Dual Functions of SPOP and ERG Dictate Androgen Therapy Responses in Prostate Cancer

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Driver genes with a mutually exclusive mutation pattern across tumor genomes are

thought to have overlapping roles in tumorigenesis. In contrast, we show here that mutually-

exclusive prostate cancer driver alterations involving the *ERG* transcription factor and the

ubiquitin ligase adaptor *SPOP* are synthetic sick. At the molecular level, the incompatible

cancer pathways are driven by opposing functions in SPOP. ERG up-regulates wild type

SPOP to dampen androgen receptor (AR) signaling and sustain ERG activity through

degradation of the bromodomain histone reader ZMYND11. Conversely, SPOP-mutant

tumors stabilize ZMYND11 to repress ERG-function and enable oncogenic androgen

receptor signaling. This dichotomy regulates the response to therapeutic interventions in the

AR pathway. While mutant SPOP renders tumor cells susceptible to androgen deprivation

therapies, ERG promotes sensitivity to high-dose androgen therapy and pharmacological

inhibition of wild type SPOP. More generally, these results define a distinct class of

antagonistic cancer drivers and a blueprint toward their therapeutic exploitation.

Normal cells transform into cancer cells by the acquisition of genetic aberrations in so

called driver genes. In some instances, the functional redundancy of mutations in different

genes result in a mutually-exclusive mutation pattern across tumor genomes because one

alteration is sufficient to activate the specific oncogenic pathway. Based on this

assumption, bioinformatic tools have been generated to search for functional redundancy

of mutated genes in larger cancer genome data sets[1,2].

In prostate cancer, recurrent gene fusions involving the *ERG* transcription factor and

point mutations in the ubiquitin ligase adaptor *SPOP* are two truncal mutations that are

mutually exclusively distributed across tumor genomes (Fig. 1a and Supplementary Fig.

1a)[3-7]. The underlying cause for this exquisite pattern remains controversial. While earlier

reports suggested a functional redundancy between mutant SPOP and ERG based on the

finding that mutant SPOP stabilizes the ERG oncoprotein[8,9], more recent studies challenge

this view by showing descriptive evidence for divergence in tumorigenesis[3,10].

RESULTS

Activation of the ERG oncogene and missense mutation in SPOP are synthetic sick

To shed light into the functional relationship of these recurrent driver genes, we

assessed the impact of SPOP mutations and ERG activation on the cellular growth of

mouse prostate epithelial organoids. To do so, we first established and validated the

organoids by the presence of multilayered structure with expression of CK5 and CK8 in

basal and luminal cells, respectively (Supplementary Fig. 1b). In agreement with recent

reports, lentiviral-transduced point mutants of SPOP (SPOP-Y87C, SPOP-W131G) or a

truncated version of ERG, which typically results from gene fusion with androgen-

regulated genes in prostate cancer (ΔERG, amino acids 33-486), promoted cell growth (Fig.

1b, Supplementary Fig. 1c)[11-14]. While SPOP mutant organoids displayed a round shape,

the over-expression of ΔERG gave rise to characteristic finger-like protrusions.

Surprisingly, the joint expression of both drivers considerably diminished cell growth and

reduced finger-like protrusions, implying a synthetic sick relationship between the two

genetic alterations. Cytological follow up analysis revealed reduced proliferation

evidenced by reduced Ki-67 and increased p16, P-HP1y positivity and vacuolization of the

cytoplasm compatible with senescence induction (Supplementary Fig. 1d).

We wondered if the observed synthetic sick relationship also applied to established

cancer cells from advanced, castration-resistant metastatic disease. Forced expression of

mutant SPOP (SPOP-Y87C, SPOP-W131G) promoted 3D growth of *ERG* fusion-negative

LAPC-4 human prostate cancer cells (Supplementary Fig. 2a). The oncogenic effect was

paralleled by an increase in the expression of the oncogenic transcription factors MYC and

HOXB13 and a decrease in the cell cycle inhibitor p21 as seen also in an organoid line

derived from $Spop^{F133V}$-mutant transgenic mice (Supplementary Fig. 2a,b)[11]. In contrast,

we observed the opposite phenotypic and molecular changes in VCaP human prostate

cancer cells harboring the recurrent *TMPRSS2-ERG* fusion (Fig. 1c, d & Supplementary

Fig. 2c, d). In this setting, mutant SPOP (SPOP-Y87C, -F102C, -W131G, -F133S)

dramatically decreased the proliferation of cancer cells in culture and the growth of

xenograft tumor models *in vivo*. Similarly, to the mouse prostate organoids, induction of

senescence was evidenced by an increase of senescence-associated β-galactosidase (SA-β-

gal) positive cells and upregulation of p21 and GDF15 protein levels. In line with this, the

transfer of conditioned medium from VCaP cells expressing mutant SPOP (SPOP-Y87C ,

SPOP-W131G)   reduced also the proliferation of parental VCaP cells, indicating a

contribution of Senescence-Associated Secretory Phenotype (SASP) to the synthetic sick

relationship (Fig. 1d and Supplementary Fig. 2f,j).

Conversely, forced expression of ΔERG significantly reduced the growth of SPOP-

Y83C mutant LuCaP-147 patient-derived xenograft (PDX) cancer cells *in vivo* and in

culture (Fig. 1e, Supplementary Fig. 3a, b)[15], adding orthogonal support for a synthetic sick

relationship between mutant SPOP and ΔERG in advanced prostate cancer. In addition,

over-expression of MYC promoted cancer cell growth in both VCaP and LuCaP-147 cells

(Supplementary Fig. 3c, d). The latter finding may suggest that the over-expression system

per se is not the underlying cause of the synthetic sick relationship mentioned above.

Next, we wondered if genetic or pharmacologic suppression of ERG signaling may

revert the growth suppressing function of mutant SPOP in VCaP cells. Indeed, knockdown

of ERG by short-hairpin RNA interference decreased the growth of VCaP control cells and

of cells over-expressing wild-type SPOP, while it promoted the growth of cells over-

expressing SPOP-W131G (Supplementary Fig. 3e). In addition, low doses of the ETS

inhibitor YK-4-279 promoted specifically the growth of VCaP cells over-expressing

mutant SPOP (Fig. 1f). We noted a similar effect when VCaP cells were co-treated with a

small molecule inhibitor of SPOP (Supplementary Fig. 3f)[16]. In aggregate, the data support

an antagonistic relationship between oncogenic activation of ERG and a loss of SPOP

function in prostate cancer cells.

Mutant SPOP-induced androgen receptor signaling antagonizes ERG activity

To assess the underlying molecular biology of the antagonistic relationship between

ERG-fused and *SPOP*-mutants tumors, we interrogated the transcriptomes from the TCGA

cohort to nominate differences across these tumor subtypes. Indeed, the unbiased principal

component analysis (PCA) revealed major differences in the transcriptional output (Fig.

2a). The differences were maintained in castration-resistant prostate cancers (CRPC) from

the (SU2C) cohort using a single-sample gene-set enrichment analysis approach (Fig. 2b).

Furthermore, derived (PDX) models also retained analogue transcriptional differences, as

demonstrated by different behavior shown by *SPOP*-Mutant (LuCaP-78, -147) and *ERG*-

Fused (LuCaP-35, -23.1, VCaP cells) models (Fig. 2b).

In *SPOP*-mutant prostate cancer, several dysregulated SPOP substrates (e.g. NCOA3,

TRIM24, BET proteins) have been shown to boost the AR pathway leading ultimately to

high levels of AR target genes (Supplementary Fig. 4a)[3,11,17-24]. In contrast, *ERG*-fused

cancer cells express typically lower levels of AR target genes as illustrated by the widely

adopted AR score (Supplementary Fig. 4a)[3]. To get more insights into this different

behavior, we further performed differential expression analysis between the two tumor

subtypes. Gene-set enrichment analysis resulted in a clear up regulation of the canonical

androgen response pathway in *SPOP*-mutant versus *ERG*-fused tumors, as defined by the

respective Hallmark gene-set, curated by the Molecular Signature Database (MSigDB)

(Supplementary Fig. 4b). In line with the divergence of transcriptome profile identified in

the PCA plot of Fig. 2a, a clear division between *SPOP*-mutant and *ERG*-fused tumors was

also reported in their respective cistrome counterpart[25]. By re-analyzing ChIP-Seq data, we

could determine that most differentially bound regions between both tumor types, are

indeed characterized by increased AR-binding in *SPOP*-mutant tumors (Fig. 2c).

Next, we analyzed the transcriptome changes of the VCaP cells overexpressing SPOP

mutants (SPOP-MTs; SPOP-Y87C, -F102C, -F133S). The unbiased hallmark analysis

showed a dramatic increase in the androgen response, recapitulating the changes identified

in primary prostate cancer (Supplementary Fig. 4b, 5a and 5b). Based on these results, we

posited that differential levels of androgen receptor (AR) signaling in *SPOP*-mutant versus

ERG-fused cancers might be at the root of the incompatibility between the driver events.

Thus, we analyzed in particular AR- and ERG-related transcription in VCaP cells, and

generated a custom signatures using ChIP-seq data and matched RNA-seq samples

(Supplementary Table 1)[25]. As expected, SPOP-MTs increased the transcription of genes

bound by AR and induced by its ligand dihydrotestosterone (DHT), whereas genes bound

by AR and repressed by DHT were further reduced (Fig. 3a, Supplementary Fig. 5c, and

Supplementary Table 1). Remarkably, we observed the opposite effect on genes bound

only by ERG. Mutant SPOP downregulated ERG-induced genes (e.g. MYC) and

upregulated ERG-repressed genes, respectively (Fig. 1d). In line with these findings, gene

ontology analysis of AR-ERG co-bound gene signature in VCaP cells indicated that the

most striking transcriptional changes were linked to cellular differentiation and cell cycle

arrest that are directly induced by DHT and repressed by ERG (e.g. HOXA genes,

CDKN1A/p21, Fig. 1d, Fig. 3b, Supplementary Fig. 5d). To reduce the number of genes

falling within our custom signatures, we used a particularly restrictive approach and

considered as co-bound only genes where AR and ERG binding sites were overlapping for

at least 1bp. As a result, some genes (i.e. CDKN1A/p21) which are bound both by AR and

ERG in their promoter region, but which bindings do not overlap, are not included in this

category despite being bona fide co-bound targets.

The dramatic upregulation of this gene set was paralleled by a downregulation of cell

cycle genes (e.g. E2F and MYC targets), implying a direct link between the induction

AR/ERG co-bound genes, the repression of ERG targets, cell differentiation and the

synthetic sick relationship of ERG and mutant SPOP (Fig. 1d, Supplementary Fig. 5a-e).

The relationship of AR- and ERG-related custom signatures to the hallmark gene sets are

highlighted in Fig. 3c in a two-dimensional network. Moreover, independently generated

signatures of senescence-associated transcripts were enriched in VCaP overexpressing

SPOP mutants, further corroborating our data of a senescence-induced cell cycle arrest (

Supplementary Fig. 5f)[26,27].

Conversely, we assessed the consequence of ERG over-expression in LNCaP cells

under low DHT levels where mutant SPOP triggers AR signaling and tumor growth

(Supplementary Fig. 6a,b)[17,21]. Over-expression of ΔERG in this setting robustly reverted

the induction of signatures related to cell proliferation (e.g. E2F and MYC targets) and AR

signaling. Taken together, the data implies a reciprocal incompatibility of mutant SPOP

induced AR signaling and the function of the ERG oncogene.

Next, we verified if corresponding transcriptional changes were found in clinical tissue

samples. Indeed, ERG-regulated genes culled from VCaP cells were up-regulated in *ERG*-

fused and down-regulated in *SPOP*-mutant primary tumors (Fig. 3d, Supplementary Fig.

4c)[3]. Importantly, the most striking changes between the two groups were found again in

the AR/ERG co-bound gene set in primary prostate cancers (Supplementary Fig. 4d)[3,6].

The results underscore both the relevance of our cell culture-based data and highlight the

transcriptional differences among *ERG*- and *SPOP*-driven tumors.

ZMYND11 is a *de novo* SPOP substrate

Using tandem mass tag (TMT)-based quantitative mass-spectrometry, we set out to

search for SPOP substrates that may influence the activity of AR and ERG and thereby

may cause to the synthetic sick relationship between mutant SPOP and ERG in VCaP cells

overexpressing mutant SPOP (SPOP-MTs; SPOP-Y87C, -F102C, -W131G, Fig. 4a).

Because recurrent loss-of-function SPOP mutants impair substrate ubiquitylation and

proteasomal degradation, we searched for proteins which expression levels increase

without concomitant increase in mRNA levels (Fig. 4b, Supplementary Fig. 7a). Overall,

we noted a strong correlation of protein with mRNA expression changes with consistent

changes of our AR and ERG custom signatures at the protein level (Fig. 4b, Supplementary

Fig. 7b, c). In addition, we found a marked upregulation of the known SPOP substrate and

AR activator TRIM24 at the protein level (Fig. 1d & 4a, Supplementary Table 2)[20,21] and

subsequently assessed if TRIM24 and more generally AR is implicated in the synthetic

sick relationship between mutant SPOP and ERG. Indeed, knockdown of TRIM24 by two

short hairpin RNAs partially reverted the growth inhibition mediated by mutant-SPOP in

VCaP cells and reduced AR signaling (Supplementary Fig. 7d-f), while over-expression of

AR was sufficient to decrease cellular growth (Supplementary Fig. 7g, h).

The most striking upregulation was noted for the bromodomain histone reader

ZMYND11 (Fig. 4b). In line with a SPOP substrate, wild type SPOP bound and decreased

the expression of HA-ZYMND11 in a proteasome-dependent manner (Fig. 4c, d). We

found two degron sequences that were required for efficient SPOP-mediated ubiquitylation

and protein degradation (Fig. 4e-g, Supplementary Fig. 8a.). As expected, SPOP mutants

failed to bind and adequately ubiquitylate HA-ZMNYD11-WT (Fig. 4h-j)[8,9,18-20,22-24].

Finally, we confirmed that expression of mutant SPOP prolonged the half-life of

endogenous ZMYND11 in VCaP cells and upregulated ZMYND11 expression in other

prostate cancer cells (Fig. 4k and Supplementary Fig. 8b).

ZMYND11 induces AR signaling pathway and represses ERG activity

Next, we assessed if ZMYND11 protein upregulation also contributed to the synthetic

sick relationship. In support, forced expression of the degron-deficient variants of

ZMYND11 (HA-ZMYND11-DMT1/DMT2) was sufficient to diminish the growth of

VCaP cells (Fig. 5a, b, Supplementary Fig. 8c), while knockdown of ZMYND11 partially

reverted the growth inhibition mediated by mutant SPOP (Fig. 5c).

We postulated that ZMYND11 up-regulation could contribute to the synthetic sick

relationship by repressing the transcriptional activity of the ERG oncogene or enhancing

AR signaling. To this end, expression changes induced by HA-ZMYND11-DMT2 largely

overlapped with genes perturbed by mutant SPOP while the opposite was noted when

ZMYND11 expression was reduced by RNA interference (Fig. 5d and Supplementary Fig.

8d). In comparison to mutant SPOP, AR and ERG target genes were similarly dysregulated

by HA-ZMYND11-DMT2 (Fig. 5e). Because the PWWP domain of ZMYND11 has been

involved in the regulation of transcription through its ability to bind H3K36me3 histone

marks[28], we tested the contribution of this domain to the overall transcriptional output.

Indeed, the PCA of VCaP cells over-expressing either HA-ZMYND11-DMT2 or a PWWP

domain deficient mutant (W294A) revealed a major contribution of this domain to the

ZMYND11 induced transcriptional changes (Supplementary Fig. 8e).

Subsequently, we mapped the genomic occupancy of ZMYND11 in VCaP cells

expressing the SPOP-Y87C mutant by chromatin immunoprecipitation sequencing (ChIP-

seq) and found an enrichment of ZMYND11 binding sites at promoter regions controlling

ERG-induced genes (e.g. MYC,) and AR/ERG co-bound genes (e.g. p21/CDKN1A) (Fig.

5f, g and Supplementary Fig. 9a-e). The data imply a critical enhancer function of

ZMYND11 in boosting AR signaling and repressing ERG signaling downstream of mutant

SPOP.

Wild type SPOP is required for ERG oncogenic function

We reasoned that ERG-driven tumors might require wild type SPOP to degrade

ZMYND11 and thereby unlock the oncogenic function of ERG. In support, over-

expression of wild type SPOP increased the 3D growth of mouse prostate epithelial

organoids and VCaP cells only when ERG was over-expressed (Fig. 1b, Supplementary

Fig. 2c, 10a, b). Remarkably, ERG-fused human tumor tissues displayed also the highest

SPOP mRNAs levels (Fig. 6a). Thus, we wondered if ERG itself may directly upregulate

SPOP transcription to support its own oncogenic activity. Indeed, mining ERG ChIP-seq

data in VCaP cells revealed ERG bindings sites in the promoter region of SPOP

(Supplementary Fig. 10c). Moreover, knockdown of ERG reduced SPOP protein levels in

VCaP cells, while forced expression of a ΔERG led to the upregulation of SPOP mRNA

and protein levels in PC3 cells (Fig. 6b, Supplementary Fig. 3e, and 10d).

We then asked if the elevated SPOP levels in the context of forced ΔERG expression

have a functional impact on the oncogenic activity of ΔERG in the androgen-independent

PC3 cells, in which ERG promotes tumor cell invasion[29]. Indeed, the reduction of SPOP

levels by RNA interference reduced the ability of ΔERG to invade into matrigel (Fig. 6c).

Similarly, knockdown of SPOP in VCaP cells reduced cell growth in 3D cell culture and

impaired ERG-mediated gene transcription (Supplementary Fig. 10e, f). In accordance

with the ability of mutant SPOP to repress the function of endogenous wild type SPOP in

a dominant-negative manner, the over-expression of mutant SPOP (SPOP-Y87C, -F102C,

-W131G, -F133S) phenocopied the effect of SPOP knockdown on ERG-mediated invasion

in PC3 cells (Supplementary Fig. 10g, h). In agreement with the established repressive

function of ZMYND11 on ERG, we found that over-expression of HA-ZMNYD11-DM2

was sufficient to repress ERG-induced invasion and established target genes in PC3 cells

(Fig. 6d, e). Taken together, the data imply the existence of a positive feed-forward loop,

in which ΔERG promotes the expression of SPOP to sustain its oncogenic activity.

ERG and mutant SPOP trigger different responses to therapeutic interventions

Based on the above-mentioned differences in tumorigenesis, we speculated that ERG

or mutant SPOP could also trigger different therapeutic responses. In light of the

dependency of ERG-driven tumors on wild type SPOP function, we hypothesized that

ERG-fusion positive cells may be particularly sensitive to pharmacological inhibition of

SPOP. We analyzed the response of the SPOP small molecule inhibitor compound 6b

(SPOP-i) in *ERG*-fused, *SPOP* mutant and other prostate cancer cell lines and patient-

derived xenograft models (PDX)[16]. The SPOP inhibitor increased the protein but not the

mRNA levels of established SPOP substrates and ZMNYD11 while the related inactive

analog compound 6c did not (Supplementary Fig. 11a-c). The latter did also not exert any

activity in 3D culture models (Supplementary Fig. 11d). In agreement with our previous

results, we found that *ERG*-fused cells (VCaP, LuCaP-23.1, -35) were more sensitive to

SPOP-i than *ERG*-negative cells (22Rv1, LNCaP, PC3), while *SPOP* mutant cells (LuCaP-

78, -147) were particularly insensitive in 3D culture models and in xenograft tumor models

in vivo (Fig. 7a-f, Supplementary Fig. 11e). We further validated our results in the mouse

prostate epithelial organoids and confirmed the increased sensitivity of ΔERG-expressing

cells to SPOP inhibition in this isogenic system (Fig. 7g).

Given the notion that wild type SPOP dampens AR function in the context of ERG to

sustain tumor growth, we asked if VCaP cells are particularly susceptible to increased DHT

levels. Indeed, exposure to high-dose of testosterone *in vivo* or DHT *in vitro* induced

similar molecular changes as for the over-expression of mutant SPOP and greatly

suppressed the growth of *ERG*-fusion positive cells but not of *SPOP* mutant cells *in vitro*

and *in vivo* (Fig. 1c, 8a-f and Supplementary Fig. 12a-e). Moreover, signatures of

senescence-associated transcripts were also enriched in VCaP cells upon treatment with

DHT, further corroborating our data of a senescence-induced cell cycle arrest

(Supplementary Fig. 12f). Strikingly, the sensitivity to SPOP-i and to high testosterone *in*

vivo correlated well with ERG protein expression levels in the respective ERG-fusion

positive cell line and PDX model (Fig. 8g, h). The data suggests a therapeutic opportunity

for SPOP inhibition or high-dose androgen therapy in prostate cancers that express high

levels of ERG.

Conversely, and because *SPOP* mutant cancers are driven predominantly by androgen

signaling and consequently display high-level activation of AR-related transcripts in

human tumor tissues, we speculated that these tumors may be particularly susceptible to

androgen deprivation or anti-androgen therapies (ADT) (Supplementary Fig. 4c). Indeed,

the prevalence of *SPOP* mutations in primary tumors -and tumors that had progressed after

initial surgery or radiotherapy- is consistently higher as compared to tumors that had

become resistant to subsequent ADT (also referred as castration-resistant prostate cancer,

CRPC, Supplementary Fig. 13a). In line with the notion that this difference may be related

to a better response of SPOP mutant tumors to ADT, SPOP mutant tumor display a trend

towards better overall survival despite progressing faster after initial therapy (Fig. 9a,b).

To functionally analyze the response of androgen deprivation or the anti-androgen

enzalutamide, we chose to ectopically expressed different SPOP variants and ΔERG in the

androgen-dependent human LAPC4 prostate cancer cells that are wild-type for both driver

genes. In accordance with the clinical observation, the presence of mutant SPOP (SPOP-

Y87C, SPOP-W131G) rendered LAPC4 cells more susceptible to either ADT or

enzalutamide in comparison to cells expressing control vector (Fig. 9c, Supplementary Fig.

13b). In contrast, ΔERG rendered the same cells more resistant to enzalutamide. In line

with the previous findings in VCaP and LuCaP-147 cells, ΔERG expression rendered

LAPC4 cells susceptible to high levels of DHT, while mutant SPOP had the opposite effect

(Supplementary Fig. 13b). Taken together, the different responses to established and

experimental therapeutic modalities observed between mutant SPOP and ERG add further

credence to their divergent roles of the AR pathway related to tumorigenesis.

DISCUSSION

Although multiple studies over recent years have uncovered different genetically-

defined subtypes of primary prostate cancer, their biological understanding and therapeutic

implications remain largely unexplored territory. Here, we report two diametrically

different paths toward tumorigenesis triggered by either highly recurrent missense

mutation in *SPOP* or gene fusion involving the *ERG* oncogene. Importantly, wild type

SPOP emerges as a critical component that enforces oncogenic ERG signaling in part

through dampening AR activity, while mutant SPOP drives tumorigenesis through

activation of AR signaling. Moreover, several studies have previously highlighted the

importance of AR target genes in the context of SPOP mutants and ERG positive tumors

[17,30,31]. Based on the incompatibility between the two tumor subtypes, our work enabled

the development of specific custom signatures related to AR and ERG transcript that are

necessary to drive proliferation and tumorigenesis in the context of *ERG*-positive and

SPOP-mutants tumors. In addition, we show that the bromodomain histone reader

ZMYND11 is a SPOP substrate implicated downstream of SPOP in the opposing

regulation of the ERG and AR pathway in the two tumor subtypes (Fig. 10). The AR and

ERG pathways have been previously reported to have a partially antagonistic

relationship[31,32], further corroborating our findings.

Because activation of the androgen receptor by androgens represents a key lineage

specific oncogenic pathway in prostate cancer, androgen deprivation/antagonization

therapies (ADT) remain the uniform treatment modality up to this very day. That said, the

responses to ADT are highly variable and may last from a few weeks up to many years.

Here, we provide functional evidence that pre-existing prostate cancer founder mutations

influence the treatment response. Most notably, *SPOP* mutations promote susceptibility to

androgen deprivations therapies. In agreement with our findings, earlier reports have

shown underrepresentation of *SPOP* mutant tumors in cohorts of castration-resistant

disease and a more favorable response to the abiraterone and enzalutamide[33,34].

Conversely, we show that the presence of the ERG oncogene increases the

susceptibility of tumor cells to high-dose androgen therapy, while cells expressing mutant

SPOP remain largely unaffected. This is of clinical interest because testosterone treatment

of patients with advanced castration-resistant disease has recently shown to trigger anti-

tumor responses in around one third of the patients[35]. It is tempting to speculate that these

insights may help to discern responders from non-responders.

In addition, we provide evidence that the antagonistic relationship between mutant

SPOP and ERG may be used towards the development of new therapeutic avenues. More

specifically, we show that ERG-driven cancer cells are particularly sensitive to the

inhibition of wild-type SPOP using recently developed small molecule inhibitors[16]. Our

preclinical data suggests that SPOP inhibition may be effective in clinical settings where

ERG is robustly expressed (e.g. neo-adjuvant setting or early metastatic disease).

More generally, our results identify another paradigm for antagonistic driver genes in

prostate cancer that has recently emerged also for other cancer types [36-38]. In analogy to

prostate cancer, truncal point mutations in *DNMT3A* and gene fusions in *PML-RARA* are

mutually exclusive drivers in acute myeloid leukemia (AML). Similarly to SPOP, intact

DNMT3A has been found to be critical for PML-RARA-driven leukemia (Supplementary

Fig. 14a, b) [39,40]. Importantly, we demonstrate here for prostate cancer that the concept of

antagonistic driver genes can be exploited to identify therapeutic opportunities.

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Author Contributions

T.B. and J.P.T. originally developed the concept, further elaborated on it and designed

the experiments together with G.E.T. and M.B. ; T.B. and G.E.T. performed experiments

and analyzed the data together with N.D.U., S.A.C., A.M., T.S., L.P.B., F.S., M.Z., V.C.,

Anna R., H.J., D.B., M.C., D.A. and R.G.. T.B. and G.E.T. performed xenograft tumor

experiments in immunodeficient mice. Z.D. and C.G.Y. provided SPOP 6b and SPOP 6c

compounds. M.S. , S.D. and S.M. performed immunohistochemical experiments and J.P.T

the subsequent analysis. H.M. and P.S. provided prostate cancer samples. M.B., A.V. and

Andrea R. analyzed genomic and RNA-Seq data. J.P.T, H.M., S.A.C., M. A. R., M.K.J.,

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interpreted the data and wrote the paper. T.B., G.E.T. and M.B. contributed equally to this

work.

Author Information

MAR is listed as a co-inventor on US and International patents in the diagnostic and

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Methods

Cell culture, Transfection and Infection

VCaP, LNCaP, PC3, 22Rv1, HEK293 cells were purchased from ATCC. LAPC-4 were a

gift from Prof. Helmut Klocker. VCaP and HEK293 were grown in DMEM with Glutamax

(Gibco); LNCaP, PC3, 22Rv1, LAPC-4 in RPMI medium (Gibco); all were supplemented

with 10% full bovine serium (FBS; Invitrogen), or 10% charcoal-stripped serum (CSS;

One Shot Fetal Bovine Serum, Charcoal Strippped, Gibco) for androgen deprivation

therapy response, and 1% Penicilin/Streptomycin. LuCaP-147 were grown in StemPro

medium (hESC SFM StemPro, Gibco) with regular supplements. All cells were incubated

at 37°C and 5% CO_2 and routinely tested for mycoplasma contamination.

For stable knockdown experiments, cells were infected with pLKO-1 vectors (Sigma) and

the following clones were used; *SPOP*: TRCN0000140431 (shSPOP_1) and

TRCN000013911 (shSPOP_2); *TRIM24*: TRCN000021262 (shTRIM24_1) and

TRCN0000195528 (shTRIM24_2); *ERG*: TRCN0000429354 (shERG_1) and

TRCN0000432394 (shERG_2); *ZMYND11*: TRCN0000275479 (shZMYND11_1) and

TRCN0000275542 (shZMYND11_2). After infection, cells were selected in the presence

of puromycin (2 μg/ml).

For SPOP, ΔERG, HA-ZMYND11-WT, HA-ZMYND11-DMT1, HA-ZMYND11-DMT2,

MYC and AR over-expression a derivate of the pLX304 vector was used throughout in

which the CMV promoter has been exchanged to a PGK promoter and the blasticidin

cassette left unchanged (ΔERG constructs) or exchanged by a puromycin resistance

cassette (SPOP constructs) (pLX_TRC_307, available at Addgene as Plasmid 41392,

pCW107). All ORFs were cloned into pLX_TRC_307 using Nhe1 and Mlu1. Tumors from

PDX LuCaP-78, -147, -35,-23.1 were collected, dissociated and cultured as previosly

described[41].

Chemicals

MG-132 (M7449) and Cycloheximide (CHX, C4859) were purchased from Sigma and

used at 20 μM and 100 μg/ml in all experiments, respectively. SPOP inhibitor (SPOP-i,

compound 6b) and its inactive analog (compound 6c), were provided by the laboratory of

C. Yang (State Key Laboratory of Drug Research, Shanghai Institute of Materia Medica).

DHT (5α-Dihydrotestosterone) was purchased from Sigma (D-073), MDV3100

(Enzalutamide) was purchased from APExBIO (A3003). YK-4-279 (ETS inhibitor) was

purchased from Selleckchem. All chemicals were used at the indicated concentration.

Dose-response curves and cell-growth assays

Cells were seeded (between $1x10^3$ and $1x10^4$ per well) in a 96-well plate. Cells were

subsequently treated with serial dilutions of DHT (in 10% CSS medium), or enzalutamide;

SPOP inhibitor, ETS inhibitor to determine dose-response curves or were left untreated for

cell-growth assays. Proliferation at corresponding time points was assessed by MTT

(Methylthiazolyldiphenyl-tetrazolium bromide) assay according to the manufacturer's

recommendations (Sigma). For each time point, absorbance (OD, 590 nm) was measured

in a microplate reader.

SA-β-galactosidase staining of VCaP cells

Senescence-associated-β-galactosidase (SA--β-gal) staining was performed using the

Senescence β-Galactosidase Staining Kit (Cell Signaling, #9860) following the

manufacturer's instructions. To avoid false-positive staining, we adapted the β-

Galactosidase Staining Solution to pH 7. Cells incubated with glycerol 70% were observed

under a bright field microscope.

Matrigel Invasion assay

Invasion assay was performed as previously described[42]. Briefly, equal number of PC3

cells were seeded into 10cm dishes and starved with a medium without fetal bovine serum

for 24 hr.; subsequently 1×10^5 cells were resuspended in 100 µl of starved medium and

seeded onto the basement of a Boyden chamber (CLS3422; Sigma) coated with Matrigel.

RPMI with 10% fetal bovine serum was added to the lower chamber. After 48hr, invaded

cells were fixed with 10% of formalin and stained with crystal violet. Absorbance was

measured at 560 nm.

Clonogenic assay in methylcellulose

Cells were seeded (between 5×10^3 and 1×10^4) in methylcellulose (Methocult H4100,

StemCell Technologies) in triplicate. Cells were left untreated for cell-growth assay. For

SPOP inhibitor assay, cells were treated with vehicle (0.1% DMSO) or drug (SPOP-i) at

corresponding concentration. For androgen therapy, cells were treated with vehicle (0.01%

Methanol) or DHT at corresponding concentration. Cells were incubated at 37°C and 5%

CO_2 for 7-28 days and colonies were stained with MTT solution at 37°C overnight and

absorbance (OD, 590 nm) was measured in a microplate reader.

Mouse Prostate Organoid Generation and Experiments

Prostate tissue was extracted from euthanized mice, digested and seeded in matrigel as

previously described[43]. To overexpress *SPOP* species and *ΔERG* genes, mouse prostate

cells were virally infected by spinoculation for 1hr at 600g at 32 °C and selected with

puromycin. For the ''organoid formation assay'' 1.5×10^4 single cells were plated per well

onto 40µl of Matrigel on day 1 and organoids were grown in "revised human prostate

organoids medium" as previously described[41]. Briefly, the medium included the following

reagents: adDMEM/F12, glutaMAX (2mM), Pen-strep 100u/ml, HEPES (10mM), B27

(1X), EGF (5ng/ml), AB3-01 (500nM), Noggin (100ng/ml), R-Spondin 1 (500ng/ml),

DHT (1nM), FGF-2 (5ng/ml), FGF-10 (10ng/ml), Prostaglandin E2 (1000nM).   The

number of formed organoids that reached 100 µM of diameter was counted on days 14 post

plating with cellSens software (Olympus). For the Dose-Response experiment 1×10^4

mouse prostate cells were plated in 40µl of Matrigel and treated with vehicle (0.1% DMSO)

or drug (SPOP-i) at indicated concentration for 7 days. Live/dead staining was performed

using Calcein AM (final concentration 1uM) and Ethidium Homodimer I (EthD-1, final

concentration (1.33uM) dissolved in the medium for 1 hour. GFP or RFP positive

organoids were analyzed under a fluorescence microscope. The percentage of dead

organoids (RFP positive) was normalized to the Etoposide-treated (positive control)

organoids.   The genetically engineered Mouse Prostate Organoids, derived from

PbCre;R26^{F133V} , were generated as previously described[11].

Immunohistochemistry

Cytoblocks were prepared from the pellets of organoids by adding plasma and thrombin in

order to obtain a solid matrix. Once solidified, the organoids were fixed in 10% formalin

(Thermo Scientific, 5701) and embedded in paraffin as a normal tissue. Sections of 4 µm

were used for IHC analyses and hematoxylin and eosin (H&E) staining (Diapath, C0303)

and (Diapath, C0363) respectively. Once dried the sections were treated with OTTIX plus

solution (Diapath, X0076) and OTTIX shaper solution (Diapath, X0096) to dewax and

rehydrate the sections. Antigen retrieval was performed using pH 6 solutions at 98°C for

20 min. Successively the endogenous peroxidases and non-specific binding sites were

blocked using 3% H2O2 (VWR chemicals, 23615.248) and Protein-Block solution (DAKO

Agilent technologies, X0909) respectively, for 10 min. Sections were then stained for anti-

p16 (ab211542, Abcam, 1:1200), anti-Ki67 (Clone SP6; Lab Vision Corporation #RT-

9106-R7, RTU) anti-Phospho-HP1y (Ser83) Antibody (CST #2600, 1:200), anti-CK8

(ab,59400, Abcam), anti-CK5 (ab52635, Abcam). IHC analyses were performed using the

Imagescope software.

***In vivo* experiments**

All animal experiments were carried out in male athymic nude mice (Balb/c nu/nu, 6-

8 weeks old), NSG mice (NOD Scid Gamma, 6-8 weeks old), and NRG (NOD Rag gamma,

6-8 weeks old) accordingly to protocol approved by the Swiss Veterinary Authority (No.

TI-14-2014, TI-38-2018, TI-39-2018 and TI-42-2018). Patient-derived xenografts (PDX)

LuCaP-147, -78, -35, -23 were provided by Eva Corey (University of Washington) and

maintained as previously described[44]. 2×10^6 VCaP cells, 5×10^6 LuCaP-147, LuCaP-23.1,

LuCaP-35 and LuCaP-78 were resuspended in 100 µl of PBS and Matrigel 1/1 and

subcutaneously injected into both of the dorsal flanks of the mice. Tumor growth was

recorded using digital caliper and tumor volumes were calculated using the formula (*L x W*

^{2}) /2, where *L*=length and *W*=width of tumor. For the testosterone propionate (25mg/kg)

and SPOP inhibitor (SPOP-i, 50mg/kg) treatment, the mice were grouped randomly and

the treatment started when the mean tumor volume reached $100m^3$. Tumor volume and

weight were measured 2 times per week. Testosterone propionate was resuspended first in

ethanol (150mg/kg) and then in Corn oil (Sigma) at a final concentration of 25mg/kg. SPOP

inhibitor was resuspended in Dulbecco's phosphate buffered saline (PBS) at a final

concentration of 50 mg/kg. At the end of the experiment, mice were euthanized, tumors

extracted and weighted. Testosterone level was measured using the Human Testosterone

ELISA Kit from Abcam (ab174569). In order to recapitulate the levels of

supraphysiological testosterone administrated in clinical trials[35], mice reaching at least 3

times the testosterone levels measured before the treatment initiated were included in the

depicted data.

Antibodies, Immunoblotting, and Immunoprecipiation

Antibodies used in immunoblotting and immunoprecipitation assays were: anti-SPOP

(ab81163, Abcam), anti-TRIM24 (Sc-271266, Santa Cruz), anti-ß-ACTIN (4967, Cell

Signaling), anti-AR (Sc-7305, Santa Cruz), anti-GADPH (Sc-47724, Santa Cruz), anti-

ERG (Sc-271048, Santa Cruz), anti-α-Tubulin (3873S, Cell Signaling), anti-ZMYND11

(NBP2-20960, Novus Biologicals), anti-HA (H3663, Sigma), anti-BRD2 (A302-583A,

Bethyl Labs), anti-NCOA3 (2126, Cell Signaling), anti-DEK (610948, BDBioscience),

anti-p21 (2947S, Cell Signaling) , anti-c-MYC (5605S, Cell Signaling), anti-HOXB13 (Sc-

28333, Santa Cruz), anti-PTEN (9559, cell signaling), anti-p21 (ab188224, Abcam), anti-

HOXB13 (NBP2-43655, Novus biologicals), anti-GDF-15 (27455-1-AP, proteintech). All

antibodies were employed at dilutions suggested by the manufacturers.

For immunoblotting, cells were washed with PBS and subsequently lysed in RIPA

buffer (Sigma) and sonicated. Protein concentration was determined using the BCA reagent

(ThermoFisher), same amounts of protein were separated by SDS-PAGE (Biorad) and

transferred onto PVDF membrane (ThermoScientific). The membrane was incubated for

one hour in 5% nonfat dry milk/TBS-T blocking buffer followed by incubation with the

primary antibody overnight at 4°C. The membrane was washed with TBS-T followed by

incubation with horseradish peroxidase-conjugated secondary antibody (Promega).

To detect interactions of SPOP and ZMYND11, cells were lysed in 1 % NP40 buffer

(50mM Tris-HCl pH 7.4, 150 mM NaCl, 1 % NP40) with 2x protease inhibitor cocktail

(Complete, Roche), sonicated, and 3 mg of lysate were incubated overnight with 2 µg of

anti-HA-tag or control mouse IgG antibody (sc-2025, Santa Cruz Biotechnology) at 4 °C.

Subsequently, antibodies were collected by 25 µl protein A/G magnetic beads (88803,

Fisher Scientific) for 2h, followed by 2 washing steps with 1 % NP40 buffer. Proteins were

eluted by addition of 1x SDS-sample buffer under reducing conditions at 95 °C for 5 min.

In Vivo **Ubiquitylation Assay**

293T cells were transiently transfected with indicated plasmids: pCW107-HA-ZMYND11-

WT or HA-ZYMND11-DMT1/DMT2 (2 µg), pCW107-SPOP-WT or SPOP-MT (2 µg),

CMV-8x Ubi-His (2 µg). 42 hours later, cells were treated with MG-132 (20µM) or DMSO

for additional 7 hours. Cells were then washed with PBS and collected by centrifugation.

Small amount of cells was lysed in RIPA buffer and the rest in Buffer C (6M guanidine –

HCL, 0.1 M Na2HPO4/NaH2PO4, 10mM Imidazole, pH=8). The whole cells extract was

sonicated and incubated with 60 µl of Ni-NTA agarose (Sigma) overnight at 4°C. Next,

Ni-NTA beads were washed once with Buffer C, twice with Buffer D (1 volume of Buffer

C: 3 volumes of Buffer E) and once with Buffer E (25 mMTris-HCL, 20 mM Imidazole,

pH=6.8). Elution of bound proteins was processed by boiling in 1x SDS loading buffer

containing 300 mM Imidazole. Samples were loaded, separated by SDS-PAGE, and

detected by immunoblotting.

Gene Expression Studies

RNA was extracted using the RNeasy kit (Qiagen) and processed by Kapa SybrFAST

one-Step qRT-PCR kit according to manufacturer`s instructions. qRT-PCR was undertaken

on an Applied Biosystems StepOnePlus System. The target mRNA expression was

quantified using ∆∆Ct method and normalized to Actin expression. The following primers

were used: *SPOP*, forward 5'- GAAATGGTGTTTGCGAGTAAACC-3', reverse 5`-

TACCTACGCTTCCAGTCTCTG-3'; *ERG*, forward 5'-

TGTATGCCAGCATTTGTTTCTT-3', reverse 5'- TTGCTGGTCTTGCCATTCCT-3';

β-ACTIN, forward 5'-AAGGAGCCCCACGAAAAAT-3', reverse 5'-

ACCGAACTTGCATTGATTCCAG-3'; *PLAU,* forward 5'-

TACGGCTCTGAAGTCACCACCAAAAT-3', reverse 5'-

CCCCAGCTCACAATTCCAGTCAA-3'; *PLAT,* forward 5'-

CACTGGGCCTGGGCAAACATA-3', reverse 5'-CACGTCAGCCTGCGGTTCTTC-3';

TMPRSS2, forward 5'-CAGGAGTGTACGGGAATGTGATGGT-3', reverse 5'-

GATTAGCCGTCTGCCCTCATTTGT-3'; *KLK2*, forward 5'- CTGCCCATTGCCTAA

AGAAG-3', reverse 5'- GTAGAGCGGGTGTGGGAAG-3'; *PSA* forward 5'-

GAGCACCCCTATCAACCCCCTATT -3' , reverse 5'-

AGCAACCCTGGACCTCACACCTAA-3'; *ZMYND11,* forward 5'-

ATGGCACGTTTAACAAAAAGACG-3', reverse 5'-

CGGTCAATGTTGGCAATCTGC-3'; *BRD2*, forward 5'-

CTACGTAAGAAACCCCGGAAG-3＇, reverse 5'-GCTTTTTCTCCAAAGCCAGTT-

3; *TRIM24*, forward 5`-CAGCCACAAATGCCTAAGCAG-3＇, reverse 5'-

GTGTTGGGAACTTGGATAACTGG-3;   *NCOA3*,   forward   5'-

AGACGGGAGCAGGAAAGTAAA-3',   reverse   5'-

GTAAAAGCGGTCCTAAGGAGTC-3';   *DEK*,   forward   5'-

AACTGCTTTACAACAGGCCAG-3＇, reverse 5'-ATGGTTTGCCAGAAGGCTTTG-

3'.

RNA-Seq of VCaP, LNCaP and LuCaP cells

RNA sequencing for all experiments involving LuCaP xenografts, VCaP and LNCaP

cells was performed at the Institute of Oncology Research using Next Ultra II Directional

RNA Library Prep Kit for Illumina and sequenced on the Illumina NextSeq500 with single-

end, 75 base pair long reads. The overall quality of sequencing reads was evaluated using

a variety of tools, namely FastQC (Andrews S., 2010), RSeQC[45], AfterQC[46] and

Qualimap[47]. Sequence alignments to the reference human genome (GRCh38) was

performed using STAR [48] (v.2.5.2a). Gene-expression was quantified at gene level by using

the comprehensive annotations made available by Gencode[49]. Specifically, we used *v27*

release of the Gene Transfer File (GTF). Raw-counts were further processed in the *R*

Statistical environment and downstream differential expression analysis was performed

using the DESeq2[50] pipeline.

Genes being expressed at very low levels were automatically filtered out through the

Independent Filtering feature embedded in DESeq2 (alpha = 0.05). Differential-expression

results were ranked according to the computed Wald-statistics values. Subsequently, gene-

set enrichment testing was performed using Camera[51] pre-ranked (inter-gene correlation

equal to 0.1, parametric test procedure). Statistical enrichments were assessed for gene-sets

belonging to the Hallmark collection, which is curated by the Molecular Signature

DataBase[52,53] (*MSigDB*), and for custom ERG and DHT-specific gene-signatures. All

enrichments were corrected for multiple testing using Benjamini and Hochberg FDR

adjusted p-value.

Identification of ERG and AR related gene signatures

We retrieved RNA-seq data from GEO Dataset *GSE83652*[14] to identify transcriptional

perturbations in VCaP cells following treatment with DHT or following silencing of ERG.

To this purpose we completely reprocessed samples SRR3713255-57, SRR3713267-72

using *STAR* and *DESeq2* as previously described for VCaP cells. In addition, to identify

direct targets, we integrated information relative to AR and ERG chromatin binding sites,

which we derived from GEO Dataset *GSE28950*[25]. To maximize the number of peaks and

to reduce false negatives, we merged results of experiments performed at different time

points, namely 2h and 18h after DHT exposure. De-multiplexed reads were aligned to *hg38*

release of the human reference genome using bwa-mem[54] (0.7.15). MACS[55] (v.2.1.0) was

used to perform peak calling procedure using a cutoff FDR q-value of 0.01 and a mappable

genome size optimized for *hg38* equal to 2.9 gigabases. Downstream analysis was

performed in R statistical environment. We identified binding sites overlapping promoters

by using bedtools[56].

Promoters were defined as DNA regions ranging from 1500 *bp* upstream to 500 *bp*

downstream of Transcription Start Sites (TSSs).

To discriminate between ERG- and AR- specific transcriptional responses we stratified

genes into three main classes: genes whose promoter regions are bound by *AR* but not by

ERG, genes whose promoters are bound by ERG but not by *AR*, and finally, genes whose

promoters are co-bound by both AR and ERG. AR bound only genes were further

subdivided into two sets, those being significantly (FDR<0.05) induced following DHT

treatment and those being significantly repressed. A similar approach was applied to ERG

bound only genes, where genes were subdivided into ERG-induced and ERG-repressed

gene-sets, if they were respectively down or up-regulated following *ERG* silencing. To be

more stringent in the definition of AR-specific and ERG-specific signatures, we excluded

genes from the ERG-induced set that were also significantly up-regulated following DHT

treatment, *vice-versa* we excluded ERG-repressed genes that were significantly down-

regulated following DHT-treatment. The same criteria were applied for DHT-specific

gene-sets. Finally, defined an additional gene-set (DHT-induced/ERG-repressed)

consisting of genes being co-bound by AR and ERG in their promoter region, which were

significantly up-regulated following DHT treatment but also significantly upregulated

following ERG-silencing. All gene-sets are detailed in Supplementary Table 1. Overlap

between custom derived gene-signatures and the most represented Hallmark's gene-sets

was assessed using GeneOverlap R package (Shen L, Sinai M, 2013). Two-dimensional

network visualization was generated with Cytoscape. [57]

Gene-set testing and RNA-Seq data processing of clinical samples

Publicly available RNA-Seq data for primary prostate cancer were obtained from The

Caner Genome Atlas[3] (TCGA) database and retrieved from Genomics Data Commons

(GDC) in form of gene-centric raw counts, using *TCGAbiolinks* package[58]. We selected

individuals characterized by either SPOP or ERG fusion, and a third group defined as

"others", which includes all remaining samples, excluding those patients exhibiting any

other ETS-rearrangement. Differential expression and gene-set enrichment between

samples harboring ERG fusions and SPOP-mutations were performed using *DESeq2* and

Camera (pre-ranked) as previously described for prostate cancer cells. Single-sample gene-

set enrichment analysis (GSVA[59] package) was applied to measure, for each individual

patient, the overall activity of the custom gene-sets that were previously generated in VCaP

cells. Following differential expression analysis between ERG-rearranged and SPOP

mutant primary tumors, we defined two gene-sets consisting of SPOP-upregulated (n =

443, log_2FC >1, FDR<0.05) and ERG-upregulated (n = 359, log_2FC >1, FDR<0.05) genes.

PolyA+ RNASeq data for metastatic prostate cancer were obtained from SU2C

cohort[60]. Normalized *RPKM* values, retrieved through *cBioportal*, were log transformed

and patient's categorization (SPOP/ERG/OTHER) was performed in the same manner as

for primary tumors. To evaluate whether transcriptional differences occurring between

ERG-rearranged and SPOP-mutant individuals were also conserved in CRPC setting, we

quantified the above mentioned SPOP-upregulated/ERG-upregulated signatures in the

SU2C 2019 cohort, using single-sample gene-set enrichment analysis. The obtained

ssGSEA scores were scaled in a range between -1 and 1 (SPOP-Upregulated) and between

1 and -1 (ERG-upregulated, inverted). Subsequently we averaged these rescaled values in

order to obtain an aggregate score.

Circular representation of interactions between gene-sets

Chord diagrams were generated using circlize[61] package in R statistical environment.

Strings, whose thickness is proportional to the number of shared elements, represent

common genes between sets.

ZMYND11 ChIP-seq in VCaP cells

ChIP-seq using anti-ZMYND11 antibody (NBP2-20960, Novus Biologicals) was

performed in VCaP cells, overexpressing either wild-type SPOP or mutant SPOP harboring

Y87C point mutation. Briefly, to isolate chromatin, cells (120.000.000 per IP) were cross-

linked using 1% Formaldehyde cross-link protein-DNA complexes and crosslinking was

terminated by the addition of 1/10 volume 1.25 M glycine for 5 min at room temperature

followed by cell lysis and sonication, resulting in an average chromatin fragment size of

200 bp. Samples lysis was performed as previously described using MNase enzyme 1000

gel units=1 uL[62]. After adding the MNase sonication buffer, the samples were sonicated

for 30 cycles, 30 sec ON and 30 sec OFF at high voltage. ChIP and input DNA (50 ng)

were used for indexed library preparation using NEBNext Ultra II DNA Library Prep kit

and subjected to 75 *bp* single-end sequencing on the Illumina NextSeq500. All procedures

were performed at the Institute of Oncology Research. De-multiplexed reads were aligned

to *hg38* release of the human reference genome using bwa-mem[54] (0.7.15). MACS[55]

(v.2.1.0) was used to perform peak calling procedure using a cutoff FDR q-value of 0.01

and a mappable genome size optimized for *hg38* equal to 2.9 gigabases. Downstream

analysis was performed in R statistical environment. ChIPseeker[63] was used to annotate

peaks and to represent the distribution of ZMYND11 binding sites relative to Transcription

Start Sites (TSSs). The R package chipenrich[64] was subsequently used to determine

enrichment or depletion of ZMYND11 peaks in regions surrounding *TSSs* of genes that are

included in Hallmarks or custom gene-set collections. Surrounding regions were defined

as ranging from 5kb upstream to 5kb downstream of their *TSSs* (locusdef = 5kb), which is

in line with the overall behavior of ZMYND11 binding sites around *TSSs* (Supplementary

Fig. 6f-g).

Identification of AR-binding sites in primary prostate cancer specimen

Publicly available ChIP-Seq data were retrieved from GSE120738[3]. ChIP-seq data were

reprocessed as described for ZMYND11 samples. Differential binding affinity of AR

between ERG-rearranged and SPOP-mutant tumors was performed using DiffBind (Stark

R and Brown G, 2011).

Frequency of SPOP mutations across patients' cohorts

We defined the percentage of SPOP-mutant and TMPRSS2-ERG positive tumors

across different patients' cohorts originating from multiple sources. Patients with

primary/loco-regional prostate tumors were derived from TCGA and MSK-IMPACT

Clinical Sequencing cohorts[34]. Patients with tumor-progression (non-castrate) were

derived from MSK-IMPACT and TCGA cohorts, by including from the latter only

individuals that showed tumor-progression based on survival information. Castration

resistant prostate cancer patients were retrieved from MSK-IMPACT, Beltran et. al[65] and

from the SU2C[60] .datasets. Neuroendocrine prostate cancer samples were retrieved from

the SU2C cohort (samples annotated with neuroendocrine features) and from Beltran et

al[65]. Total number of *SPOP*-mutant and *TMPRSS2-ERG* tumors were determined based

on the clinical annotations of the individual studies and integrated with fusion information

from TCGA Fusion Gene Database (www.tumorfusions.org). Survival analysis was

performed in R statistical environment using the TCGA and MSK-IMPACT clinical

sequencing cohort.

Quantitative liquid-chromatography-mass spectrometry (LC-MS/MS)

In solution digestion VCaP cell pellets were lysed at 4 ºC in 8 M urea, 50 mM Tris-

HCl pH 8.0, 150 mM NaCl, 1 mM EDTA, 2 µg/µl aprotinin (Sigma-Aldrich), 10 µg/µl

leupeptin (Roche), and 1 mM phenylmethylsulfonyl fluoride (PMSF) (Sigma). Protein

concentration was determined using a bicinchoninic acid (BCA) protein assay (Pierce).

Proteins were reduced with 5 mM (DTT) for 45 min at room temperature (RT), followed

by alkylation with 10 mM iodoacetamide for 30 min at room temperature in the dark. The

urea concentration was reduced to 2 M using 50 mM Tris-HCl, pH 8. Samples were

digested for 2 h at 25 ºC with endoproteinase Lys-C (Wako Laboratories) at an enzyme-to-

substrate ratio of 1:50. Samples were subsequently digested overnight at 25 ºC with

sequencing grade trypsin (Promega) at an enzyme-to-substrate ratio of 1:50. Following

overnight digestion, samples were acidified to a final concentration of 1% formic acid.

Peptide samples were desalted on a 100 mg tC18 Sep-Pak SPE cartridge (Waters).

Cartridges were conditioned with 1 mL of 100% MeCN, 1 mL of 50% MeCN/0.1% FA,

and 4x with 1 mL of 0.1% TFA. The sample was loaded, and washed 3x with 1 mL of 0.1%

TFA, 1x with 1 mL of 1% FA, and eluted 2x with 600 µl of 50% MeCN/0.1% FA.

TMT labeling of peptides

736 Peptides were labeled with TMT 10-plex isobaric mass tagging reagents (Thermo

737 Fisher Scientific). Each TMT reagent was resuspended in 41 µL of MeCN. Peptides were

738 resuspended in 100 µL of 50 mM HEPES and combined with TMT reagent. Samples were

739 incubated at RT for 1 h while shaking. The TMT reaction was quenched with 8 µL of 5%

740 hydroxylamine at RT for 15 min with shaking. TMT labeled samples were combined, dried

741 to completion, reconstituted in 100 µL of 0.1% FA, and desalted on StageTips or 100 mg

742 SepPak columns as described above.

744 **Basic Reverse Phase (bRP) Fractionation**

745 The TMT labeled samples were fractionated using offline high pH reversed-phase

746 chromatography (bRP) as previously described [Mertins et al Nat Prot]. Samples were

747 fractionated using Zorbax 300 Extend C18 column (4.6 × 250 mm, 300 Å, 5 µm, Agilent)

748 on an Agilent 1100 series high-pressure liquid chromatography (HPLC) system. Samples

749 were reconstituted in 900 µL of 4.5 mM ammonium formate (pH 10) in 2% (vol/vol)

750 acetonitrile (MeCN) (bRP solvent A). Samples were injected with Solvent A at a flow rate

751 of 1 mL/min and separated using a 96 min gradient. The gradient consisted of an initial

752 increase to 16% solvent B (90% MeCN, 5 mM ammonium formate, pH 10), followed by

753 60 min linear gradient from 16% solvent B to 40% B and successive ramps to 44% and

754 60% at a flow rate of 1 mL/min. Fractions were collected in a 96-deep well plate (GE

755 Healthcare) and pooled in a non-contiguous manner into final 24 proteome fractions.

756 Pooled fractions were dried to completeness using a SpeedVac concentrator.

758 **Liquid chromatography and mass spectrometry**

Desalted peptides were resuspended in 3% MeCN/0.1% FA and analyzed by online

nanoflow liquid chromatography tandem mass spectrometry (LC-MS/MS) using Q-

Exactive plus mass spectrometer (Thermo Fisher Scientific) coupled on-line to a Proxeon

Easy-nLC 1200 as previously described [Mertins et al Nature Protocols]. Briefly, 1 ug of

each sample was loaded onto a microcapillary column (360 μ m outer diameter $\times$ 75 μ

m inner diameter) containing an integrated electrospray emitter tip (10 μ m), packed to

approximately 22 cm with ReproSil-Pur C18-AQ 1.9 μ m beads (Dr. Maisch GmbH) and

heated to 50 $^{\circ}$C. Samples were analyzed with 110 min LC-MS method. The 110 min

method contained a mobile phase with a flow rate of 200 nL/min, comprised of 3%

acetonitrile/0.1% formic acid (Solvent A) and 90% acetonitrile/0.1% formic acid (Solvent

B), with the following gradient profile: (min:%B) 0:2; 1:6; 85:30; 94:60; 95:90; 100:90;

101:50; 110:50 (the last two steps at 500 nL/min flow rate). The Q-Exactive plus MS was

operated in the data-dependent mode acquiring HCD MS/MS scans (r =35,000) after each

MS1 scan (r = 70,000) on the 12 most abundant precursor ions using an MS1 target of 3

$\times$ 10^6 and an MS2 target of 5 $\times$ 10^4 . The maximum ion time utilized for MS/MS

scans was 120 ms; the HCD-normalized collision energy was set to 30; the dynamic

exclusion time was set to 20 s, isotope exclusion function was enabled, and peptide match

function was set to preferred. Charge exclusion was enabled for charge states that were

unassigned, 1 and >6.

MS Data Analysis

All data were analyzed using Spectrum Mill software package v 6.1 pre-release

(Agilent Technologies). Similar MS/MS spectra acquired on the same precursor m/z within

+/- 60 s were merged. MS/MS spectra were excluded from searching if they were not

within the precursor MH+ range of 750-4000 Da or if they failed the quality filter by not

having a sequence tag length >0. MS/MS spectra were searched against UniProt human

database. All spectra were allowed +/- 20 ppm mass tolerance for precursor and product

ions, 30% minimum matched peak intensity, and "trypsin allow P" enzyme specificity with

up to 4 missed cleavages. The fixed modifications were carbamidomethylation at cysteine,

and TMT at N-termini and internal lysine residues. Variable modifications included

oxidized methionine and N-terminal protein acetylation. Individual spectra were

automatically designated as confidently assigned using the Spectrum Mill autovalidation

module. Specifically, a target-decoy based false-discovery rate (FDR) scoring threshold

criteria via a two-step auto threshold strategy at the spectral and protein levels was used.

First, peptide mode was set to allow automatic variable range precursor mass filtering with

score thresholds optimized to yield a spectral level FDR of 1 %. A protein polishing

autovalidation was applied to further filter the peptide spectrum matches using a target

protein-level FDR threshold of 0. Following autovalidation, a protein-protein comparison

table was generated, which contained experimental ratios. For all experiments, non-human

contaminants and reversed hits were removed. Furthermore, data were filtered to only

consider proteins with 2 or more unique peptides and was median normalized.

Statistical analysis

GraphPad Prism version 8.3 (GraphPad Software) was used for statistical analysis.

Data are depicted as mean + s.e.m. unless otherwise specified. The number of independent

experiments or mice used is indicated in each figure legends. Unpaired Student's t-test was

used for comparisons between two groups, one-way analysis of variance (ANOVA) with

multiple comparisons for two groups or more, and two-way ANOVA with multiple

comparisons for repeated measurements. Multiple comparisons tests were corrected by

controlling the False Discovery Rate (FDR) using Benjamini and Hochberg's method. .

Correlation analyses were performed using Pearson correlation coefficients.

Data Availability

The original mass spectra have been deposited in the public proteomics

repository MassIVE and are accessible at ftp://MSV000082915@massive.ucsd.edu when

providing the dataset password: prostate. If requested, also provide the username:

MSV000082915. This data will be made public upon acceptance of the manuscript.

RNA-Seq data generated have been deposited in the *ArrayExpress* database at EMBL-

EBI and were assigned the accessions *E-MTAB-7165, E-MTAB-7170, E-MTAB-7173*.

These data are accessible when providing the datasets username and password as follow:

Username: *Reviewer_E-MTAB-7165* Password: *4sJqpicv*

Username: *Reviewer_E-MTAB-7170* Password: *AAA65MMy*

Username: *Reviewer_E-MTAB-7173* Password: *sIpcob0o.*

CHiP-Seq data generated have been deposited in the *ArrayExpress* database at EMBL-

EBI and were assigned the accessions *E-MTAB-7174*. These data are accessible when

providing the dataset username and password as follow:

Username: *Reviewer_E-MTAB-7174* Password: *onwwwwni*

All these data will be made public upon acceptance of the manuscript.

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Fig 1. Genetic alterations in *SPOP* and *ERG* are synthetic sick. a Distribution of genetic alterations in *SPOP* and *ERG* across 333 primary prostate cancers in TCGA database[1]. **b** 3D growth of mouse prostate epithelial organoids derived from C57BL/6 mice over-expressing the indicated SPOP and ERG species (bar represents 20 µm) (n=3, technical replicates). Representative bright field pictures and haematoxylin and eosin (H&E) stained sections are shown. **c** *In vivo* growth of VCaP xenografts over-expressing the indicated SPOP species in immune-compromised mice (each group, n=10). **d** Immunoblot of VCaP cells over-expressing the indicated SPOP species and corresponding quantification of the indicated protein levels depicted as heatmap. Protein expression changes were normalized to β-ACTIN and Control cell line. **e** Tumor growth kinetics of xenografts established from LuCaP-147 PDX (SPOP-Y83C) stably overexpressing ΔERG or Control vector (each group, n=10). Corresponding immunoblot and quantification depicted as heatmap. Protein expression changes were normalized to Vinculin (VCL) and Control cell line. **f** Dose-response curve of VCaP cells overexpressing the indicated SPOP species and treated with the ETS-inhibitor YK-4-279. All error bars, mean + s.e.m. P values were determined by one-way ANOVA (b) or two-way ANOVA (c,e,f) with multiple comparisons and adjusted using Benjamini-Hochberg post-test *$P < 0.05$, **$P < 0.01$, ***$P < 0.001$. Molecular weights are indicated in kilodaltons (kDa).

Fig. 2 Transcriptome analysis of primary prostate cancer patients reveals major differences between *SPOP* mutant and *ERG*-fused tumor subtypes. a PCA-analysis based on RNA-Seq derived mRNA expression levels (TCGA cohort). *ERG*-fused (violet) and *SPOP*-mutant (green). Individuals were annotated into subtypes as described in Material and Methods. **b** Boxplots representing the transcriptional activity of SPOP integrated-signature (see Materials and Methods) applied to CRPC samples (SU2C-2019 cohort, left) and PDX-models. Scores are determined genes-signatures derived from primary prostate tumors (TCGA-cohort). ERG-fused samples are depicted in violet, SPOP-mutant samples are depicted in green. Samples not harboring SPOP mutations or ERG rearrangements are represented in grey. *P*-values were determined using Wilcoxon-test. **c** Boxplots depicting the amount of normalized reads per binding site across all the differentially bound (DB) regions (FDR = 0.05) resulting from the comparison between *ERG*-fused (violet) and *SPOP*-mutant (green) samples.

Fig. 3 Mutant SPOP-induced androgen receptor signaling antagonizes ERG activity.
a Gene-set enrichment analysis of VCaP cells overexpressing SPOP mutant (SPOP-MTs; SPOP-Y87C, -F102C, -W131G) versus SPOP-wild type (-WT), based on RNA-seq data. Experiments were performed using three replicates for each condition. Enrichments are determined on custom gene-sets of direct androgen receptor (AR) and ERG target genes (Supplemental Table 1). Enrichments and FDR-adjusted p-values are computed with *Camera* (pre-ranked) **b** Venn Diagram and heatmap depicting the expression of genes included in the custom gene-set of AR/ERG co-bound genes that are repressed by ERG and induced by DHT in VCaP cells overexpressing SPOP-MTs (SPOP-Y87C, F102C, W131G), SPOP-WT and vector Control. Genes (rows) and samples (columns) were clustered using Euclidean distance. Gene expression values were normalized using variance stabilizing transformation (vst) and subsequently scaled and centered by row prior of clustering. Columns represent average expression of three replicates for each condition. **c,** Two-dimensional network representing overlaps between the 10 most significantly enriched Hallmark and custom gene-sets, identified when comparing SPOP-MTs (SPOP-Y87C, F102C, W131G) to SPOP-wild type (-WT) overexpressing VCaP cells. Thickness of edges is proportional to the significance of the overlap of the connected nodes measured by Fisher test. Only edges with FDR value <0.05 are shown. Size of nodes is proportional to gene-set enrichment significance and equals to -10 x $\log_{10}$ (FDR). **d,** Heatmap representing gene-set activity stratified according to tumor subtype, derived from TCGA cohort. For each tumor group, the average value of single sample GSEA scores was considered. Values were scaled and referenced to samples that did not harbor any ETS-fusion (*ERG, ETV1,* and *ETV4)* or point mutations in *SPOP*[1].

Fig. 4 ZMYND11 is a *de novo* SPOP substrate. a Schematic illustration showing the design of the proteomics experiments. Tandem Mass Tag (TMT)-based quantitative mass-spectrometry (n=2, biological replicates) was used in VCaP cells overexpressing Control vector (Control), SPOP-WT, or three different SPOP mutants (SPOP-Y87C, SPOP-F102C and SPOP-W131G). **b** Scatter plot comparing transcriptomic and proteomic derived fold changes resulting from the comparison between SPOP-MTs (average across SPOP-Y87C, -F102C, -W131G) and SPOP-WT VCaP cells. (n=3 biological replicates for transcriptome, n=2 biological replicates for proteome). Genes belonging the DHT-Induced/ERG-Repressed gene-signature (AR+ERG co-bound) are highlighted in red. TRIM24 and ZMYND11 are the most upregulated proteins without changes at mRNA levels and are highlighted in green. CDKN1A (p21) upregulated at both mRNA and protein levels, and MYC downregulated at both mRNA and protein levels, are highlighted in black. **c** Over-expression of HA-ZMYND11 and SPOP-WT in 293T cells and subsequent expression analysis of the indicated proteins by immunoblotting. **d** Whole cell extracts (WCE) of 293T cells over-expressing HA-ZMYND11-WT and different SPOP species and corresponding anti-HA-immunoprecipitation (HA-IP). Expression of the indicated proteins was analyzed by immunoblotting. **e** Domain structure of ZMYND11 with indicated SPOP-degron and ubiquitin sites. **f** Forced expression of SPOP-WT together with HA-ZMYND11-WT or two degron deficient mutants (DMT1 & DMT2) in 293T cells. **g** *In vivo* ubiquitylation assay of HA-ZMYND11 in 293T cells. Cell were transiently transfected with the indicated constructs and histidine-tagged (his-tag), ubiquitylated proteins were pulled down using nickel beads. Ubiquitylated HA-tagged ZMYND11 was detected by immunoblotting. **h** Over-expression of HA-ZMYND11 and SPOP-Y87C in 293T cells and subsequent expression analysis of the indicated proteins by immunoblotting after proteasomal inhibition with MG132. **i** Whole cell extracts (WCE) and corresponding anti-HA-immunoprecipitation (HA-IP) of 293T cells over-expressing HA-ZMYND11-WT and different SPOP-MTs species as indicated. Expression of the indicated proteins was analyzed by immunoblotting. **j** *In vivo* ubiquitylation assay of HA-ZMYND11 in 293T cells. Cell were transiently transfected with the indicated constructs and histidine-tagged (his-tag), ubiquitylated proteins were pulled down using nickel beads. Ubiquitylated HA-tagged ZMYND11 was detected by immunoblotting. **k** Immunoblots of indicated proteins in VCaP, LNCaP and LAPC4 human prostate cancer cells overexpressing the indicated SPOP species. Molecular weights are indicated in kilodaltons (kDa).

Fig. 5 ZMYND11 induces AR signaling pathway and represses ERG activity. a,b Immunoblot of indicated proteins (**a**) and corresponding 2D proliferation assay (b) of VCaP cancer cells overexpressing HA-ZMNYD11-WT and derived degron-deficient mutants (DMT1/2) (n=3). Correlation between cell viability and ZMYND11 protein expression changes (Prot. Exp. Changes), as quantified by immunoblot in the same cell lines. *P* values were calculated using Pearson rank correlation. **c** Fold-change cell viability of VCaP cancer cells over-expressing the indicated SPOP species with and without

ZMYND11 knockdown using two different short hairpin RNAs, at day 16 (n=3). Protein expression of the indicated proteins was analyzed by immunoblotting. **d** Chord-diagram of transcriptionally regulated genes by either SPOP-MTs or HA-ZMYND11-DMT2 in VCaP cells (FDR<0.05). Strings, whose thickness is proportional to the number of shared elements, represent common genes between sets. **e** Gene-set enrichment analysis of VCaP cell overexpressing HA-ZMYND11-DMT2 compared to Control, based on RNA-seq data. Enrichments are performed on custom gene-sets of direct androgen receptor (AR) and ERG target genes. FDR-adjusted p-values are computed with *Camera* (pre-ranked). **f** Heatmap of ChIP-seq signals around TSS regions (+/- 4kb) at which ZMYND11 bindings were identified by peak calling procedure (Macs2) in VCaP cells overexpressing the indication constructs. **g** IGV-derived screenshots representing loglikelihood ratio of ZMYND11 bindings in mutant SPOP (SPOP-Y87C) vs wild-type SPOP over-expressing VCaP cells. Reported are *MYC* (up) and *CDKN1A* (bottom). All error bars, mean ± s.e.m. *P* values were determined by two-way ANOVA (**b**) or one-way ANOVA with multiple comparisons and adjusted using Benjamini-Hochberg post-test (**c**). **$P < 0.01$, ***$P < 0.001$. Molecular weights are indicated in kilodaltons (kDa).

Fig. 6 SPOP-WT is an ERG target gene and required for ERG-mediated cell invasion. **a** *SPOP* mRNA expression levels in 333 primary prostate cancer tissues stratified according to the indicated driver mutations[1]. Error bars, mean ± s.d. **b** SPOP mRNA and protein levels in response to forced expression of ΔERG in PC3 prostate cancer cells by qPCR and immunoblotting, respectively. Error bars, mean + s.e.m. (n=3). *P* values were determined by unpaired, two-tailed Student's t-test. #*P* < 0.05; Control versus ΔERG for *SPOP* expression levels. ***P* < 0.001; Control versus ΔERG for *ERG* expression levels. **c** Transwell Matrigel invasion assay of PC3 cells with forced expression of ΔERG and knockdown of SPOP using two different short hairpin RNAs. Protein expression of the indicated proteins was assessed in parallel by immunoblotting. Error bars, mean ± s.e.m. (n=3). **d** Transwell Matrigel invasion assay of PC3 cells with forced expression of ΔERG and HA-ZMYND11-DMT2 and corresponding immunoblot analysis. Error bars, mean ± s.e.m. (n=3). **e** Analysis of the ΔERG- and HA-ZMYND11-DMT2-induced transcriptional changes in the ERG target genes *PLAU* and *PLAT*. All error bars, mean ± s.e.m. *P* values were determined by one-way ANOVA with multiple comparisons and adjusted using Benjamini-Hochberg post-test (**a,c,d,e**). NS, not significant. ***P* < 0.01, ****P* < 0.001. Molecular weights are indicated in kilodaltons (kDa).

Fig. 7 *ERG*-positive tumor cells are particularly sensitive to SPOP inhibition. **a** SPOP inhibitor (SPOP-i, compound 6b) mediated 3D growth inhibition in methylcellulose in the indicated prostate cancer cell lines. **b** Tumor growth kinetics with (n =10) or without (vehicle; n = 10) SPOP-i treatment in xenografts established from LuCaP-147 (SPOP-Y83C) PDX cells. **c** Tumor growth kinetics with (n = 4) or without (vehicle; n = 4) SPOP-i treatment in xenografts established from LuCaP-78 (SPOP-W131G) PDX cells. **d** Tumor growth kinetics with (n = 11) or without (vehicle; n = 11) SPOP-i treatment in xenografts established from VCaP. **e** Tumor growth kinetics with (n = 8) or without (vehicle; n = 10) SPOP-i treatment in LuCaP-35 (ERG-positive) PDX. **f** Tumor growth kinetics with (n = 6) or without (vehicle; n = 8) SPOP-i treatment in LuCaP-23.1 (ERG-positive) PDX. All SPOP-i treatment initiated when tumors reached an average of 100mm^3. **g** Dose-response curves to SPOP-i treatment of Mouse Prostate Organoids overexpressing ΔERG, SPOP-Y87C and Control vector**.** All error bars, mean + s.e.m. *P* values were determined by two-way ANOVA with multiple comparisons and adjusted using Benjamini-Hochberg post-test. NS, not significant. ***P* < 0.01, ****P* < 0.001.

Fig. 8 ERG and mutant SPOP trigger different responses to therapeutic interventions a Dose-response curves to DHT treatment of VCaP, LuCaP-35, LuCaP-78 and LuCaP-147 PDX cancer cells. Prior to DHT treatment, PDX were grown in standard media without DHT. VCaP were starved for 24h in CSS medium (RPMI + 10% charcoal-stripped serum). Cell viability was assessed after 2 weeks. **b** Tumor growth kinetics with (n = 10) or without (vehicle; n = 10) testosterone treatment in xenografts established from LuCaP-147 (SPOP-Y83C). **c** Tumor growth kinetics with (n = 4) or without (vehicle; n = 4) testosterone treatment in xenografts established from LuCaP-78 (SPOP-W131G) cells. **d** Tumor growth kinetics with (n = 6) or without (vehicle; n = 10) testosterone treatment in xenografts

established from VCaP (ERG positive) cells. **e** Tumor growth kinetics with (n =12) or without (vehicle; n = 12) testosterone treatment in xenografts established from LuCaP-23.1 (ERG positive) cells. **f** Tumor growth kinetics with (n = 10) or without (vehicle; n = 10) testosterone treatment in xenografts established from LuCaP-35 (ERG positive) cells. **g** Sensitivity to Testosterone and SPOP-i treatment in xenograft and PDX models. LuCaP-23.1, LuCaP-35 and VCaP are *ERG*-positive cancer cells. LuCaP-147 and LuCaP-78 are *SPOP* mutant cancer cells (respectively SPOP-Y83C and SPOP-W131G). Growth inhibition is calculated using the last tumor measurements as shown in b-f and Fig. 7b-f. **h** Correlation of sensitivity to SPOP-i or testosterone treatment shown in Extended Data Fig. 8f-j and 9e-i, with ERG protein levels, as quantified by immunoblot, in PDX models and xenografts. *P* values were calculated using Pearson rank correlation. Corresponding immunoblot and quantification of AR and ERG protein levels depicted as a heatmap. Protein expression changes were normalized to GAPDH and LuCaP-78. All error bars, mean + s.e.m. *P* values were determined by two-way ANOVA with multiple comparisons and adjusted using Benjamini-Hochberg post-test (**b-e**) or by unpaired, two-tailed Student's t-test (**g**), NS, not significant. **P < 0.01, ***P < 0.001. Molecular weights are indicated in kilodaltons (kDa).

Fig. 9 *SPOP* mutant tumors are particularly susceptible to androgen deprivation therapies (ADT). a Progression-free survival of prostate cancer patients derived from the TCGA-cohort. Curves representing TMPRSS2-ERG rearranged and SPOP-mutant patients are indicated in violet and green, respectively. The area around the curves represents 80% confidence interval. The bar plot in the lower left corner indicates the percentage of SPOP-mutant tumors within all patients who were diagnosed with prostate cancer (DIAG) and within the individuals who developed a progression of the disease (PROG). **b** Overall survival of prostate cancer patients derived from the MSK-IMPACT cohort. Curves representing *TMPRSS2-ERG* rearranged and *SPOP*-mutant patients are indicated in violet and green, respectively. The area around the curves represents 80% confidence interval. The bar plot in the lower left corner indicates the percentage of SPOP-mutant tumors within all patients who were diagnosed with prostate cancer (DIAG), within individuals who developed a metastatic progression of the disease (PROG), and within individuals who developed castration-resistant prostate cancer (CRPC). P values for Kaplan-Meier curves were determined using log-rank test. **c** Enzalutamide sensitivity of LAPC4 cells overexpressing ΔERG or SPOP mutant species (Y87C, W131G). All error bars, mean + s.e.m. *P* values were determined by unpaired, two-tailed Student's t-test (**c**), NS, not significant. *$P < 0.05$, **$P < 0.01$. Molecular weights are indicated in kilodaltons (kDa).

Fig. 10 Schematic representation of the proposed model for the aversive relationship between mutant *SPOP* and *ERG* in prostate cancer.

References

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