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Drug-Induced Epigenomic Plasticity Reprograms Circadian Rhythm Regulation to Drive Prostate Cancer toward Androgen Independence

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10.1158/2159-8290.CD-21-0576

Publication date 2022 Document Version

Accepted author manuscript **Published in** Cancer discovery

Citation (APA)

Linder, S., Hoogstraat, M., Stelloo, S., Eickhoff, N., Schuurman, K., de Barros, H., Alkemade, M., Bekers, E. M., Wessels, L. F. A., & More Authors (2022). Drug-Induced Epigenomic Plasticity Reprograms Circadian Rhythm Regulation to Drive Prostate Cancer toward Androgen Independence. *Cancer discovery*, *12*(9), 2074-2097. https://doi.org/10.1158/2159-8290.CD-21-0576

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1 Drug-induced epigenomic plasticity reprograms circadian rhythm regulation to

2 drive prostate cancer towards androgen-independence

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- 47 **Running Title:** Drug-induced epigenomic plasticity drives PCa cell survival

48 Abstract

49 In prostate cancer, and rogen receptor (AR)-targeting agents are very effective in various stages of the 50 disease. However, therapy resistance inevitably occurs and little is known about how tumor cells adapt to 51 bypass AR suppression. Here, we performed integrative multi-omics analyses on tissues isolated before 52 and after 3 months of AR-targeting enzalutamide monotherapy from high-risk prostate cancer patients 53 enrolled in a neoadjuvant clinical trial. Transcriptomic analyses demonstrated that AR inhibition drove 54 tumors towards a neuroendocrine-like disease state. In addition, epigenomic profiling revealed massive 55 enzalutamide-induced reprogramming of pioneer factor FOXA1 - from inactive chromatin binding sites 56 towards active cis-regulatory elements that dictate pro-survival signals. Notably, treatment-induced FOXA1 57 sites were enriched for the circadian rhythm core component ARNTL. Post-treatment ARNTL levels 58 associated with poor outcome, and ARNTL suppression decreased cell growth in vitro. Our data highlight 59 a remarkable cistromic plasticity of FOXA1 following AR-targeted therapy, and revealed an acquired 60 dependency on circadian regulator ARNTL, a novel candidate therapeutic target.

61

62 Significance

63 Understanding how prostate cancer cells adapt to AR-targeted interventions is critical for identifying 64 novel drug targets to improve the clinical management of treatment-resistant disease. Our study revealed 65 an enzalutamide-induced epigenetic plasticity towards pro-survival signaling, and uncovered circadian 66 regulator ARNTL as an acquired vulnerability after AR inhibition, presenting a novel clinical lead for 67 therapeutic development.

68

69 Keywords

Prostate cancer, neoadjuvant clinical trial, Androgen Receptor, enzalutamide, epigenetic plasticity,
FOXA1, circadian rhythm, ARNTL

72

73 Introduction

Hormonal ablation is the mainstay treatment for patients with metastatic prostate cancer (PCa), ever
 since the direct critical connection between androgens and prostate tumor progression was first described

76 (1). The androgen receptor (AR) is the key driver of PCa development and progression, and multiple 77 therapeutic strategies have been developed over the years to effectively block the activity of this hormone-78 driven transcription factor. Upon androgen binding. AR associates with the chromatin at distal *cis*-regulatory 79 enhancer elements, where it regulates the expression of genes through long-range chromatin interactions 80 in three-dimensional genomic space (2,3). AR does not operate in isolation, but rather recruits a large 81 spectrum of coregulators and other transcription factors to promote expression of genes that drive cancer 82 cell proliferation (4). Critical AR interactors in the transcription complex are HOXB13 and FOXA1, which 83 are both upregulated in primary PCa (4-6) and demarcate enhancers that drive not only primary 84 tumorigenesis but also metastatic disease progression (7). Mechanistically, FOXA1 acts as a pioneer 85 factor, rendering the chromatin accessible for AR to bind (8-11). FOXA1 is frequently mutated in PCa (12-86 16) which was shown to alter its pioneering capacities, perturb luminal epithelial differentiation programs, 87 and promote tumor growth, further highlighting the critical role of FOXA1 in human prostate tumors (17,18). 88 Most patients are diagnosed with organ-confined PCa, which can potentially be cured through locoregional 89 therapies, such as surgery (radical prostatectomy), radiotherapy and/or brachytherapy (19). However, 90 approximately 30% of these patients experience a biochemical recurrence (BCR) - a rise in prostate-91 specific antigen (PSA) serum levels - indicating PCa relapse (20). At this stage of the disease, suppression 92 of androgen production is a commonly applied therapeutic intervention that can delay further cancer 93 progression for years (21,22). Nevertheless, the development of resistance to androgen deprivation is 94 inevitable, resulting in castration-resistant prostate cancer (CRPC) for which there is no cure (23). Most 95 CRPC tumors acquired molecular features that enable active AR signaling despite low circulating androgen 96 levels, a finding that led to the development of several highly effective AR-targeted therapies. Enzalutamide 97 (ENZ) is one of the most frequently used AR-targeting agents, which functions through a combined 98 mechanism of blocked AR nuclear import, diminished AR chromatin binding and decreased transcription 99 complex formation, effectively impairing AR-driven PCa growth (24), ENZ's potent anti-tumor activity has 100 been demonstrated in multiple clinical trials, which led to its FDA approval in various PCa disease stages 101 - from metastatic CRPC (25,26), to metastatic hormone-sensitive (27), and even non-metastatic CRPC 102 (28) – illustrating how AR-targeted therapies are being progressively introduced earlier in clinical practice. 103 A clinical benefit of ENZ monotherapy as a neoadjuvant treatment prior to prostatectomy for patients with

104 localized disease, has not been established. Although effective, resistance to AR pathway inhibition will 105 ultimately develop, and the management of advanced PCa with this acquired resistance remains a major 106 clinical challenge, especially since the underlying mechanisms are still not fully elucidated (29). Therefore, 107 furthering our understanding of how ENZ affects PCa biology may lead to the identification of acquired 108 cellular vulnerabilities that could be therapeutically exploited.

109 To study global drug-induced transcriptional and epigenetic plasticity in human prostate tumors and identify 110 cellular adaptation mechanisms to evade drug treatment, we designed a phase 2 clinical trial to perform 111 multi-omics studies in pre- and post-treatment samples from high-risk localized PCa patients, treated with 112 neoadjuvant ENZ monotherapy. We identified transcriptional reprogramming after treatment, with 113 deactivation of AR signaling and an activation of cell plasticity with neuroendocrine (NE)-like features upon 114 3 months of AR suppression. Post treatment, these tumors harbored a distinct set of 1,430 de novo 115 occupied FOXA1-positive cis-regulatory elements, positive for - yet independent of - AR activity, which are 116 dictated by circadian clock core regulator ARNTL to drive tumor cell proliferation instead. Using ARNTL 117 knockdown experiments we could further enhance ENZ sensitivity in cell line models, revealing an 118 unexpected biological interplay between hormonal resistance and circadian rhythm regulation, and 119 identifying a novel highly promising candidate drug target in the clinical management of primary high-risk 120 PCa.

121

122 Results

123 Neoadjuvant ENZ therapy for patients with high-risk localized PCa

124 To study how early ENZ intervention affects prostate tumor biology in a non-castrate environment, we 125 performed integrative multi-omics analyses as part of a single-arm, open-label phase 2 clinical trial: the 126 DARANA study (Dynamics of Androgen Receptor Genomics and Transcriptomics After Neoadjuvant 127 Androgen Ablation; ClinicalTrials.gov number, NCT03297385). In this trial, 56 men with primary high-risk 128 (Gleason score ≥ 7) PCa were enrolled (Fig. 1A). Patient demographics and disease characteristics are 129 summarized in Table 1, and clinical outcomes of this study are discussed in Supplementary Data 130 (Supplementary Fig. S1: Supplementary Table S1). Prior to ENZ therapy, magnetic resonance imaging 131 (MRI)-guided core needle tumor biopsies were taken - hereafter referred to as the pre-treatment setting.

132 Subsequently, patients received neoadjuvant ENZ treatment (160 mg/day) without androgen deprivation 133 therapy for three months, followed by robotic-assisted laparoscopic prostatectomy. Based on baseline MRI 134 information and palpation, additional tumor-targeted core needle biopsies were taken ex vivo – representing 135 the post-treatment setting. This pre- and post-treatment sampling allowed us to study the epigenetic, 136 genomic, transcriptomic and proteomic effects of neoadjuvant ENZ therapy in individual patients (Fig. 1A). 137 We generated chromatin immunoprecipitation (ChIP-seq) profiles of the prostate cancer drivers AR and 138 FOXA1, as well as the histone modification H3K27ac before and after ENZ treatment, and integrated these 139 cistromic findings with pre- and post-treatment gene expression (RNA-seq), copy number (CNV-seq) and 140 immunohistochemistry (IHC) data from the same tumors. Stringent quality control (QC) analyses were 141 performed on all data streams (Supplementary Fig. S1B), and the following number of samples passed 142 all QC measures (Fig. 1B): AR ChIP-seq (pre: n=10; post: n=12), FOXA1 ChIP-seq (pre: n=17; post: n=17), 143 H3K27ac ChIP-seq (pre: n=24; post: n=23), CNV-seq (pre: n=24; post: n=24), RNA-seq (pre: n=42; post: 144 n=52) and IHC (post: n=51).

145 Collectively, we performed integrative multi-omics analyses as part of a clinical trial that enabled us to 146 examine ENZ-induced oncogenomic changes to identify early epigenetic steps in treatment response, but 147 also therapy-induced resistance.

148

149 Characterization of tissue ChIP-seq data

150 To assess how neoadjuvant ENZ treatment affects the *cis*-regulatory landscape in primary PCa, we 151 generated human tumor ChIP-seg profiles for the transcription factors AR and FOXA1, along with the active 152 enhancer/promoter histone mark H3K27ac before and after neoadjuvant intervention. ChIP-seq quality 153 metrics are summarized in Supplementary Fig. S2 and Supplementary Table S2. Visual inspection at 154 known AR target genes showed high-quality data for all ChIP-factors in both clinical settings (Fig. 2A). On 155 a genome-wide scale, the H3K27ac ChIP-seg profiles were highly distinct from the transcription factors 156 (TFs) and divided the samples into two main clusters irrespective of their treatment status (Fig. 2B and 157 2C). Notably, AR and FOXA1 ChIP-seq datasets were intermingled in the clustering analysis, suggesting 158 largely comparable binding profiles which is in line with FOXA1's role as a canonical AR pioneer factor 159 (Supplementary Fig. S3) (5,30). As described previously (31), highest Pearson correlation was found

160 between H3K27ac samples, indicating comparable histone acetylation profiles among primary PCa 161 samples (Fig. 2B; Supplementary Fig. S3). Much greater heterogeneity in chromatin binding was 162 observed for the TFs AR and FOXA1, which is further supported by the steep decrease in the number of 163 overlapping AR and FOXA1 peaks with increasing number of samples compared to H3K27ac (Fig. 2D; 164 Supplementary Fig. S3). In order to maintain the high-confidence peaks that have been reproducibly 165 identified in multiple patients without losing too much binding site heterogeneity between samples, we 166 decided to generate consensus peaksets. To this end, we only considered binding sites that were present 167 in at least 3 out of 22 AR samples, 7 out of 34 FOXA1 samples and 13 out of 47 H3K27ac samples, which 168 corresponds to ~25% of all binding sites identified for each factor (Fig. 2D). Genomic distribution analyses 169 of these consensus sites revealed distinct enrichments for annotated genomic regions: While AR and 170 FOXA1 were almost exclusively found at intronic and distal intergenic regions, H3K27ac peaks were also 171 enriched at promoters (Fig. 2E), which is in line with previously published genomic distributions of AR 172 (5,31), FOXA1 (5,9), and H3K27ac (31,32). In addition, motif enrichment analyses at AR and FOXA1 173 consensus peaks identified, as expected, androgen and Forkhead response elements among the top-174 ranked motifs, respectively (Fig. 2F).

Taken together, we generated multiple high-quality tissue ChIP-seq data streams that now allowed us tostudy ENZ-induced changes in primary PCa patients.

177

178 ENZ treatment enriches for newly acquired FOXA1-bound regulatory regions

179 To identify ENZ-induced TF reprogramming and epigenetic changes, we performed differential binding 180 analyses comparing the pre- and post-treatment tissue ChIP-seq samples. Therefore, we first ran 181 occupancy-based unsupervised principal component analyses (PCA) to detect whether ENZ treatment led 182 to differences in TF chromatin binding. While the sample size of the AR ChIP-seq data stream was not 183 sufficient to observe significant differences in peak occupancy pre-versus post-treatment (Supplementary 184 Fig. S4A), the FOXA1 data did show such differences, with a clear separation of pre- and post-treatment 185 FOXA1 samples in the second principal component (Fig. 3A). Subsequent supervised analysis (pre vs. 186 post) revealed a total of 1,905 genomic regions (475 pre-enriched, 1,430 post-enriched; Supplementary 187 **Table S3**) that showed significant differential FOXA1 binding between both clinical settings (FDR < 0.05;

Fig. 3B and 3C; Supplementary Fig. S4B and S4C). Further characterization of these differential FOXA1 regions showed that both sets of binding sites were still preferentially located in intronic and distal intergenic regions (with a slight enrichment for promoters at the post-enriched sites; Supplementary Fig. S4D). In addition, Forkhead domain family motifs were the top enriched motifs at both pre- and post-enriched sites, illustrating that treatment does not alter FOXA1 motif preference and still occupies canonical FOXA1 binding sites (Supplementary Fig. S4E).

194 To examine whether structural variations are underlying these differential FOXA1 binding events, we 195 performed CNV-seq on the same tumor specimens and then projected onto the differential FOXA1 196 cistromics the structural copy-number data. These analyses revealed a comparable level of CNV at pre-197 and post-treatment enriched FOXA1 sites before and after ENZ treatment, with an overall trend towards 198 less CNV upon treatment (Supplementary Fig. S5A-S5C). However, in none of the matched sample pairs 199 (pre and post CNV-seq, and FOXA1 ChIP-seq; n=15) a strong correlation between copy number difference 200 and ChIP-seg signal difference was observed (R = 0.11; Supplementary Fig. S5D). In total, at only 44 out 201 of 1,905 differential FOXA1 binding sites (< 2.5%), we observed copy number differences between post-202 and pre-treatment samples that could potentially explain binding site occupancy in 3 or more patients, 203 indicating that the vast majority of these differential binding events is based on treatment-induced 204 transcription factor reprogramming, rather than structural variation (Supplementary Fig. S5E).

205 As FOXA1 dictates AR chromatin binding capacity (5), epigenetic plasticity of FOXA1 induced by treatment 206 may be associated with alterations in the AR cistrome. To assess this, and to explore the epigenetic 207 landscape surrounding the differentially bound FOXA1 regions, we compared the ChIP-seg signal of all 208 three factors (AR, FOXA1, H3K27ac) at differential (pre- / post-enriched) and consensus (shared by \geq 30 209 patients; n=338) FOXA1 sites before and after ENZ therapy. While the FOXA1 ChIP-seq signal was highest 210 at consensus binding sites, the pre- and post-treatment enriched regions followed the expected trend and 211 showed significantly higher signal in the corresponding settings (Fig. 3D). Notably, we also observed less 212 binding of FOXA1 to consensus sites when treated with ENZ, although the differences are much milder 213 compared to the effects seen at pre-enriched FOXA1 sites ($P_{adj} = 3.62 \times 10^{-22}$ at consensus vs. 3.76 x 10⁻²² 214 ¹³⁰ at pre-enriched sites, Mann-Whitney U-test; Fig. 3D; Supplementary Fig. S6A). This could possibly be 215 explained by decreased FOXA1 gene expression levels upon ENZ treatment (Supplementary Fig. S6B).

216 The AR ChIP-seq signal followed the same patterns as observed for FOXA1, suggesting that relocated 217 FOXA1 upon treatment functionally drives alterations in the AR cistrome (Fig. 3D). Unexpectedly, the pre-218 enriched FOXA1 sites were completely devoid of any H3K27ac signal in both pre- and post-treatment 219 samples, while the post-enriched counterparts were positive for this active enhancer/promoter mark with a 220 significant increase post-ENZ (P_{adj} = 5.59 x 10⁻⁴, Mann-Whitney U-test; Fig. 3D; Supplementary Fig. S6C 221 and S6D), suggesting that pre-ENZ FOXA1 sites are inactive. To validate these observations in an 222 independent cohort, we analyzed previously published AR (n=87), H3K27ac (n=92) and H3K27me3 (n=76) 223 ChIP-seq data from a cohort of 100 primary treatment-naïve PCa samples (31). Supporting our previous 224 analyses, the vast majority of post-enriched FOXA1 sites were H3K27ac-positive and their histone 225 acetylation status positively correlated with AR binding (R = 0.78) (Fig. 3E; Supplementary Fig. S6E). The 226 pre-enriched FOXA1 sites, however, were again H3K27ac-negative, while the repressive histone 227 modification H3K27me3 was present, which further points towards an inactive epigenetic state of these 228 regulatory regions (Fig. 3E).

229 Recently, we reported that prostate cancers can reactivate developmental programs during metastatic 230 progression (7). These sentinel enhancers appeared to be premarked by FOXA1 from prostate gland 231 development, and albeit inactive in normal and primary tumor specimens, the sites get reactivated by AR 232 during metastatic outgrowth. Given the inactivity of the pre-enriched FOXA1 sites, we hypothesized that 233 FOXA1 might be decommissioned at such developmental enhancers prior to hormonal intervention. To test 234 this, we overlapped the differential FOXA1 binding sites with the metastasis-specific AR binding sites (met-235 ARBS; n=17,655), which revealed a strong enrichment for these developmental regulatory elements at pre-236 treatment FOXA1 sites ($P = 2.13 \times 10^{-16}$, Fisher's exact test; Supplementary Fig. S6F). But are the inactive 237 pre-enriched FOXA1 sites solely epigenetically suppressed, or are these regions intrinsically incapable of 238 being active in this cellular context? To address this guestion and to further elucidate the role of AR at these 239 differentially bound FOXA1 sites, we integrated our tissue ChIP-seg findings with previously identified 240 tumor-specific AR binding sites (n=3,230) (5) that were functionally characterized using Self-Transcribing 241 Active Regulatory Regions sequencing (STARR-seq), a massive parallel reporter assay to systematically 242 annotate intrinsic enhancer activity (33). With this, three distinct classes of AR binding sites (ARBS) were 243 identified (Supplementary Table S4): enhancers that were active regardless of AR stimulation

244 (constitutively active; n=465), ARBS with no significant enhancer activity (inactive; n=2,479) and inducible 245 AR enhancers that increase activity upon androgen treatment (inducible: n=286). Interestingly, we found 246 that post-treatment FOXA1 sites were enriched for constitutively active ARBS, which further supports the 247 high enhancer activity and H3K27ac positivity observed at these sites, but also illustrates that this activity 248 is constitutive and AR-independent (Fig. 3F). Consistent with our postulated inactivity of the pre-treatment 249 enriched FOXA1 sites, these regions overlapped highly significantly with inactive ARBS ($P = 8.60 \times 10^{-9}$, 250 Fisher's exact test), which implies that these DNA elements are intrinsically inactive and incapable to act 251 as functional enhancers, and possibly explains why these AR-bound sites did not show active regulatory 252 marks (Fig. 3E and 3F). As no enrichment of our differential FOXA1 sites was observed with inducible 253 ARBS (Pre-enriched: 4/475; Post-enriched: 2/1,430), these data further support a conclusion that AR itself 254 is not a driver at FOXA1 sites that are differentially occupied after ENZ exposure in patients.

Overall, these results suggest that prior to hormonal intervention, FOXA1 is decommissioned at inactive developmental enhancer elements, which based on their primary DNA sequence are intrinsically incapable of being active – at least in the tested hormone-sensitive disease setting. However, upon ENZ treatment, FOXA1 gets reprogrammed to highly active *cis*-regulatory regions, which act in an AR-independent manner.

259

260 Transcriptional rewiring upon neoadjuvant ENZ

261 Having assessed the cistromic and epigenomic changes in response to neoadjuvant ENZ, we next 262 determined how transcriptional programs were affected by this hormonal intervention. Principal component 263 analysis (PCA) across both treatment states revealed that three months of ENZ therapy has a major effect 264 on global gene expression profiles (Fig. 4A). Subsequently, we performed differential gene expression 265 analysis, in which we compared pre- and post-treatment RNA-seq samples. Gene set enrichment analysis 266 (GSEA) showed that AR signaling, along with mitosis and MYC signals, was strongly decreased upon 267 treatment (Fig. 4B and C: Supplementary Fig. S7A). Since ENZ blocks the AR signaling axis, we analyzed 268 the androgen-response pathway in more detail, which revealed a strong downregulation of AR target genes 269 in almost every patient (Fig. 4D). In contrast to this, TNF α signaling, IFN- γ response and epithelial-270 mesenchymal transition (EMT) signals were most upregulated (Fig. 4B; Supplementary Fig. S7B).

271 Previously, we identified three distinct subtypes of primary treatment-naïve PCa (31), which we named 272 Cluster 1-3 (Cl1-3). While Cl1 and Cl2 were mainly dominated by their ERG fusion status - with Cl1 273 expressing high ERG levels (ERG fusion-positive) and Cl2 expressing low ERG levels (ERG fusion-274 negative) - Cl3 was enriched for neuroendocrine (NE)-like features, including low AR activity and a high 275 NE gene expression score. To assess the impact of neoadjuvant ENZ therapy on these PCa subtypes, we 276 performed unsupervised hierarchical clustering in the pre- and post-treatment setting using the originally 277 identified top 100 most differentially expressed genes per cluster. Prior to hormonal intervention, we could 278 robustly assign the samples into all three clusters (Cl 1: n=23, Cl2: n=11, Cl3: n=8) with highly comparable 279 distributions as we previously reported in another cohort of patients (31) (Supplementary Fig. S8A). Our 280 pre- and post-treatment sampling now allowed us to investigate how individual tumors were affected by 281 neoadjuvant therapy. This revealed that three months of ENZ therapy pushed almost all of the tumors 282 towards our NE-like cluster 3 (Fig. 4E; Supplementary Fig. S8B). To assure that the observed effects are 283 not solely driven by the treatment-induced reduction in AR activity (Fig. 4C and 4D), we used a well-284 established neuroendocrine PCa (NEPC) signature (34) to calculate gene expression fold changes pre-vs. 285 post-ENZ, which confirmed an induction of NE-like signaling upon treatment (Fig. 4F).

286 Collectively, these results demonstrate that three months of neoadjuvant ENZ therapy not only uniformly 287 diminishes AR signaling, but also pushes practically all of our primary PCa samples towards a NE-like gene 288 expression state independently of their original subtype.

289

290 Post-treatment FOXA1 sites drive pro-survival gene programs, dictated by circadian clock 291 component ARNTL

Having examined the global cistromic and transcriptomic changes upon ENZ therapy, we next characterized the biological consequences of the observed FOXA1 reprogramming using integrative analyses. We hypothesized that the newly acquired FOXA1 sites would be driving expression of genes associated with tumor cell survival programs. Using H3K27ac HiChIP data generated in LNCaP cells (35), pre- and post-treatment FOXA1 sites were coupled to their corresponding gene promoters (**Supplementary Table S5**). Subsequently, genome-wide CRISPR knockout screen data from Project Achilles (DepMap 20Q1 Public; VCaP) were used to identify those genes essential for prostate cancer cell proliferation

299 (36,37). While genes associated with pre-treatment FOXA1 sites were not enriched for essentiality as 300 compared to all other CRISPR-targeted genes in the library, genes under control of post-treatment FOXA1 301 sites showed a significant enrichment ($P = 7.54 \times 10^{-12}$, Welch's t-test) for critical drivers of tumor cell 302 proliferation (**Fig. 5A**).

303 Based on our STARR-seg and RNA-seg data, we concluded that AR is likely not driving enhancer activity 304 at post-treatment FOXA1 sites (Fig. 3F; Fig. 4C and 4D). Therefore, we sought to identify transcription 305 factors involved in the activation of these regulatory regions that are selectively occupied by FOXA1 306 following treatment. To this end, we overlaid the genomic coordinates of the post-treatment enriched 307 FOXA1 binding sites with those identified in publicly available ChIP-seq datasets (n= 13,976) as part of the 308 Cistrome DB transcription factor ChIP-seg sample collection (38,39). Besides FOXA1 and AR, which were 309 expected to bind at these regions (Fig. 3D), we also identified the glucocorticoid receptor (encoded by the 310 NR3C1 gene), which has previously been described to be upregulated upon antiandrogen treatment and 311 able to drive the expression of a subset of AR-responsive genes, conferring resistance to AR blockade (40-312 42). Unexpectedly, the second most enriched transcription factor after FOXA1 was circadian rhythm core 313 component ARNTL (Aryl Hydrocarbon Receptor Nuclear Translocator Like; also known as BMAL1) which 314 has not previously been implicated in PCa biology (Fig. 5B). Interestingly, ARNTL transcript levels were 315 upregulated upon ENZ treatment ($P = 6.4 \times 10^{-3}$, Mann-Whitney U-test; Fig. 5C), which was accompanied 316 by increased H3K27ac ChIP-seq signals at the ARNTL locus (Supplementary Fig. S9A). Consistent with 317 this, tissue microarray IHC analysis also revealed elevated ARNTL protein levels after treatment when 318 comparing the prostatectomy specimens post-ENZ with those of matched untreated control patients (P =319 6.89 x 10⁻¹⁹, Fisher's exact test; Fig. 5D). To assess whether ARNTL levels are also associated with patient 320 outcome, we compared the average ARNTL gene expression of patients that did not experience a BCR 321 (responders, n=29) with those that experienced an early BCR within \leq 6 months post-surgery (non-322 responders. n=8; Supplementary Table S1). While pre-treatment ARNTL levels were not significantly 323 different between ENZ responders and non-responders, high ARNTL levels after treatment were associated 324 with poor clinical outcome ($P = 4.79 \times 10^{-3}$, Mann-Whitney U-test; Fig. 5E). Notably, the CLOCK protein, 325 which forms a heterodimer with ARNTL to activate transcription of core clock genes, didn't show differential 326 expression upon ENZ treatment (Supplementary Fig. S9B) and is also not associated with clinical

327 outcome (**Supplementary Fig. S9C**), hinting towards a treatment-induced role of ARNTL that is 328 independent of its canonical function in the circadian machinery.

329 Taken together, these data suggest that the circadian clock regulator ARNTL may be functionally involved

- in ENZ resistance by driving tumor cell proliferation processes.
- 331

332 Acquired ARNTL dependency in ENZ-resistant PCa cells

To further investigate the relevance of ARNTL as a transcriptional driver at post-treatment FOXA1 sites, we performed *in vitro* validation experiments. To this end, we used hormone-sensitive LNCaP PCa cells, which we either cultured in full medium alone (Pre^{LNCaP}) or with ENZ for 48 hours (Post^{LNCaP}), mimicking our clinical trial setting (**Fig. 6A**). Based on the acquisition of NE-like gene expression profiles post-ENZ (**Fig. 4E and 4F**), we also included the ENZ-resistant LNCaP-42D model (43) that possesses NE-features (Res^{LNCaP-42D}; **Fig. 6A**), allowing us to further validate our patient-derived findings in cell lines recapitulating the transcriptional features of post-treatment clinical specimens.

We performed FOXA1 ChIP-seq experiments in all three cell line conditions (**Supplementary Fig. S10A-S10D**; **Supplementary Table S6**), which revealed highly similar FOXA1 chromatin binding dynamics as observed in our clinical samples: While the pre-enriched FOXA1 sites identified *in vivo* showed less binding upon treatment, we observed that merely 48 h of ENZ exposure was sufficient to strongly induce binding at post-enriched sites, which was further increased in the long-term exposed, treatment-resistant LNCaP-42D cell line (**Fig. 6B**; **Supplementary Fig. S10E**). Similarly, genome-wide correlation analyses indicated that short-term ENZ treatment in cell lines induced FOXA1 reprogramming to regions that are FOXA1-

bound in treatment-resistant but not in treatment-naïve cells (Supplementary Fig. S10F and S10G).

Having shown that differential FOXA1 chromatin binding in tumors could be recapitulated *in vitro*, we next sought to further assess the role of ARNTL in these pre-clinical models. Therefore, we first confirmed that treatment with ENZ increased ARNTL protein levels in LNCaP and LNCaP-42D models (**Fig. 6C**), recapitulating the clinical observations (**Fig. 5C and 5D**). Since cistromic ARNTL profiling has to date not been reported in PCa models, we generated ARNTL ChIP-seq data (**Supplementary Fig. S11A-S11D**) to validate its binding at post-treatment FOXA1 sites. Interestingly, while we already observed ARNTL binding to these regulatory regions in the pre-treatment setting, this was strongly enhanced upon ENZ exposure

355 (Fig. 6D; Supplementary Fig. S11E). To identify functional differences in ARNTL cistromes induced upon 356 treatment, we overlapped the ARNTL peaks identified in all tested cell line conditions, which revealed a 357 massive cistromic reprogramming upon ENZ treatment (Fig. 6E: Supplementary Fig. S11F and S11G). 358 Notably, ~70% of ENZ-gained ARNTL peaks (n=1,752) in LNCaP cells were captured by the ARNTL 359 cistrome in treatment-resistant cells. Subsequent pathway over-representation analyses of genes coupled to these Post^{LNCaP}-Res^{LNCaP-42D}-shared ARNTL binding sites revealed a treatment-induced enrichment for 360 361 gene sets implicated in cell cycle progression and cell division, further supporting a possible functional 362 involvement of ARNTL in sustaining tumor cell proliferation when AR is blocked by ENZ (Fig. 6E). To 363 challenge this hypothesis, we assessed whether ARNTL-knockdown affects the viability of hormone-364 sensitive and in particular of long-term ENZ-exposed cell lines. While ARNTL-targeting had minimal effect 365 on LNCaP cell proliferation (with or without ENZ), ARNTL knockdown significantly suppressed cell growth 366 of ENZ-resistant LNCaP-42D cells in the absence (P = 0.031, two-way ANOVA) and even more so in the 367 presence of ENZ ($P = 7 \times 10^{-4}$, two-way ANOVA), indicating that targeting ARNTL also partially restores 368 ENZ-sensitivity in this treatment-resistant cell line model (Fig. 6F).

Overall, these data confirm the ENZ-induced FOXA1 reprogramming as observed in PCa patients upon
 neoadjuvant antiandrogen therapy, and revealed an acquired dependency on circadian rhythm regulator
 ARNTL to drive tumor cell growth – positioning ARNTL as a highly promising new drug target in combination
 with ENZ for the treatment of high-risk PCa.

373

374 Discussion

In medicine, the evolutionary selection pressure as imposed by drug treatment has been a well-known clinical challenge, ever since the first antibiotics were discovered in the early 20th century. Also in oncology, clear escape mechanisms for both targeted therapeutics and systemic treatments are known for many years, involving *ESR1* mutations in metastatic breast cancer (44), *EGFR* mutations in lung cancer (45), *KRAS* mutations in metastatic colorectal cancer (46), but also somatic amplification of the *AR* locus and/or an upstream *AR* enhancer in castration-resistant PCa (47,48). Apart from genetic alterations, also epigenetic rewiring (7,41) or transdifferentiation are reported as mechanisms of resistance, including

treatment-emergent neuroendocrine (NE) prostate cancers that occur as an adaptive response under the
 pressure of prolonged AR-targeted therapy (49,50).

Our unique clinical trial design with paired pre- and post-treatment biopsies of high-risk primary PCa treated with ENZ monotherapy, allowed us to unravel global ENZ-induced alterations in gene regulation. We report that large-scale treatment-induced dedifferentiation in PCa may be a gradual process, of which the early signs are identified on transcriptomic level within the first months of treatment onset. While complete adenocarcinoma-NE transdifferentiation was not observed in any of our samples, cellular plasticity characterized by transcriptomic features of NE disease may not only be present prior to treatment (31), but also become enriched upon short-term endocrine treatment exposure.

391 In PCa development (5,51) and progression (7), AR has been reported to expose substantial plasticity in 392 its enhancer repertoire, and we now illustrate this is also the case in primary disease upon short-term 393 treatment. Besides AR, FOXA1 is considered a master transcription factor and critical prostate lineage 394 specific regulator acting in PCa, that upon overexpression during tumorigenesis gives rise to a tumor-395 specific AR cistrome. Also in NEPC, FOXA1 cistromes are reprogrammed (52), which indicates a direct 396 AR-independent role of FOXA1 in PCa progression. Our study confirms these observations and shows that, 397 while co-occupied by AR, the pre- and post-ENZ enriched FOXA1 sites appeared indifferent to AR 398 signaling.

The functional implications of the pre-treatment FOXA1 sites remain unclear, as those regions were inactive, both in primary tissues as well as in reporter assays. A subset of these *cis*-regulatory elements demarcates developmental epigenomic programs, that we previously reported as being occupied by FOXA1 from prostate development to tumorigenesis and metastatic progression (7), whereas others may be relevant for different physiological processes.

The treatment-induced cistromic repositioning of FOXA1 initiated a thus far unknown transcriptional rewiring, in which ARNTL, a classical circadian rhythm regulator and dimerization partner of CLOCK, compensates for AR inhibition and becomes essential to rescue cellular proliferation signals. Recently, it has been reported that CRY1 – a transcriptional coregulator of ARNTL – is AR-regulated in PCa, and modulates DNA repair processes in a circadian manner (53). The current data illustrate that circadian rhythm may have a potential impact on drug response, as most clock components are regulated on

410 transcriptional level, in time. Our data now show that AR blockade forces tumor cells to adapt epigenetically, 411 upon which these cells – over time – become dependent on ARNTL as a transcriptional regulator of 412 proliferation processes. This acquired cellular vulnerability appears to be dependent on whether or not AR 413 activity is inhibited and cells have had time to achieve full epigenetic reprogramming, explaining the limited 414 effect of ARNTL knockdown in hormone-sensitive PCa cells, as compared to the long-term ENZ-exposed 415 treatment-resistant model.

416 With the identification of ARNTL as a rescue mechanism for tumor cells to evade AR blockade, the next 417 question presents whether ARNTL could serve as a novel therapeutic target, which should be further 418 pursued in future drug development and clinical research. Being critically relevant for circadian rhythm 419 regulation, it would be imperative to balance ARNTL targeting in relation to any adverse side-effects. 420 Additionally, we demonstrate that the surprisingly dynamic enhancer repertoire of FOXA1 is not only critical 421 in prostate tumorigenesis (5) and neuroendocrine differentiation (52), but also appears crucial in evading 422 AR therapy-induced growth inhibition, further supporting the rationale to intensify efforts in targeting this 423 highly tissue-selective, yet critical transcriptional regulator, directly or indirectly (54).

424

425 Methods

426 Study design

427 Primary PCa tissues before and after enzalutamide (ENZ) treatment were acquired as part of the phase 428 2, prospective, single-arm DARANA study (ClinicalTrials.gov #NCT03297385) at the Netherlands Cancer 429 Institute Antoni van Leeuwenhoek hospital. To allow sample size calculation, we performed a survey into 430 the surgical margins of 1492 in-house prostatectomy specimens (Gleason \geq 7), not treated with 431 antihormonal therapy prior to surgery, which revealed 34% not-radical resections. Earlier randomized 432 studies on neoadjuvant androgen ablation showed reductions in positive surgical margin rate of at least 433 50% (55-57). To detect a reduction of positive surgical margins from 34% to 17% with a power of 80% and 434 an alpha set at 0.05, 55 patients needed to be included. Inclusion criteria were over 18 years of age, 435 Gleason ≥ 7 PCa and planned for prostatectomy. Prior to treatment a multi-parametric MRI scan was made 436 to identify tumors in the prostate (cT-stage) and pelvic lymph node metastasis (cN-stage). Patients were 437 treated with ENZ, once daily 160 mg P.O. without androgen deprivation therapy, for three months prior to

438 RALP and a pelvic lymph node dissection. 55 patients completed therapy without dose adjustments, while 439 one patient (DAR37) discontinued ENZ three weeks prematurely. The resection specimen was assessed 440 for prostate tumor stage (vpT-stage) and the pelvic lymph nodes (vpN-stage). Primary clinical outcome 441 measure was positive margins rate and secondary endpoints were differences in pre- and post-treatment 442 T (tumor) and N (lymph node) stage and median time to biochemical recurrence, defined as two consecutive 443 rises of serum PSA with a minimal level of ≥ 0.2 ng/mL. The trial was approved by the institutional review 444 board of the Netherlands Cancer Institute, informed consent was signed by all participants enrolled in the 445 study, and all research was carried out in accordance with relevant guidelines and regulations.

446

447 Pre- and post-treatment sampling

448 Prior to ENZ intervention, 4 pre-operative MRI-guided 18-gauche core needle tumor biopsies were taken 449 per patient. Directly after prostatectomy, 8 additional tumor-targeted core needle biopsies (4x 14 gauche, 450 4x 5 mm) were taken from prostatectomy specimens ex vivo, using previous MRI information and palpation. 451 Biopsy and prostatectomy specimens were fresh frozen (FF) or formalin-fixed paraffin-embedded (FFPE) 452 for ChIP-seg and CNV-seg, or RNA-seg and immunohistochemistry analyses, respectively. Prior to ChIP-453 seq experiments, FF material was cut in 30 µm sections, while FFPE material was cut in 10 µm sections 454 prior to RNA extraction. Tissue sections were examined pathologically for tumor cell content and only 455 samples with a tumor cell percentage of \geq 50% were used for further downstream analyses.

456

457 ChIP-seq

458 Chromatin immunoprecipitations on PCa tissue specimens and cell line models were performed as 459 previously described (58). In brief, cryosectioned tissue samples were double-crosslinked in solution A 460 (50 mM HEPES-KOH, 100 mM NaCl, 1 mM EDTA, 0.5 mM EGTA), first supplemented with 2 mM 461 disuccinimidyl glutarate (DSG; CovaChem) for 25 min at room temperature. Then, 1% formaldehyde 462 (Merck) was added for 20 min and subsequently guenched with a surplus of 2.5 M glycine. Cell lines were 463 crosslinked using single-agent fixation. Therefore, 1% formaldehyde was added to the cell culture medium 464 and incubated at room temperature for 10 min, followed by glycine-guenching as described above. Tissue 465 and cell line samples were lysed as described (59) and sonicated for at least 10 cycles (30 sec on; 30 sec

off) using a PicoBioruptor (Diagenode). For each ChIP, 5 μg of antibody were conjugated to 50 μL magnetic
protein A or G beads (10008D or 10009D, Thermo Fisher Scientific). The following antibodies were used:
AR (06-680, Merck Millipore), FOXA1 (ab5089, Abcam), H3K27ac (39133, Active Motif), and ARNTL
(ab93806, Abcam).

470 Immunoprecipitated DNA was processed for library preparation using a KAPA library preparation kit 471 (KK8234, Roche) and generated libraries were sequenced on the Illumina HiSeg2500 platform using the 472 single end protocol with a read length of 65-bp, and aligned to the human reference genome hg19 using 473 Burrows-Wheeler Aligner (v0.5.10) (60). Reads were filtered based on mapping quality (MAPQ \geq 20) and 474 duplicate reads were removed. Peak calling over input controls (per tissue sample or cell line) was 475 performed using MACS2 (v2.1.1) and Dfilter (v1.6) for tissues, and MACS2 (v2.1.2) for cell lines (61,62). 476 For tissue samples, only the peaks shared by both peak callers were used for downstream analyses. 477 DeepTools (v2.5.3) was used to calculated read counts in peaks (FRiP) (63). Read counts and the number 478 of aligned reads, as well as normalized strand coefficient (NSC) and relative strand correlation (RSC), which 479 were calculated using phantompeaktools (v1.10.1) (64), are shown in Supplementary Table S2 for tissue 480 ChIP-seq data and Supplementary Table S6 for cell line ChIP-seq data. Tissue ChIP-seq samples that 481 passed the following quality control measures were included in the final analyses; tumor cell percentage \geq 482 50%, ChIP-qPCR enrichment, and more than 100 peaks called (Supplementary Fig. S1B).

For visualization of cell line ChIP-seq data, an average enrichment signal was generated by merging
mapped reads of replicate samples using SAMtools (v1.10-3) (65).

485 Genome browser snapshots, tornado and average density plots were generated using EaSeg (v1.101) (66). 486 Genomic distribution and motif enrichment analyses were performed using the CEAS and the SeqPos motif 487 tools on Galaxy Cistrome (67), respectively. Cistrome Toolkit was used to probe which TFs and chromatin 488 regulators have a significant binding overlap with the differential FOXA1 peak sets (39). For this, genomic 489 coordinates of high-confidence binding sites (FC \geq 1.2) were converted between assemblies (from hg19 to 490 hq38), using the UCSC genome browser liftOver tool (68). The DiffBind R package (v2.10) was used to 491 generate correlation heatmaps and PCA plots based on occupancy, to perform differential binding analyses 492 using a false discovery rate (FDR) < 0.05, and to generate consensus peaklists (69).

493 ChIP-seq signal of various datasets (FOXA1, AR and H3K27ac from this study; AR, H3K27ac and 494 H3K27me3 from a previously reported study (31)) at differential and consensus FOXA1 sites was 495 investigated by counting mapped reads in FOXA1 peak regions using bedtools multicov (v2.27.1) (70). 496 Readcounts were subsequently z-transformed and visualized using the aheatmap function from the R 497 package NMF (v0.21.0) (71) with a color scheme from RColorBrewer (v1.1-2; https://CRAN.R-498 project.org/package=RColorBrewer). To determine significance in binding site occupancy differences 499 between pre- and post-treatment FOXA1 sites, median z-transformed readcounts were calculated per 500 sample and compared using a Mann-Whitney U-test. These median readcounts per sample were also used 501 to assess the correlation between ChIP-seq signals of AR, FOXA1 and H3K27ac at pre-enriched, post-502 enriched and consensus FOXA1 binding sites.

503 Bedtools intersect (v2.27.1) (70) was used to determine overlap of differential FOXA1 binding sites and 504 inactive, constitutively active and inducible AR-binding sites.

To assign FOXA1 and ARNTL binding regions to potential target genes, we overlapped differential FOXA1 binding sites with H3K27ac HiChIP data (35) using bedtools intersect. To assess whether or not genes coupled to FOXA1 binding sites were considered to be essential for the VCAP prostate cancer cell line, we used the DepMap (Broad 2020) 20Q1 Public gene effect dataset (36). Gene set overlaps between genes linked to ChIP-seq binding sites and the Molecular Signatures Database (v7.4) were computed using Gene Set Enrichment Analysis (GSEA) (72) with an FDR q-value cutoff \leq 0.05.

511

512 RNA-seq

Prior to RNA isolation, FFPE material was pathologically assessed. The expert pathologist scored tumor cell percentage and indicated most tumor-dense regions for isolation on a hematoxylin and eosin (H&E) slide. RNA and DNA from FFPE material were simultaneously isolated from 3-10 sections (depending on tumor size) of 10 µm using the AllPrep DNA/RNA FFPE isolation kit (80234, Qiagen) and the QIAcube according to the manufacturer's instructions. cDNA was synthesized from 250 ng RNA using SuperScript III Reverse Transcriptase (Invitrogen) with random hexamer primers.

519 For RNA-seq, strand-specific libraries were generated with the TruSeq RNA Exome kit (Illumina) and 520 sequenced on the Illumina HiSeq2500 platform using the single end protocol with a read length of 65-bp.

521 Sequencing data was aligned to the human reference genome hg38 using TopHat (v2.1.0 using bowtie 522 1.1.0) (73) and the number of reads per gene was measured with HTSeq count (v0.5.3) (74).

523 For QC purposes, total readcounts per sample were determined and hierarchical clustering based on the 524 Euclidean distance was applied. Samples with a readcount \geq 2 standard deviations below the mean of all

sample readcounts were removed, as well as samples that clustered in a separate branch.

526 Global gene expression differences between pre- and post-treatment samples passing QC were

527 determined using DESeq2 (v1.22.2) (75). Significance of expression level differences between pre- and

528 post-treatment samples was determined using a paired t-test.

529 Gene set enrichment was performed using pre-ranked GSEA (72) based on the Wald statistic provided by 530 DESeq2. For visualization purposes, the data were Z-transformed per gene. Heatmaps of gene expression 531 values were created using the aheatmap function from the R package NMF (v0.21.0) (71) with a color 532 scheme from RColorBrewer (v1.1-2; https://CRAN.R-project.org/package=RColorBrewer).

To assign samples to previously described PCa subtypes (31), the z-transformed expression levels of the top ~100 most differentially expressed genes (n=285) in each of the three clusters were investigated. Using these values, samples were clustered based on their Pearson correlation. The resulting tree was divided into 3 clusters, corresponding to the previously published PCa subtypes. Potential transitioning of samples from one cluster to another after treatment was visualized using a riverplot (v0.6; <u>https://CRAN.R-</u> <u>project.org/package=riverplot</u>).

To calculate fold changes of neuroendocrine scores upon treatment, expression of 70 neuroendocrine signature genes were obtained from castration-resistant neuroendocrine and prostate adenocarcinoma samples as published previously (34). The expression of 5 of the 70 neuroendocrine signature genes were not included in the analysis (KIAA0408, SOGA3, LRRC16B, ST8SIA3, SVOP) because the genes are not expressed in these samples. Expression fold changes between paired pre- and post-treatment samples were calculated (n=39) and concordance in gene expression differences (fold change sign) were measured using Pearson correlation.

546

547 CNV-seq

548 Low-coverage whole-genome samples (ChIP-seq inputs), sequenced single-end 65-bp on a HiSeq 2500 549 system were aligned to hg19 with Burrows-Wheeler Aligner backtrack algorithm (v0.5.10) (60). The 550 mappability per 20-kb on the genome, for a samples' reads, phred quality 37 and higher, was rated against 551 a similarly obtained mappability for all known and tiled 65-bp subsections of hg19. Sample counts were 552 corrected per bin for local GC effects using a non-linear loess fit of mappabilities over 0.8 on autosomes. 553 Reference values were scaled according to the slope of a linear fit, forced to intercept at the origin, of 554 reference mappabilities after GC correction. Ratios of corrected sample counts and reference values left 555 out bins with mappability below 0.2 or overlapping ENCODE blacklisted regions (76).

556 Copy number log ratios were smoothed and segmented using the R package DNACopy (v1.50.1; 557 https://bioconductor.org/packages/release/bioc/html/DNAcopy.html) with the parameters set to 558 alpha=0.0000000001, undo.SD=2, and undo.splits="sdundo". Bedtools intersect (v2.27.1) (70) was used 559 to determine overlap between copy number segments and differential FOXA1 binding sites. These data 560 were subsequently visualized using the aheatmap function from the R package NMF (v0.21.0) (71) with a 561 color scheme from RColorBrewer (v1.1-2; https://CRAN.R-project.org/package=RColorBrewer).

To correlate FOXA1 ChIP-seq signal with copy number status at differential FOXA1 sites, we employed the z-transformed FOXA1 ChIP-seq readcounts as described in the ChIP-seq section. The difference in transformed ChIP-seq readcounts and the difference in normalized segmented copy number data between matched post-treatment and pre-treatment samples was calculated for every patient. Subsequently, the Pearson correlation between these two sets of differences was calculated.

567 Immunohistochemistry

568 For immunohistochemistry (IHC) analysis, we matched our ENZ-treated patient cohort (n=51) in a 1:2 569 ratio to untreated control patients (not receiving ENZ prior to prostatectomy; n=110) based on 570 clinicopathological parameters (initial PSA, Gleason score, TNM stage, age) using the R package MatchIt 571 (v.4.1.0) (77).

Tissue microarrays (TMAs) were prepared containing 3 cores per FFPE tumor sample. Tumor-dense areas
in FFPE megablocks were marked by an expert pathologist on a H&E slide. Cores were drilled in a receptor
block using the TMA grandmaster (3D Histech/Sysmex). Next, cores were taken from the donor block and

placed in the receptor block using the manual tissue arrayer (4508-DM, Beecher instruments). The filled
receptor block was placed in a 70°C stove for 9 minutes and cooled overnight at RT.

577 Immunohistochemistry was applied to TMA slides using a BenchMark Ultra autostainer (Ventana Medical

578 Systems). In brief, paraffin sections were cut at 3 µm, heated at 75°C for 28 minutes and deparaffinized in

579 the instrument with EZ prep solution (Ventana Medical Systems). Heat-induced antigen retrieval was

580 carried out using Cell Conditioning 1 (CC1, Ventana Medical Systems) for 64 minutes at 95°C.

581 For ARNTL IHC, TMAs were stained with an anti-ARNTL antibody (ab230822, Abcam) for 60 minutes at

582 36°C using a 1:1000 antibody dilution. Bound antibody was detected using the OptiView DAB Detection Kit

583 (Ventana Medical Systems). Slides were counterstained with hematoxylin and bluing reagent (Ventana584 Medical Systems).

585 ARNTL staining intensity (weak, moderate, strong) in tumor cells was scored by an expert pathologist.

586 Tissues scored for at least two cores were analyzed and used for statistical analysis.

587

588 Cell lines and cell culture

589 LNCaP human PCa cell line was purchased from the American Type Culture Collection (ATCC). 590 Enzalutamide-resistant LNCaP-42D cells were described previously (43). LNCaP cells were maintained in 591 RPMI-1640 medium (Gibco, Thermo Fisher Scientific) supplemented with 10% FBS (Sigma-Aldrich). 592 LNCaP-42D cells were maintained in RPMI-1640 medium containing 10% FBS and 10 µM ENZ. Cell lines 593 were subjected to regular mycoplasma testing and all cell lines underwent authentication by short tandem 594 repeat profiling (Eurofins Genomics). For hormone stimulation with synthetic and rogen, cells were treated 595 with 10 nM R1881 (PerkinElmer) for 48 h. For in vitro AR blockade, cells were treated with 10 µM ENZ 596 (MedChemExpress) and harvested at the indicated time points.

597

598 Transient cell line transfections

599 Transient transfections of cell lines were performed according to the manufacturer's instructions using 600 Lipofectamine 2000 (Invitrogen) or Lipofectamine RNAiMAX (Invitrogen) for overexpression or siRNA 601 knockdown experiments, respectively. ARNTL containing expression plasmid was obtained from the

602 CCSB-Broad Lentiviral Expression Library. siRNA oligos targeting ARNTL (M-010261-00-0005), and the 603 non-targeting control (D-001206-14) were purchased from Dharmacon.

604

605 Western blotting

Total proteins were extracted from cells using Laemmli lysis buffer, supplemented with a complete protease inhibitor cocktail (Roche). Per sample, 40 µg of protein was resolved by SDS-PAGE (10%) and transferred on nitrocellulose membranes (Santa Cruz Biotechnology). The following antibodies were used for Western blot stainings: ARNTL (ab93806, Abcam), PSA (5365, Cell Signaling Technology), and ACTIN (MAB1501R, Merck Millipore). Blots were incubated overnight at 4°C with designated primary antibodies at 1:1000 (ARNTL, PSA) or 1:5000 (ACTIN) dilution, and visualized using the Odyssey system (Li-Cor Biosciences).

613

614 Cell viability assay

For cell viability assays, LNCaP or LNCaP-42D cells were seeded at 2×10^3 cells per well in 96-well plates (Greiner) ± 10 µM ENZ, and reverse transfected with 50 nM siRNA (Dharmacon) using Lipofectamine RNAiMAX (Invitrogen). Cell viability was assessed 7 days post-transfection using the CellTiter-Glo Luminescent Cell Viability Assay kit (Promega), as per the manufacturer's instructions.

619

620 Statistical analysis

621 For differential binding and differential gene expression analyses (pre- vs. post-ENZ), an FDR cutoff < 622 0.05 (P < 0.01) and $P_{adj} < 0.01$ was used, respectively. A Mann-Whitney U-test was used to determine 623 differences in region readcounts (adjusted for multiple testing using FDR) and differences in gene 624 expression levels before and after ENZ treatment. For peak set and gene set overlaps, Fisher's exact or 625 Welch Two Sample t-tests were applied. A Fisher's exact test was used to determine differences in IHC 626 staining intensity. Differences in cell viability were tested using a two-way ANOVA followed by Tukey's 627 multiple comparisons test (GraphPad Prism 9). Corresponding bar chart shows the mean with error bars 628 representing the SD of three independent experiments. All boxplots indicate the median (center line), upper-629 (75) and lower - (25) quartile range (box limits) and 1.5 x interquartile range (whiskers). Significance is

630 indicated as follows: ns, P > 0.05; *, P < 0.05; **, P < 0.01; ***, P < 0.001; ****, P < 0.0001. Further details 631 of statistical tests are provided in the figure legends.

632

633 Declaration of Potential Conflicts of Interest

634 W. Zwart, A.M. Bergman and H. van der Poel received research funding from Astellas Pharma B.V.

635 (Leiden, the Netherlands). No potential conflicts of interest were disclosed by the other authors.

636

637 Acknowledgements

The authors would like to thank the NKI-AVL Core Facility Molecular Pathology & Biobanking (CFMPB) for technical assistance, the NKI Genomics Core Facility (GCF) for next generation sequencing and bioinformatics support, and the NKI Research High Performance Computing facility (RHPC) for computational infrastructure. We also thank all Zwart/Bergman lab members for fruitful discussions and technical advice. Finally, we would like to thank all patients and clinical staff who were involved in the DARANA trial.

644

645 Grant Support

- 646 This work was supported by Movember (NKI01 to A.M. Bergman and W. Zwart), KWF Dutch Cancer
- 647 Society (10084 ALPE to A.M. Bergman and W. Zwart), KWF Dutch Cancer Society/Alpe d'HuZes Bas
- 648 Mulder Award (NKI 2014-6711 to W. Zwart), Netherlands Organization for Scientific Research (NWO-VIDI-
- 649 016.156.401 to W. Zwart), and Astellas Pharma (to W. Zwart, A.M. Bergman, H. van der Poel).
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869 Tables

870 Table 1: Characteristics of the DARANA cohort (n=56).

Table summarizing the patient baseline demographics, and pre- and post-treatment disease characteristics of the DARANA cohort.

- 872 Shown are age (years), initial Prostate-specific antigen (PSA) serum levels (ng/mL) and International Society of Urological Pathology
- 873 (ISUP) grade at diagnosis (with associated Gleason scores (GS)). In addition, T-stage (T) and Lymph node status (N) before (pre =
- at diagnosis) and after (post = at surgery) neoadjuvant ENZ therapy, as well as the surgical margin status of the prostatectomy
- 875 specimens are shown. Pre-treatment measures are based on histological evaluation of biopsy material and radiographic evaluation
- 876 (clinical grading; c), while post-treatment assessments are based on histological evaluations of prostatectomy specimens (pathological
- grading after neoadjuvant therapy; yp). Biochemical recurrence (BCR) was defined as a rise in PSA of ≥ 0.2 ng/mL. Mean time to
- 878 BCR (months) and time to last follow-up (months) are indicated. For continuous variables (age, baseline PSA, time to BCR, and time
- to last follow-up) the mean and 95% confidence interval (CI) are shown. For categorical variables (baseline ISUP, T-stage, N-status,
- surgical margins, BCR) the number of patients (no.) and percentages (%) are indicated.

DARANA cohort (n= 56)			
Age – years (95% CI)	67 (65-68)		
Baseline PSA level – ng/mL (95% CI)	12.8 (10.4-15.2)		
Baseline ISUP grade – no. (%)			
ISUP 1 (GS 3+3)	0 (0)		
ISUP 2 (GS 3+4)	16 (28)		
ISUP 3 (GS 4+3)	9 (16)		
ISUP 4 (GS 4+4, 3+5, 5+3)	20 (36)		
ISUP 5 (GS 4+5, 5+4, 5+5)	11 (20)		
T-stage (T) – no. (%)	Pre (cT)	Post (ypT)	
T1	1 (2)	1	
T2	25 (44)	20 (36)	
Т3	29 (52)	36 (64)	
T4	1 (2)	0 (0)	
Lymph node status (N) – no. (%)	Pre (cN)	Post (ypN)	
NO	53 (95)	39 (70)	
N1	3 (5)	17 (30)	
Surgical margins – no. (%)			
Negative	39 (70)		

Positive	17 (30)
Biochemical recurrence (BCR) – no. (%)	PSA ≥ 0.2 ng/mL
No	32 (57)
Yes	23 (41)
Unknown	1 (2)
Mean time to BCR – months (95% CI)	12 (8-15)
Mean time to last follow-up – months (95% CI)	37 (33-42)

881

883 **Figures**

Figure 1



884

885 Figure 1: Clinical trial design and omics data sample collection.

886	(A)	Study design of the DARANA trial (NCT03297385). Multi-omics profiling, consisting of (I) Androgen Receptor (AR) ChIP-
887		seq, (II) FOXA1 ChIP-seq, (III) H3K27ac ChIP-seq, (IV) DNA copy number sequencing (CNV-seq), (V) gene expression
888		profiling (RNA-seq) and (VI) immunohistochemistry (IHC) analysis, was performed on MRI-guided biopsy samples prior
889		to ENZ treatment (Pre) and tumor-target prostatectomy specimens after 3 months of neoadjuvant ENZ therapy (Post).
890	(B)	Overview of data availability and quality control analyses for each sample. Individual data streams are indicated
891		separately with ChIP-seq for AR (red), FOXA1 (blue), H3K27ac (green), CNV-seq, RNA-seq and IHC (all black). The
892		ENZ treatment status indicates the pre-treatment (top) and post-treatment samples (bottom) per omics dataset. Samples
893		not passing QC (light gray) were successfully applied for focused raw data analyses. Blank spots for ChIP-seq or CNV-
894		seq samples indicate that the fresh-frozen material didn't pass the tumor cell percentage cut-off of \geq 50%.

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Figure 2



897 Figure 2: Characterization of tissue ChIP-seq data streams.

898	(A)	Representative example snapshots of AR (red), FOXA1 (blue) and H3K27ac (green) ChIP-seq data for four genomic
899		loci in one patient. Pre- (light colors) and post-ENZ treatment (dark colors) is indicated. Y-axis indicates ChIP-seq
900		signal in fragments per kilobase per million reads mapped (FPKM).
901	(B)	Correlation heatmap based on peak occupancy. Clustering of the samples is based on all called peaks and
902		represents Pearson correlations between individual ChIP-seq samples. The column color bars indicate the ChIP-seq
903		factor (AR, FOXA1, H3K27ac) and treatment status (Pre, Post).
904	(C)	Principal component analysis (PCA) plot based on peak occupancy. Each dot represents a ChIP-seq sample that is
905		colored per factor.
906	(D)	Elbow plot depicting the peak overlap between ChIP-seq samples per factor. Shown is the percentage of overlapping
907		peaks with increasing number of samples. Consensus peaksets were designed by using a cut-off of peaks present
908		in at least 3 AR, 7 FOXA1, or 13 H3K27ac samples.
909	(E)	Pie charts showing the genomic distribution of AR (left), FOXA1 (middle) and H3K27ac (right) consensus peaks.
910	(F)	Word clouds show motif enrichment at AR (left) and FOXA1 (right) consensus sites. The font size represents the z-
911		score and colors correspond to transcription factor families.
912		





914	Figure 3: Di	ifferential FOXA1 binding upon ENZ treatment.
915	(A)	Principle component analysis (PCA) plot based on peak occupancy of FOXA1 ChIP-seq data. Color indicates pre-
916		treatment (light blue) and post-treatment (dark blue) FOXA1 samples.
917	(B)	Coverage heatmap depicting differential FOXA1 binding sites, selectively enriched in the pre-treatment (n=475) or post-
918		treatment (n=1,430) setting.
919	(C)	Representative example snapshots of FOXA1 ChIP-seq signal at two pre-enriched (left) and two post-enriched (right)
920		FOXA1 sites in one patient. Pre- (light blue) and post-ENZ treatment (dark blue) is indicated. Y-axis indicates ChIP-seq
921		signal in FPKM.
922	(D)	Boxplots indicating ChIP-seq signal (z-scaled readcounts) at pre-enriched (n=475), post-enriched (n=1,430) and
923		consensus FOXA1 peaks (shared by ≥ 30 patients; n=338) for FOXA1 (blue), AR (red), and H3K27ac (green) ChIP-seq
924		datasets before (Pre; light colors) and after (Post; dark colors) ENZ treatment. **, P < 0.01; ***, P < 0.001; ****, P <
925		0.0001 (Mann-Whitney U-test adjusted for multiple testing using FDR).
926	(E)	Coverage heatmap showing occupancy of differential (pre-/post-enriched) and consensus FOXA1 peaks in an external
927		ChIP-seq dataset consisting of 100 untreated primary tumors (31). Heatmap color indicates region read counts (z-score)
928		at pre-enriched, post-enriched and consensus FOXA1 sites (rows) in the AR (red), H3K27ac (green) and H3K27me3
929		(gray) ChIP-seq data streams (columns).
930	(F)	Bar chart representing the overlap between differential FOXA1 sites (pre-enriched or post-enriched) and constitutively
931		active (left) or inactive (right) AR binding sites (ARBS), based on STARR-seq. *, P < 0.05; ****, P < 0.0001 (Fisher's
932		exact test).
933		





935	Figure 4: No	eoadjuvant ENZ deactivates AR signaling and induces neuroendocrine (NE)-like gene expression signatures.
936	(A)	Principle component analysis (PCA) plot based on gene expression data. Color indicates pre-treatment (gray) and post-
937		treatment (black) samples. Ellipses are based on the 80% confidence interval.
938	(B)	Gene set enrichment analyses (GSEA) for Hallmark gene sets. Shown are the top differentially enriched pathways upon
939		ENZ treatment. Y-axis indicates the normalized enrichment score (NES).
940	(C)	Enrichment plot of the Hallmark Androgen Response pathway, for genes up- or downregulated after ENZ. Y-axis
941		indicates enrichment score (ES). GSEA statistics (FDR, ES, NES) are indicated.
942	(D)	Unsupervised hierarchical clustering of pre- and post-treatment RNA-seq samples based on the expression of AR-
943		responsive genes. Color scale indicates gene expression (z-score).
944	(E)	River plot showing state transitions between Clusters 1 (dark blue), Cluster 2 (green) and Cluster 3 (light blue) for paired
945		pre-treatment and post-treatment RNA-seq samples (n=39). Number of samples assigned to each cluster before and
946		after treatment as well as the hallmarks per cluster are indicated.
947	(F)	Waterfall plot depicting the Pearson correlation of neuroendocrine gene expression signature fold changes upon ENZ
948		treatment per patient. Colors indicate the patients cluster affiliations after treatment.
949		



950 951



952 (A) Boxplot showing DepMap (20Q1) genome-wide loss-of-function CRISPR screen data for VCaP PCa cells, separately
 953 analyzing the gene effect score of genes associated with post-enriched FOXA1 sites (top), pre-enriched FOXA1 sites
 954 (middle) or all other tested genes (bottom). Differential FOXA1 binding sites were coupled to their respective target genes
 955 using H3K27ac HiChIP data. Indicated as controls are PCa-relevant driver genes: oncogenes MYC, FOXA1, AR, TP53

956		and tumor suppressor PTEN. The recommended gene effect score cutoff of -0.5 is shown (dotted vertical line). ns, $P > 1$
957		0.05; ****, <i>P</i> < 0.0001 (Fisher's exact test).
958	(B)	Dot plot representing ranked GIGGLE similarity scores for transcriptional regulators identified at post-treatment FOXA1
959		sites. The top 20 factors identified are shown, and the 5 most enriched factors are labeled.
960	(C)	Boxplot showing normalized ARNTL gene expression before and after 3 months of neoadjuvant ENZ treatment. **, P <
961		0.01 (Mann-Whitney U-test).
962	(D)	Representative ARNTL immunohistochemistry (IHC) stainings (left) and quantification of ARNTL staining intensity (right)
963		in tissue microarrays consisting of prostatectomy specimens from untreated patients (not receiving neoadjuvant ENZ;
964		n=110) and DARANA patients post-ENZ (n=51). Scale bars, 100 μ m. ****, <i>P</i> < 0.0001 (Fisher's exact test).
965	(E)	Boxplots depicting normalized ARNTL gene expression in ENZ non-responders (biochemical recurrence (BCR) ≤ 6
966		months; n=8) and responders (no BCR; n=29) in the pre- (left) and post- (right) treatment setting separately. ns, P > 0.05;
967		**, <i>P</i> < 0.01 (Mann-Whitney U-test).
968		





969 970

Figure 6: Treatment-induced dependency on ARNTL in ENZ-resistant PCa cells.

- 971 (A) Experimental setup for *in vitro* validation experiments.
- (B) Tornado plots (left) and average density plots (right) visualizing FOXA1 ChIP-seq signal (in FPKM) at pre-enriched (top)
 and post-enriched (bottom) FOXA1 binding sites in untreated (Pre^{LNCaP}), short-term ENZ-treated (Post^{LNCaP}), and ENZ-

- 974 resistant NE-like LNCaP cells (Res^{LNCaP-42D}). Data are centered at differential FOXA1 peaks depicting a 5-kb (heatmaps)
 975 or 1-kb (density plots) window around the peak center. Heatmap color depicts the ChIP-seq signal compared to the
 976 untreated condition (Pre^{LNCaP}), with blue indicating lower peak intensity and orange indicating higher peak intensity.
 977 Average of two biological replicates is represented.
- 978 (C) Western blot showing ARNTL protein levels in LNCaP and LNCaP-42D cells following treatment with synthetic androgen
 979 (R1881) and/or ENZ (ENZ) for 48 h. ARNTL overexpression (in LNCaP-42D cells) as well as stainings for PSA and
 980 ACTIN are included as controls for antibody staining, hormonal treatment and protein loading, respectively. Images are
 981 representative of three independent experiments.
- (D) Tornado plots (left) and average density plot (right) visualizing ARNTL ChIP-seq signal (in FPKM) at post-enriched
 FOXA1 binding sites in untreated (Pre^{LNCaP}), short-term ENZ-treated (Post^{LNCaP}), and ENZ-resistant NE-like LNCaP cells
 (Res^{LNCaP-42D}). Data are centered at post-treatment FOXA1 peaks depicting a 5-kb (heatmaps) or 1-kb (density plots)
 window around the peak center. Heatmap color depicts the ChIP-seq signal compared to the untreated condition
 (Pre^{LNCaP}), with blue indicating lower peak intensity and orange indicating higher peak intensity. Average of two biological
 replicates is represented.
- (E) Venn diagram (top) indicating the overlap of ARNTL binding sites in all tested cell line conditions (Pre^{LNCaP}, Post^{LNCaP}, 989
 Res^{LNCaP-42D}). For each condition, only peaks present in both replicates were included. Gene ontology terms for ARNTL bound gene sets uniquely shared between Post^{LNCaP} and Res^{LNCaP-42D} conditions are presented below. Overlapping
 ARNTL binding sites (n=1,752) were coupled to their respective target genes using H3K27ac HiChIP data. Color
 indicates the gene set enrichment (FDR q-value) and size depicts the number of genes that overlap with the indicated
 gene sets. Cell cycle-related gene ontology terms are highlighted.
- 994(F) Bar chart (top) showing relative cell viability of LNCaP (left) and LNCaP-42D (right) cells upon transfection with non-
targeting siRNA (siNT) or siARNTL, and exposure to ENZ. Cell viability assays were performed 7 days post-transfection.996ENZ treatment is indicated and the average of three biological replicates is shown relative to the untreated (- ENZ) siNT
orndition per cell line. Western blots (bottom) indicate ARNTL protein levels in LNCaP (left) and LNCaP-42D (right) cells
following siRNA-mediated silencing of ARNTL for 48 h. Transfection with siNT and staining for ACTIN are included as
controls for siRNA treatment and protein loading, respectively. Images are representative of three independent
experiments. ns, P > 0.05; *, P < 0.05; ***, P < 0.001 (two-way ANOVA followed by Tukey's multiple comparisons test).