The genomic landscape of metastatic castration-resistant prostate cancers using whole

genome sequencing reveals multiple distinct genotypes with potential clinical impact

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Keywords: prostate cancer, metastasis, cancer genomics, homologous recombination deficiency

Abstract

Here we present whole-genome sequencing (WGS) analysis of fresh-frozen metastatic biopsies from

197 castration-resistant prostate cancer patients. Using hierarchical unsupervised clustering based on

genomic aberrations only, we defined eight different clusters. We detected four distinct and potentially

clinically relevant genotypes harboring unique genomic features, including: 1) Microsatellite Instability;

2) Homologous Recombination Deficiency (HRD) with enriched genomic deletions and *BRCA2*

aberrations; 3) tandem duplication phenotype associated with biallelic *CDK12* mutations; and 4) a

subgroup enriched for chromothripsis events. Our data suggest that classifying patients using WGS

characteristics may improve classification of HRD patients. Moreover, we confirmed that important

regulators of *AR*-mediated signaling are located in non-coding regions. Using ChIP sequencing data,

we showed that the amplified *AR* and *MYC* promoter regions contain open chromatin and bind AR,

suggesting a role in AR mediated biology. Thus, high-resolution WGS may be used to improve patient

stratification.

Revision 0.9

Main text

Prostate cancer is known to be a notoriously heterogeneous disease and the genetic basis for this interpatient heterogeneity is poorly understood. The ongoing development of new therapies for metastatic prostate cancer that could potentially only be effective in molecularly defined subgroups, further increases the need for accurate patient classification and stratification[1–3]. Comprehensive genomic analyses in primary prostate cancer was able to classify 74% of analyzed patients into seven predefined subtypes based on ETS fusions and mutations in *SPOP*, *FOXA1* and *IDH1*[4]. More recently, whole-genome sequencing (WGS) of metastatic prostate cancer demonstrated that structural variations (SVs) arose from specific alterations such as *CDK12*[-/-] and *BRCA2*[-/-] genotypes[5–7]. As example, the predominance of tandem duplications was strongly associated with biallelic *CDK12* mutations. Thus, WGS enables the identification of patterns of DNA aberrations (i.e. 'genomic scars') that may profoundly improve classification of tumors that share a common etiology and that may be targeted using different therapies.

We analyzed fresh-frozen metastatic tumor samples and matched blood samples from 197 castration-resistant prostate cancer (CRPC) patients using WGS generating to date the largest WGS dataset for mCRPC and combine it with AR ChipSeq data (**Figure 1a).** Clinical details on biopsy site, age and previous treatments of the included patients are described in **figure 1** and **supplementary table 2**. An overview of the sequencing quality is provided in **supplementary figure 1**. The median tumor mutational burden on coding regions (TMB) was 2.54/Mb in our mCRPC cohort; this is roughly twice as high as compared to primary prostate cancer and furthermore, 14 patients had high TMB (>10)), which in other tumor types has been associated with a high sensitivity to check-point inhibitors (**supplementary figure 2**)[8]. We analyzed the somatic genomic aberrations and found a median of 6621 single-nucleotide variants (SNVs; IQR: 5048-9109), 1008 small insertions and deletions (InDels; IQR: 739-1364), 55 multi-nucleotide variants (MNVs; IQR: 34-86) and 224 SVs (IQR: 149-370) per patient (**supplementary figure 3a-c**). We observed a highly complex genomic landscape consisting of multiple driver mutations and structural rearrangements in our cohort. We confirmed that known key driver genes of prostate cancer were enriched for nonsynonymous mutations, including, *TP53, AR, FOXA1, SPOP, RB1* and *PTEN* (**Figure 2** and **supplementary figure 3d-e**). Distinct amplified genomic regions included 8q, 10q and Xq; deleted regions affected 8p, 10q, 13q and 17p (**supplementary figure 3d**). In addition to large-scale chromosomal copy-number alterations, we

could also further pinpoint narrow genomic regions targeted with recurrent copy-number alterations

which could potentially reveal important genes within or near the proximity of these events

(**supplementary table 3**).

TMPRSS2-ERG gene fusions were the most common fusions in our cohort (*n* = 84 (91.3% of

all ETS fusions); **Figure 2** and **supplementary figure 4**) and is comparable to localized prostate

cancer[4,9]. The predominant deletion site was located upstream of the second exon of *ERG*, which

preserves its ETS domain in the resulting fusion gene. In 42 patients (21.3%) we observed regional

hypermutation ("kataegis") (**Figure 2** and **supplementary figure 5**); this seems to be comparable to

kataegis rates in primary prostate cancer and thus is not an obvious driving force in metastatic

progression[9].

Several studies have shown that metastatic disease significantly differs from localized

prostate cancer and that disease progression towards CRPC is mainly driven by increased androgen

receptor signaling[10,11]. In-depth analysis of the AR-pathway revealed that aberrant AR signaling

occurred in up to 80% of our patients. In 57.3% of patients both *AR* and the *AR*-enhancer (located

about 631 kB upstream of the *AR* gene[6]) were affected (**Figure 3a**). In an additional 6.6% and 14.7%

of patients only *AR* alterations or *AR*-enhancer amplification occurred, respectively. Concurrent

amplification of *AR* and *AR*-enhancer was not necessarily of equal magnitude, which resulted in

differences in copy number enrichment of these loci (**Figure 3b**). ChIP-seq data of two mCRPC

patients and prostate cancer cell-lines (LNCaP and VCaP) revealed active enhancer regions

(H3K27ac), coupled with actively bound *AR and FOXA1,* at the detected amplification peaks, which

was found to be enriched in CRPC settings (**Figure 3c**). This indicates that *AR*-enhancer amplification

could be associated with increased AR-signaling for this genomic region, which is supported by

previous studies demonstrating that this amplification ultimately resulted in significantly elevated

expression of *AR* itself[6,7]. Furthermore, a recurrent focal amplification at a non-coding area was

observed at 8q24.21 near *PCAT1*. This locus bears similar epigenetic characteristics to the *AR*-

enhancer with regard to H3K27ac and, to a lesser extent, binding of AR and/or FOXA1 (**Figure 3c**).

This locus could represent a somatically-acquired putative enhancer affecting *MYC* expression

(**Figure 3d**), however functional follow-up studies should be performed to further this hypothesis[12]. In

addition, *PCAT1* is a long non-coding RNA which is known to be upregulated in prostate cancer and

negatively regulates *BRCA2* expression while positively affecting *MYC* expression[13,14]. These data

show that most prostate cancers reactivate the AR pathway either directly or indirectly when

progressing toward (m)CRPC.

Our comprehensive WGS data and sample size enabled us to perform unsupervised

clustering to identify genomic scars that can define subgroups of mCRPC patients. We clustered our

genomic data using total number of SVs, relative frequency of SV categories, TMB and tumor ploidy.

This analysis defined eight distinct subgroups (**Figure 4-5** and **supplementary figure 6-8**): A)

Microsatellite Instability (MSI) signature with high TMB and associated with mismatch repair

deficiency; B) Tandem duplications (>100 kb) phenotype associated with biallelic *CDK12* inactivation;

D) Homologous Recombination Deficiency (HRD) features with many (>100kb) deletions and

association with (somatic) mutations in BRCAness-associated genes; F) chromothripsis; C, E, G, H);

non-significant genomic signature without any currently known biological association. **Table 1**

summarizes the key features of each subgroup. Cluster A and B represent previously identified

genomic subgroups[5,7,10,15] and in both groups only a minority of the patients was allocated to these

subgroups without a specific mutation in the corresponding genes. Interestingly, 2 out of 13 patients in

cluster B (Tandem duplications (>100 kb) phenotype) did not show a bi-allelic (somatic) mutation in

CDK12, suggesting that tandem duplications may arise in patient without *CDK12* mutations. Cluster D

shows significant features of HRD, specifically biallelic *BRCA2* inactivation, mutational signature 3,

enrichments of deletions (<100 kb) and is supported by high HR-deficiency scores (CHORD)[16,17]

(**supplementary figure 6**). Although this is a known association[7,18], our clustering analysis potentially

refines patient classification, as 32% of this subgroup (7/22) does not have a defining biallelic *BRCA2*

(somatic) mutation of which four of these patients show at least one (deleterious) aberration in other

BRCAness-related genes[19]. In addition, 4 patients in other clusters show non-synonymous mutations

in *BRCA2* without corresponding genomic scars, mutational signature and/or HR-deficiency scores

(**figure 4**, **supplementary figure 9**). Only a single sample in cluster A harbored a *BRCA2* mutation

with known pathogenic effect (p.T3030fs; RCV000031792)[20] and was identified with a HR-deficiency

score which was clustered based on MSI-profile in our analysis. These patients might be deemed

false-negative or false-positive when using FDA-approved assays (BRCAnalysis™ and

FoundationFocus™) to predict response to poly(ADP-ribose) polymerase inhibitors (PARPi) or

alkylating drugs like platinum compounds based on the presence of *BRCA* mutations. In cluster F, we

detect significantly more chromothripsis events in comparison to the other clusters (80% vs 20%).

However, the overall frequency of chromothripsis (23.3%) was comparable with previous findings[7,9].

We failed to reproduce a previous finding suggesting chromothripsis to be associated with inversions

and p53 inactivation in prostate cancer[7]. Apart from the chromothripsis events, no clear gene

aberration was associated with this cluster. In the remaining patients, we could not identify a distinct

genomic signature or biologic rationale (cluster C, E, G, H). In cluster C, conjoint aberrations of

BRCA1 and *TP53* were observed in one patient with a high HR-deficiency prediction score (CHORD),

which is known to lead to a small tandem duplication phenotype (<100 kb)[21]. Two other patients within

cluster C display a weak CHORD scoring associated with HR-deficiency, however no additional

evidence was found for a BRCA1 loss-of-function within these patients.

146   The classification of patients using WGS has the advantage of being, in theory, more precise

and less prone to bias compared to analyses using targeted panels consisting of a limited number of

genes. Overall, our study describes the complete genomic landscape of metastatic CRPC and

confirms the central role of *AR* signaling in this disease. We identify distinct CRPC subgroups based

on phenotypic characteristics encompassing genomic signatures, including MSI, BRCAness and

CDK12 inactivity, which may be clinically relevant.[5,19] Moreover, we show the added value of WGS-

based unbiased clustering in identifying additional patients with genomic scars who are eligible for

specific therapies and we could classify patients even if WGS (or our methodology) did not find

conclusive evidence for a bi-allelic mutation in the proposed gene-of-interest.

155   This study also showed that a large population of mCRPC patients do not fall into an as-of-yet

clinically-relevant or biologically-clear genotype and further research can help elucidate the oncogenic

driver events and provide new therapeutic options. In addition, further analysis using whole-genome

sequencing data allows us to gain more insight into the role of non-coding regions of the genome in

prostate cancer.

Methods

Methods, including statements of data availability and any associated accession codes and

references, are available in the supplementary information file.

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Acknowledgements

This publication and the underlying study have been made possible partly on the basis of the data

that Hartwig Medical Foundation and the Center of Personalized Cancer Treatment (CPCT) have

made available to the study. We would like to thank the local principal investigators of all contributing

centers for their help with patient enrollment (listed in **supplementary table 1**). We would also like to

thank Tesa M. Severson for her help with the computational analyses of the ChIP-seq data, Suzan

Stelloo for providing ChIP-seq results on cell lines and Arne van Hoeck for providing the CHORD

(HR-deficiency) prediction scores. In addition, we would like to thank the Barcode for Life foundation

for making this research possible.

Grants

This work was supported in parts by a KWF-Alpe d'HuZes project [NKI 2014-7080], a grant from

Astellas Pharma [Lolkema/NL-72-RG-11] and a Johnson & Johnson grant [212082PCR3014].

Author contributions

LFVD, JVR, MS, MPL and HJGVDW wrote the manuscript, which all authors critically reviewed. JVR

and HJGVDW performed the bioinformatics analyses. LFVD, MPL and NM managed clinical data

assessment. PW and WZ performed ChIP-seq experiments and data analyses. NM, MPL, PH, AMB,

MSVDH, IMVO, and RDW are clinical contributors. MSVDH, EV, NS, JWMM and SS are part of the

CPCT-02 study. EPJGC coordinated the sequencing of samples and contributed to the bioinformatics

analyses.

Competing interests

All authors declare no competing interests.

Additional information

Supplementary information is available for this paper in the supplementary information file.

 FIGURE LEGENDS

Figure 1 - Overview of study design and patient cohort (*n* = 197).

(a) **Flowchart of patient inclusion.** From the CPCT-02 cohort, patients with metastatic prostate

cancer were selected. Patients were excluded if data from metastatic samples were not

available and if clinical data analysis showed that patients had hormone-sensitive or neuro-

endocrine prostate cancer or unknown disease status at the time of analysis.

(b) **Overview of the biopsy sites.** Number of biopsies per metastatic site analyzed with WGS.

(c) **Age of patients at biopsy.** Bee-swarm boxplot with notch of the patient age distribution.

Figure 2 - mCRPC remains a genetically heterogeneous disease but shows several recurrent

somatic alterations in key genes affecting several oncogenic pathways.

Based on dN/dS (q ≤ 0.1) and GISTIC2 focal peak (q ≤ 0.1) criteria, we show the genes and

focal genomic foci which are most often recurrently mutated, amplified or deleted in our

mCRPC cohort of 197 patients. The upper track (top bar plot) displays tumor mutational

burden (TMB) per SNV, InDel and MNV category on coding regions (square root scale).

Samples are sorted based on mutual-exclusivity of the depicted genes and foci. The heatmap

displays the type of mutation(s) per sample, (light-)green or (light-)red backgrounds depict

copy-number aberrations whilst the inner square depicts the type of (coding) mutation(s).

Relative proportions of mutational categories (coding mutations (SNV, InDels and MNV), SV,

deep gains (high-level amplifications resulting in many additional copies) and deep deletions

(high-level losses resulting in (near) homozygous losses) per gene and foci are shown in the

barplot next to the heatmap. Narrow GISTIC2 peaks covering ≤ 3 genes were reduced to

gene-level rows if one of these genes is present in the dN/dS (q ≤ 0.1) analysis or is a known

oncogene or tumor-suppressor. For GISTIC2 peaks covering multiple genes, only deep

amplifications and deep deletions are shown. Recurrent aberrant focal genomic foci in gene

deserts are annotated with their nearest gene. Significance scores (-1*log10(q)) of the dN/dS

and GISTIC2 analysis are shown on the outer-right barplots; bars in the GISTIC2 significance

plot are colored red if these foci were detected as a recurrent focal deletion and green if

detected as a recurrent focal gain. Per sample, the presence of (predicted) ETS fusions,

kataegis, chromothripsis, CHORD predictive score (HR-deficiency), MSI status and biopsy

location are shown as bottom tracks

Figure 3 - Whole Genome Sequencing reveals novel insight into the various molecular (non-

coding) aberrations affecting AR regulation.

(a) Mutational overview of top recurrently-mutated genes affecting AR regulation and their

putative enhancer foci (as detected by GISTIC2). The first track represents the number of

genomic mutations per Mb (TMB) per SNV, InDels and MNV category (square root scale).

Samples are sorted based on mutual-exclusivity of the depicted genes and foci. The heatmap

displays the type of mutation(s) per sample, (light-)green or (light-)red backgrounds depict

copy-number aberrations whilst the inner square depicts the type of (coding) mutation(s).

Relative proportions of mutational categories (coding mutations (SNV, InDels and MNV), SV,

deep gains and deep deletions) per gene and foci are shown in the barplot next to the

heatmap. The presence of (predicted) ETS fusions, kataegis, chromothripsis, CHORD

predictive score (HR-deficiency), MSI status and biopsy location are shown as bottom tracks.

(b) Overview of the copy-number deviations between putative enhancer and gene regions for *AR*

and *MYC*. Samples were categorized as enhancer- or gene-enriched if enhancer-to-gene

ratio deviated >1 studentized residual (residual derived from a linear model without the

respective observation) from a 1:1 ratio.

(c) Copy-number and ChIP-seq profiles surrounding the *AR* and *PCAT1/MYC* gene loci (with

1.25 additional Mb up-/downstream). The upper track displays the selected genomic window

and the overlapping genes. The first and second track display the aggregated mean copy-

number (per 1000bp window) of the enhancer- and gene-enriched samples, respectively.

These profiles identify distinct amplified regions (indicated by red asterisk) in proximity to the

respective gene bodies. The 3th to 8th tracks represent AR ChIP-seq profiles (mean read-

coverage per 1Mb windows) in two mCRPC patients, LNCaP and LNCaP with R1881

treatment, VCaP and bicalutamide-resistant VCaP. The 9th to 11th tracks represent FOXA1

ChIP-seq profiles (mean read-coverage per 1Mb windows) in two mCRPC patients and

LNCaP with R1881 treatment. The 12th to 14th tracks represent H3K27ac ChIP-seq profiles

343	(mean read-coverage per 1Mb windows) in two mCRPC patients and LNCaP with R1881
344	treatment reflecting active enhancer regions.
345	ChIP-seq peaks (MACS/MACS2; q < 0.01) are shown as black lines per respective sample.
346	

Figure 4 - Unsupervised clustering of mCRPC reveals distinct genomic phenotypes.

(a) Dendogram of unsupervised clustering (Pearson correlation; ward.D) with optimal leaf
ordering. Top eight clusters are highlighted and denoted based on order of appearance (left
to right): A to H. Y-axis displays clustering distance (Pearson; ward.D).

(b) Number of genomic mutations per Mb (TMB) of SNV, InDels and MNV categories. All
genome-wide somatic mutations were taken into consideration (square root scale).

(c) Absolute number of unique structural variants per sample. Cumulative frequency of
inversions, tandem duplication, deletions, insertions and translocations.

(d) Relative frequency per structural variant category, Tandem Duplications and Deletions
are subdivided into >100kb and <100kb categories. This track shows if an enrichment for
particular category of (somatic) structural variant can be detected, which in turn, can be
indicative for a specific mutational aberration.

(e) Relative genome-wide ploidy status, ranging from 0 to ≥7 copies. This track shows the
relative percentage of the entire genome which is (partially) lost (ploidy < 2/diploid) or
amplified (> 2/diploid).

(f) Relative contribution to mutational signatures (COSMIC) summarized per proposed
etiology. This track displays the proposed etiology of each SNV based on their mutational
contexts.

(g) Relative frequency of SNV mutational changes.

(h) HR-deficient prediction score as assessed by CHORD. The binary prediction score of
CHORD (ranging from 0 to 1) is shown, in which higher scores reflect more evidence for HR-
deficiency in a given sample.

(i) MSI status as determined using a stringent threshold of MSI characteristics[22].

Figure 5 - Distinct molecular phenotypes in mCRPC are enriched by mutually-exclusive
aberrations in key pathways or large-scale somatic events.

 (a) **Cluster-specific enrichment of mutated genes, chromothripsis, gene fusions and**

 kataegis (Fisher's Exact Test ≤ 0.05). Percentages to the left of the black line represent the

 relative mutational frequency in mCRPC samples which are not present in the respective

 cluster, whilst the percentages to the right of the black line represent the relative mutational

 frequency present in the samples from the tested cluster.

 (b) **Genomic overview with biologically-relevant genes in the clusters with mutational**

 enrichment of genes or large-scale events (A, B, D and F). The first track represents the

 number of genomic mutations per Mb (TMB) per SNV, InDels and MNV category (square-root

 scale). The second track represents the absolute number of unique structural variants per

 sample. The third track represents the relative frequency per structural variant category,

 Tandem Duplications and Deletions are subdivided into >100kb and <100kb categories. The

 fourth track represents relative genome-wide ploidy status, ranging from 0 to ≥7 copies. The

 fifth track represents the relative contribution to mutational signatures (COSMIC) summarized

 per proposed etiology. The sixth track displays somatic mutations in the relevant genes found

 in at least one cluster. The lower tracks represent presence of ETS fusions, chromothripsis,

 kataegis, HR-deficiency prediction scores and MSI status based on a threshold of MSI

 characteristics.

 Table 1 – Cluster characteristics

 Overview of the distinctive characteristics for each cluster (A-H).

a

(CPCT-02 study)
Metastatic prostate cancer
patients included for biopsy
n = 341

Non-metastatic biopsy site
(i.e. prostate)
n = 3

No biopsy taken
n = 24

Fresh-frozen biopsy and
blood control taken
n = 314

Successful biopsy
(TC% ≥ 30%)
n = 218

Failed biopsy
(TC% < 30%)
n = 96

No WGS due to insufficient
tumor DNA quantity
n = 1

WGS biopsy (114x) and
blood control (38x)
n = 217

non-mCRPC
(Neuro-endocrine/HSPC/Unknown)
n = 20

Final inclusion of mCRPC
(adenocarcinoma)
n = 197

b

Lung — 3

Bone — 70

Liver — 29

Soft tissue — 14

Lymph node — 81

c

Age at biopsy

mCRPC cohort(n = 197)

	Cluster A
Number of patients (n; % of cohort)	13 (6.6)
Tumor mutational burden (CDS)	36.88
InDel/SNV ratio	0.99
Number of structural variants	149
Main structural variant category or differentiating category	None
Ploidy status	1.92
Main mutational signature	MSI
Top 3 cluster-specific abberations (% of cluster)	MSH6 (69.2) JAK1 (69.2) CIC (58.3)
ETS-fusions (n)	3
Chromothripsis (n)	1
Kataegis (n)	6

All numbers are median of the cluster, unless otherwise indicated.
CDS, coding sequence

Table 1: Cluster characteristics

Cluster B	Cluster C	Cluster D
13 (6.6)	15 (7.6)	22 (11.2)
2.44	3.00	4.39
7.07	6.73	7.28
669	237	323
Tandem duplications (>100 kb)	Tandem duplications (<100 kb)	Deletions (>100kb)
2.39	3.19	2.16
N/A	N/A	BRCA
CDK12 (84.6) FGF3 (69.2) FGF4 (69.2)	None	BRCA2 (68.2)
2	7	7
0	1	5
1	2	5

Cluster E	Cluster F	Cluster G
55 (27.9)	20 (10.15)	34 (17.3)
2.12	2.51	2.12
7.13	6.15	6.13
178	399.5	221.5
None	None	None
3.24	3.35	2.98
N/A	N/A	N/A
None	Chromothripsis (80)	None
25	10	23
8	16	8
13	7	5

Cluster H
25 (12.7)
2.30
5.81
201
Insertions
1.97
N/A
None
16
7
3

Mutations
BRCA2

32900000

Cluster G - chr13:32900730T>G (c.838 - Missense)

Cluster D - chr13:32907202TA>T (c.1815 - Frameshift)

Cluster E - chr13:32912615G>T (c.4350 - Truncating)

Cluster D - chr13:32913166T>A (c.4901 - Missense)

Cluster D - chr13:32914157A>AT (c.5893 - Frameshift)

Cluster H - chr13:32915094C>G (c.6829 - Missense)

Cluster D - chr13:32915120G>C (c.6855 - Missense)

Cluster D - chr13:32915237GCCACACATTCT>G (c.6973 - Frameshift)

32920000

Cluster D - chr13:32915254TACATGTCCCG>T (c.6990 - Frameshift)

32940000

Cluster E - chr13:32945129C>T (c.8751*† - Missense)

Cluster D - chr13:32954022C>CA (c.9317† - Frameshift) Pathogenic

Cluster A - chr13:32954022CA>C (c.9317† - Frameshift) Pathogenic

Cluster D - chr13:32954022C>CA (c.9317† - Frameshift) Pathogenic

32960000

Exon   Frequency of variant allele
UTR   Frequency of reference allele

Supplementary table 1: Participating centers	
Organization	**Local principal investigator**
Radboud UMC, Nijmegen	Carla van Herpen
Erasmus MC, Rotterdam	Martijn Lolkema
Franciscus Gasthuis & Vlietland, Rotterdam	Paul Hamberg
NKI-AVL, Amsterdam	Neeltje Steeghs
Isala, Zwolle	Jan Willen de Groot
Martini Ziekenhuis, Groningen	Johan van Rooijen
Medisch Centrum Leeuwarden	Hiltje de Graaf
Maastricht UMC, Maastricht	Vivianne Tjan-Heijnen
Noordwest Ziekenhuisgroep, Alkmaar	Mathijs Hendriks
UMC Utrecht, Utrecht	Els Witteveen
Amphia Ziekenhuis, Breda	Bert Jan ten Tije
Reinier de Graaf Gasthuis, Delft	Annelie Vulink
Treant Zorggroep, Hoogeveen	Sophia van den Boogerd
Zuyderland Medisch Centrum, Geleen	Frans Erdkamp
ETZ Elisabeth, Tilburg	Laurens Beerepoot
Leids Universitair Medisch Centrum, Leiden	Hans Gelderblom
Maasstad Ziekenhuis, Rotterdam	Rineke Leys
Meander Medisch Centrum, Amersfoort	Haiko Bloemendal
St. Antonius Ziekenhuis, Utrecht	Maartje Los
VUmc, Amsterdam	Henk Verheul
ZGT, Almelo	Esther Siemerink

Included patients for this study (n)
91
38
20
15
3
3
3
3
3
3
2
2
2
2
1
1
1
1
1
1
1

CPCT-02 (mCRPC)

Martini Ziekenhuis (n=3)

3)

)

ZGT (n=1)

2)

Supplementary table 2: Patient characteristics		
Patients (n=197)		
	n	**%**
Age at biopsy		
Median	68	
Range (min-max)	48-83	
Prior ADT		
Yes	197	100.0
Drug-based	181	91.9
Surgery-based (orchiectomy)	3	1.5
With Docetaxel	6	3.0
No clear documentation of ADT type	7	3.6
Prior systemic therapy (other than ADT)		
0 previous treatments	27	13.7
≥ 1 previous treatments	170	86.3
1 previous treatment	45	22.8
2 previous treatments	69	35.0
3 previous treatments	31	15.7
4 previous treatments	19	9.6
5 previous treatments	6	3.0
Type of prior systemic therapy (other than ADT)		
Hormonal therapy only	20	10.2
Chemotherapy only	37	18.8
Radionucleotide therapy only	4	2.0
Immunotherapy only (Dendritic cell therapy)	4	2.0
Targeted therapy only	0	0.0
Hormonal and chemotherapy	68	34.5
Hormonal and radionucleotide therapy	3	1.5
Chemotherapy and radionucleotide therapy	3	1.5
Hormonal and immunotherapy	3	1.5
Chemotherapy and immunotherapy	3	1.5
Hormonal, chemotherapy and radionucleotide therapy	15	7.6
Hormonal, chemotherapy and immunotherapy	4	2.0
Hormonal, radionucleotide and immunotherapy	2	1.0
Hormonal, chemotherapy and targeted therapy (Olaparib)	2	1.0
Hormonal, chemotherapy, radionucleotide and immunotherapy	1	0.5
Unknown at time of analysis	1	0.5
Prior radiotherapy		
Yes (curative radiotherapy of the prostate and/or palliative radiotherapy of metastases)	117	59.4
No	77	39.1
Unknown at time of analysis	3	1.5
Started therapy after biopsy for whole-genome sequencing		
Yes	138	70.1
Hormonal therapy	53	26.9

Chemotherapy	56	28.4
Radionucleotide therapy	12	6.1
Immunotherapy (Pembrolizumab)	6	3.0
Targeted therapy	3	1.5
Combinational therapy	6	3.0
Other*	2	1.0
No	19	9.6
Unknown at time of analysis	40	20.3
Biopsy site		
Liver	29	14.7
Lymph node	81	41.1
Bone	70	35.5
Lung	3	1.5
Soft tissue/Other**	14	7.1
*Boneregulating agent		
**Soft tissue/other: (sub)cutis, muscle, peritoneum, kidney, bladder, adrenal gland		

Sheet G - Exclusive mutations

Gene	Cluster	mutationsInCluster	nomutationsInCluster	mutationsInOtherCluster
CDK12	Cluster B	11	2	5
BRCA2	Cluster D	15	7	12
MSH6	Cluster A	9	4	4
Chromoth	Cluster F	16	4	30
JAK1	Cluster A	9	4	11
FGF3	Cluster B	9	4	11
FGF4	Cluster B	9	4	11
MDM4	Cluster B	10	3	21
CIC	Cluster A	7	6	5
CCND1	Cluster B	9	4	16
EPAS1	Cluster A	7	6	6
MSH2	Cluster A	7	6	7
SLC45A3	Cluster B	9	4	18
FBXO11	Cluster A	6	7	5
ELK4	Cluster B	9	4	22
NSD1	Cluster B	6	7	6
BCL2L12	Cluster B	5	8	3
FGFR4	Cluster B	5	8	3
HIST2H2I	Cluster B	8	5	18
ASXL1	Cluster A	7	6	12
SIX2	Cluster A	6	7	7
ZFP36L2	Cluster A	8	5	19
NBN	Cluster B	11	2	48
ARID1A	Cluster A	6	7	8
ERF	Cluster A	6	7	8
MLH1	Cluster A	6	7	8
CNTNAP2	Cluster B	9	4	28
POTEE	Cluster A	5	8	4
SETD1B	Cluster A	7	6	14
TBX3	Cluster B	6	7	9
FRS2	Cluster B	7	6	15
ACVR1	Cluster A	5	8	5
CHD2	Cluster A	5	8	5
KMT2B	Cluster A	5	8	5
PBRM1	Cluster A	5	8	5
ZFHX3	Cluster A	9	4	31
ECT2L	Cluster B	5	8	5
PCBP1	Cluster A	4	9	2
IDH2	Cluster B	4	9	2
BCL11A	Cluster A	6	7	11
TCF7L2	Cluster A	5	8	6
CDK4	Cluster B	5	8	6
IL2	Cluster B	5	8	6
RAB35	Cluster B	5	8	6
KMT2C	Cluster A	8	5	26
EML4	Cluster A	6	7	12
KMT2D	Cluster A	6	7	12
ACVR2A	Cluster A	5	8	7
BAX	Cluster B	5	8	7
EREG	Cluster B	5	8	7

Sheet G - Exclusive mutations

nomutationsInOtherCluster	p	p.adj		
179	9.91E-13	7.29E-09		
163	1.63E-10	5.99E-07		
180	4.62E-10	1.13E-06		
147	1.84E-08	3.39E-05		
173	9.44E-08	9.93E-05	**Test used:**	
173	9.44E-08	9.93E-05	Two-sided Fisher's Exact Test	
173	9.44E-08	9.93E-05		
163	4.77E-07	4.39E-04		
179	5.75E-07	4.70E-04		
168	1.04E-06	7.63E-04		
178	1.21E-06	8.10E-04		
177	2.35E-06	1.33E-03		
166	2.28E-06	1.33E-03		
179	8.98E-06	4.42E-03		
162	9.00E-06	4.42E-03		
178	1.74E-05	8.00E-03		
181	2.76E-05	1.07E-02		
181	2.76E-05	1.07E-02		
166	2.64E-05	1.07E-02		
172	2.99E-05	1.09E-02		
177	3.12E-05	1.09E-02		
165	3.66E-05	1.22E-02		
136	3.89E-05	1.24E-02		
176	5.29E-05	1.44E-02		
176	5.29E-05	1.44E-02		
176	5.29E-05	1.44E-02		
156	4.87E-05	1.44E-02		
180	5.99E-05	1.57E-02		
170	6.50E-05	1.65E-02		
175	8.53E-05	2.09E-02		
169	9.25E-05	2.20E-02		
179	1.16E-04	2.30E-02		
179	1.16E-04	2.30E-02		
179	1.16E-04	2.30E-02		
179	1.16E-04	2.30E-02		
153	1.00E-04	2.30E-02		
179	1.16E-04	2.30E-02		
182	1.63E-04	3.08E-02		
182	1.63E-04	3.08E-02		
173	1.97E-04	3.42E-02		
178	2.04E-04	3.42E-02		
178	2.04E-04	3.42E-02		
178	2.04E-04	3.42E-02		
178	2.04E-04	3.42E-02		
158	2.49E-04	4.07E-02		
172	2.86E-04	4.49E-02		
172	2.86E-04	4.49E-02		
177	3.38E-04	4.97E-02		
177	3.38E-04	4.97E-02		
177	3.38E-04	4.97E-02		