1. IRE1 activity regulates tumour and microenvironment cell lineage states while stratifying localised and

2. metastatic prostate cancer.

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Abstract

Prostate cancer (PCa) is an androgen receptor (AR) driven, high-incidence disease significantly

contributing to cancer mortality. PCa is in need of better risk stratification at diagnosis and treatment

outcomes in patients at high risk of metastasis. The unfolded protein response (UPR) is an AR-

dependent process. However, the impact of the UPR transducer IRE1 on AR-dependent biology and

treatment resistance has not been defined. We use diverse pre-clinical models of stress response to

describe IRE1 activity impact on multiple disease stages and demonstrate its involvement with poor

prognosis (RB1 loss), and cell lineage determination (club phenotypes). Integrating clinical

transcriptomic datasets, we chart IRE1 activity throughout PCa evolution by developing a PCa-specific,

IRE1 activity gene set (IRE1_18) reflecting both tumoral and micro-environmental niches. IRE1_18 can

determine tumoral identity, inform androgen deprivation treatment suitability, prognosticate localised

and metastatic disease independently from AR activity, and guide IRE1 modulation as a novel

combination therapeutic.

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Introduction

Prostate cancer (PCa) is a heterogenous, multifocal, high-incidence disease that affects one in eight

men globally. PCa evolution depends on transcriptional events and epigenetic modifications, forming

an adaptive response to intrinsic/extrinsic stresses, culminating in lineage reprogramming and

treatment resistance[1]. PCa results in ~360000 deaths annually[2] due to metastatic progression,

supported by treatment/stress-resistant cancer cells which experience high proliferative rates and

androgen receptor (AR)-driven high metabolic demand.

These events require high protein turnover, leading to misfolded protein accumulation, causing

Endoplasmic Reticulum (ER) stress. The unfolded protein response (UPR) resolves this stress through

three main transducers, Inositol Requiring Enzyme alpha (IRE1α), protein kinase RNA-like endoplasmic

reticulum kinase (PERK) and activating transcription factor 6α (ATF6). IRE1α, hereafter referred to as

IRE1 and encoded by the gene *ERN1*, is the most evolutionarily conserved transducer of the UPR and

signals through two distinct downstream functional outputs[3]. Firstly, through the unconventional

splicing of the transcript encoding XBP1, a transcription factor, resulting in a non-canonical splice

variant lacking a 60-nucleotide sequence which encodes the active protein (XBP1s). Secondly, through

the targeted degradation/decay (RIDD) of a subset of mRNA (e.g., *DGAT2*[4], *BLOC1S1*[5]) and miRNA

transcripts to ease transcriptional load[6]. Both functions require IRE1 endoribonuclease activity and can

be thought of as a continuum of early (XBP1s) and late (RIDD) overlapping responses that are designed

to adapt to stress.

Using short (up to 24 hour) pre-clinical model treatments we have previously shown that the IRE1-XBP1

axis contains AR-induced genes, notably *ERN1* itself[7]. Targeting IRE1 in androgen-dependent pre-

clinical models can restrain tumorigenesis[8]. However, it remains unclear which IRE1-dependent genes

or biological processes underpin these responses and how IRE1 activity may change in response to

sustained stress, reflecting PCa evolution. Most studies so far have focused on the role of IRE1 as an

activator of XBP1 impacting tumour development[9] largely ignoring IRE1-RIDD impact on RNA stability

which could be restraining the emergence of cancers that are more difficult to treat.

In this study we used bulk, single cell (SC) and spatial (ST) RNA sequencing data from two-dimensional

(cell lines) and three-dimensional models (patient derived explants (PDEs)), as well as patient biopsy

samples to describe how IRE1 activity changes throughout PCa evolution. Crucially, we found that

adaptation to IRE1 loss correlated with cell identities usually driven by prognostically poor genomic

alterations such as RB1 loss. We also observed that IRE1-deficient cells develop a phenotype

resembling club cells, epithelial cells that interact with immunosuppressive microenvironmental cells.

We produced a novel IRE1 activity gene set, IRE1_18, representing significant ERN1 transcriptomic

perturbations in PCa-specific models. IRE1_18 is a proxy of multiple signalling pathways linked to PCa

progression and can chart PCa evolution from early stage to treatment resistant disease while

prognosticating large clinical cohorts. Current clinical practices impact therapeutic effectiveness on

PCa mortality[10] and therefore, using IRE1 activity to identify tumours with treatment resistance traits

early, could be transformative of patient care. IRE1 activity could be used to stratify patients for therapy

intensification, surveillance or suitability of IRE1/UPR modulation as novel combination therapeutics

alongside conventional treatments.

Results

Changes in AR activity correlate with significant changes in IRE1 activity in PCa cell lines and

clinical datasets.

To capture steady state adaptations to AR-induced physiological stresses we cultured LNCaP cells in

charcoal stripped serum media (CSM) or full serum media (FM) for 72 hours or treated them with the

synthetic androgen R1881 or EtOH for 12 hours. Whole transcriptome sequencing of these conditions

followed by gene set enrichment analysis (GSEA) revealed biological pathways negatively enriched in

CSM vs FM while positively enriched in the R1881 vs EtOH conditions including the androgen response,

fatty acid metabolism, MTORC1 signalling, oxidative phosphorylation and cholesterol homeostasis

(Figure 1A, B, Supplementary tables S1, S2)). Canonical AR targets reflecting AR activity were

significantly upregulated in the R1881 treatment whilst significantly downregulated in the CSM

condition (Supplementary Figure S1A, Supplementary tables S3, S4).

Some Hallmarks were only significant upon androgen treatment or androgen deprivation. These

included a positive enrichment of protein secretion and negative enrichment of the Unfolded Protein

Response (UPR) in R1881 vs EtOH and CSM vs FM conditions respectively (Figure 1A, B). The UPR

Hallmark is not IRE1 specific containing target genes of all UPR transducers (ATF4, ATF6 and XBP1s)

while RIDD may not be represented within it. Consequently, we used a curated list of XBP1, and RIDD

target genes to correlate alterations in AR signalling with IRE1 activity. Whereas XBP1 target genes were

largely androgen-induced, RIDD targets were not consistently androgen dependent (Figure 1C,

Supplementary figure S1A).

UPR responses are tissue dependent and therefore, we determined how AR and IRE1 target gene

expression changed over 72-hour CSM culture of LNCaP compared to FM. A decrease in IRE1 activity

(*XBP1s* downregulated, RIDD targets *DGAT2*, *BLOC1S1* upregulated) was seen as early as 8 hours,

mirroring full length AR levels and activity (Figure 1D). Similar profiles were observed at the protein level

(Supplementary Figure S1B). Simultaneously we observed that levels of *ATF4* and *eIF2α* either did not

significantly or consistently change (Supplementary Figure S1C).

Hallmarks including E2F targets, G2M checkpoint and epithelial mesenchymal transition EMT (Figure

1A) were partially androgen dependent. Therefore, we hypothesised that they may be regulated by other

AR-dependent global mechanisms such as the IRE1 axis. We expanded our pre-clinical model

repertoire and assessed canonical AR and IRE1 target gene expression in patient derived explants

(PDEs)[11] treated with, or deprived of, dihydrotestosterone (DHT) for 12 hours. AR response markers

KLK3 and *CAMKK2* were significantly upregulated upon DHT treatment. *ERN1* itself and *DGAT2*,

BLOC1S1 were not significantly altered but XBP1s was significantly upregulated (Figure 1E,

Supplementary figure S1D).

We then tested if this AR-IRE1 activity correlation could be replicated within a clinical cohort

representing AR positive and AR negative tumours. Regulon analyses of 48 castration resistant prostate

cancer (CRPC) samples split into CRPC-Adenocarcinoma (n=35, AR positive) versus CRPC-

neuroendocrine prostate cancer (NEPC, n=13, AR negative)[12] showed that AR and XBP1 regulons were

both less active in NEPC samples while reprogramming markers pertaining to stem-like lineages such

as SOX2 and POU3F2 were increasingly active (Figure 1F, G). As such, transcription factor (TF) activity

correlates with IRE1 activity and PCa lineage.

Figure 1. *Changes in AR activity correlate with significant changes in IRE1 activity in pre-clinical models and clinical datasets.*

Figure 1. Changes in AR activity correlate with significant changes in IRE1 activity in pre-clinical

models and clinical datasets. Barplots displaying GSEA normalized enrichment scores (NES) for significantly

enriched MSigDB Hallmark gene sets in: LNCaP cells grown in charcoal stripped media (CSM) for 72 hours

compared to LNCaP cells grown in full media (FM) for 72 hours (A), and LNCaP cells treated with R1881 for 12

hours compared to LNCaP cells treated with ethanol (EtOH) for 12 hours (B). Bar length indicates the normalized

enrichment score and bar colour signifies the statistical significance. Hallmarks activated in B) but suppressed in

A) are underlined in green. Hallmarks activated or suppressed in both A) and B) are underlined in orange. Hallmarks

representing protein homeostasis are underlined in red. C) Expression heatmaps of selected "IRE1 induced" (left)

and "IRE1 repressed" (right) genes according to the literature in LNCaP cells in CSM (green) vs FM (purple), (top);

and R1881 (yellow) vs EtOH (orange) (bottom). Colour scale represents expression z-score from black (lowest)

to red (highest). D) mRNA fold change of androgen receptor activity markers (full length AR, CAMKK2, KLK3) and

IRE1 activity markers (ERN1, XBP1s, DGAT2) in LNCaP parental cells grown over 72 hours in CSM, normalised to

untreated FM control. Statistical analysis was performed using an ordinary one-way ANOVA with Dunnett's

multiple comparisons test and single pooled variance. Four stars signify adjusted p value of <0.0001. E) mRNA

fold change of AR and IRE1 activity markers KLK3, CAMKK2, XBP1s and DGAT2 in patient derived explants (PDEs)

treated with dihydrotestosterone (DHT), normalised to PDEs treated with vehicle control. Statistical analysis was

performed using a T-test per condition. Each star signifies an order of magnitude of significance <0.05. F, G)

Heatmaps representing regulons significantly altered in neuroendocrine (NEPC, pink) castration resistant (CRPC)

tumours compared to adenocarcinoma CRPC (light blue). Regulon levels are represented as single-sample

enrichment scores with individual regulons denoted by the name of their corresponding transcription factor (TF).

Significantly activated and suppressed regulons are shown, with AR, XBP1 (AR, IRE1 activity), SOX2 and POU3F2

(reprogramming) in red boxes. Colour scale represents regulon enrichment z-scores from lowest in greyscale to

highest in red-scale.

Knocking out IRE1 induces changes in lineage characteristics of AR dependent PCa cells.

Having established that IRE1 activity changes with PCa progression alongside the AR we investigated

the adaptation of AR positive cells (LNCaP) to stable IRE1 ablation, using an IRE1 CRISPR-knockout

(LNCaP ERN1[-/-])[13], lacking IRE1 at steady state or chemical ER stress (tunicamycin/thapsigargin)

(Supplementary Figure S2A, B). Bulk RNAseq confirmed multiple canonical "IRE1 induced" and "IRE1

repressed" genes were downregulated and upregulated respectively in ERN1[-/-] cells (Supplementary

figure S2C).

Any observed biology in the ERN1[-/-] cells could be attributed to the cell line generation selection

pressure. To definitively link physiological observations to the absence of IRE1 activity we stably

expressed an IRE1 mutant in the ERN1[-/-] cells carrying a truncation mutation at tryptophan 780[14] thus

lacking the effector ribonuclease domain. These new ERN1[-/-] Q780* cells were IRE1 activity deficient

at steady state and ER stress conditions (Supplementary Figures S2D-G).

We then documented the growth of LNCaP, LNCaP ERN1[-/-] and LNCaP ERN1[-/-] Q780* in FM or CSM.

ERN1[-/-] and ERN1[-/-] Q780* doubling time was 25% and 32% lower than parental LNCaP in FM,

respectively. In CSM, ERN1[-/-] and ERN1[-/-] Q780* doubling time was increased by 50% compared to FM

whilst parental cells stopped growing (Figure 2A) entirely. To document the biological functions

underlying this phenotype we performed pathway analysis on whole transcriptome data from these

conditions. E2F targets, G2M checkpoint, and the EMT were significantly enriched in ERN1[-/-] samples

(Figure 2B, Supplementary Table S5).

Consequently, androgen deprivation-induced growth arrest and DNA repair abrogation on PCa cells

might be a consequence of both the repression of XBP1 activity and the stabilization of RIDD targets.

We thus compared significantly altered Hallmarks induced by androgen deprivation with those altered

by knocking out IRE1. ERN1[-/-] cells were still sensitive to CSM as E2F targets, G2M checkpoint and AR

Response were negatively enriched compared to ERN1[-/-] cells grown in FM (Figure 2C, Supplementary

Table S6). However, these Hallmarks were not affected by androgen deprivation to the same degree as

DNA repair, AR response and MTORC1 were upregulated in ERN1[-/-] compared to parental cells grown

in CSM (Figure 2D, Supplementary Table S7).

We observed that low *XBP1* correlated with high activity of reprogramming TFs (*SOX2, POU3F2*) in

clinical samples (Figure 1F, G) and that IRE1 loss leads to cell cycle de-regulation in PCa cell lines

(Figure 2A-D). Consequently, we measured mRNA of lineage reprogramming[15,16] and cell cycle

regulation[17] genes with known clinical association in PCa. *RB1* was significantly downregulated whilst

BRCA2, SOX2 EZH2 and *POU3F2* mRNA was significantly upregulated in ERN1[-/-] cells in FM (Figure

2E). Therefore, adaptation to IRE1 activity loss changed the capacity of LNCaP cells to respond to

androgen deprivation whilst retaining their AR axis.

We then mapped TF repertoires (regulons) significantly affected by *ERN1* knockout. Stress response

(CREB3L1),   inflammatory   signalling   (NFATC3,   IRF6),   cell   cycling   (E2F3)   and

embryogenesis/stemness/differentiation regulons (SOX4, PITX2, HOXB13) were significant. RBL1 was

positively enriched (Figure 2F, Supplementary table S8) which has been shown to regulate cell cycle in

response to RB1 loss[18].

Gene signatures associated with RB1 loss, NEPC and basal phenotypes (Supplementary Table S9)

segregated ERN1[-/-] from parental cells, in a similar manner to the regulons (Figure 2G). Thus, knocking

out IRE1 in cells that are AR-dependent, and retain some luminal epithelial character, could change

their lineage characteristics. Comparing luminal, basal and club cells within a PCa SC dataset[19] we

observed that TFs with regulons activated upon IRE1-loss were significantly upregulated in basal and

club cells compared to luminal epithelial cells (Supplementary Figure 3A). We then documented the

enrichment of genes upregulated in tumoral, luminal, basal and club cells[20] (Supplementary Table S10)

in the ERN1[-/-] transcriptome. ERN1[-/-] cells were positively enriched for club[21–23] while negatively

enriched for tumoral and luminal cell markers, showing decreases of *AMACR*, ACPP and NTN4 but

increases in *MMP7, PIGR, MSMB* and *KRT19* (Figure 2H, Supplementary Figure S3B).

Figure 2. *Knocking out IRE1 induces lineage changes in AR dependent PCa cells.*

Figure 2. Knocking out IRE1 induces lineage changes in AR-dependent PCa cells. A) Growth curves of LNCaP

parental, LNCaP ERN1-/- and LNCaP ERN1-/- Q780* cells grown over 120 hours in FM or CSM. Confluence fold

change was normalised to Incucyte readings taken at the 0-hour time point of each condition. Exponential

(Malthusian) growth curve equations were used to estimate the doubling time of each cell line in FM or CSM.

GSEA normalized enrichment scores (NES) for significantly enriched MSigDB Hallmark gene sets in LNCaP ERN1-

/- cells compared to LNCaP parental cells grown in FM (B), LNCaP ERN1-/- cells grown in CSM compared to

LNCaP ERN1-/- cells grown in FM (C), and LNCaP ERN1-/- cells compared to LNCaP parental cells grown in CSM

(D). Bar length indicates the normalized enrichment score (NES) and bar colour signifies the statistical

significance. E) mRNA fold change of ERN1, XBP1, cell cycle markers RB1, BRCA2, and lineage reprogramming

markers EZH2, POU3F2, SOX2 in LNCaP ERN1-/- normalized to parental LNCaP. Statistical analysis was

performed using an ordinary two-way ANOVA and Sidak's multiple comparisons test with single pooled variance.

Heatmap representing regulons significantly altered in LNCaP ERN1-/- (IKO, light blue) compared to parental

LNCaP (pink). Top 30 statistically significant regulons shown (p<0.001). Regulon levels are represented as single-

sample enrichment scores with individual regulons denoted by the name of their corresponding transcription

factor (TF). Colour scale represents regulon enrichment z-scores (from lowest in greyscale to highest in red-

scale). G) GSVA single-sample gene set enrichment scores enrichment scores in LNCaP ERN1-/- cells (IKO,

yellow) compared to LNCaP parental cells (mustard) shown from left to right for gene sets including: TFs with

regulons activated in ERN1-/- cells, genes upregulated in the RB1-loss signature from Ertel et al., a basal cell

signature from Aggarwal et al, and the genes upregulated in the NEPC signature from Beltran et al. Statistical

testing was performed using limma. H) Bulk RNAseq log2 fold change expression levels in ERN1[-/-] cells compared

to parental, of genes widely reported to be associated with tumour, luminal epithelial, basal epithelial and club

epithelial cells. Downregulated genes in blue, upregulated genes in red. Created using Biorender.com. Each

experiment was performed at least in duplicate. Four stars signify adjusted p value of <0.0001.

IRE1 perturbation affects cell cycle and lineage reprogramming markers and responses to

physiological androgen deprivation therapy (ADT).

Adaptation to IRE1 loss leads to an epithelial club-like lineage shift histologically associated with

inflammatory atrophy[21], that can result from AR blockade[23]. Low IRE1 activity correlates with ADT

resistant clinical presentations (Figure 1) and IRE1 deficient cells express high RBL1 and low RB1 RNA

levels. Lineage plasticity is facilitated by knocking RB1 out, or knocking RB1 out alongside TP53 in

LNCaP cells [24,25] which are reported to be resistant to replication stress, have high proliferative rates

and increased DNA repair capacity[26]. Intriguingly, Hallmarks affected by RB1 loss or RB1 and TP53 loss

were similar to IRE1 loss (Figures 3A, 3B, Supplementary Tables S11, S12) showing upregulation in E2F,

G2M checkpoint and MYC targets.

We then tested whether IRE1 activity was altered in RB1 deficient cells. *ERN1, BRCA2*, *RB1*, *EZH2* and

POU3F2 were significantly downregulated in RB1$^{-/-}$ cells (Figure 3C). RB1$^{-/-}$/TP53$^{-/-}$ cells had

significantly lower levels of both *ERN1*, *XBP1s, BRCA2* and *RB1* (Figure 3D) while higher levels of *EZH2*,

POU3F2 and *SOX2* (Figure 3E) compared to parental.

ERN1 siRNA silencing significantly upregulated *POU3F2*, *EZH2*, *SOX2* (Figure 3F). Treatment with the

salicylaldehyde IRE1 inhibitor MKC8866[27] caused significant upregulation of *EZH2*, *POU3F2*, *SOX2* and

BRCA2 (Figure 3G) while treatment with the IRE1 activator IXA4[28], caused significant upregulation of

RB1 and downregulation of *BRCA2* with reprogramming markers showing little change (Figure 3H).

We correlated resistance to ADT with IRE1 activity loss. Therefore, we tested the effect of

pharmacological IRE1 activity upregulation on the ability of treatment resistant cells to grow in CSM.

CSM did not affect growth of either RB1$^{-/-}$ or RB1$^{-/-}$/TP53$^{-/-}$ LNCaP cells. However, addition of subtoxic

doses of the IRE1 activator IXA4, slowed their growth in CSM. RB1-/- doubling time increased by 40%

after 24 hours whilst RB1$^{-/-}$/TP53$^{-/-}$ only changed by 10% after 72-hours compared to FM and DMSO

(Figure 3I, Supplementary Figure S4A).

Then, in a metastatic, castration resistant PCa (mCRPC, 98 tumours from 55 men)[29] cohort, we

investigated whether the patterns of enrichment of gene signatures comprising TFs with regulons

significantly activated or suppressed in ERN1$^{-/-}$ (Figure 2F) could differentiate AR positive from AR

negative mCRPC alongside several AR and NEPC gene signatures. The set of TFs with regulons

suppressed in ERN1-/- - clustered with AR activity signatures whilst the TFs with regulons activated in

ERN1-/- clustered with RB1-loss and ARV-7 activity signatures[30,31], distinguishing AR naïve from AR

agnostic or NEPC/DNPC PCa. Overall, ERN1-/- cells clustered with NEPC and double-negative for

NEPC and AR PCa (DNPC) as well as some AR-low (ARLPC) mCRPC samples (Figure 3J).

Consequently, IRE1 transcriptomic repertoires may correlate with AR driven and agnostic phenotypes

by playing a key role in lineage plasticity, cell cycle and ultimately treatment response.

Figure 3. *IRE1 activity impacts AR responses in RB1 loss.*

Figure 3. IRE1 activity impacts AR responses in RB1 loss. GSEA normalized enrichment scores (NES) for

significantly enriched MSigDB Hallmark (H) gene sets in: LNCaP RB1$^{-/-}$TP53$^{-/-}$ cells compared to LNCaP parental

cells grown in FM (A), and LNCaP RB1-/- cells grown in CSM compared to LNCaP parental cells grown in FM (B).

Bar length indicates the normalized enrichment score (NES) and bar colour signifies the statistical significance.

C, D, E) mRNA fold change of *ERN1, XBP1, RB1, BRCA2, EZH2, POU3F2, SOX2* in C) LNCaP RB1-/- normalised

to parental LNCaP and D, E) LNCaP RB1$^{-/-}$TP53$^{-/-}$ cells normalised to parental LNCaP. F) mRNA fold change of

ERN1, EZH2, POU3F2, SOX2 in LNCaP parental cells treated with siRNA against ERN1 normalised to scramble

control. mRNA fold change of *XBP1s, EZH2, POU3F2, SOX2, RB1, BRCA2* in LNCaP parental cells treated with

G) MKC8866 or H) IXA4 normalised to DMSO. Statistical analysis was performed using an ordinary two-way

ANOVA and Sidak's multiple comparisons test with single pooled variance. I) Growth curves of LNCaP RB1$^{-/-}$ cells

grown over 120 hours in FM (green), CSM (red) or in CSM in the presence of subtoxic (25μM) doses of IXA4

(purple). Confluence fold change was normalised to Incucyte readings taken at the 0-hour time point of each

condition. T-tests were used for individual time point comparisons between FM and CSM; CSM and CSM plus

IXA4. Exponential (Malthusian) growth curve equations were used to estimate the doubling time of each condition.

J) Heatmap of GSVA enrichment scores for 11 selected PCa gene signatures shown across samples from the

Labrecque patient cohort (N=98) together with LNCaP parental and ERN1-/- cell line samples (N=5). Colour

scale represents GSVA enrichment z- scores (from lowest in greyscale to highest in red-scale). Each experiment

was performed at least in duplicate. Four stars signify adjusted p value of <0.0001.

Enzalutamide treatment induces a UPR downregulation which may act as a negative feedback

loop that leads to treatment resistance to ADT in PCa.

Adaptation to IRE1 activity loss may push AR responsive cells towards an ADT resistant phenotype,

marking the starting point of IRE1 activity saturation and the shift from an active phenotype to a passive,

adapted phenotype. ADT could catalyse this key event in the shift from early stage to late-stage disease

progression. Consequently, we investigated the effect of increasing concentrations of Enzalutamide

(ENZA) on LNCaP, LNCaP ERN1$^{-/-}$ and LNCaP ERN1$^{-/-}$Q780* cells over 12 hours. Both IRE1-deficient

cell lines had significantly higher ENZA IC50s (Figure 4A, Supplementary Figure S4B).

ENZA treatment of parental LNCaP, induced suppression of AR response, DNA repair, MTORC1, E2F

targets and the UPR (Figure 4B, Supplementary Table S13). Comparing ERN1-/- to parental LNCaP

treated with ENZA, we observed positive significant enrichment for E2F targets, G2M checkpoint and

DNA repair (Figure 4C, Supplementary Table S14), corroborating our CSM results (Figures 1,2). In

addition, we compared ERN1$^{-/-}$ responses to ENZA, to publicly available RB1$^{-/-}$/TP53$^{-/-}$ LNCaP ENZA

response data observing strong similarities between the two. Only inflammatory, hypoxia and MTOR

signalling significantly differentiating the responses of these two cell lines to ENZA (Supplementary

Figure S4C, Supplementary Tables S15, S16).

Thus, IRE1 activity suppression may play a pivotal role in acquired treatment resistance (Figure 4D)

through lineage adaptation to ADT. We then tested whether IRE1 activation could sensitise RB1

deficient cells to ENZA. Whereas the IRE1 activator IXA4 had no effect on viability when used alone

(not shown), it sensitised both ENZA resistant (Supplementary Figure S4D, E) RB1$^{-/-}$ and RB1$^{-/-}$TP53$^{-/-}$

cells to physiologically relevant doses of ENZA over a 12-hour treatment (Figure 4E).

Consequently, treatment resistant PCa pre-clinical models have low IRE1 activity recapitulating clinical

observations but a short-term increase in IRE1 activity enhances the response to AR antagonism.

Figure 4. *Enzalutamide treatment downregulates UPR responses which contribute to acquired ADT resistance.*

A.

B.

LNCaP Parental ENZA vs DMSO

C.

LNCaP ERN1-/- ENZA vs Parental ENZA

D.

E.

Figure 4. Enzalutamide treatment downregulates UPR responses which contribute to acquired ADT

resistance. A) Enzalutamide (ENZA) concentration gradient measuring cytotoxicity through ATP MTT assays in

LNCaP parental, LNCaP ERN1$^{-/-}$ and LNCAP ERN1$^{-/-}$ Q780* cells treated for 12 hours with increasing

concentrations of ENZA. Logarithmic concentration vs normalised response curves were fit to calculate the IC50

of ENZA for each cell line. Bar chart represents ENZA IC50 comparison between the cell lines. Maximum variance

used to calculate significance by ordinary one-way ANOVA with multiple comparisons. B) GSEA normalized

enrichment scores (NES) of LNCaP parental cells treated with 10μM ENZA for 12 hours compared to DMSO or

C) of LNCaP ERN1$^{-/-}$ cells treated with 10μM ENZA for 12 hours compared to parental LNCaP cells treated with

the same conditions. Bar length indicates the normalized enrichment score (NES) and bar colour signifies the

statistical significance. D) Schematic summarising observations of figures 1-4. During AR driven tumour growth

represented by the R1881 dataset, multiple key mechanisms are upregulated including the UPR. In the eventuality

of ADT represented by the CSM and ENZA datasets, these processes are reversed/ downregulated. However,

removing IRE1 activity in isolation leads to an increase in processes that directly contradict the effects of ADT

leading to partial treatment resistance. Created using Biorender.com. E) Cell titer glo assay, fluorescence fold

change of treatment resistant LNCaP RB1$^{-/-}$ and LNCaP RB1$^{-/-}$TP53$^{-/-}$ cells treated with 50μM and 100μM ENZA

in the presence or absence of 25μM IXA4. Statistical analysis was performed using an ordinary one-way ANOVA

with Dunnett's multiple comparisons test and single pooled variance. Each experiment was performed at least in

duplicate. Four stars signify adjusted p value of <0.0001.

Integrative analysis of sequencing datasets gives rise to a novel IRE1 activity gene set (IRE1_18)

that represents multiple biological functions that drive PCa dynamic disease progression.

We identified PCa-specific, IRE1-responsive genes, labelling them "IRE1 activity induced" (significantly

downregulated in ERN1$^{-/-}$) or "IRE1 activity repressed" (significantly upregulated in ERN1$^{-/-}$)

(Supplementary Table S17) by integrating results from multiple RNAseq datasets (41 genes, Figure 5A,

Supplementary Figure S5A). This gene set represents key PCa progression biological functions

comprising cell adhesion, motility, cell cycle, DNA repair, and protein trafficking, while correlating with

changes in cell cycle, EMT and metabolism PCa-associated genes (Supplementary Figure S5B, S5C

Supplementary Tables S18, S19).

We then investigated whether this approach yielded direct IRE1 activity targets within IRE1_41. XBP1

ChIP-seq analysis of LNCaP cells treated with R1881 for 24 hours showed that multiple "IRE1 induced"

genes displayed XBP1 peaks directly in their promoter (Supplementary figure S6A). RNA silencing of

IRE1 or XBP1 downregulated "IRE1 induced" and upregulated "IRE1 repressed" genes (Figure 5B). The

same effect was achieved by MKC8866 (IRE1 RNAse inhibitor) treatment (Figure 5C) while the opposite

with IXA4 (Figure 5D) or thapsigargin (Figure 5E) treatment. To ensure that these effects were not cell-

line specific, the same analyses were also performed in C4-2B cells and similar effects were observed

(Supplementary Figure S6B-F).

We then assessed AR/IRE1 activity canonical genes in a compendium of integrated RNA-seq datasets

comprising over 1000 clinical tissue samples ranging from benign to NEPC (Prostate Cancer Atlas[32]).

We also applied signatures of basal and luminal epithelial cell lineages to these data[33]. The basal

signature expression mirrored AR repression while AR induction signature mirrored both the luminal

signature and IRE1 induction (Supplementary Figure S7A). We then investigated the new IRE1 activity

gene set in these samples, documenting the pseudo-time disease progression trajectory attributed to

the "IRE1 induced" and "IRE1 repressed" components (Supplementary Figure S7B). A particular

subgroup of genes (*MYL9, THBS1, TRANK1, TGM2, RND3 and CHI3L2)* was consistently

downregulated compared to normal tissue (Figure 5F).

The IRE1 activity gene set was derived from significant differential expression upon IRE1 perturbation

in pre-clinical cancer models. However, some of these genes under IRE1 control could be enriched in

other cell types. ST[34] data expression maps encompassing benign, stromal and localised grade group

(GG)4 histological annotations, showed that signal was concentrated exclusively in the benign/stromal

region; either ubiquitously (*RND3, MYL9*) or in very specific benign sub-compartments (*TGM2*). Some

genes were tumour-region specific with the majority of "IRE1 induced" genes (e.g., SNHG8) enriched

in GG4 (Figure 5G, Supplementary Figure S7C) while also significantly upregulated in tumoral compared

to normal epithelial cells in two separate SC datasets of primary tumour and matched benign samples

[19,35] (Supplementary Figure S7D). Recapitulating multicellularity, IRE1 activity could distinguish

between SC luminal, basal and club cells.

Using these ST and SC sequencing filters, we selected refined IRE1 activity gene sets that identify

tumoral and peri-tumoral regions. We applied further redundancy filters (Supplementary Figure S8A)

based on expression patterns across clinical designations in the Prostate Cancer Atlas, which

segregated AR+ from AR- tumours (*VSTM2L, TRPM8*) or displayed increasing gradient expression

(*HELLS, XRCC2, BRCA2*) and decreasing gradient expression patterns (*SNHG8, EEF2*). We thus

produced shorter filtered gene sets that were applied to segregating mCRPC (Labrecque) or localised

(PRAD TCGA) cohorts (Supplementary Figure S8B, S8C). The final set of 18 genes (labelled IRE1_18)

still represented diverse biological functions and cell types (Supplementary Figure S8D, Supplementary

Table S20). Both "IRE1 induced" and "IRE1 repressed" parts of IRE1_18 differentiated AR-positive and

AR-negative tumours from normal tissue (Supplementary Figure S8E, Supplementary Tables S21, S22).

We next assessed how IRE1_18 clustered metastatic (Labrecque) and localised (TCGA-PRAD) patient

cohorts compared to AR signatures. The "IRE1 induced" part of IRE1_18 was significantly reduced in

NEPC/DNPC mCRPC samples compared to AR positive mCRPC samples showing similarity to a 10-

gene AR induced signature[36]. Within the AR positive mCRPC groups, "IRE1 repressed" subset of

IRE1_18 most negatively enriched in the amphicrine group whilst most positively enriched in the AR-

low group (Figure 5H). In the TCGA-PRAD, the "IRE1 induced" gene set was significantly decreased in

primary disease compared to benign tissue. The exact opposite was observed in the "IRE1 repressed"

gene set (Figure 5I). Encouragingly, both parts of IRE1_18 segregated AR-positive from AR-negative

tumours and primary tumours from benign tissue in a manner comparable to AR activity signatures..

Since IRE1_18 represents different cell types and biological functions it should have specific prognostic

power in different clinical settings. Indeed, splitting IRE1_18 in two subgroups comprising both

"induced" and "repressed" genes significantly prognosticated both mCRPC (Figure 5I) and localised

disease (Figure 5J) for overall and progression free survival respectively. In the PCF/SU2C cohort,

patients with significantly differentially expressed mRNA of IRE1_18 had worse probability of overall

survival compared to patients without (Supplementary Table S23). In the TCGA/PRAD cohort, patients

with significantly differentially expressed mRNA of IRE1_18 showed significantly worse probability of

progression free survival, compared to patients without (Supplementary Table S24).

Figure 5. *Multiple sequencing datasets give rise to a novel IRE1 activity gene set that represents multiple key biological functions and dynamic disease progression.*

Figure 5. Multiple sequencing datasets give rise to a novel IRE1 activity gene set that represents key PCa

biological functions and dynamic disease progression. A) Gene expression heatmaps of genes in the IRE1

activity signature defined as "IRE1 induced" (top) when downregulated in LNCaP ERN1[-/-] cells (yellow) compared

to parental (mustard) and as "IRE1 repressed" (bottom) when upregulated in LNCaP ERN1[-/-] cells compared to

parental. mRNA fold change of *ERN1, XBP1s, DGAT2, VSTM2L, SSR2, SNHG8, HELLS, LGI2* in LNCaP parental

cells treated with B) siRNA against ERN1 or XBP1 for 48 hours normalised to scramble control, C) 5μM MKC8866

for 5 hours normalised to DMSO, D) 25μM IXA4 for 5 hours normalised to DMSO and E) 100nM Thapsigargin for

5 hours normalised to DMSO. Statistical analysis was performed using an ordinary two-way ANOVA and Sidak's

multiple comparisons test with single pooled variance. F) Heatmap of the IRE1 activity gene set expression fold

change in primary, AR positive CRPC, NEPC and DNPC patient cohorts compared to normal tissue (Prostate

Cancer Atlas). Fold changes calculated using "Normal" as control in the DESeq2 Prostate Cancer Atlas function.

Downregulated genes in blue, upregulated in red. G) Histological annotations of 6.5 mm^2 tissue slide where benign

(blue), stroma (green) and grade group 4 (GG4) cribiform tumour is represented (red). Loupe browser generated

distribution maps of individual genes (*MYL9, TGM2, RND3, SNHG8*) across these histological designations

showing either generalised or specific localisation to benign or tumoral tissue. Box plots of GSVA enrichment

scores for AR activity and IRE1_18 gene sets, benchmarked in H) mCRPC (Labrecque) and I) localised disease

(TCGA PRAD) datasets. Patient populations represented include AR positive (dark blue), AR low (light blue),

amphicrine (grey), NEPC (red), double negative (DNPC, orange), normal tissue (light green) and primary tumour

(dark green). Statistical testing was performed using limma (lmFit function). Shortened versions of the IRE1

activity signature ((*VPS13A, EEF2, TRPM8, POLQ, TGM2, THBS1, LGI2, STX5, TRPC1*), (*VSTM2L, HELLS, CHI3L2,*

SNHG8, DDX11, SLITRK3, CRYL1, NEMP1, MYL9)) tested for prognostication power showing significance in I)

probability of overall survival and J) progression free survival in an mCRPC (PCF/SU2C) and primary disease

(PRAD TCGA) cohort respectively. Altered expression defined as J) significant log-transformed mRNA z-scores

compared to expression distribution of all samples (RNA Seq FPKM) and K) significant z-scores of tumour

samples compared to the expression distribution of all log-transformed mRNA expression of adjacent normal

samples in the cohort. Statistical analysis was performed using a Logrank (Mantel-Cox test). Each experiment

was performed at least in duplicate. Four stars signify adjusted p value of <0.0001. Heatmap in F) was created

using Biorender.com.

Novel IRE1 activity gene set distinguishes between histological grades, between benign and

tumour-bearing virtual biopsies and correlates with multiple clinical attributes.

IRE1_18 is a proxy of multiple biological functions, cell types and treatment responses and identifies

or prognosticates both localised and metastatic disease. Recently, we described how gene expression

profiling of virtual tumour and benign biopsies reveals differences between grade groups and tumour

clones not reflected in bulk analyses. Therefore, bulk analyses of whole biopsies or tumour-only areas,

as used in clinical practice, may provide inaccurate assessments[37] as tissue type, tumour grade and

clonal composition all influence gene expression.

We then tested if IRE1_18 could be confounded by heterogeneity. Pseudo bulk analysis of tumour

bearing virtual biopsy using IRE1_18 showed transcriptomic changes attributed to both tumoral or peri-

tumoral tissue. Upregulation of *CHI3L2*, *TGM2* and *TRPM8*, *SNHG8*, *VSTM2L* (Supplementary Figure

S9A, B) reflected peri-tumoral (*TGM2*, *CHI3L2*) or tumoral (*SNHG8*, *TRPM8, VSTM2L*) regions

(Supplementary Figure S9C). We proceeded to determine if IRE1_18 could identify high grade tumour,

profile a bulk virtual biopsy and place this profile in an exact subgroup of patients within a larger external

clinical cohort. IRE1_18 genes separating GG4 from local benign in a 6.5mm[34], showed the same

pattern in a virtual biopsy containing both GG2 and GG4. Since this patient solely underwent radical

prostatectomy, we expected their virtual biopsy to correspond to primary disease and indeed, the

expression profile of IRE1_18 corresponds to the primary disease subgroup of the Prostate Cancer

Atlas (Figure 6A).

Thereafter we correlated IRE1_18 expression with increasing grade group in this radical prostatectomy

using GG1, GG2 and GG4 spots to recapitulate the full spectrum of multifocality and heterogeneity. The

"IRE1 induced" component (*TRPM8, EEF2, SNHG8, VSTM2L*) increased whilst the "IRE1 repressed"

component (*THBS1, DDX11, HELLS, NEMP1, LGI2*) decreased with tumour grade (Figure 6B).

To expand on this correlation, we explored data from 118 samples (DKFZ, Cancer Cell cohort) and

observed significant differential expression changes in IRE1_18 genes with increasing Gleason score

(Figure 6C). Furthermore, in the PCF/SU2C cohort, significant differential expression of these genes

correlated with worsening pathology classification (NEPC/small cell disease) (Figure 6D). Indeed,

change in mRNA status of these genes significantly correlated with clinical attributes in multiple

datasets including NEPC scoring, BCR status and mutation count (Supplementary Table S25).

Therefore, IRE1 activity represented by IRE1_18 showcases the importance of stress response biology

in PCa pathophysiology, progression and treatment response.

Figure 6. *IRE1_18 may act as a proxy of multiple clinical attributes.*

Figure 6. IRE1 activity may act as proxy of multiple clinical attributes. A) Schematic of whole radical

prostatectomy axial disc with coded dashed squares signifying corresponding spatial transcriptomic 6.5mm2

squares. As previously described three rectangles show virtual biopsy placements corresponding to completely

benign or tumour bearing 1.2mm x 18 mm standard transurethral (TRUS) sampling. From right to left, log fold

change expression of the IRE1 activity signature genes is represented in heatmaps locally comparing GG4 to

benign; comparing a benign virtual biopsy to a tumour bearing virtual biopsy carrying GG2 and GG4; and comparing

primary, AR positive CRPC, DNPC and NEPC bulk samples to normal tissue. B) Histological annotations (left)

from three separate 6.5 mm^2 squares from the same radical prostatectomy representing multifocality and

heterogeneity. Grade groups (GG) 1 (yellow), 2 (orange) and 4 (red) can be seen. On the right, log fold change in

expression of "IRE1 induced" (*EEF2, TRPM8, SNHG8, VSTM2L*) and "IRE1 repressed" (*THBS1, LGI2, DDX11,*

HELLS, NEMP1) genes when comparing increasing GG to each corresponding benign region. Each locally

distinguishing comparison involves 331 GG1 spots, 938 GG2 spots, 1169 GG4 spots and 942, 703 and 1078

benign spots respectively. C) Correlation of expression pattern of the IRE1 activity gene set to radical

prostatectomy Gleason score in the DKFZ Cancer Cell 2018 cohort. Statistical analysis was performed using Chi-

squared test. Altered (red) status is defined as differentially significant z-score expression in tumour samples

compared to the expression distribution of all log-transformed mRNA expression in adjacent normal samples in

the cohort. D) Correlation of expression pattern of the IRE1 activity signature to pathology classification including

adenocarcinoma (green), small cell (orange) and adenocarcinoma with NEPC features (purple). Statistical

analysis was performed using Chi-squared test. Altered gene expression is defined as differentially significant log-

transformed mRNA z-scores compared to mRNA signal from all samples in the cohort. Heatmaps in A) were

created using Biorender.com.

Discussion

We have previously reported that the IRE1-XBP1 axis of the unfolded protein response is regulated by

androgen receptor activity and that inhibiting IRE1 restricts tumorigenesis in androgen-dependent

models of prostate cancer. Sequencing well-annotated clinical samples has revealed that whilst AR

activity is maintained in many cases as PCa progresses to metastasis, AR variants and target genes that

support progression change. In addition, NEPC/DNPC PCa subtypes do not seem to rely on AR function

or expression. By contrast, the UPR/IRE1 axis is capable of adapting as these disease states emerge.

Given that the UPR-IRE1 axis is a feature of a wide range of cell types we have sought to determine how

IRE1 activity relates to the emergence of acquired treatment resistance and to a spectrum of PCa sub-

types or cell lineages. To assign IRE1 activity to cell lineages and PCa subtypes we derived a PCa-

specific gene set (IRE1_18) by initially relying on data derived from PCa cell-lines and pre-clinical

models. When refining IRE1_18 we selected genes that could discriminate between PCa sub-types in

patient datasets and to prognosticate both localized and metastatic disease. De-convoluting IRE1_18

spatially, we determined that in clinical samples these genes are associated with a variety of cell types

and biological functions likely to reflect niches in which treatment resistant disease can emerge.

Our cell-line/pre-clinical models cannot capture the cellular heterogeneity or biological complexity of

these niches, but a potentially significant finding was that knocking out IRE1 in an androgen-dependent

PCa cell-line results in the increased expression of lineage markers associated with club-like epithelial

progenitor (*MMP7, PIGR, CP, LCN2*), and lineage reprogramming (*SOX2, EZH2, POU3F2*). IRE1 activity

may impact cellular differentiation being associated with the acquisition of cancer stem cell

characteristics in glioblastoma through XBP1s-driven repression of mir-148, which de-represses

stemness TFs such as SOX2[38]. In PCa, *SOX2* has been associated with stemness and treatment

resistance arising from knocking out TP53 and RB1 in androgen-dependent pre-clinical models[39].

Intriguingly, knocking out IRE1 also leads to transcriptional changes resembling RB1 loss and oncogenic

TF coregulation[40] while correlating with AR loss and NEPC genotypes in mCRPC datasets. Importantly,

pharmacologically increasing IRE1 activity in RB1 deficient cells sensitised them to physiological and

pharmacological AR deprivation.

In the lifetime of PCa evolution and treatment, UPR/IRE1 activity has been shown to increase as an

early stress-response, adaptive, pro-survival mechanism due to mounting metabolic stress and protein

folding demand, parallel to the AR. But what happens when AR activity starts to decrease

physiologically, or as a response to therapy? ADT-induced IRE1 signalling ablation may lead to

adaptation that pushes a subset of tumours towards an AR+ CRPC or NEPC/DNPC phenotype, thus

placing IRE1 activity at the centre of acquired treatment resistance.

Our study thus introduces two major novel themes for further investigation. Firstly, in more effectively

stratifying patients for ADT and perhaps in the future, modulators of IRE1/UPR activity (ONC201[41],

NXP800[42]) by mapping IRE1 responses to AR antagonism (STAMPEDE), to metastatic site (Hartwig

Medical Foundation) and to RB1-loss, BRCA2-amplification phenotypes lacking somatic RB1/BRCA2

changes (TCGA, SU2C). Secondly, for further developing models of lineage plasticity that can capture

transition states between high AR activity, differentiated luminal epithelial cells and low AR activity,

club-like epithelial progenitors aiming to predict the emergence of AR-negative basal or stem-like cells.

IRE1 activity proxies (e.g., IRE1_18) representing multicellular tumoral niches are not dependent on

accurate tumour sampling and can identify peri-tumoral tissue in false negative biopsies or inform risk

stratification in histologically benign tissue with pre-tumoral genotypes. Indeed, clinically approved

genomic tests (Decipher) contain genes attributed to multiple cell types and thus significantly associate

with risk of metastasis and PCa specific mortality[43].

The majority of significant work in this area of lineage plasticity has thus far focused on transcription

factors (*ACSL1, POU3F2*) and chromatin-modifying enzymes (*EZH2*). Against this backdrop it will be

interesting to determine whether changes in IRE1 activity create permissive states in which these more

established drivers of lineage plasticity can function more effectively, or whether IRE1's contribution is

mechanistically independent.

Summary

In conclusion, we explored PCa dynamic disease progression through the homeostatic mechanism of

the UPR. By focusing on the UPR transducer IRE1, we produced an IRE1 activity gene set that can act

as a proxy of AR activity, stress response, histological grade, cell type, clinical designation and

potentially treatment response. IRE1 activity is commonly thought to be a driver of tumorigenesis. We

here, report for the first time that adaptation to sustained stress and suppression of the IRE1 axis may

induce lineage characteristics that mimic a club-like pre-cursor to RB1 loss and treatment resistant

disease. This has major implications for IRE1 modulation in treatment resistant PCa as a differentiation

therapy and for use of IRE1 activity as early identifier of lethal disease.

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Dr Doultsinos and Prof. Mills designed the study, analysed data and wrote the manuscript. Dr Doultsinos carried out all experiments presented (qPCR, Western Blot, growth assays, viability assays, RNAseq) supported by Ms Abusamra, Ms Bridges and Prof. Hester. Ms Tomljanovic (bulk preclinical and clinical datasets), Dr Pilalis (bulk preclinical datasets), Dr Parmentier (single cell datasets), Dr Figiel (spatial datasets), Dr Zekri (ChIP-seq) and Dr Doultsinos (bulk clinical/preclinical, spatial datasets) carried out data analysis. Dr Leach and Prof. Bevan generated and provided PDEs. Prof. Chatziioannou, Dr Lamb, Dr Le Magnen and Prof. Zwart provided study design input and access to datasets and samples.

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Conflicts of interest: AC, EP are co-founders of e-NIOS Applications PC (https://e-nios.com/).

Materials & Methods

Cell lines & Culture Conditions

LNCaP and C4-2B cells were purchased from ATCC and cultured in RPMI supplemented with

10% foetal bovine serum (FBS) or DMEM supplemented with 10% foetal bovine serum (FBS)

respectively. LNCaP knocked out for RB1 or knocked out for both RB1 and TP53 were kindly

donated by Professor Peter Nelson (Fred Hutchinson Cancer Centre) and were cultured in RPMI

supplemented with 10% foetal bovine serum (FBS). LNCaP knocked out for ERN1 were kindly

donated by Professor Wilbert Zwart (The Netherlands Cancer Institute) and were cultured in

RPMI supplemented with 10% foetal bovine serum (FBS). No antibiotics were used in the

culturing of these cells as they may have the capacity to bind a variety of kinases including IRE1

as we previously reported[44]. For androgen deprived media, Charcoal Stripped FBS was used

instead of normal FBS to supplement RPMI.

Primers and qPCR

RNA was extracted from cell pellets using TRIzol reagent (Fisher Scientific, #12034977)

according to manufacturer's instructions. cDNA was reverse transcribed from 2000ng of RNA

(measured by Nanodrop) using Maxima Reverse Transcriptase (ThermoFisher Scientific

EP0743) and associated reagents (RiboLock-Life Technologles-EO0381; DNTPs-Thermo

Scientific-18427013; Random Hexamer-Life Technologies-SO142). qPCR was executed on a

QuantStudio5 (Applied Biosystems) using Power SYBR Green PCR master mix (ThermoFisher

Scientific) and results were analysed using the Design & Analysis Software 2.8.3 (Applied

Biosystems).

Gene	Direction	Sequence	Manufacturer
DGAT2	Forward	GGCTGGTGTTTGACTGGAA	Eurofins Genomics
	Reverse	TCAGCAGGTTGTGTGTCTTC	
ACTIN	Forward	AGAGCTACGAGCTGCCTGAC	
	Reverse	AGCACTGTGTTGGCGTACAG	
XBP1s	Forward	TGCTGAGTCCGCAGCAGGTG	
	Reverse	GCTGGCAGGCTCTGGGGAAG	
ERN1	Forward	GCCACCCTGCAAGAGTATGT	
	Reverse	ATGTTGAGGGAGTGGAGGTG	
POLQ	Forward	CCTGTGGCATCTCCTTGGAGCA	
	Reverse	AATCCCTTGGCTGGTCTCCATC	
KLK3	Forward	CGTGACGTGGATTGGTGC	
	Reverse	ACTGCCCTGCCACGAGAG	
HELLS	Forward	GATTTTGGATCGAATGCTGCCAG	
	Reverse	ATGGACCCATCAAGCCTGCTGA	
AR	Forward	AAGACCTGCCTGATCTGTGG	
	Reverse	CGAAGACGACAAGATGGACA	
CAMKK2	Forward	TGAAGACCAGGCCCGTTTCTACTT	
	Reverse	TGGAAGGTTTGATGTCACGGTGGA	
BRCA2	Forward	AATGTCAGACAAGCTCAAAG	Merck
	Reverse	TCATGTATTTTTCAGGTGGC	

RB1	Forward	GACCCAGAAGCCATTGAAATCT	Eurofins Genomics
	Reverse	GGTGTGCTGGAAAAGGGTCC	
SSR2	Forward	CTTCACCTCGGCAACAATTACT	Sigma Kicqstart
	Reverse	GGGGAGAATCGC CTGTCAAAC	
SOX2	Forward	GAGCTTTGCAGGAAGTTTGC	Eurofins Genomics
	Reverse	GCAAGAAGCCTCTCCTTGAA	
POU3F2	Forward	ACACTGACCGATCTCCACGCAGTA	
	Reverse	GAGGGTGTGGGACCCTAAATATGAC	
EZH2	Forward	AATAAAAGCGATGGCGATTGG	
	Reverse	TCCACTGCCTTCTGAGTCC	
ATF4	Forward	GTTCTCCAGCGACAAGGCTA	
	Reverse	ATCCTGCTTGCTGTTGTTGG	
eIF2α	Forward	GCCTTTCTGTCCTCAGTCAAGC	
	Reverse	CTCTCCATCTGCCTCGCAAGTT	
ATF6	Forward	CTCCGAGATCAGCAGAGGAA	
	Reverse	AATGACTCAGGGATGGTGCT	
LGI2	Forward	TGGAGAACTGCATGGTGCTGGA	
	Reverse	CTTACAGCCCACGATGGACTGA	
VSTM2L	Forward	GATCCAGTGGTGGTATGTACGG	
	Reverse	ACACTTATTTTGGTTGCCTCCTTC	
SNHG8	Forward	AAGTTTACAAGCATGCGCGG	
	Reverse	TCAAACTGACGGTTCTCGGG	

Drug Treatments

734   Cells were seeded in 96, 24 or 6 well plate format depending on the assay (96 well plate for

735   viability, 96 or 24 well plate for growth, 6 well plate for RNA extraction) and drugs were added

736   using the range of concentrations below.

Compound	Manufacturer	Lot number	Concentrations
IXA4	Cayman Chemical	0653659-8	10 - 25 µM
MKC8866	Cayman Chemical	0664887-3	5 µM
Thapsigargin	Sigma	0000120608	100 nM
Tunicamycin	Sigma	079M4004V	500 nM
Enzalutamide	Selleck Chemicals	MDV3100	10 µM

Cell Titer Glo sensitivity assays

739   Cells were seeded into 96-well plates and incubated at 37 °C for 24 hours. They were then

740   treated with drug combinations for a further 12 hours. Percentage survival was measured

741   photometrically using the CellTiter-Glo 2.0 viability assay (Promega) in a microplate reader

742   (FLUOstar OMEGA 415-2812). Survival was normalized to the control treatment.

Growth assays

744   Cells were seeded into 96-well plates and incubated at 37 °C for 24 hours. Complete media

745   change followed to either full (FM) or charcoal stripped media (CSM) for 120 hours. Cell growth

746   was monitored using Incucyte technology taking confluence measurements every 24 hours.

747   Growth was normalised to the initial measurement at time 0.

Transfection assays and RNA silencing.

For stable expression of IRE1 mutant Q780*, LNCaP ERN1$^{-/-}$ cells were transfected with the

Q780* construct[14] using LTX Plus reagent (ThermoFisher Scientific, 15338100) according to

the manufacturer's instructions. Successfully transfected cells were selected by cell sorting

for GFP expression using a BD FACSAriaIII cell sorter. Thereafter, constant GFP expression was

monitored using the ZOE™ Fluorescent Cell Imager. Lipofectamine RNAiMax transfection

reagent (Thermofisher Scientific, 13778150) was used for transfecting siRNAs. Cells were

incubated at 37 °C for 48 hours before extracting RNA for downstream analysis.

siRNA target	Sequence		Manufacturer
	Sense	Antisense	
XBP1	CCUGCCUACUGGAUGCUUAtt	UAAGCAUCCAGUAGGCAGGaa	Ambion
IRE1	GAAACUUCCUUUUACCAUCtt	GAUGGUAAAAGGAAGUUUCgt	Ambion
Control	Silencer Select Negative Control #1		Ambion

Immunoblotting and Antibodies

Antibody	Manufacturer	Catalog number	Incubation dilution
Anti-IRE1 (phospho S724) antibody (rabbit)	Abcam	ab48187	1 in 1000
Anti-XBP1 antibody (rabbit)	Abcam	ab37152	1 in 1000
Anti-CAMKK2 antibody (rabbit)	Sigma-Aldrich	HPA017389	1 in 1000
Anti-alpha tubulin (DM1A) antibody (mouse)	Santa Cruz Animal Health	sc-32293	1 in 5000
Monoclonal beta Actin antibody (mouse)	Abcam	ab6276	1 in 5000
Anti-rabbit IgG, HRP-linked Antibody (goat)	Cell Signaling	#7074	1 in 2000
Anti-mouse IgG, HRP-linked Antibody (horse)	Cell Signaling	#7076	1 in 2000

Online Tools/Databases

- <u>cBioportal</u>: RNA sequencing datasets investigated included the Metastatic Prostate Adenocarcinoma (SU2C/PCF Dream Team, PNAS 2019[45]), Prostate Cancer (DKFZ, Cancer Cell 2018[46]), Prostate Adenocarcinoma (MSK, Cancer Cell 2010[47]) and Prostate Adenocarcinoma (TCGA, Firehose Legacy or PanCancer). The comparisons were based on log-transformed mRNA expression z-scores compared to expression distribution of all samples (RNA Seq FPKM) or of tumour samples compared to the expression distribution of all adjacent normal samples in the cohort when available with a z score threshold of +/- 2.0. Altered mRNA was defined as mRNA significantly high or low compared to all or normal samples in each dataset using the z score threshold. Kaplan Meier curves were generated for the Prostate Adenocarcinoma (SU2C/PCF Dream Team, PNAS 2019) and Prostate Adenocarcinoma (TCGA, PanCancer) datasets. Only samples from patients with survival data or progression data were used in the generation of these curves respectively. Samples with any significant alteration in mRNA expression were compared to samples with zero significant changes in the former whilst samples with significant alterations in at least three genes were compared to samples with zero significant changes in the latter.

- <u>Prostate Cancer Atlas</u>[32]: DESeq2 differential expression analysis was carried out foreach of sample groups designated as Primary, AR+ CRPC, NEPC and DNPC compared to the sample group designated as normal tissue. Differentially significant mRNA expression changes ($padj<$/=0.01) were documented to mine for genes of interest representing IRE1 activity. Pseudotime disease progression plots were generated for "IRE1 induced" and "IRE1 repressed" sub-groups of the IRE1 activity gene set using the in-built functionality of the Prostate Cancer Atlas

(https://prostatecanceratlas.org/app/home). Further in-built functionality was used to

analyse gene set enrichment of our IRE1 activity gene set in the normal, primary, AR+

CRPC, NEPC and DNPC samples groups.

• Correlation AnalyzeR[48]: The online interface of this tool was used to verify the

correlation of the IRE1 activity gene set with other significantly differentially expressed

genes in prostate cancer and the biological processes it represents.

• CAMCAPP[49]: This tool was used to verify gene by gene correlation of the IRE1 activity

gene set with clinical co-variates in three separate clinical cohorts (Stockholm,

Cambridge, MSKCC).

RNA extraction, library prep and sequencing

RNA was extracted from cell line pellets using the RNeasy Plus Mini kit. Samples were QC'd

using Nanodrop and the Qubit Flex High Sensitivity ID kit. Libraries were prepared using the

TruSeq Stranded Total RNA with Illumina Ribo-Zero plus rRNA Depletion workflow

(https://emea.support.illumina.com/downloads/truseq-stranded-total-rna-with-ribo-zero-

plus-rrna-depletion-ref-guide.html) . Index adapters, rRNA depletion kit were acquired in

addition   to   the   TruSeq   Stranded   Total   Library   Prep   Gold   kit

(https://emea.illumina.com/products/by-type/sequencing-kits/library-prep-kits/truseq-

stranded-total-rna.html#tabs-bb5d9c771f-item-01fd0b3d08-order). cDNA library QC was

carried out on Tapestation and Libraries were pooled, diluted to ~10 nM, denatured, and further

diluted for paired-end sequencing (NovaSeq6000 platform, Illumina, NovaSeq 6000 S2/S4

reagent kit v1.5, 300 cycles). A test run prior to full sequencing was executed on a MiSeq

system. Bulk paired-end RNA sequencing was performed to obtain at least 20mn reads per

sample. Raw data were processed using a standardized computational workflow to obtain

gene-level raw counts. FASTQ files were quality-checked using FastQC to assess read quality

and adapter contamination. High-quality reads were aligned to the human reference genome

(GRCh38, Ensembl release 108) using STAR aligner (Dobin et al., 2013). Gene-level

quantification was performed using HTSeq-count (Anders et al., 2015).

RNA Sequencing Data analysis

1. Spatial transcriptomics data were obtained from a previously published dataset derived

from radical prostatectomy tissue of a patient diagnosed with multifocal prostate cancer

(Nature). Each spatial transcriptomics spot was reviewed by two independent pathologists

as part of a consensus approach for pathological annotation.

• Virtual Biopsy: The analysis was performed using the same approach as previously

described (Molecular Cancer). Gene expression data were extracted from the barcode ID

of designed biopsies (LoupeBrowser, v8.0.0) using R (v4.4.0). Libraries were normalised

using the spaceranger aggr (v2.0) function, which corrects for batch effects to facilitate the

comparison of gene expression data across samples and minimise technical variability. The

analysis included seven distinct tissue sections, representing a total of 3721 spatial

transcriptomics spots. To maintain data integrity, spots with fewer than 500 Unique

Molecular Identifier (UMI) counts were removed. Fold changes were calculated using

EdgeR (v3.40.1). as previously described[37].

Single Visium 6.5mm squares: The raw sequencing reads were directly processed using

Space Ranger and mapped using reference genome GRCh38, release 93 as previously

described[34]. Each histological square was treated as a distinct spatial transcriptomic

experiment to determine if, in an independent manner, gene sets could be used as

distinguishing tools between histological characteristics. Histological annotations (benign,

GG1, GG2, GG4, stroma) were compared using either the locally distinguishing or globally

distinguishing functions of the Loupe Browser which was also used to generate ST images

displayed.

2. Single Cell

PCa tissue data from Chen et al. were accessed via the GEO online portal (GSM4203181) as a raw count matrix. Quality control filtering was applied based on the criteria described in the original publication: cells with fewer than 200 or more than 5,218 detected genes, more than 20% mitochondrial content, and fewer than 56 housekeeping genes were discarded. The filtered data were then converted into an SCE object, separated by patient ID, and normalized using a cluster-based normalization factor computed after each sample was first partitioned into biologically relevant clusters using the quickCluster(min.size = 100) function from the scran R package (v1.26.2).

Next, to correct for differences in sequencing depth across batches, all SCE objects were rescaled using the MultiBatchNorm(min.mean = 0.1) function from the Batchelor R package (v1.14.1). Epithelial cells were retrieved using the singleR package (v2.0.0), leveraging the BlueprintEncodeData reference accessed via the celldex package (v1.16.0). Using scran (v1.26.2), variance modeling of gene expression was performed for each sample individually with the modelGeneVar() function; results were merged with combineVar(), and the top 1,000 most highly variable genes were selected using getTopHVGs() for subsequent analysis. Batch correction was applied using the fastMNN(k = 300, d = 20) function from the Batchelor package.

To separate luminal malignant cells from healthy epithelial cell types, clustering was performed on the integrated data using the clusterCells() function from the scran R package (v1.26.2). The clustering was conducted with the Walktrap algorithm, exploring different values of k = 10, which resulted in 14 distinct clusters. All clusters were considered tumor except cluster 7 and 14, which were negative of both tumor and luminal markers (e.g AMACR, PCA3, KLK3) and were positive for basal markers such as KRT19, KRT18 and KRT14.

PCa tissue data from Hirz et al. were accessed via the GEO online portal (GSE181294) as a raw

count matrix. Cells passing quality thresholds and cell type annotation, as defined in the

original publication, were retrieved from the annotation table provided by the authors and

stored in an SCE object. Similar to the Chen et al. dataset, each sample first underwent cluster-

based normalization followed by inter-sample correction to account for sequencing depth

differences between batches.

3. Bulk

The RNA sequencing data (raw counts) of the Labrecque mCRPC patient cohort (accession

number GSE126078) and the RB1$^{-/-}$/TP53$^{-/-}$ cell line dataset (accession number GSE147250)

was retrieved from GEO. TCGA data was accessed via the GDC Data Portal

(https://portal.gdc.cancer.gov/projects/TCGA-PRAD, TCGA-PRAD.2022-04-27). The Beltran

patient cohort FPKM data was obtained from the supplementary material of Beltran et al.,

2016.

Data visualization and statistical analyses of the bulk RNA sequencing datasets were

performed using R statistical software (v4.4.1). Batch effect correction was executed using the

sva package (v.3.54.0). Differential gene expression analysis was performed using the DESeq2

package (v1.40.2), with subsequent LFC shrinkage using the 'apeglm' estimator. Gene set

enrichment analysis (GSEA) was conducted on a subset of the Molecular Signatures Database

(MSigDB) v2023.2.Hs using the clusterProfiler package (v4.8.2). Gene Set Variation Analysis

(GSVA) was used for single-sample pathway enrichment as implemented in the GSVA package

(v1.48.3). Differential expression at the pathway level was assessed using the lmFit function

from limma (v3.62.2). For visualization and clustering purposes, the variance stabilizing

transformation (VST) was applied using the DESeq2 package. Visualization was performed

using ggplot2 (v3.5.0) and heatmaps were generated using the pheatmap package (v1.0.12).

The viper R package (v1.4.0) was used for single-sample regulon enrichment and differential

testing of regulon activity levels between sample groups. The regulons of 1695 human

transcription factors were previously inferred using ARACNE -AP on an in-house set of mCRPC

patient tumors. Unless otherwise specified, the statistical significance threshold across all

analyses was set at 0.05 and a Benjamini–Hochberg correction was applied to account for

multiple testing.

891