difference was found between the tumor and normal samples, while PCBM were hypermethylated compared to the primaries (North shelf $P$ = 0.001, North shore $P$ = 0.0022, South shore $P$ = 0.0034, South shelf $P$ =

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0.0015, Wilcoxon tests). These data suggest hypermethylation of CpG islands and surrounding regions is a characteristic of the metastatic samples in our cohort.

As we had identified distinct methylomes associated with the *SPOP* mutant and *TMPRSS2-ERG* fusion metastases, we asked whether this global hypermethylation of CpG islands was more strongly associated with one of these genetic backgrounds. We found that CpG island methylation was significantly higher in the *SPOP* mutant metastases than *TMPRSS2-ERG* fusion PCBM or PCBM with neither an *SPOP* mutation nor the *TMPRSS2-ERG* fusion (*P* = 0.004 and *P* = 0.034, respectively, Wilcoxon tests; **Fig. 4B**). The *TMPRSS2-ERG* fusion PCBM showed no significant difference in CpG island methylation compared to PCBM with neither the fusion nor *SPOP* mutation (*P* = 0.14, Wilcoxon test). Together these findings suggest that CpG island hypermethylation is, in fact, predominantly a feature of the *SPOP* mutant PCBMs.

We asked whether this *SPOP* mutant-specific CpG island hypermethylation was unique to PCBM or whether it was also observed in metastases to other locations (i.e., non-brain metastases) which harbored *SPOP* mutations. Using a previously published dataset (13), we examined the methylation status of the same CpG islands in *TMPRSS2-ERG* fusion or *SPOP* mutant metastases. While this dataset only included five *SPOP* mutant metastases, we found no significant difference in the level of CpG island methylation between these and the *TMPRSS2-ERG* fusion non-brain metastases (*P* = 0.29, Wilcoxon test; **Supplementary Fig. S7**). This may suggest that the *SPOP* mutant-specific CpG island hypermethylation is a specific characteristic of PCBM.

We performed gene set enrichment analysis using the CpG-level DM data to understand which processes might drive the aberrant methylation. Comparing primaries against normal, metastases against normal, and metastases against primaries, highlighted regions of known histone 3 lysine 27 trimethylation (H3K27me[3]), as well as targets of the PRC2 complex,

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which catalyzes the formation of this histone modification complex, and its subunits (EZH1/2, SUZ12, and EED, **Fig. 4C**). While the enrichment for H3K27me$^3$ regions and PRC2 targets (e.g., Benporath_ES_with_H3K27me3 and Benporath_PRC2_targets) was significant for all three comparisons, it was most pronounced in the metastasis vs. normal comparison, and this pattern persists for all examined gene sets related to its subunits examined. Likewise, when examining the methylation changes specifically associated with either the *TMPRSS2-ERG* fusion or *SPOP* mutant backgrounds, in comparison to metastases with neither alteration, the same PRC2-associated pathways were enriched in the *SPOP* mutant metastases and not enriched, or enriched to a far lesser extent, in the *TMPRSS2-ERG* fusion metastases **(Fig. 4D)**. Using targeted RNA-seq data from these samples, we observed a significant increase in the expression of *EZH2,* a component of PRC2, in the metastases compared to primary samples (*P* = 0.0037, Wilcoxon test; **Fig. 4E**). Furthermore, when stratifying the samples by *TMPRSS2-ERG* fusion or *SPOP* mutation, we found that only the *SPOP* mutant metastases showed a slight numerical increase in *EZH2* expression compared to metastases with neither *TMPRSS2-ERG* fusion nor *SPOP* mutation (**Fig. 4F**).

Together these findings demonstrate that CpG island hypermethylation is prevalent in PCBM and even higher in *SPOP* mutant metastases. Our results suggest that aberrant methylation in PCBM is driven by PRC2 activity and may be associated with H3K27me$^3$.

*PCBM DNA methylation changes may suggest mechanisms driving PCBM*
While we have, so far, focused on the different DNA methylation landscapes of PCBM with different mutational drivers, it is clear that these samples share the crucial characteristic of having metastasized to the brain and so may share epigenetic alterations associated with this. Indeed, overlapping the differentially methylated regions called in the *SPOP* mutant or *TMPRSS2-ERG* fusion primary and metastatic samples confirmed a substantial overlap of DMRs (**Fig. 5A**). We, therefore, selected DMRs from the comparison of all metastases

against the normal prostate samples falling within 1500 bp of TSS', identified their corresponding genes and performed a gene ontology analysis using the KEGG pathways, to identify the aberrantly methylated biological pathways which may permit, or drive, metastasis to the brain. This highlighted the neuroactive ligand-receptor interaction, thyroid hormone synthesis, cell adhesion molecules, and pluripotency of stem cells gene sets as the significantly enriched pathways ($q < 0.05$, overrepresentation test, Wallenius' noncentral hypergeometric distribution; **Fig. 5B**). Interestingly, the neuroactive ligand-receptor interaction gene set was identified in a transcriptomic analysis of brain-specific breast cancer metastases (34), while cell adhesion molecules play an important role in cancer migration, immune response, and metabolism (35). Moreover, thyroid hormones (36,37) and the pluripotency of stem cells (38) may play a role in the progression of prostate cancer. We further examined the genes in these sets. While their associated promoter-DMRs were frequently hypermethylated, we found hypomethylation of key genes in these pathways, such as *GABRB3, VIPR1, CLDN8,* and *CLDN4* (**Fig. 5C**). As promoter hypomethylation may be associated with increased transcription, this is in line with studies showing upregulation of these genes in prostate cancer, other primary cancers, and metastases (34,39,40). Notably, the methylation status of the CpGs associated with these genes was similar in normal brain tissue to normal prostate tissue (**Supplementary Fig. S8**), suggesting that these hypomethylation events may be specific to prostate cancer and PCBM. To address this, we examined the CpGs failing in the hypomethylated DMRs at the promoters of *GABRB3*, *CLDN8,* and *CLDN4* in the dataset of non-brain metastases (13) and TCGA cohort of primary prostate cancers (28) (CpGs associated with the promoter of *VIPR1* were not present in the external datasets). In PCBM, the promoters of *GABRB3* and *CLDN4* were significantly hypomethylated compared to the non-brain metastases ($P = 0.015$ and $P = 0.012$ respectively, Wilcoxon test; **Fig. 5D.** While methylation of the promoter of *CLDN8* varied in all samples, there was a noticeable subpopulation of PCBM with hypomethylation of this promoter (**Fig. 5D**). We observed some variation at the CpG level in these comparisons (**Supplementary Fig. S9**). However, most CpGs (present in all three datasets)

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supported the above results, and some showed even more striking differences between PCBM and non-brain metastases (**Supplementary Fig. S9**). Interestingly, there was no difference in methylation status at these loci when comparing primary samples between cohorts ($P < 0.05$).

These data suggest there are consistencies in the DNA methylation patterns of PCBM. These are related to the methylation status of genes in the neuroactive ligand-receptor and cell adhesion molecules pathways implicated in other cancers. Our findings suggest that methylation changes affecting these genes may also be specific to brain metastasis in prostate cancer.

**Discussion**

Here, we present the DNA methylation landscapes of a cohort of 42 patients with brain metastases from prostate cancer, complementing our recent study on the genetic landscapes of these patients (17). Our findings highlight the inter-patient and intra-patient epigenetic heterogeneity present in PCBM, demonstrate that DNA methylation landscapes associated with early driver genetic events persist in PCBM, and provide potential explanations for the mechanisms underlying PCBM-associated DNA methylation patterns.

Epigenetic subtypes of advanced prostate cancers have been identified (13,41), and epigenetic stability is observed between metastases at different sites within a patient (15). With this, we have shown that epigenetic patterns linked to genetic driver alterations persist through the process of metastasis to the brain, supporting the idea that epigenetic stability may be exploited in the effective, systematic treatment of mCRPC, even when involving rare metastatic sites such as the brain. Further, the multi-region sampling employed in this study allowed us to examine intra-metastatic epigenetic heterogeneity at fine resolution. Interestingly, the methylation landscape of intra-patient brain metastases was relatively

homogeneous compared to intra-patient primary tumors, potentially suggesting clonal selection in the metastatic process.

We identified salient characteristics of PCBM, particularly global hypermethylation affecting CpG islands, which was most prominent in PCBM with *SPOP* mutations, and appears to be linked to the activity of the PRC2 complex. PRC2-associated aberrant DNA methylation are well-studied characteristic of advanced prostate cancer. They have been linked to prostate cancer progression using single-cell transcriptomics (42) and are particularly prevalent in *TMPRSS2-ERG* fusion-negative prostate cancer (12). However, the relevance of PRC2-associated aberrant DNA methylation events in metastatic prostate cancers is less well studied. Although the number of normal prostate samples used in this study was limited, our findings related to PRC2 activity in PCBM suggest that findings from studies on primary prostate cancers hold true in this specific metastatic setting and again reinforce the importance of early driver mutations in shaping the DNA-methylation landscape of metastatic prostate cancer.

PRC2-associated aberrant DNA methylation was a characteristic of this dataset and was associated with the two different genetic backgrounds on which we focussed. However, examining the DNA methylation changes consistent across the metastatic samples provided some interesting insight into how DNA methylation changes, and hypomethylation, in particular, might promote brain metastases, as these changes appeared to be specific to PCBM when compared to non-brain metastases. While transcriptomic data was lacking from this study, overexpression of neuroactive ligand-receptors (*GABRB3*) has previously been linked to brain metastases from other cancers (34), possibly suggesting that metastasizing cells from disparate primary tumors may require similar reprogramming to establish brain metastases.

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The study represents an initial characterization of the DNA methylation features of PCBM. Future work will focus on elucidating whether CpG island hypermethylation, driven by the activity of PRC2 or hypomethylation of genes such as *GABRB3, CLDN4,* and *CLDN8,* confer brain-specific metastatic potential to prostate cancer cells (39,40,43). Integration of these data with transcriptomics from PCBM will improve understanding of how epigenetically driven transcriptional changes may drive PCBM and provide insights into therapeutic strategies for the systematic treatment of mCPRC, encompassing less common metastatic sites such as the brain.

## Acknowledgments

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**Figure legends**

**Fig. 1. The methylome of brain metastases from prostate cancer is largely inherited from primary tumors and is driven by genomic background.**

**A**) Principal component analysis using the 1% most variably methylated CpG sites from the Illumina EPIC array (8,038 sites). **B**) Spearman's correlation between eigenvectors of PCs 1-10, with sample type (i.e., primary tumor vs. metastasis), patient, and mutational status (\*=$P$ < 0.05, \*\*=$P$ < 0.01, \*\*\*=$P$ < 0.001). **C**) Euclidean distance between primary samples and normal prostate samples and between metastatic and normal samples. **D**) Euclidean distance between primary samples within each patient, metastatic samples within each patient, and between primary and metastatic samples within each patient. **E**) As in D, but comparison done between patients. PCBM primary n=57, PCBM metastasis n=95, normal prostate n=2.

**Fig. 2. Intrapatient variation between selected primary and metastatic samples.**

The samples from each patient were clustered using the 1% most variably methylated CpG sites from the Illumina EPIC array (8,038 sites). Heatmap shows β values for primary and metastatic samples. Class of CpG (in relation to CpG island) is shown in blue/green heatmap on the left. Average $\log_2$ fold change of the primary tumors and metastases compared to normal prostate tissue are shown on the right. Sample type (primary or metastatic), histology, and genetic alterations from whole exome sequencing are annotated below.

**Fig. 3. *SPOP* mutant and *TMPRSS2-ERG* fusion PCBM have distinct methylomes.**

**A**) Unsupervised hierarchical consensus clustering of metastatic samples from 42 patients. Samples from each patient were clustered using 1% most variably methylated CpG sites from the Illumina EPIC array (8,038 sites). The heatmap shows the β values. Class of CpG (in relation to CpG island) is shown in blue/green heatmap on the left, along with methylation

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status of CpG sites in normal prostate tissue. Mutational burden (mutations/Mb) is shown in the barplot on top. Sample type (primary or metastatic), histology, and genetic alterations from whole exome sequencing are annotated below. **B**) Enrichment in CpG types among the variably methylated CpGs, showing CpGs with $\log_2$ fold difference to mean β value of normal prostates below -2 and above 2 (left column), and below -10 and above 3 (right column). Values and colors indicate Pearson correlation coefficients. *P*-values were $< 2.22\times10^{-16}$ in all cases. **C**) Number of DMRs in primary cancers compared to normal prostate and metastases compared to normal prostate (DMR ≥ 5 CpGs, >|20%| change in methylation, *q* < 0.05). **D**) Number of DMRs in metastases with *SPOP* mutation vs. samples with neither *SPOP* mutation nor *TMPRSS2-ERG* fusion, and metastases with *TMPRSS2-ERG* fusion vs. samples with neither *SPOP* mutation nor *TMPRSS2-ERG* fusion (DMR ≥ 5 CpGs, >|20%| change in methylation). **E**) Enrichment for differentially methylated CpGs between all metastases and normal prostate tissue, *SPOP* mutant metastases and normal prostate tissue, *TMPRSS2-ERG* fusion metastases, and normal prostate tissue, *SPOP* mutant metastases compared to non-*TMPRSS2-ERG/SPOP* mutant metastases, and *TMPRSS2-ERG* fusion metastases compared to non-*TMPRSS2-ERG/SPOP* mutant metastases. Values indicate the log-odds ratio from Fisher's exact tests.
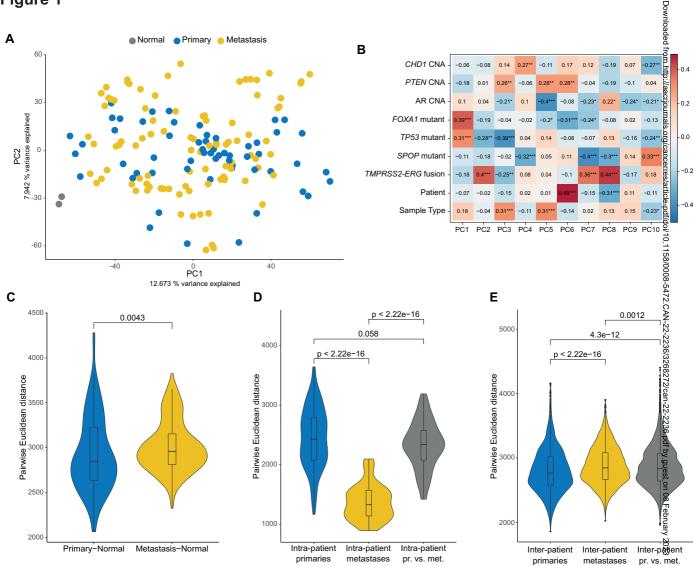
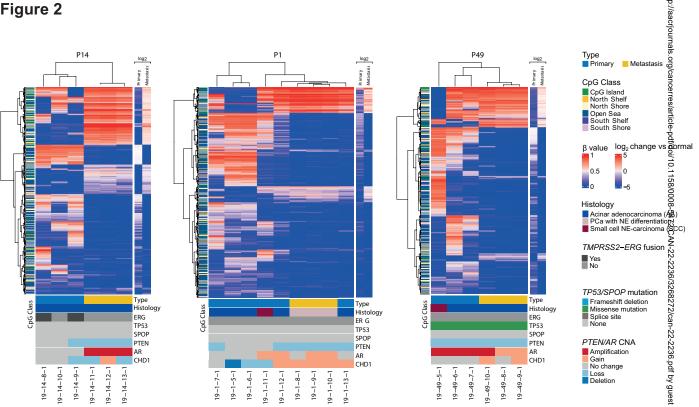**Fig. 4. Global CpG island hypermethylation in PCBM is associated with the PRC2 complex.**

**A**) Mean methylation of CpG sites in relation to CpG islands in normal prostate, primary, and metastatic samples. *P* values computed from Wilcoxon tests. **B**) Mean methylation of CpG sites in relation to CpG islands in normal prostate, primary, and metastatic samples, stratified by *TMPRSS2-ERG* fusion, *SPOP* mutation, or neither. *P* values computed from Wilcoxon tests. **C**) Gene set enrichment analysis on DM CpG sites between primary tumors and normal prostates, metastases and normal prostates, and metastases and primary tumors, using curated gene sets from MSigDB. Pathways shown are from the union of the top 10 DM pathways from each of the three comparisons. **D**) Same as **C** but showing the
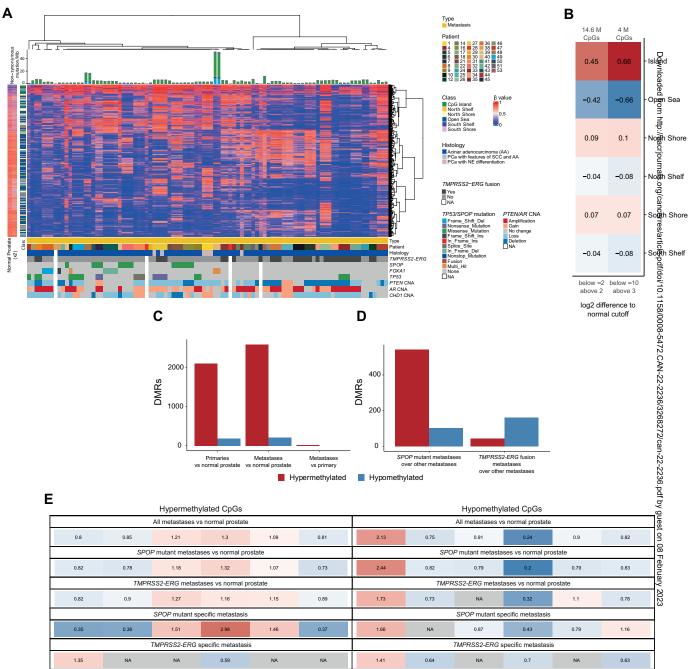
union of the 10 most DM pathways between *SPOP* mutant or *TMPRSS2-ERG* fusion metastases against metastases with neither alteration. **E**) Expression of *EZH2* in primary tumors and metastases from targeted-RNA sequencing. *P* values computed from Wilcoxon tests. **F**) Same as **E**, with samples stratified by *TMPRSS2-ERG* fusion, *SPOP* mutation, or neither.

**Fig. 5. PCBM DNA methylation changes may suggest mechanisms driving PCBM.**

**A**) Overlap between DMRs in *SPOP* mutant primaries, *SPOP* mutant metastases, *TMPRSS2-ERG* fusion primaries, and *TMPRSS2-ERG* fusion metastases, compared to normal prostates. Dots and lines show sets being intersected, bar plots on top show the intersection size (Mb), and bar plots on the right show the size of set (Mb). **B**) Gene ontology analysis on genes with promoter-associated DMRs in PCBM compared to normal prostate tissue. **C**) Heatmap showing mean difference in methylation level at DM promoters of genes in the neuroactive ligand-receptor interaction and cell adhesion molecules gene sets for primary tumors and metastases, compared to normal prostates. White indicates the absence of a DMR at a given promoter. **D**) Mean level of methylation at promoter DMRs of *GABRB3*, *CLDN4,* and *CLDN8* in primary samples from the PCBM cohort, TCGA primary prostate cancers, metastatic samples from the PCBM cohort, and non-brain metastases from Zhao *et al.* 2021. *P* values computed from Wilcoxon tests. PCBM primary n=57, TCGA primary n=502, PCBM metastasis n=95, non-brain metastasis n=100.

# Figure 1

# Figure 2

# Figure 3

# Figure 4

# Figure 5