

Interplay between Genome and Epigenome of Prostate Cancer

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In this study, we synthesized the first

Chinese Prostate Cancer Genome and Epigenome Atlas (CPGEA).

Data items

Figure 1A: Genomic and epigenomic abnormality across all tumor specimen (multi-panel bar graph).

Figure 1B: Mutation signature.

Figure 2: The Big Big Circle.

Figure 3: Non-coding SNV and recurrent ncSNV.

Figure 4: FOXA1

Figure 5: DNA methylation

Figure 6: Subtyping

Figure 7: Pathway

Table 1: Clinical and pathological characterization of specimen

Supplementary Figure 1: Diagram of the study design/scope

Supplementary Figure 2A: GISTIC results

Supplementary Figure 2B: Examples of amplification/deletion (METTL24 or a new one)

Supplementary Figure 3ABC: Example of fusion

Supplementary Figure 4: SMG waterfall plot

Supplementary Figure 5: FOXA1 validation

Supplementary Figure 6: Non-CG methylation differences

Supplementary Figure 7: DMR stats

Supplementary Figure 8: DMR enrichment

Supplementary Figure 9: epi-mutation and IDH/TET genes

Supplementary Table 1: Molecular and pathological estimates of specimen

Supplementary Table 2: Datasets generated in this study

Supplementary Table 3: Statistics underlying Figure 1

Supplementary Table 4: GISTIC result, arm level

Supplementary Table 5: GISTIC, focal (with gene names)

Supplementary Table 6: Fusion events

Supplementary Table 7: RT-PCR primers for fusion detection

Supplementary Table 8: Description of 13 cohorts

Supplementary Table 9: Statistics underlying Figure 2

Supplementary Table 10: Genomic distribution of non-coding SNV

Supplementary Table 11: Potential functional ncSNV

Supplementary Table 12: FOXA1 validation

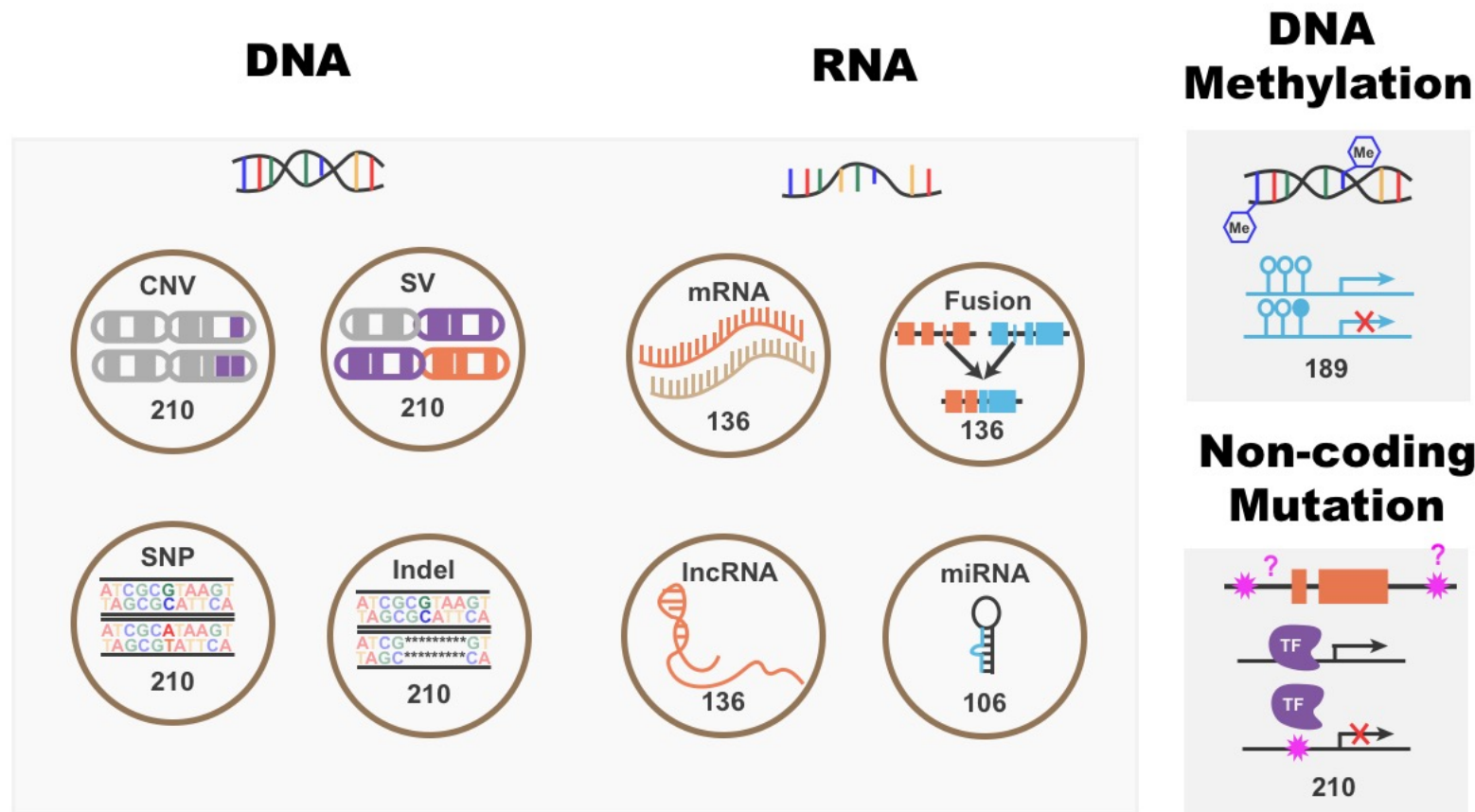
First Goal : direct comparison between Chinese and Western

We integrated existing genomic datasets from 2,775 prostate tumors representing 13 western cohorts

Clinical samples and histopathologic data

Table 1. Clinical characteristics comparison		
	CPGEA	TCGA ⁵
Clinical Feature		
Age	69 (50-88)	61 (43-76)
Pre-operative PSA	18.72 (1.25-1161.92)	7.4 (1.6-87.0)
Gleason Score		
3+3	17	65
3+4	56	102
4+3	41	78
≥8	95	88
Pathologic Stage		
pT2a/b	28	18
pT2c	78	111
pT3a	61	110
pT3b	33	82
pT4	10	6
PSA Recurrence		
Yes	61	33
No	128	248
Not available	21	47
Lymph Nodes Metastasis		
Positive	25	
Negative	154	
Not available	31	
Bone Metastasis		
Positive	15	
Negative	195	
Not available	/	
Margin Status		
Positive	83	69
Negative	125	193
Not available	2	71

We applied four omics technologies to characterize isolated biomolecules from the Chinese prostate cancer cohort: whole genome sequencing (WGS, 112.03× on average per sample), whole genome bisulfite sequencing (WGBS, 29.5× average coverage per CpG per sample), RNA sequencing (138.51M mapped paired-end reads on average) and miRNA sequencing (19.75M mapped reads on average). Overall, our first Chinese Prostate Cancer Genome and Epigenome Atlas was consisted of 420 WGS, 272 RNA-seq, 378 WGBS, and 212 miRNA-seq datasets (Supplementary Figure 1, Supplementary Table 2).



Landscape

CNV

- CNV arm level
- CNV gene level
- CNV comparison

Fusion

- Fusion
- Fusion comparison

SNV

- SMG
- SMG comparison

Cluster

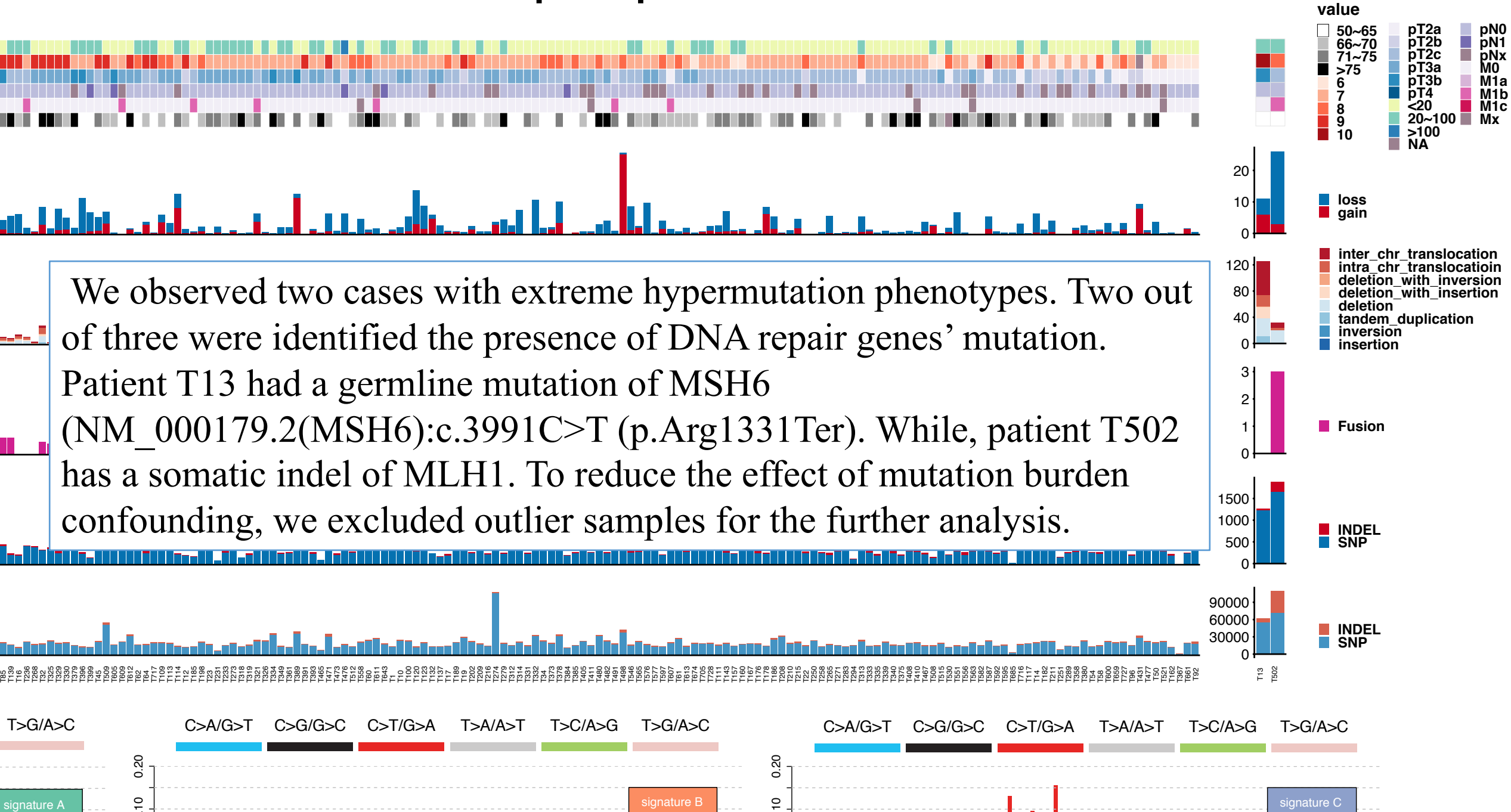
Non-coding

DNA methylation

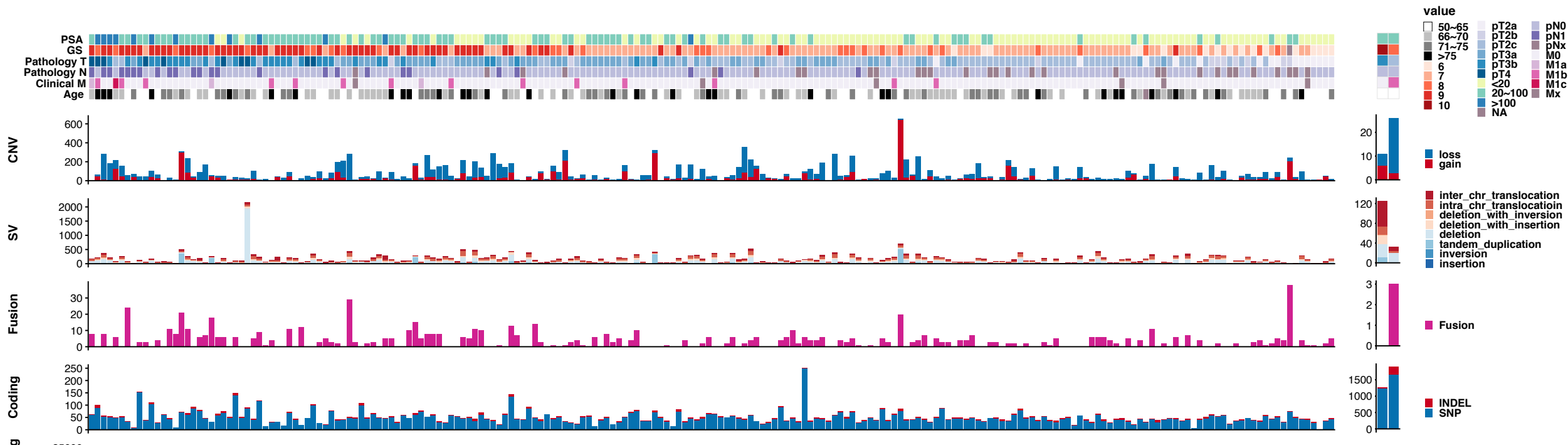
Somatic mutation landscape of prostate cancer of Chinese and Western men



Somatic mutation landscape of prostate cancer of Chinese and Western men



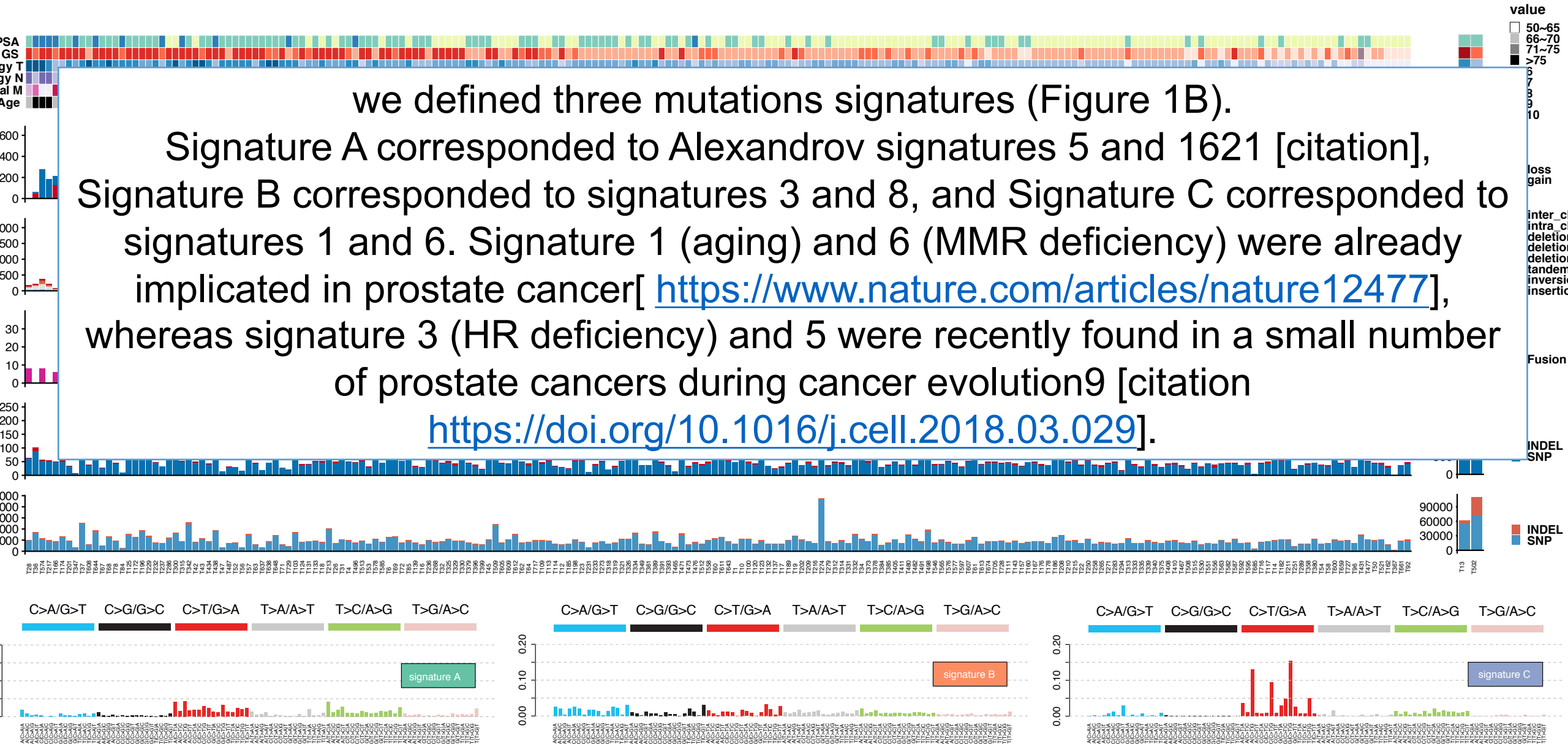
Somatic mutation landscape of prostate cancer of Chinese and Western men



The average genome-wide substitution rate was **1.5197 per Mb** (range 0.1728-7.4079 per Mb). The mutational burden across our cohort was **1.10750 per Mb** (range 0.02778-4.58333 per Mb), which corresponded to **39.87 non-synonymous mutations per tumor genome (median:36; 25th:29 and 75th:46 percentiles respectively)**.

This estimate was remarkably consistent with mutation burden estimated from exome sequencing from western cohort¹⁹

Somatic mutation landscape of prostate cancer of Chinese and Western men



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- Fusion comparison

SNV

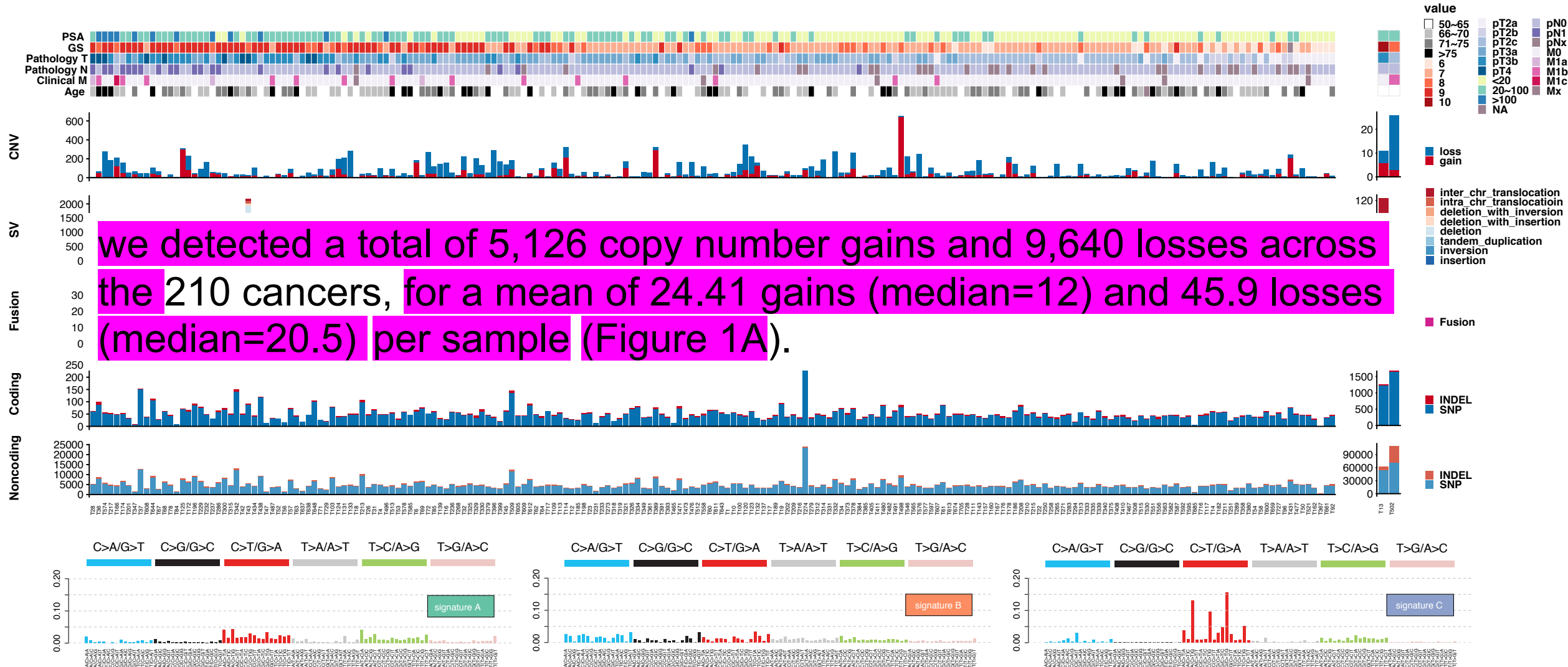
- SMG
- SMG comparison

Cluster

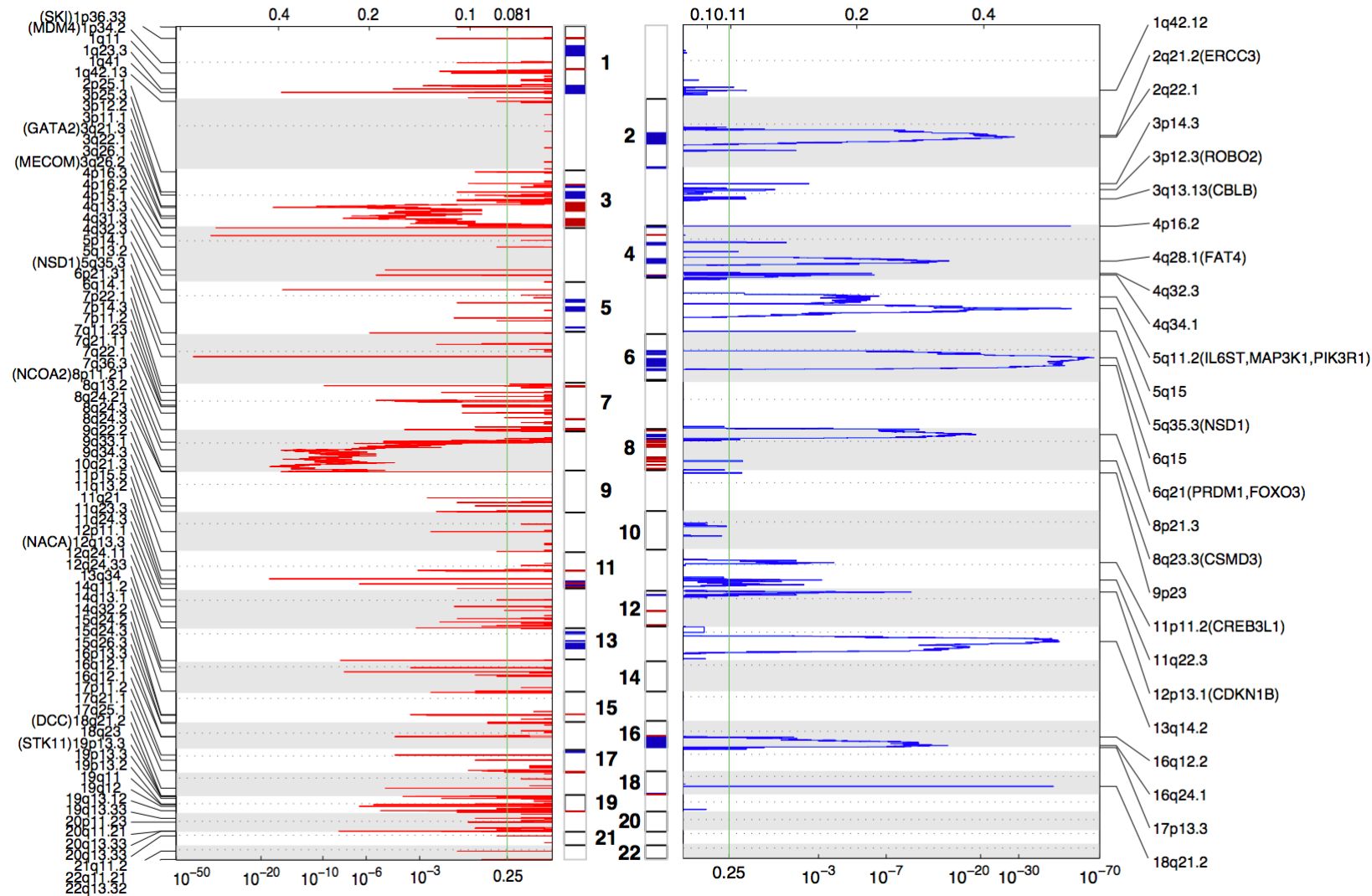
Non-coding

DNA methylation

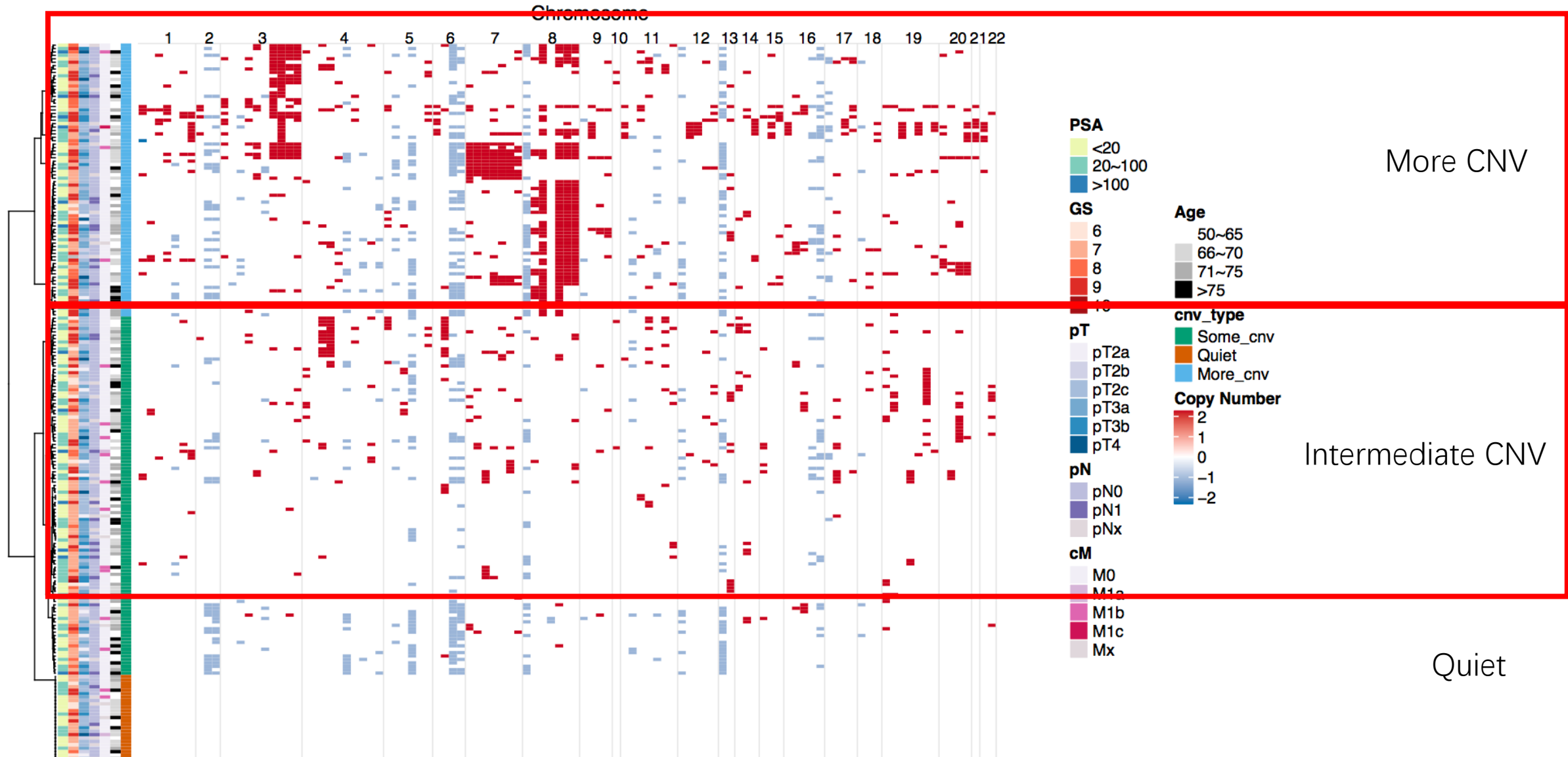
Somatic mutation landscape of prostate cancer of Chinese and Western men



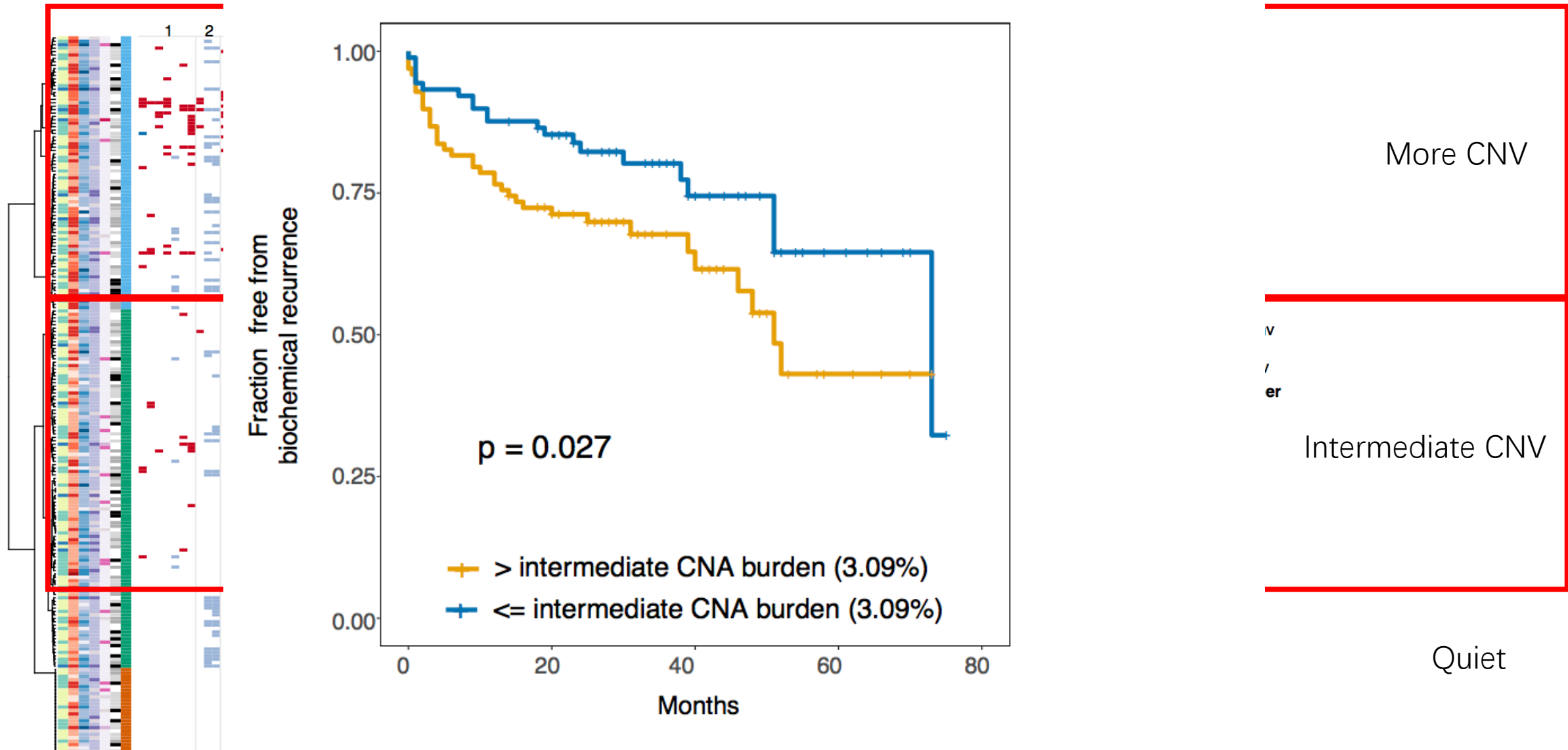
The most significantly recurrent ($q < 10^{-6}$) arm-level gains occurred on chromosome arms 8q (16%), 3q (8%), 3p (7%), 7q (7%), 8p (7%) and 7p (6%). The most significant losses were of 8p (16%) and 16q (6%)



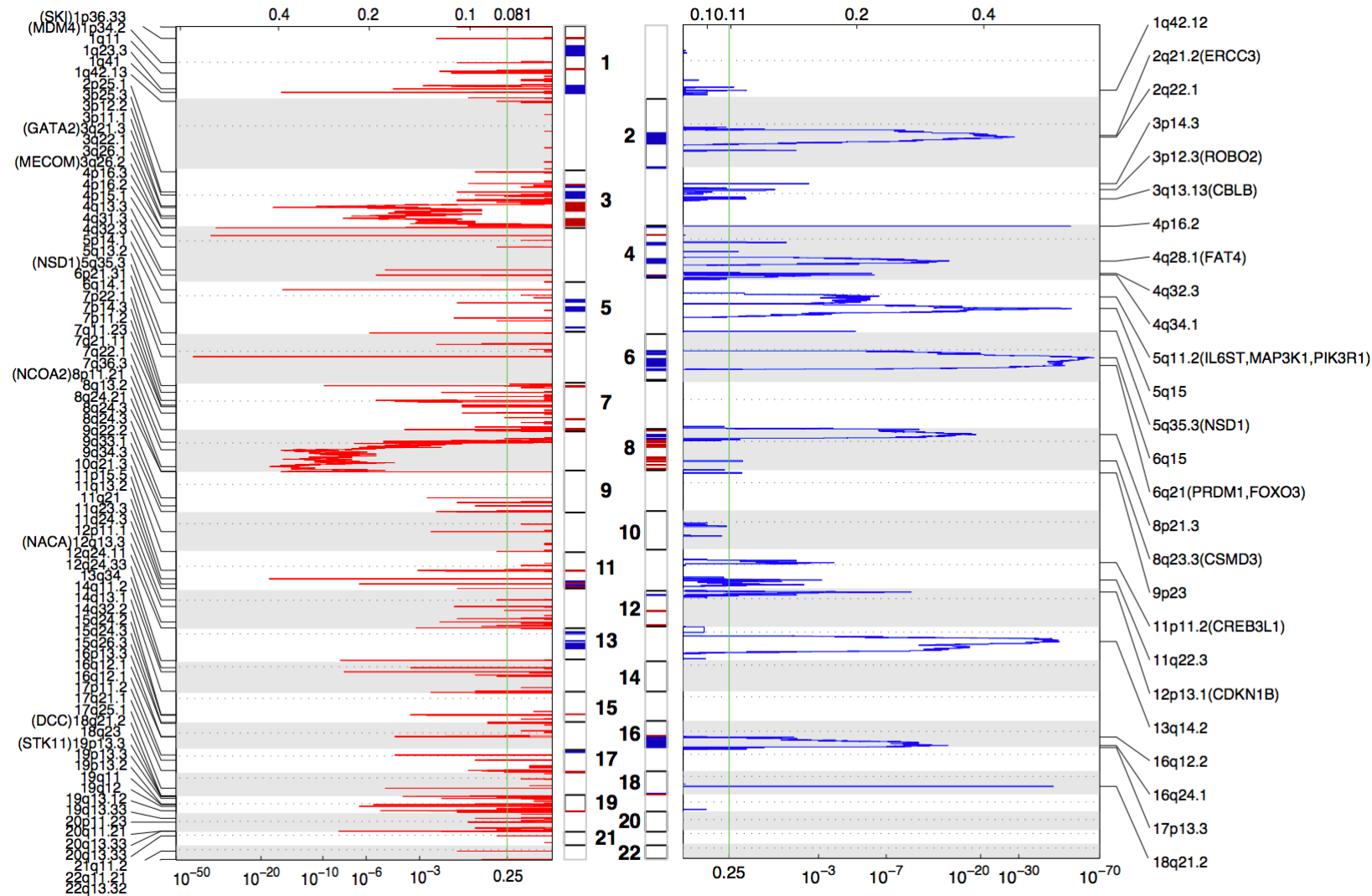
We did the clustering based on chromosomal arm level alterations as the previous TCGA paper⁵. Consistent with TCGA result, we could clearly see the trend of “more SCNA”, “Some SCNA” and “Quiet” (Supplementary Figure 2C).

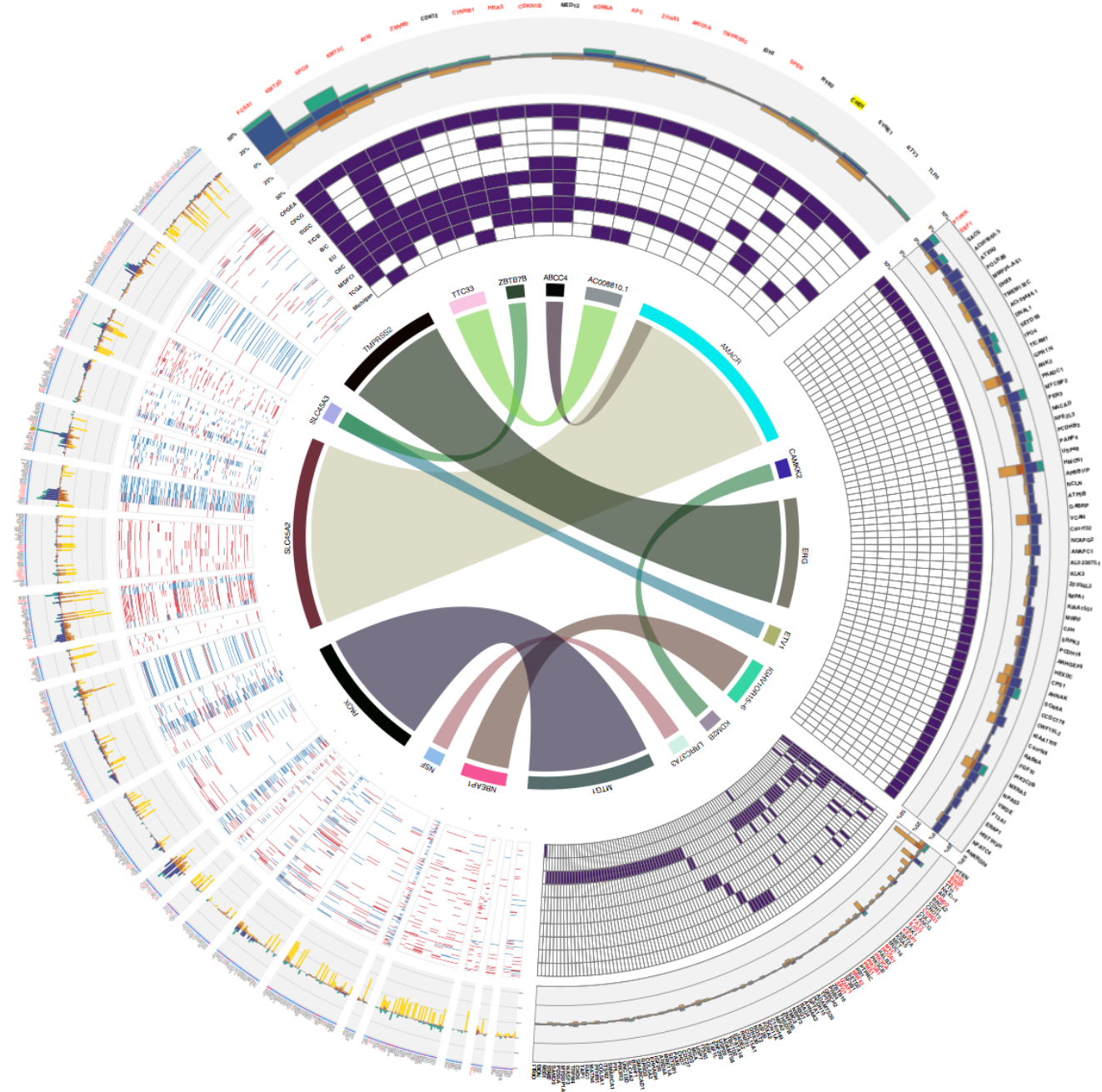


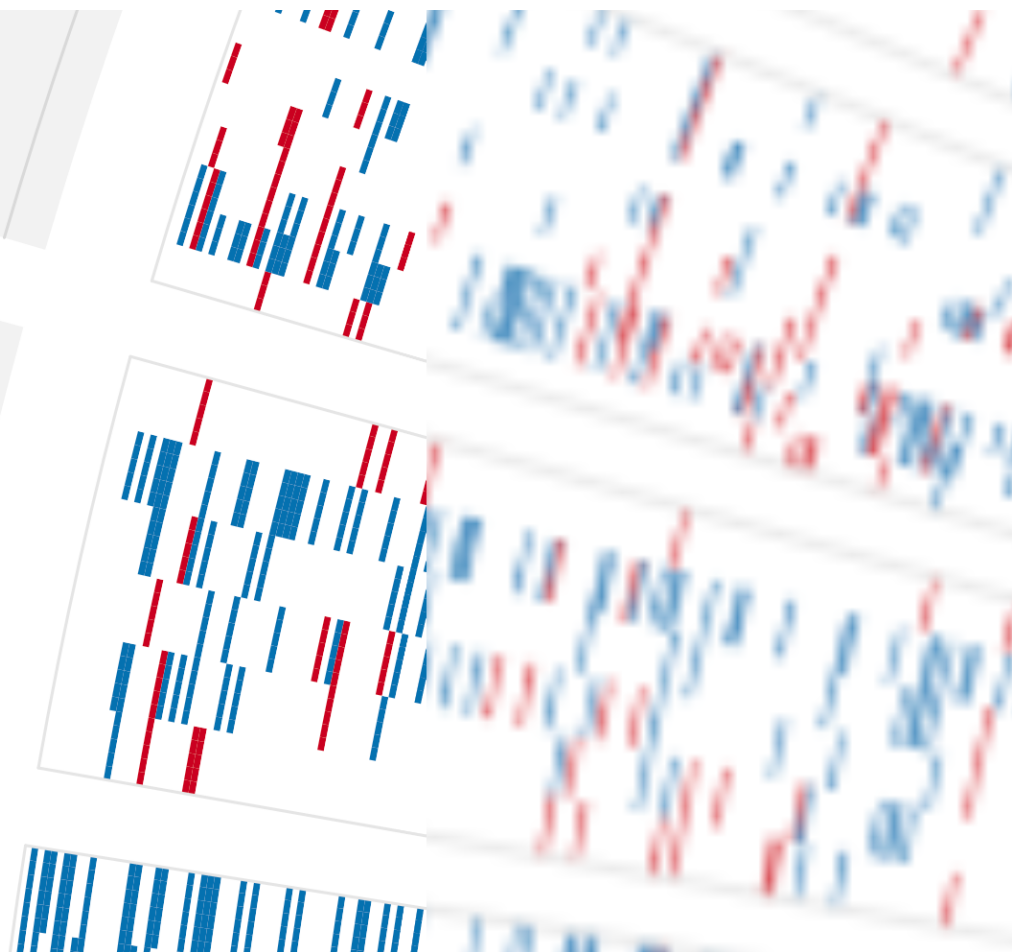
We did the clustering based on chromosomal arm level alterations as the previous TCGA paper⁵. Consistent with TCGA result, we could clearly see the trend of “more SCNA”, “Some SCNA” and “Quiet” (Supplementary Figure 2C).



These results were consistent with previous studies²⁵ [Recurrent copy number alterations in prostate cancer: an in silico meta-analysis of publicly available genomic data, citation]. We identified 79 gains and 26 losses of significant focal copy-number alteration (Supplementary Figure 2A, Supplementary Table 5).







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SNV

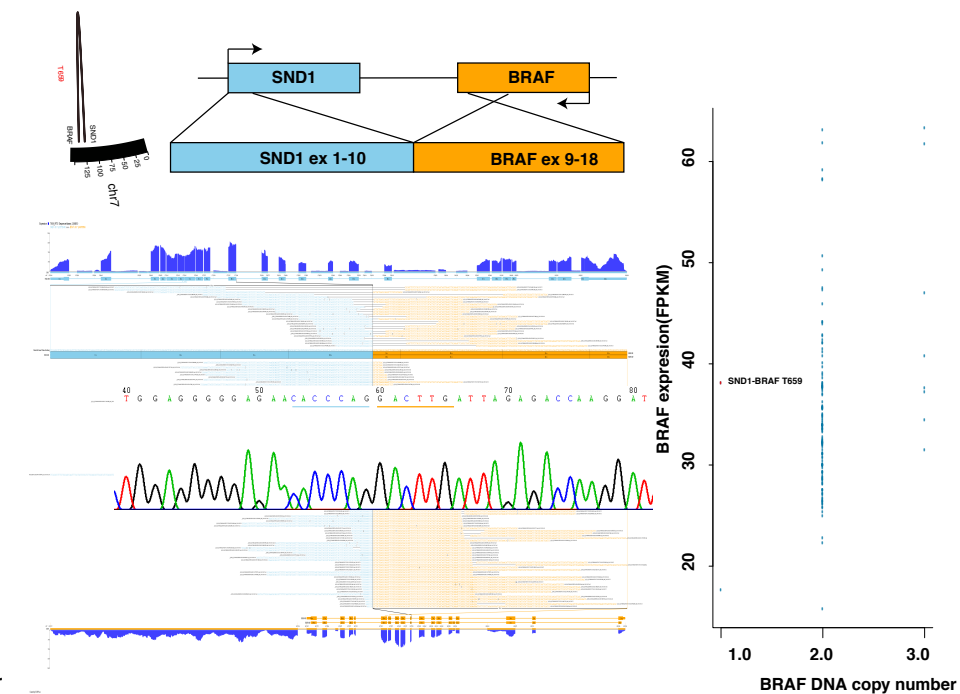
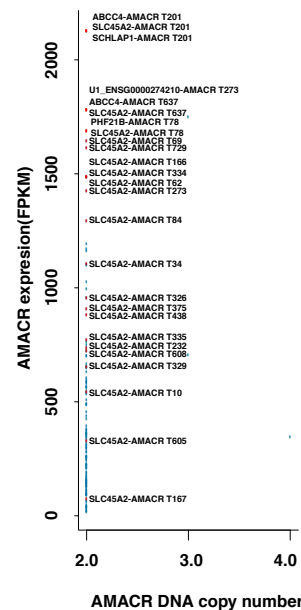
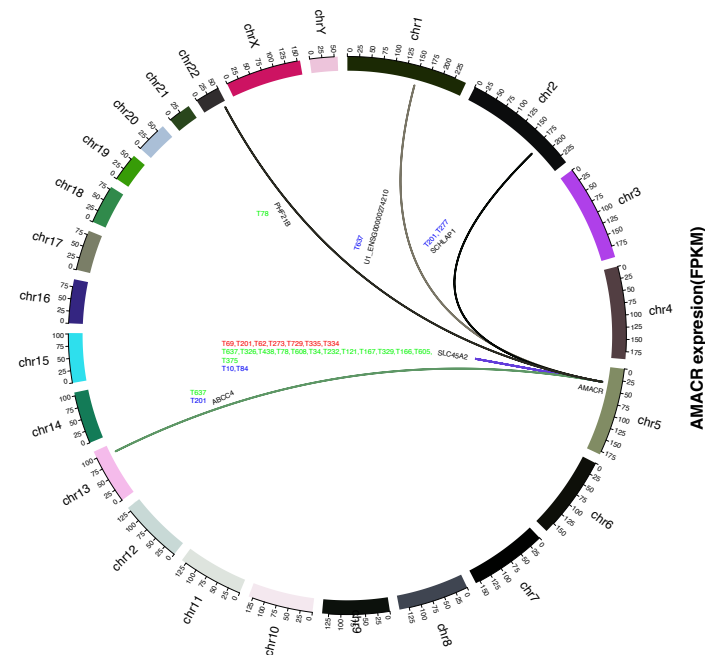
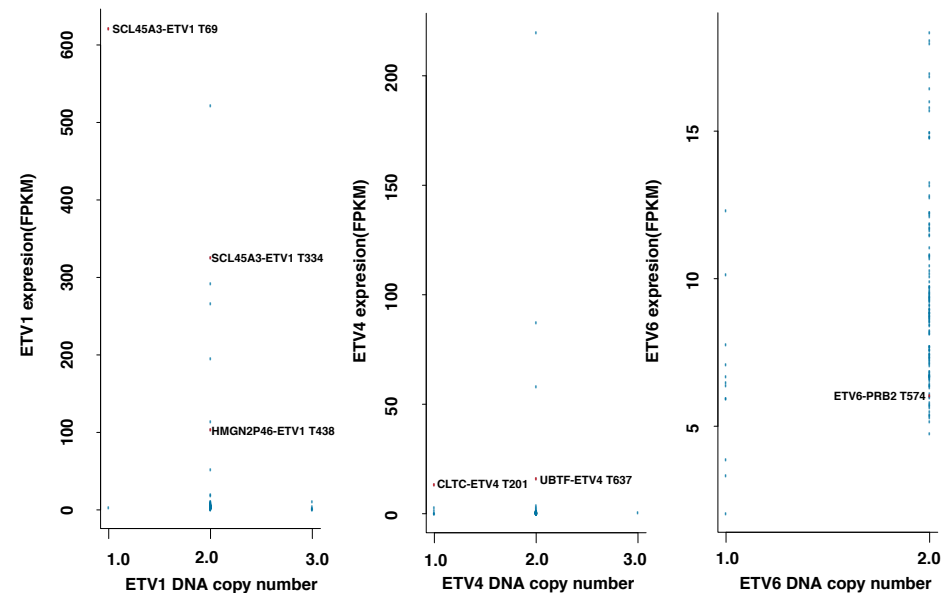
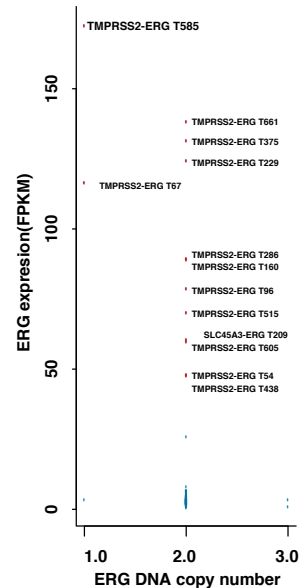
- SMG
- SMG comparison

Cluster

Non-coding

DNA methylation

We used SOAPfuse²⁶ [SOAPfuse: an algorithm for identifying fusion transcripts from paired-end RNA-Seq data, 10.1186/gb-2013-14-2-r12] and PRADA²⁷ [PRADA: pipeline for RNA sequencing data analysis, 10.1093/bioinformatics/btu169] for fusion detection with a stringent filtering workflow (Methods). We were able to detect 182 gene-to-gene fusions (Supplementary Table 6), 46 of which were previously reported in different cancers, and 136 were novel to our study (Supplementary Table 6). 148 pairs were on the same chromosome, whereas 34 pairs were trans-chromosomal. Because fusion detection usually had high false positive rate, we experimentally validated 20 gene pairs using a total of 129 sets of RT-PCR probes (Methods, Supplementary Table 7). 91 sets were scored positive, validating 19 out of the 21 fusion candidates. In stark contrast to the Western prostate cancers, the most well-known hallmark TMPRSS2-ETS gene fusion was much lower in the Chinese cohort (12/136, 8.82%). The most prominent gene fusion event in our Chinese cohort was the AMACR rearrangement, occurring in 19.12% (26/136) of the cases (Supplementary Figure 3). SLC45A2-AMACR was reported in progressive prostate cancers (15.44%, 21/136)



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DNA methylation

To detect significantly mutated genes (SMGs), we applied MuSiC²⁸ [MuSiC: identifying mutational significance in cancer genomes 10.1101/gr.134635.111] and MutSigCV²⁹ [Mutational heterogeneity in cancer and the search for new cancer-associated genes 10.1038/nature12213], and found 78 SMGs including well-known cancer genes such as SPOP, FOXA1, KDM6A, and ZMYM3 (Supplementary Figure 4A). The striking difference between TMRPSS2-EGR fusion events motivated us to comprehensively compare mutation signatures between our Chinese cohort and those established in previous studies. We collected a total of 13 cohorts, including a pilot cohort of Chinese samples published by our own group¹⁸ [citation], and the remaining was mostly consisted of Caucasians⁵⁻¹⁶ [citations] (Supplementary Table 8). We integrated 9 studies with large western cohorts to compile a list of 180 SMGs. We also used 699 consensus cancer genes from and Cancer Gene Census (CGC)³⁰ [A census of human cancer genes 10.1038/nrc1299 citation] as an additional confidence filter. Altogether, we were able to define 11 core prostate cancer genes, 9 important Chinese prostate cancer genes, and 31 important Western prostate cancer genes (Supplementary Table 9). Because the SMGs we compiled from previous studies included metastatic tumors, we carefully curated their metadata and were able to separate primary and metastatic samples. Thus, we were able to calculate mutation frequency of these prostate cancer genes in the Chinese cohort and the Western cohorts. Altogether we determined that 135 genes were significantly differentially mutated in primary prostate cancers in Chinese cohort than in Western cohorts (Figure 2). We also reported differences in genes affected by focal copy number changes and gene fusion between Chinese and Western cohorts (Figure 2).

MUSIC

78

MutSigCV

7

CP

FOXA1
SPOP
KDM6A
ZMYM3
DNAL1
C4orf32
MRRF

EU (blue)

CGC (yellow)

CPGEA (orange)

TLR6 (pointing to EU only region)

47 (EU only)

638 (CGC only)

60 (CPGEA only)

6 (EU & CGC)

31 (CGC & CPGEA)

1 (EU & CPGEA)

11 (EU & CGC & CPGEA)

2 (EU & CGC)

7 (EU & CPGEA)

3 (CGC & CPGEA)

4 (EU & CGC & CPGEA)

2 (CGC & CPGEA)

1 (EU & CPGEA)

TP53
BRAF
AKT1
PTEN (pointing to CGC only region)

TTN
NKX3-1 (pointing to CPGEA only region)

RUNX1T1
PREX2
CSF1R
NCOR2
CAMTA1
MUC4 (pointing to EU only region)

PTPRK
EBF1 (pointing to EU & CGC overlap)

KDM6A
APC
ZFHX3
ARID1A
TMPPRSS2
IDH1
SPEN (pointing to EU & CPGEA overlap)

FOXA1
KMT2D
SPOP
ATM
KMT2C
CDK12
ZMYM3
CTNNB1
HRAS
CDKN1B
MED12 (pointing to EU & CGC & CPGEA overlap)

CHD1
SYNE1
ETV3 (pointing to CGC & CPGEA overlap)

RYR2 (pointing to CGC & CPGEA overlap)

TLR6

EU

CGC

AR	IL6ST	PALB2
ARID2	JAK1	PIK3CA
BRCA2	KEAP1	PIK3CB
CDH1	KMT2A	PIK3R1
CNOT3	KRAS	PMS1
CSMD3	MET	PTPRC
CUL3	MUC16	RB1
FANCG	MYC	RNF43
FAT3	NCOR1	SETD2
		SF3B1
		U2AF1
		XPO1
		ZBTB16

Western Cohorts

C

Western Cohorts

MSKCC
Cancer Cell 2010

Primary 67.29%
181

Metastatic 13.75%
Cell Line 2.97%
Xenograft 2.97%
Unknown 13.01%

MSKCCC
Cell 2014

Metastatic 100%
12

CRC
Nat Med 2016

Metastatic 87.5%
154

MSKCCC
PNAS 2014

Primary 100%
104

TCGA
Cell 2015

Primary 100%
333

SU2C
Cell 2015

Metastatic 100%
150

Broad/Cornell
Nature Genet 2012

Primary 94.64%
106

Unknown 5.36%

CPC-Gene
Nature 2017

Primary 51.57%
246

Unknown 48.43%

MSKCC
JCO 2017

Metastatic 43.45%
Primary 54.56%

Broad/Cornell
Cell 2013

Primary 96.49%
55

Metastatic 3.51%

Neuroendocrine 1.98%

MSKCC/DFCI
Nat Genet 2018

Metastatic 32.87%
Primary 67.13%
680

T/C/B
Cell 2016

Metastatic 61.4%
70

Neuroendocrine 38.6%

Legend:

- Primary
- Metastatic
- Cell Line
- Xenograft
- Unknown

MSKCC
JCO 2017

Broad/Cornell
Cell 2013

Cell Type	Count	Percentage
Primary	55	66.43%
Metastatic	3	3.66%
Unknown	22	26.91%

CPC-Gene
Nature 2017

Category	Percentage	Count
Primary	51.57%	246
Unknown	48.43%	

CRC
Nat Med 2016

MSKCCC
Cell 2014
Metastatic
100% 

MSKCC
Cancer Cell 2010

Cancer Type	Percentage	Count
Primary	67.29%	181
Metastatic	13.75%	
Cell Line	2.97%	
Xenograft	2.97%	
Unknown	13.01%	

Broad/Cornell
Nature Genet 2012

Response	Percentage	Count
At least one	94.64%	106
Unknown	5.36%	

Histological Type	Percentage	Count (n)
Metastatic	61.4%	70
Neuroendocrine	38.6%	44

Tumor Type	Percentage
Metastatic	61.4%
Neuroendocrine	38.6%

TCGA
Cell 2015

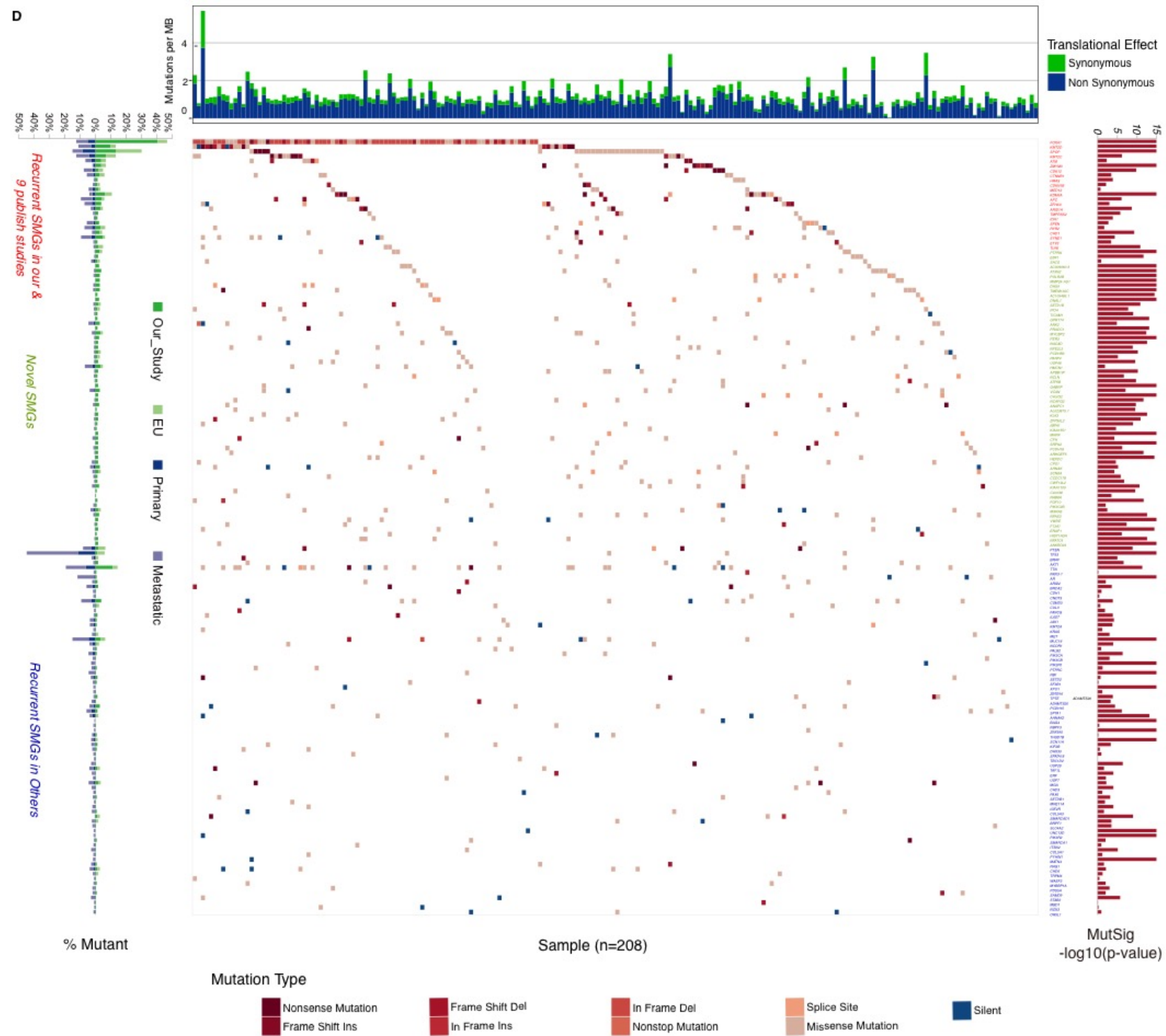
MSKCC
PNAS 2014

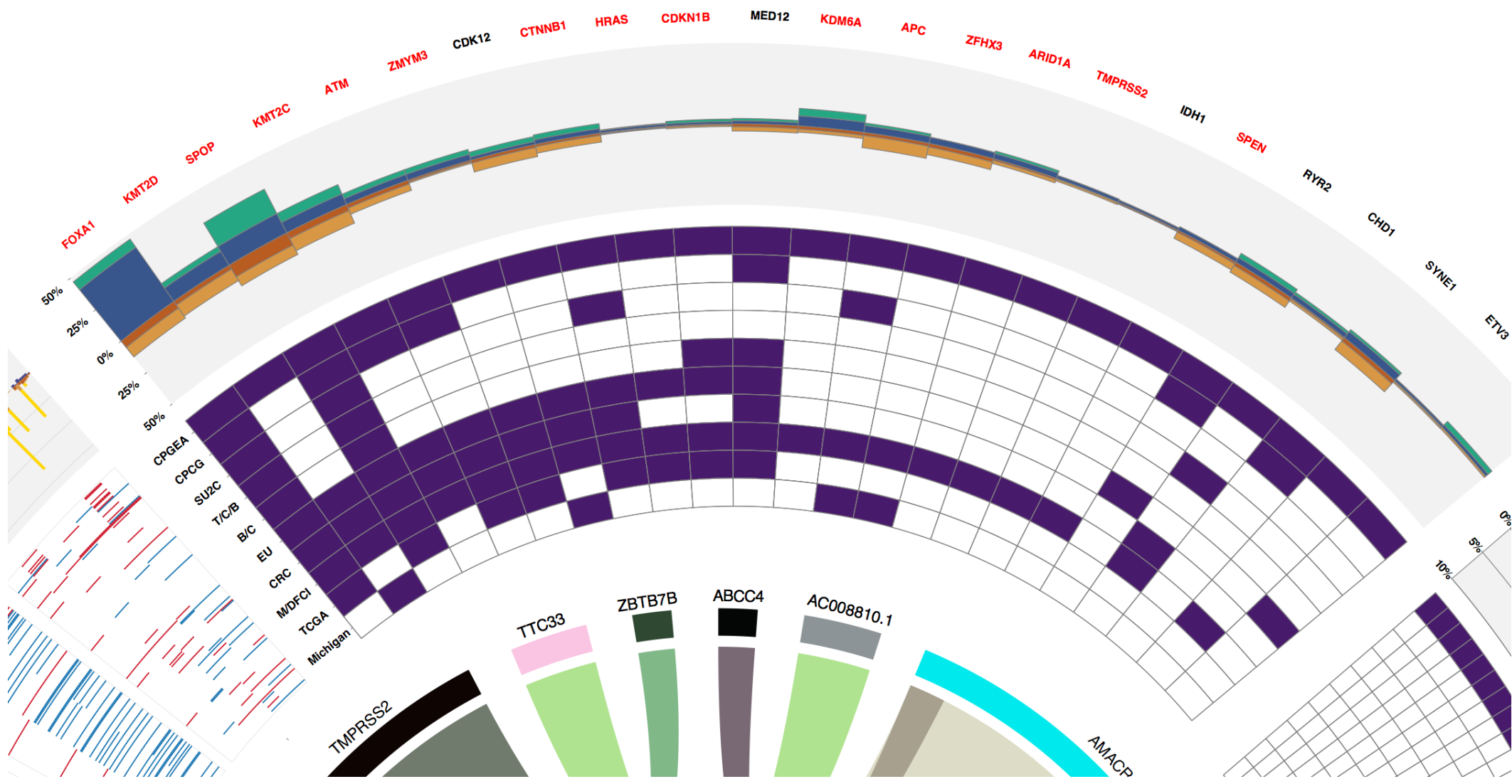


Primary
100%
104

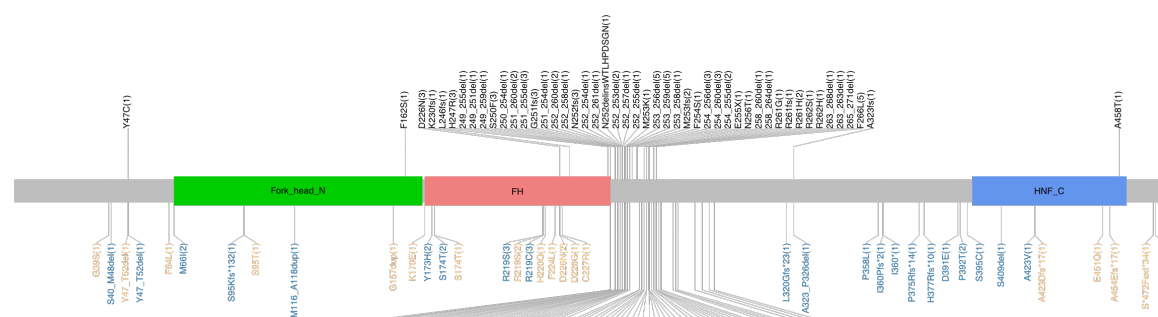
SU2C
Cell 2015

150
Metastatic
100%

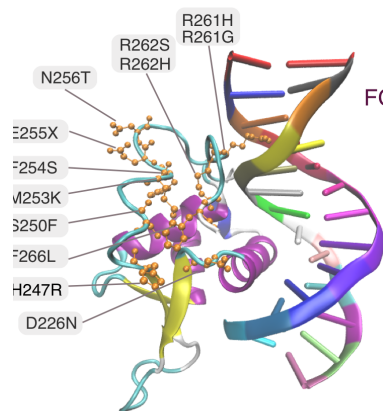
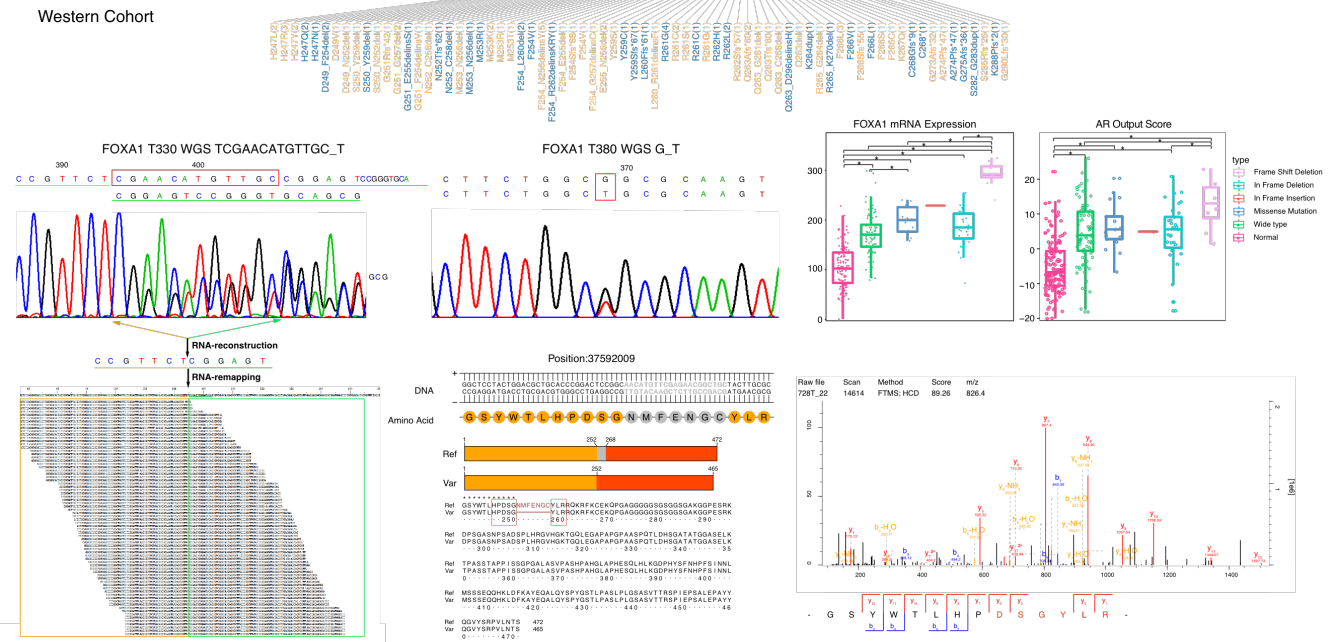




CPGEA

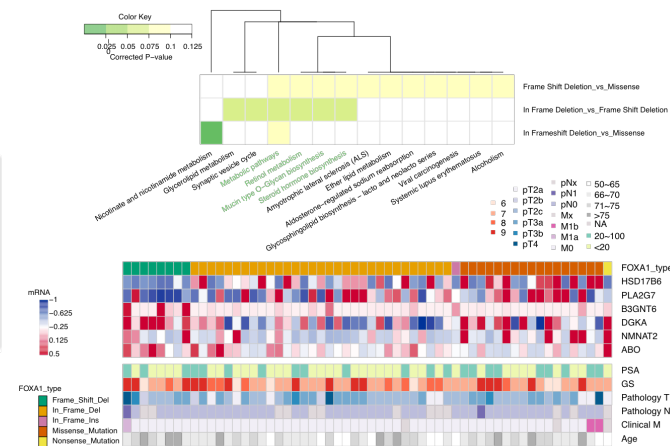
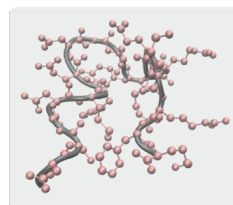


Western Cohort



FOXA1

In Frame Deletion
Residue 249-267aa



Landscape

Non-coding

CNV

- CNV arm level
- CNV gene level
- CNV comparison

DNA methylation

Fusion

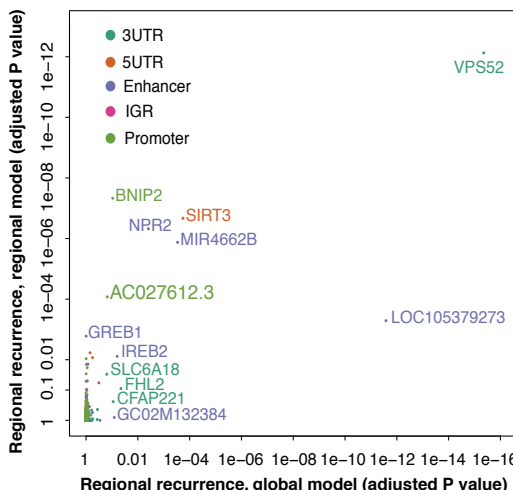
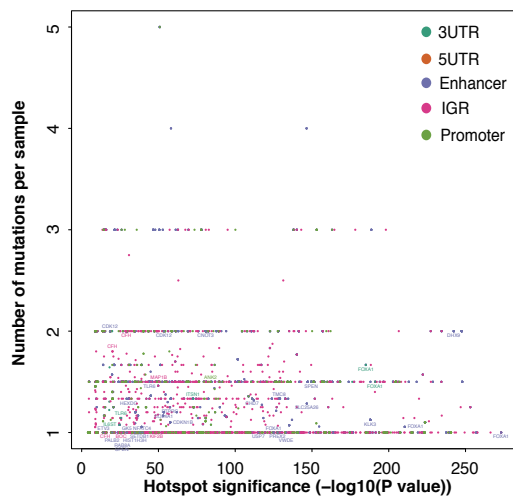
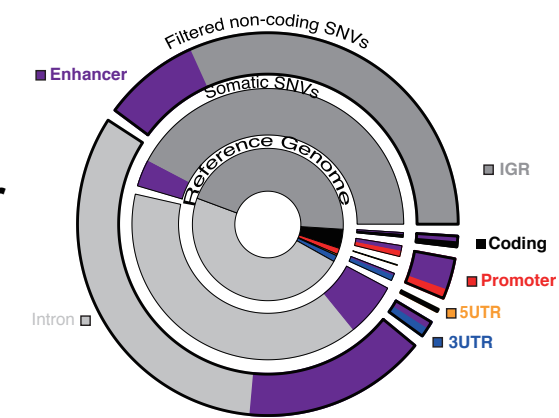
- Fusion
- Fusion comparison

SNV

- SMG
- SMG comparison

Cluster

Recurrently mutated non-coding regulatory elements in prostate cancer



0	0	0	0	14(8)	0	KMT2C*14
8	0	0	0	1	0	CFH*7
1(1)	0	0	4(4)	0	2(2)	FOXA1*5
0	0	0	0	5	0	RYR2*5
0	0	0	0	5	0	MDGA2*5
0	0	0	0	5	0	CSMD1*5
0	0	0	0	5	0	COL22A1*5
0	0	0	0	5	0	ANK2*5
0	0	0	0	3	0	USP28*3
0	0	0	0	3	0	PCDH15*3
0	0	0	0	3	0	CAMTA1*3
1	0	0	0	2	0	NPAS3*3
1	0	0	0	2	0	FAT3*3
1	1(1)	0	0	1	0	PREX2*2
3	0	0	0	0	0	KIF2B*3
0	0	0	1	1	0	AHNAK*1
1(1)	0	0	1	0	1	TLR6*2
0	0	0	1	0	1	PSAPL1*1
0	0	0	1(1)	0	1(1)	ITSN1*1
0	0	0	1(1)	0	1(1)	IL6ST*1
0	0	0	0	2	0	SYNE1*2
1(1)	0	0	0	2	0	PLXNA4*3
0	0	0	0	2	0	COL5A1*2
1(1)	1	0	0	1	0	ANAPC1*2
1	0	0	0	1	0	THSD7B*2
1	0	0	0	1	0	GRIA2*2
4(4)	0	0	0	1	0	CDK12*6
2	0	0	0	0	0	NFE2L3*2
2	0	0	0	0	0	MAP1B*2
0	0	0	0	1(1)	0	ZBTB16*1
2(2)	0	0	0	1	0	VWDE*3
0	0	0	0	1	0	TRPM4*1
0	0	0	0	1	0	TRIP12*1
0	0	0	0	1	0	SRPK2*1
0	0	0	0	1(1)	0	SLC25A26*1
0	0	0	0	1	0	SEMA5A*1
1(1)	0	0	0	1	0	RAB8A*2
0	0	0	0	1	0	PDS5A*1
0	0	0	0	1	0	PARP4*1
1(1)	0	0	0	1(1)	0	GK5*2
0	0	0	0	1	0	GABRP*1
0	0	0	0	1	0	F13A1*1
0	0	0	0	1	0	CSMD3*1
0	0	0	0	1	0	CCDC178*1
0	0	0	0	1	0	ATM*1
0	0	0	0	1	0	ADAMTS20*1
1	0	0	0	0	0	SMARCA1*1
1	0	0	0	0	0	SCN8A*1
1	0	0	0	0	0	PTPRC*1
1	0	0	0	0	0	MYC*1
1(1)	0	0	0	0	0	KLK3*2
1	0	0	0	0	0	C4orf32*1
1	0	0	0	0	0	BOC*1
1(1)	0	0	0	0	0	USP7*1
2(2)	0	0	0	0	0	USP48*2
1(1)	0	0	0	0	0	U2AF1*1
1(1)	0	0	0	0	0	TP53*1
1(1)	0	0	0	0	0	TMC8*1
4(4)	0	0	0	0	0	SPEN*4
1(1)	0	0	0	0	0	SETDB1*1
1(1)	0	0	0	0	0	RB1*1
3(3)	0	0	0	0	0	PIK3R1*3
1(1)	0	0	0	0	0	PALB2*1
1(1)	0	0	0	0	0	NFATC4*1
1(1)	0	0	0	0	0	NCOR1*1
1(1)	0	0	0	0	0	NCAPG2*1
2(2)	0	0	0	0	0	MUC4*2
1(1)	0	0	0	0	0	MRRF*1
1(1)	0	0	0	0	0	MBD1*1
1(1)	0	0	0	0	0	MATN4*1
2(2)	0	0	0	0	0	KIAA1551*2
2(2)	0	0	0	0	0	IMPA1*2
1(1)	0	0	0	0	0	HIST1H3H*1
1(1)	0	0	0	0	0	HEXDC*1
1(1)	0	0	0	0	0	ETV3*1
2(2)	0	0	0	0	0	DHX9*2
2(2)	0	0	0	0	0	CUL3*2
1(1)	0	0	0	0	0	COL4A1*1
2(2)	0	0	0	0	0	CNOT3*2
1(1)	0	0	0	0	0	CHD7*1
1(1)	0	0	0	0	0	CDKN1B*1
1(1)	0	0	0	0	0	BRAF*1
1(1)	0	0	0	0	0	ATXN2*1
1(1)	0	0	0	0	0	ARID2*1
1(1)	0	0	0	0	0	ARID1A*1

IGR Promoter 5UTR Exon Intron 3UTR

Landscape

CNV

- CNV arm level
- CNV gene level
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Fusion

- Fusion
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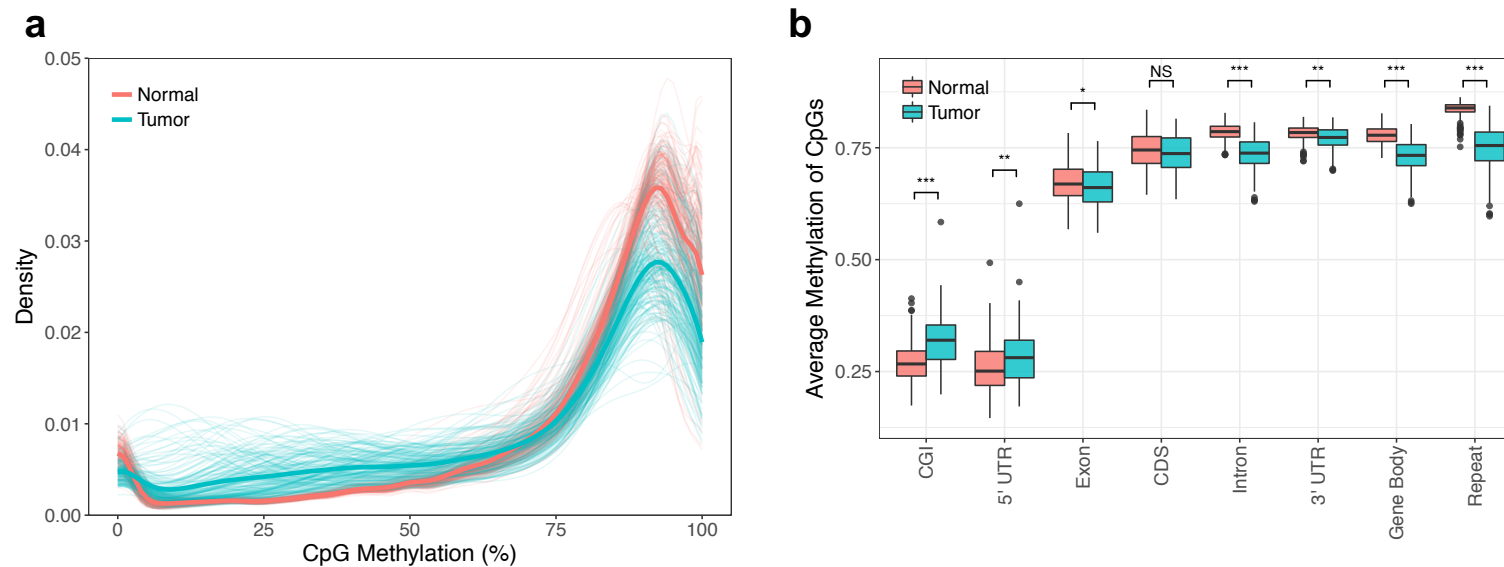
SNV

- SMG
- SMG comparison

Cluster

Non-coding

DNA methylation



Compared to normal prostate tissues, prostate tumor DNA methylomes were significantly hypomethylated (mean methylation levels of normal: 78.3%; tumor: 68.8%; p-value = 1.3×10^{-32} , Wilcoxon signed-rank test). Global CpG methylation level of normal prostates exhibited classical bimodal distribution, whereas CpG methylation of tumors was strongly shifted towards lower levels (Figure 5a, Supplementary Fig. 6a). Hypomethylation was significant across most genomic features including exons, introns, 3'UTRs, and all families of repetitive elements (Figure 5b, Supplementary Fig. 6b).

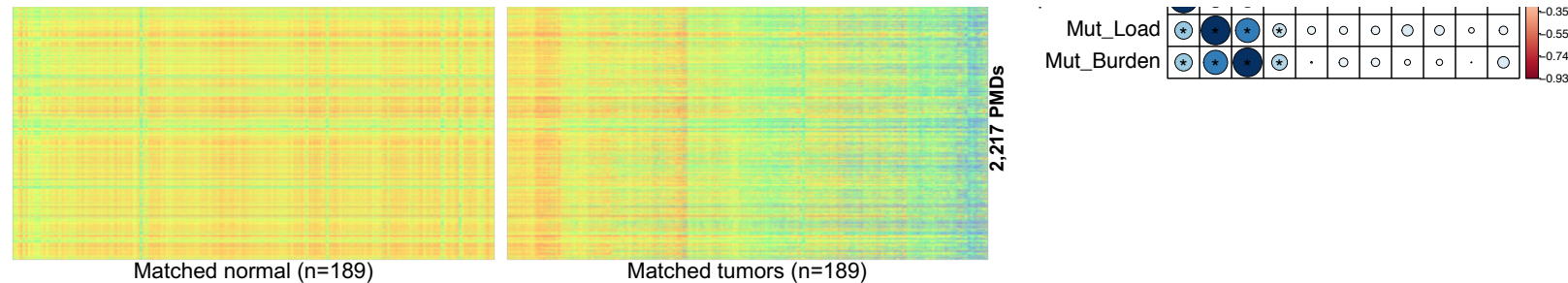


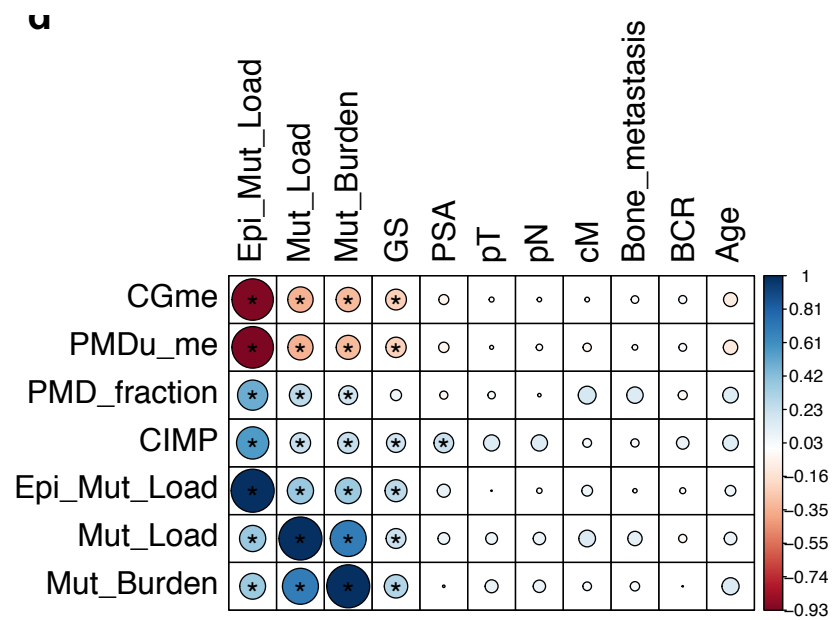
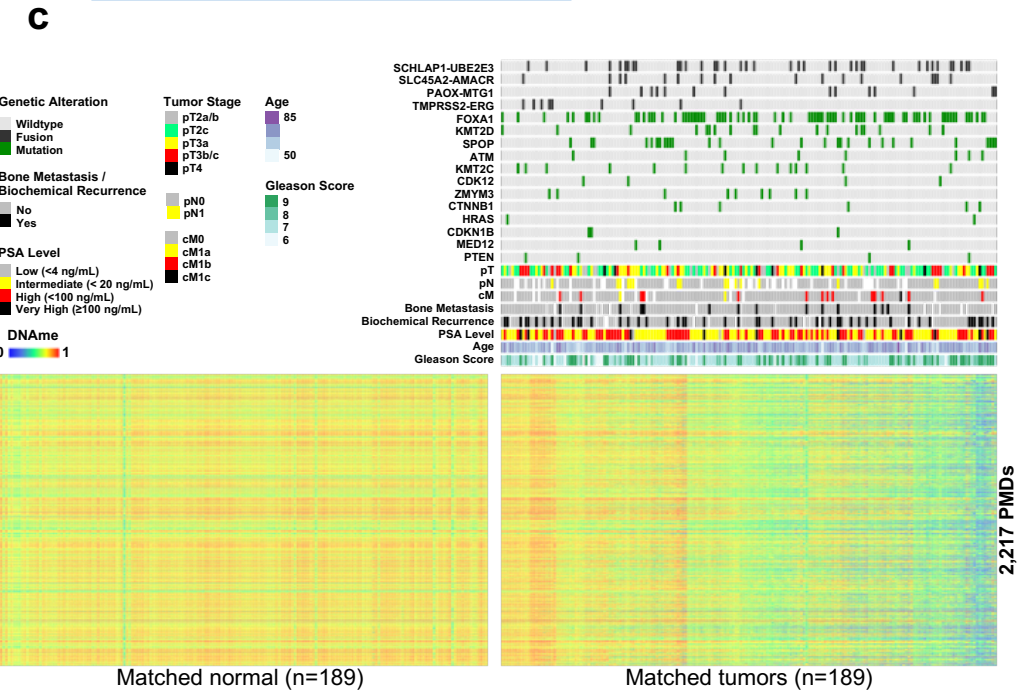
Fig5

a Partially methylated domains (PMD) have previously been characterized in cultured cells and cancer cells and are associated with gene repression and inactive chromatin marks [citations]. Using MethPipe [citations], we defined the pattern of PMD in our samples and found that PMDs were widespread in prostate cancers. These PMDs [description, size 940kb mean, methylation level, etc]. Strikingly, tumors with Gleason scores greater than 6 exhibited much lower DNA methylation levels in their PDMs than those with a Gleason score of 6 (Figure 5). [concluding remark]

<https://doi.org/10.1038/ng.865> (Irizarry / Feinberg group, 2011)

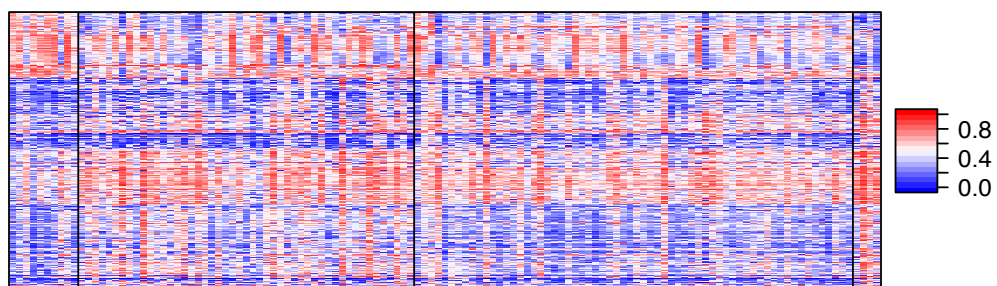
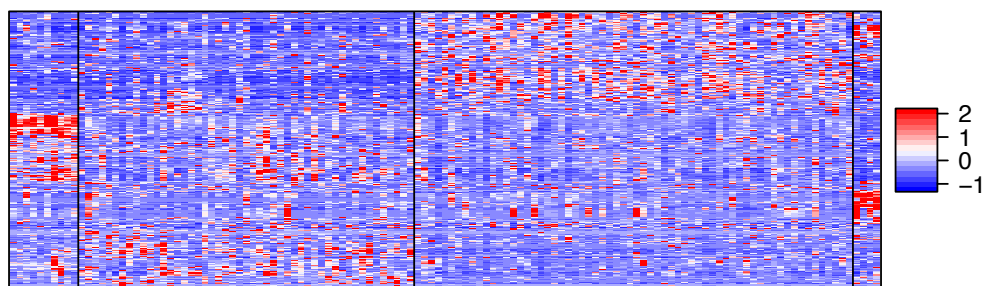
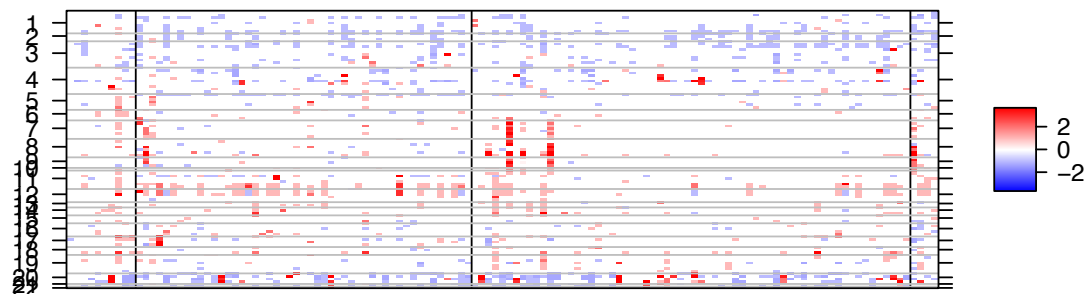
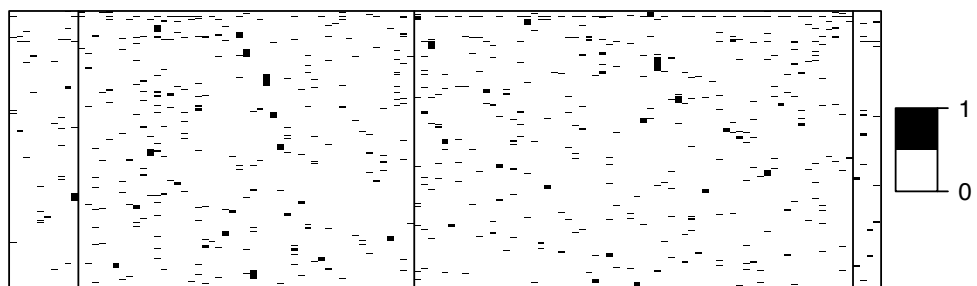
<https://doi.org/10.1101/gr.125872.111> (Ren group, 2012)

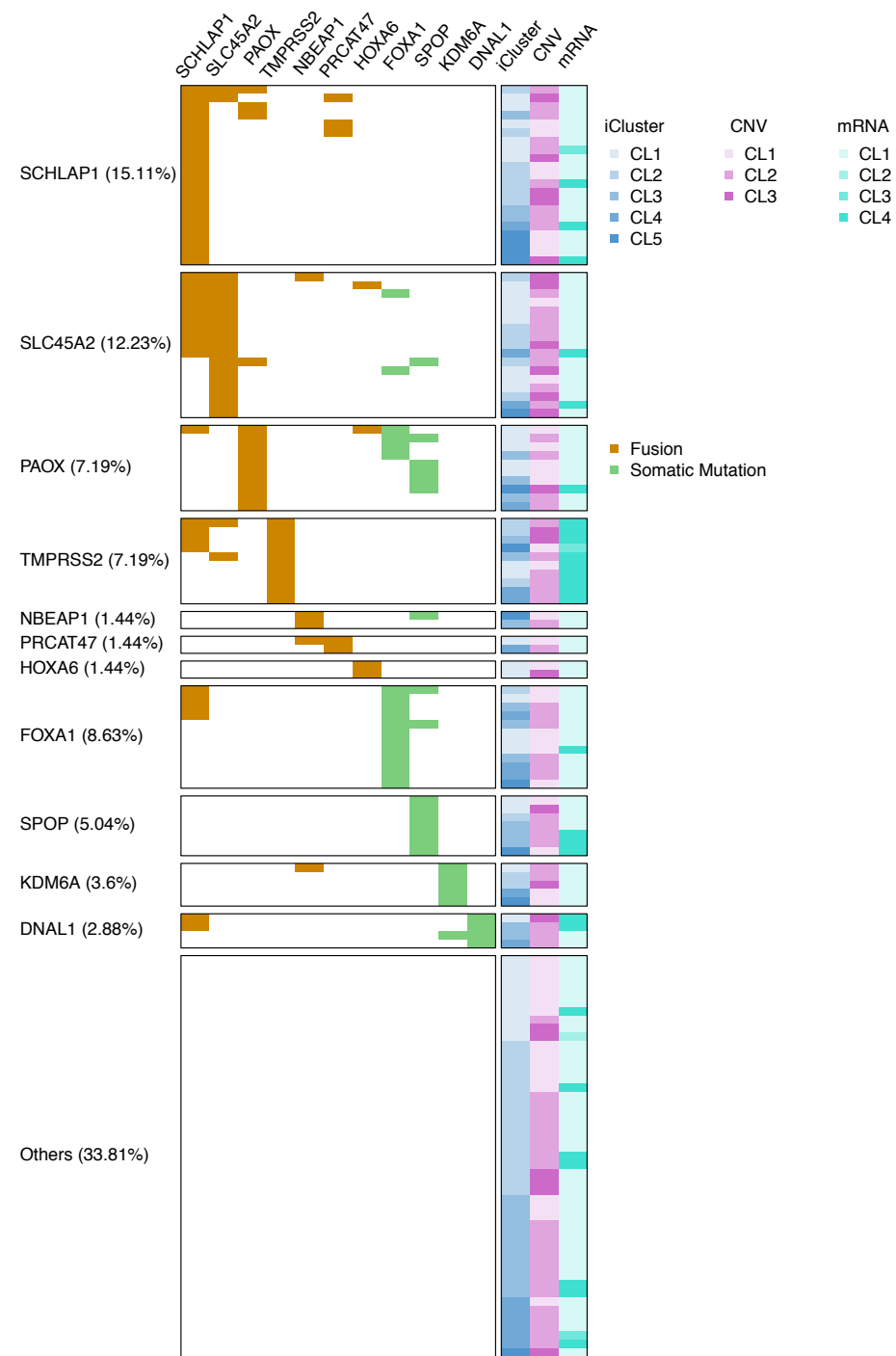
<https://doi.org/10.1038/ng.969> (Peter Laird group, 2012)



Ongoing...

- lcluster (CNV problem)
- pathway





Pathway....