Pol II degradation activates cell death independently from the loss of transcription

Nicholas W. Harper[1], Gavin A. Birdsall[1], Megan E. Honeywell[1], Athma A. Pai[3], and

Michael J. Lee[1,2,*]

[1] Department of Systems Biology, UMass Chan Medical School; Worcester, MA, USA.

[2] Program in Molecular Medicine, and Department of Molecular, Cell and Cancer Biology, UMass Chan Medical School; Worcester, MA, USA.

[3] RNA Therapeutics Institute, UMass Chan Medical School; Worcester, MA, USA.

* Corresponding author. Email: michael.lee@umassmed.edu.

ABSTRACT

Pol II-mediated transcription is essential for eukaryotic life. While loss of transcription is thought

to be universally lethal, the associated mechanisms promoting cell death are not yet known.

Here, we show that death following loss of Pol II is not caused by dysregulated gene

expression. Instead, death occurs in response to the loss of Pol II protein itself. Loss of Pol II

protein exclusively activates apoptosis, and using functional genomics, we identified a

previously uncharacterized mechanism, which we call the **P**ol II **D**egradation-dependent

Apoptotic **R**esponse (**PDAR**). Using the genetic dependencies of PDAR, we identify clinically

used drugs that owe their efficacy to a PDAR-dependent mechanism. Our findings unveil a

novel apoptotic signaling response that contributes to the efficacy of a wide array of anti-cancer

therapies.

INTRODUCTION

Accurate control of gene expression is critical for all cellular processes. For nuclear protein-coding genes, expression depends on the transcriptional activity of RNA polymerase II (hereafter referred to as Pol II). Pol II activity is considered a life-essential function, and prolonged inhibition of Pol II activity is expected to be universally lethal (*1*). In recent years, drugs designed to target the transcriptional machinery have been investigated in clinical settings, especially in the treatment of various cancers (*2*). However, despite the importance of Pol II-mediated transcription and clinical interest in drugs targeting the transcriptional cycle, the mechanisms of lethality following transcriptional inhibition remain unclear.

The mechanisms of lethality following Pol II inhibition remain molecularly uncharacterized largely due to the presumption that death in this context is not regulated. Regulated or "active" cell death refers to death induced by active signaling through specific effector enzymes or defined regulatory pathways (*3*). In contrast, transcriptional inhibition is generally assumed to cause death due to passive mechanisms; this begins with mRNA decay and the subsequent loss of protein, leading to some unavoidable catastrophic event. Cell death of this type is sometimes referred to as "accidental cell death", highlighting the use of passive mechanisms, such as decay (*4*).

In contrast with the notion that transcriptional inhibition causes an accidental death, recent studies highlight that cells are remarkably capable of buffering against perturbations to the mRNA pool (*5–9*). For instance, in normal conditions mRNA concentration scales with cell size; variations in cell size are buffered by tuning Pol II activity or mRNA decay to maintain stable mRNA concentrations (*7, 10*). Likewise, in perturbed conditions with reduced mRNA production, the effects are also blunted by a reduction in mRNA degradation rates (*7, 8*). Thus, if cells have the capacity to buffer against the downstream effects of transcriptional inhibition,

why do they instead die? To address this question, we aimed to formally characterize the mechanism(s) of lethality following Pol II inhibition in different contexts.

RESULTS

Pol II inhibition exclusively activates an apoptotic cell death

To investigate the mechanisms of cell death by transcriptional inhibition, we focused on triptolide and α-amanitin, two well-validated transcriptional inhibitors that result in the degradation of Pol II (*11*, *12*). We first measured the levels of Rpb1, the largest essential subunit of Pol II. As expected, U2OS cells exposed to a high dose of triptolide rapidly and completely degraded Rpb1, and Rpb1 degradation kinetics could be titrated in a drug dose-dependent manner (Fig. 1A and B, and fig. S1 A and B). To validate that Rpb1 protein expression could be used to interpret the transcriptional state, we also measured incorporation of 5-ethynyl uridine (EU) into newly synthesized RNAs, which confirmed the kinetics of transcriptional inhibition following triptolide exposure (fig. S1 C to E). Similar dose and time behavior was also observed following exposure to α-amanitin (fig. S1F).

To explore cell death in response to Pol II inhibition, we used 1 μM triptolide, which completely inactivated transcription within ~4 hours (Fig. 1A and B, and fig. S1D). We used the Scalable Time-lapse Analysis of Cell death Kinetics (STACK) assay, which quantifies both live and dead cell populations (*13*). We observed that live cells proliferated normally for only the first 4 hours following triptolide exposure, followed by a complete loss of proliferation, coinciding with the initiation of cell death (Fig. 1C). Cell death kinetics at lower doses of triptolide were similar, but with a delay in death onset that matched the later degradation of Pol II (fig. S2 A and B). Similar results were observed with α-amanitin, suggesting that rapid onset of lethality is a general feature of drugs that degrade Pol II (fig. S2 C and D). Taken together, these data reveal

that Pol II inhibition activates cell death much more rapidly than previously appreciated, and that

cell death is the primary phenotypic outcome of complete transcriptional inhibition.

Apoptosis following loss of transcription results from an active signaling process

5   Since inhibition of Pol II initiated cell death more rapidly than expected, we next aimed to

characterize which death mechanism(s) are activated by loss of transcription (fig. S3A). At least

14 distinct forms of regulated cell death have been identified (*14–16*). We reasoned that if cell

death following loss of transcription resulted from passive protein loss, rather than an active

signaling process, then inhibition of any single death pathway should fail to rescue viability. To

10   test this, we genetically or chemically inhibited each death pathway, one at a time, and

observed the effect on cell death following triptolide exposure. To inhibit apoptosis, we knocked

out the mitochondrial pore-forming proteins, BAX and BAK (*BAK1*) (*17*, *18*). Remarkably,

BAX/BAK double-knockout (DKO) cells were completely resistant to the lethal effects of

triptolide (Fig. 1D, and fig. S3 B and C). While BAX/BAK DKO cells did not die, these cells were

15   rendered completely non-proliferative within 4 hours of triptolide exposure, suggesting that

triptolide had completely inhibited transcription in these cells (Fig. 1D, and fig. S3D and E).

Triptolide-induced death was also suppressed using the caspase inhibitor, z-VAD, whereas

selective inhibitors of other death pathways did not alter the lethality of triptolide (Fig. 1E).

Similar effects were observed following α-amanitin treatment in U2OS cells (Fig. 1E, and fig. S3

20   F and G). Furthermore, inhibiting apoptosis suppressed the lethality of transcriptional inhibitors

in a panel of genetically unrelated cell lines (Fig. 1E, and fig. S3H). These data reveal that

transcriptional inhibition – even complete and prolonged transcriptional inhibition – exclusively

activates a single cell death pathway, the cell-intrinsic apoptotic response.

Pol II was completely inhibited at the doses we used, and as expected, RNA levels

25   decreased rapidly following loss of Pol II activity (fig. S4 A to C). However, at the time of death

initiation, cellular behaviors did not appear to be dysregulated by the loss of RNAs. For instance, we observed similar RNA decay rates U2OS and U2OS-BAX/BAK DKO cells (fig. S4D). However, 4 days following triptolide exposure, when U2OS were completely dead, BAX/BAK DKO cells display no overt signs of dysregulation, despite having lost more than 90% of the RNA pool (Fig. 1F to H, and fig. S4 E to G). Even 4 days after triptolide exposure, BAX/BAK DKO cells: were capable of re-adhering to plates following trypsinization, produced roughly normal levels of ATP, and could actively signal following exposure to external stresses, such as DNA damage (fig. S5 A to G). Normal behaviors were maintained in BAX/BAK DKO cells through ~10 days of triptolide exposure, after which, cells began dying, with the entire population dying by 24 days (Fig. 1G). Notably, the passive death observed in BAX/BAK DKO cells was coincident with the complete loss of all mRNAs, as evaluated using absolute mRNA quantification (Fig. 1H, and fig. S6 A and B). These data reveal that accidental cell death following loss of mRNA does indeed occur, but is temporally distinct from the rapid death activated by Pol II loss in cells that are proficient for activating apoptosis.

One potential explanation for these unexpected results is that the lethality induced by triptolide and α-amanitin stems from an off-target mechanism, unrelated to Pol II inhibition. To address this hypothesis, we evaluated the effects Pol II degradation in HAP1 cells, in which the endogenous RPB1 gene (POL2RA) is tagged with an auxin-inducible degron (HAP1-RPB1-AID) (Fig. 1G, top) (19). While auxin had no effect on wild-type HAP1 cells, HAP1-RPB1-AID cells displayed rapid cell death that initiated shortly after Pol II degradation (Fig. 1I, and fig. S7A). Furthermore, the lethality of Rpb1 degradation could be potently rescued by co-treatment with z-VAD, consistent with the results observed for triptolide and α-amanitin (Fig. 1I). Taken together, these data reveal that apoptosis following Pol II inhibition occurs in response to an active process that is distinct from the effects of global loss of mRNA.

Loss of Pol II protein, not loss of mRNA, activates apoptosis

While apoptotic-proficient cells die before experiencing global loss of mRNA, these cells still experience a complete loss of extremely short-lived transcripts (Fig. 1H, and fig. S4G). Mcl-1 is a potent negative regulator of apoptosis, that is unique among apoptotic regulators in that it is rapidly turned over at both the mRNA and protein levels (20–23). Indeed, Mcl-1 is rapidly lost following transcriptional inhibition at both the mRNA and protein levels (fig. S8 A and B). However, U2OS cells were insensitive to chemical inhibition or genetic knockout of *MCL1*, and loss of *MCL1* did not alter sensitivity to triptolide (fig. S8 C and D).

More generally, the loss of short-lived mRNAs did not predict the activation of apoptosis following Pol II inhibition. For instance, across transcriptional inhibitors, we observed significant variation in the kinetics of Pol II inhibition, with some drugs, such as the Cdk9 inhibitor flavopiridol, completely inhibiting Pol II activity within minutes, while other indirect inhibitors required several hours to completely inactivate Pol II (fig. S9 A to D) (24–27). However, the fastest-acting transcriptional inhibitors were not particularly fast at activating cell death. The two fastest transcriptional inhibitors, flavopiridol and THZ1, were, in fact, the two slowest activators of cell death (fig. S10 A to B). Overall, we found no correlation between the timing of Pol II inhibition and timing of cell death (fig. S10 C to D). These data suggest that the death induced by Pol II inhibition does not result from the loss of short-lived mRNAs.

The effects of transcriptional inhibitors could be phenocopied by degrading Rpb1 using an Auxin-inducible degron (Fig. 1I). Thus, we considered other mechanisms by which loss of Pol II could activate apoptotic death. Notably, while loss of Pol II activity was not correlated with activation of death, we observed a strong positive correlation between timing of cell death and timing of degradation for the *inactive* Pol II protein (fig. S10 E to G). Furthermore, the relationship between inactive Pol II decay and cell death was directly proportional, such that loss of inactive Pol II consistently preceded cell death by a fixed amount of time (fig. S10F).

Thus, we next tested the hypothesis that death was not caused by the loss of Pol II activity, but rather due to degradation of the Pol II protein itself.

A critical challenge for determining causality in this context is the mutual dependence between Pol II protein and its enzymatic function: degrading Pol II invariably inhibits Pol II activity, and the existence of Pol II protein relies on Pol II-mediated transcription. Indeed, we were unable to identify any compound capable of completely inhibiting transcription without also inevitably degrading Pol II protein. To overcome this challenge, we developed three complementary strategies for uncoupling the inhibition of Pol II activity and the degradation of Pol II protein (Fig. 2A, i-iii). First, we acutely degraded Pol II without inducing subsequent loss of mRNAs, by exposing cells to triptolide while concurrently inhibiting key mediators of RNA degradation (Fig. 2A-i). NUP93 and EXOSC5 have been shown to be critical mediators of mRNA concentration homeostasis, and acute knockdown of these proteins leads to an accumulation of mRNA (*7*). As expected, the mRNA pool in cells lacking NUP93 was stabilized, and significantly more resistant to triptolide-induced mRNA loss than control cells (Fig. 2B). Despite the stabilized mRNA pool, however, triptolide activated cell death to similar levels – and with similar kinetics – in control and NUP93-KO cells (Fig. 2C). Similar results were observed in EXOSC5-KO cells (fig. S11 A and B). These data again suggest that cell death following Pol II inhibition cannot be explained by the passive loss of mRNA alone.

Second, we sought to buffer against triptolide-induced degradation of Pol II, while maintaining transcriptional inhibition (Fig. 2A-ii). Triptolide inhibits XPB, the helicase subunit of TFIIH (*11*). Triptolide-induced degradation of Pol II has been shown to require Cdk7 activity, presumably because Cdk7 phosphorylation of Pol II facilitates removal of the pre-initiation complex, resulting in the release of Pol II from chromatin, enabling Pol II degradation (*27*). Consistent with this, co-treatment with the Cdk7 inhibitor, THZ1, blocks Pol II degradation following high-dose triptolide exposure (Fig. 2D and E). Importantly, THZ1 is itself a transcriptional inhibitor, and the combination of triptolide and THZ1 was as good at inhibiting Pol

II activity as either drug alone (Fig. 2D and F). Despite the potently inhibited transcriptional activity, we find that the addition of THZ1 suppresses the lethality of triptolide while maintaining potent transcriptional inhibition (Fig. 2D and G). This paradox – in which the lethal effects of transcriptional inhibition can be suppressed by adding another transcriptional inhibitor – is in line

5   with a model whereby lethality results from the degradation of Pol II protein.

         Finally, we asked if exogenous expression of a functionally inactive Pol II mutant protein was sufficient to rescue viability following degradation of the endogenous Pol II pool. To accomplish this, we designed a Pol II "switchover" system (Fig. 2H). This system features a doxycycline-inducible *RPB1* transgene completely devoid of the regulatory carboxy-terminal

10   domain (CTD) that is required for productive transcription (*28–31*). In addition, the *RBP1* fragment contains a known α-amanitin resistance mutation, N792D, allowing the selective degradation of endogenous Pol II (*12*). Conceptually, this system allows us to directly compare cell death outcomes following transcriptional inhibition, using isogenic cells that either have, or do not have, a pool of non-functional Pol II protein (Fig. 2I). To test this system, we stably

15   expressed the doxycycline-inducible *RPB1*-N792D-ΔCTD construct in U2OS cells expressing a Tet-On trans-activation protein. These cells were pretreated with- or without doxycycline for two days to enable expression of Rpb1-N792D-ΔCTD, followed by acute degradation of endogenous Pol II using 10 µM α-amanitin. As expected, complete loss of endogenous Pol II was observed following α-amanitin exposure, in both doxycycline-pretreated and untreated

20   conditions (Fig. 2J). However, only cells pretreated with doxycycline expressed a truncated Rpb1 protein that did not respond to α-amanitin (Fig. 2J). Expression of Rpb1-N792D-ΔCTD did not inhibit the loss of mRNA following α-amanitin treatment, validating that the CTD is required for productive transcription (Fig. 2K, and fig. S12A). Importantly, expression of Rpb1-N792D-ΔCTD potently suppressed α-amanitin-induced cell death (Fig. 2L). Expression of Rpb1-N792D-

25   ΔCTD had no effect on the lethality of unrelated apoptotic agents, and rendering the Rpb1-ΔCTD protein sensitive to α-amanitin abolished its ability to suppresses cell death (fig. S12 B

and C). These experiments demonstrate that cell death following transcriptional inhibition is caused by the degradation of Pol II protein, and not loss of Pol II activity.

Apoptosis following Pol II degradation is regulated by a unique set of genetic dependencies

We next aimed to characterize how decreased Pol II levels are sensed and communicated to initiate apoptosis. Considering the lack of prior knowledge for which processes respond to loss of Pol II protein, we evaluated this question using an unbiased genome-wide screening approach. We used Cas9-expressing U2OS cells, and the TKOv3 sgRNA library, to evaluate the effect of all single gene knockouts on triptolide-induced lethality (*32*, *33*). To identify genes that specifically regulate cell death, rather than proliferation, we developed a simple experimental approach to mechanically separate live cells from apoptotic corpses, such that both populations could be sequenced independently (Fig. 3A, and fig. S13 A to F).

Our screen identified more than 100 genes whose knockout significantly reduced the level of triptolide-induced cell death (Fig. 3B). Remarkably, this cohort of genes included every gene encoding an essential pro-apoptotic regulatory protein (Fig. 3B and C). We also identified key negative regulators of apoptosis, such as *XIAP* and BCL-xL (*BCL2L1*), which increase triptolide-induced lethality when deleted (Fig. 3B and C). Our screen also identified triptolide-specific genetic dependencies, including *CDK7*, in line with our finding that the Cdk7 inhibitor, THZ1, also rescues triptolide-induced cell death (Fig. 3B and 2G).

To clarify the mechanism by which Pol II degradation activates apoptosis, we began by exploring if the regulators of triptolide-induced cell death were enriched for genes involved in known stress response pathways. As expected, given that Pol II degradation exclusively activates apoptosis, the hits from our screen were strongly enriched for apoptotic regulatory genes (Fig. 2D); however, we did not identify any other known stress response pathways that

might facilitate activation of apoptosis (fig. S14A). Due to the proximity of Pol II to DNA, we paid

special attention to regulators of the DNA damage response (DDR). However, the DDR did not

appear to be involved in triptolide-induced cell death: no regulators of the DDR were hits in our

screen, triptolide failed to activate DDR signaling, and knocking out TP53, an essential regulator

of DDR-induced apoptosis (*34*), had no effect on triptolide-induced cell death (fig. S14 B to D).

We next explored if Pol II degradation activates apoptosis through a unique and yet

undescribed process. Surprisingly, the two genes whose deletion most strongly suppressed

triptolide-induced cell death were not core apoptotic effectors, but PTBP1 and BCL2L12 (Fig.

2B, and fig. S15A). BCL2L12 is a poorly understood protein with an undefined role in apoptotic

activation, even though the Bcl2 family has been extensively studied in the context of apoptosis

(*35*). PTBP1 is a multi-functional RNA binding protein, with well-established roles in alternative

splicing and activation of internal ribosome entry segments (IRESs) (*36–38*). To validate the role

of these genes in triptolide-induced cell death, we generated *PTBP1* and *BCL2L12* knockout

cells in the U2OS genetic background. Knocking out either *PTBP1* or *BCL2L12* suppressed

triptolide-induced cell death, validating the results of our screen (Fig. 2E).

Notably, our screen identified every essential apoptotic effector downstream of

mitochondrial outer membrane permeabilization; however, *BCL2L12* was the only Bcl-2

homology (BH) domain-containing gene that suppressed death (fig. S15B). These data suggest

that PTBP1 and BCL2L12 may play a pro-apoptotic role, specifically in the context of Pol II

degradation. To test this idea, we explored the effect of *PTBP1* or *BCL2L12* knockout in other

drug contexts. As we observed for triptolide, knocking out *PTBP1* or *BCL2L12* suppressed

lethality following exposure to other Pol II inhibitors, such as a-amanitin (Fig. 2F, and fig. S15 C

and D). In contrast, knocking out *PTBP1* or *BCL2L12* failed to suppressed lethality in the

context of canonical apoptotic agents, such as the BH3 mimetic, ABT-199, or Staurosporine

(Fig. 3F, and fig. S15 E and F). As expected, all apoptotic drugs failed to activate death in

BAX/BAK DKO cells (Fig. 2F, and fig. S15 C to F). Together, these data reveal that apoptotic death induced by Pol II degradation requires PTBP1 and BCL2L12.

Nuclear-to-cytoplasmic translocation of PTBP1 regulates death independently from its roles in RNA splicing

Knocking out *PTBP1* did not alter Pol II degradation, nor the triptolide-induced loss of transcriptional activity, or the decay of mRNAs following triptolide exposure (fig. S16 A to C). These data rule out the trivial explanation that loss of *PTBP1* inhibits triptolide activity. Furthermore, triptolide-induced changes in gene expression in U2OS and PTBP1-KO cells were nearly indistinguishable, and the shared expression changes between U2OS and PTBP1-KO cells included apoptotic regulatory mRNAs, such as *MCL1*, which were also rapidly lost at the protein level in both cell types (fig. S16 D to F).

Thus, we next explored the mechanisms by which PTBP1 could regulate apoptosis following Pol II degradation. PTBP1 is an RNA binding protein. Prior studies have reported a post-transcriptional function for PTBP1, which can promote TNF-induced apoptosis by activating IRESs to facilitate translation of apoptotic regulatory proteins, such as APAF1 (*39*, *40*). However, we observed that APAF1 protein expression was not altered during triptolide exposure, and APAF1 was expressed to similar levels in WT and PTBP1-KO cells (fig. S16F). The best described functions for PTBP1 are in regulating alternative mRNA splicing (*41–44*). Intriguingly, the influence of PTBP1 on triptolide-induced lethality also does not appear to be related to its role in splicing. Other regulators of alternative splicing did not rescue the lethality of triptolide (fig. S17A). Additionally, although we observed alternative splicing of known PTBP1-dependent splicing events in PTBP1-KO cells, triptolide did not affect PTBP1-dependent splicing, and splicing changes induced by triptolide were unaffected in PTBP1-KO cells (fig. S17 B to F) (*36*).

Many RNA binding proteins, such as PTBP1, are localized to the nucleus, and accumulate in the cytoplasm following transcriptional inhibition, though the precise reason for this is unclear (*45*, *46*). Curiously, triptolide-induced lethality was rescued by knocking out the nuclear exporter *XPO1*; and conversely, knocking out nuclear import factors, *RANGAP1* or *KPNB1*, sensitized cells to triptolide-induced lethality (Fig. 3G). Furthermore, we observed PTBP1 translocation to the cytoplasm shortly following triptolide or a-amanitin-induced Pol II degradation (Fig. 3H, and fig. S18 A and B). Finally, expression of inactive Pol II (Rpb1-N792D-DCTD) blocked the cytoplasmic translocation of PTBP1 and suppressed the lethality of a-amanitin (Fig. 3I, and fig. S18C). These data suggest that the cytoplasmic trafficking of PTBP1 is important for its death regulatory function.

We performed RNAseq on cytoplasmic and nuclear fractions separately in WT and PTBP1-KO cells, and found no evidence that the altered localization of PTBP1 is associated with a change in mRNA localization that can causally explain triptolide-induced cell death (fig. S18 D to F). A cross-comparison of our RNAseq and chemo-genetic profiling datasets further revealed no potential PTBP1-dependent transcriptional effects that could causally explain triptolide-induced cell death (fig. S18 G and H). These data further support our general finding that loss of mRNA cannot causally explain cell death following transcriptional inhibition. Moreover, these data support the notion that PTBP1 is a genuine regulator of apoptosis, specifically in the context of Pol II degradation, and that its mechanism in this context is likely unrelated to its currently known molecular functions.

Multiple anti-cancer drugs owe their lethality to Pol II degradation

We next aimed to investigate if activation of a Pol II degradation-dependent cell death could be valuable in a therapeutic context. To address this question, we focused on clinically relevant compounds – including those not conventionally considered transcriptional inhibitors –

and determined the degree to which the lethality of these drugs depended on genes required for triptolide-induced cell death. We generated a simple "Transcriptional Inhibition Similarity" (TIS) score to quantitatively evaluate the functional similarity of a given drug to the canonical Pol II degrader, triptolide (Fig. 4A and B). We applied this signature-based strategy across 46 diverse compounds, covering 16 clinically relevant drug classes (Fig. 4C, and fig. S19 A to I). Strikingly, a wide array of TIS scores were observed across the dataset, with variation both across and within drug classes (Fig. 4C, and fig. S20). High TIS scores were observed for most, but not all, drugs that directly target Pol II or transcriptional Cdk proteins (Fig. 4C). Additionally, drugs that induce proteotoxic stress, such as Brefeldin A, Thapsigargin, or Tunicamycin, were quite lethal, but featured low TIS scores, further confirming that Pol II degradation activates death in a manner that is distinct from general dysregulation of protein levels (fig. S20). Collectively, these data highlight that clinically relevant drugs covering disparate annotated mechanisms share the genetic dependencies observed for transcriptional inhibitors.

Because each drug was evaluated across a range of doses, our data revealed an unexpectedly large degree of variation in the TIS score across doses of a drug. To more precisely identify drug-dose pairs that owe their lethality to a triptolide-like mechanism, we developed a binary classifier to evaluate our TIS data. We used a conservative probabilistic nearest neighbors-based classification approach, in which a drug-dose pair was classified as "Transcriptional inhibition-like" only if its proximity to validated Pol II degrading drugs is within the neighborhood occupied by the Pol II degraders, triptolide and a-amanitin (fig. S21A) (*47, 48*). Our classifier identified significant variation in drug mechanisms of lethality, particularly across drugs that induce DNA damage (Fig. 4D). For instance, Idarubicin, a Topoisomerase II inhibitor, was lethal at the highest 4 tested doses, but was only "TI-like" at the highest dose (Fig. 2D, and fig. S20). Conversely, the UV radiation-mimetic, 4-NQO, was lethal at nearly all tested doses, but strikingly, was only TI-like at intermediate doses (Fig. 4D and fig. S20).

To validate our classifier, we next tested whether drugs classified as TI-like also degraded Pol II, and if this behavior was restricted to the appropriate doses. We focused on Abemaciclib, given the unexpected dose dependence, in which our classifier predicts Pol II degradation at only a single intermediate dose, and paradoxically, not at high doses (fig. S21B).

5 Consistent with our classifier, we found that Pol II is degraded following exposure to 1 μM Abemaciclib, but not when exposed to a ten-fold higher dose (Fig. 4E). These distinctions are particularly surprising given that Abemaciclib is lethal at both doses. Similar paradoxical dose-dependent Pol II degradation was observed for other drugs, including 4NQO, the Cdk2/4/6 inhibitor PF-06873600 (PF3600), and Idarubicin, (fig. S21 C to I).

10 Due to its widespread clinical use, we next focused on Cisplatin, a DNA cross-linking drug used in the treatment of bladder, head and neck, lung, ovarian, and testicular cancers, among others (*49*). Strikingly, our classifier predicts that Cisplatin has a TI-like mechanism of lethality at all doses that activate high levels of cell death (Fig. 4D, and fig. S22A). This was surprising, given that DNA damage can independently activate apoptosis through the DDR,

15 which is observed for other drugs in this dataset (fig. S20 and S22A). Consistent with our classifier, we observe Pol II degradation following Cisplatin exposure at all doses that induce substantial lethality (Fig. 4F, and fig. S22 C and D). These data suggest that the lethality of several DNA damaging agents may hinge upon a Pol II degradation-dependent apoptotic response in both a drug- and dose-dependent manner.

20 Finally, to further validate that Pol II degradation is responsible for the lethality of TI-like compounds, we explored other features that are unique to this mechanism. In addition to the dependence on PTBP1, BCL2L12, and the degradation of inactive Pol II protein, death kinetics represent another critical feature of this mechanism. We demonstrated that canonical transcriptional inhibitors activate death with consistent timing following loss of the inactive form

25 of Pol II (fig. S10F). For drugs classified as "TI-like", cell death kinetics were well-correlated with the kinetics of inactive Pol II degradation (Fig. 4G). Cell death kinetics for negative control drugs

showed no relationship, and no correlation was observed between cell death timing and loss of

Pol II activity (Fig. 4G). These data reveal that the mechanisms of drug-induced lethality vary in

unexpected and non-trivial ways across dose, and more specifically, that Pol II degradation is a

common mechanism of lethality across many unrelated drug classes.

DISCUSSION

The fact that transcriptional inhibition is lethal to cells is hardly surprising. However, the

common assumption to date has been that prolonged loss of transcription is toxic to cells due to

global loss of mRNA and protein. We show that death following transcriptional inhibition initiates

10   much more rapidly than previously appreciated, well before cells are experiencing stress from

the loss of mRNA or protein. Furthermore, we demonstrate that death occurs in response to an

active signaling process, which unexpectedly, is initiated by loss of the Pol II protein, not the

loss of Pol II transcriptional activity, nor the general decay in mRNA or protein levels. Therefore,

we propose that this previously uncharacterized process be termed the **P**ol II **D**egradation-

15   dependent **A**poptotic **R**esponse (**PDAR**).

Why has this phenotype not been observed previously? Notably, the central issue is

simply related to analytical resolution. Specifically, while apoptotic-deficient cells remain 100%

viable following Pol II degradation, these cells also cannot proliferate. Common drug response

metrics are likely to misinterpret the level of lethality in this context, because a non-proliferative

20   drug-treated population is already very small when compared to a rapidly dividing untreated

population (fig. S24A) (*50*). Furthermore, given the lack of mechanistic insight provided by

conventional analysis methods, any observed difference could reasonably have been

interpreted as evidence that apoptosis makes a partial and/or negligible contribution to the

lethality of transcriptional inhibition. Our initial evaluations of transcriptional inhibitors used the

25   drug GRADE analysis method (*51*), and the mechanistic clarity provided by GRADE-based

analysis revealed that apoptotic-deficient cells completely arrest cell proliferation without activating any cell death (fig. S24B).

Our study also identifies two key regulators of this process: PTBP1 and BCL2L12. PTBP1 is a ubiquitously expressed RNA binding protein with several established roles in RNA and translational regulation (*38, 40*). Notably, these prior functions appear not to contribute to lethality following Pol II degradation. Our study, however, confirms that PTBP1 can translocate to the cytoplasm in response to stress, which in the context of Pol II degradation, appears to communicate information about Pol II protein level from the nucleus to the cytoplasm, facilitating the activation of apoptosis. Additionally, *BCL2L12* is also a ubiquitously and highly expressed gene. This feature distinguishes *BCL2L12* from other members of the BCL2 family, which generally have low and varied expression across cell types. Furthermore, BCL2L12 is generally considered an anti-apoptotic protein, and has not been described to have any pro-apoptotic functions (*52*). Thus, it appears that PTBP1 and BCL2L12 have unique roles in response to Pol II degradation that are distinct from their previously annotated functions. Central remaining questions include developing a detailed molecular mechanism by which these proteins activate apoptosis following Pol II degradation, and exploring the context in which this pathway evolved to be beneficial.

Prior to this discovery, substantial attention had already been focused on targeting the transcriptional cycle in the context of cancer treatment. New drugs that target transcriptional Cdk proteins appear to be promising. However, the mechanisms allowing for the selective killing of cancer cells remain unclear, and are generally presumed to be dependent on loss of rapidly turned over proteins, such as Myc or Mcl-1 (*53*). Our study highlights that many of these drugs owe their lethality to the Pol II degradation-dependent mechanism that we describe. Notably, the definitions used in this study for defining a Pol II degradation-dependent cell death are conservative, as a drug was only characterized as PDAR-dependent if no other backup mechanisms existed to facilitate death in the absence of this pathway. Importantly, our study

also highlights that selective killing is possible for PDAR-dependent drugs, as we identified

several conventional DNA damaging chemotherapies, which have long-standing clinical utility,

that also unexpectedly score as PDAR-dependent using our conservative criteria. Thus, future

studies of the PDAR pathway are likely to reveal new biomarkers that improve our ability to use

5 a wide array of anti-cancer drugs, and also may reveal new ways to selectively kill cancer cells.

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Author contributions:

N.W.H. and M.J.L. conceptualized this project, including selection of methodology, analytical

strategies, data visualization, and interpretation. N.W.H. performed all experiments and

20 analyses in this study. G.A.B. assisted N.W.H. with the evaluation of cell death mechanisms

activated by transcriptional inhibitors and canonical cell death activating drugs. M.E.H.

generated the materials used in the chemo-genetic profiling of triptolide, and assisted with the

design and analysis of chemo-genetic profiling data. A.A.P. designed the experiments and

analytical strategy used in evaluating PTBP1- and triptolide-dependent changes in RNA

localization and RNA splicing. The original draft of this manuscript was written by N.W.H. and M.J.L. All authors participated in manuscript review and editing.

Competing interests:

The authors declare that they have no competing interests.

5 **Data and materials availability:**

RNA sequencing data generated in this study can be found as FASTQ files from the Gene Expression Omnibus (GEO) with the following accession numbers: GSE283148 (triptolide exposure time course in U2OS and BAX/BAK DKO cells), GSE283149 (triptolide exposure for 8 hours in NUP93 or EXOSC5 KO cells), GSE283150 (RNA sequencing from nuclear and 10 cytoplasmic fractionation in U2OS or PTBP1 KO cells), GSE283151 (triptolide exposure following a-amanitin in Pol II switchover system). Chemo-genetic profiling data can be found as raw FASTQ files (GSE283147).

MATERIALS AND METHODS

Cell lines and reagents

U2OS, HT-29, H1650, WI-38, Hs578T, MDA-MB-453, MCF7 and MDA-MB-157 cells were obtained from the American Type Culture Collection (ATCC). A375 cells were a gift from the Green laboratory (UMass Chan Medical School). HAP1 and HAP1-RPB1-AID cells were a gift from the Dekker laboratory (UMass Chan Medical School) (*19*). RPE-1 cells were a gift from the Pazour laboratory (UMass Chan Medical School). PC9 cells were a gift from the Pritchard laboratory (Penn State University). U2OS, MDA-MB-157, A375, Hs578T, MDA-MB-453, and MCF7 cells were grown in DMEM (Corning, 10-017-CV) supplemented with 2mM glutamine (Corning, 25-005-CI). HT-29 cells were grown in McCoy's 5A medium (Corning, 10-050-CV). WI-38 cells were grown in MEM medium (Thermo Fisher Scientific, 10370088). H1650 and PC9 cells were grown in RPMI medium (Thermo Fisher Scientific, 11875119). RPE-1 cells were grown in low glucose DMEM (Gibco, 11885-092) and Hams F-12 (Thermo Fisher Scientific, 11765054) mixed at a 1:1 ratio. HAP1 cells were grown in IMDM medium (Thermo Fisher Scientific, 12440053). Each media was supplemented with 10% FBS (Peak Serum, PS-FB2, lot no. 21E1202) and penicillin-streptomycin (Corning, 30-002-CI). Hs578T cells were supplemented with 10µg/mL insulin (Thermo Fisher Scientific, 12585014). Cell lines were cultured in incubators at 37C with 5% CO_2. Cell lines were maintained at low (<25) passage number from the original vial.

U2OS-BAX/BAK1 double knockout cells (DKO) were generated previously (*54*) . All mKate2+ cells were generated by lentiviral infection with NucLight Red Lentivirus (Sartorius, 4627), followed by FACS selection. HAP1-RPB1-AID cells were treated with 450 µg/mL hygromycin for one week prior to use. U2OS-Cas9 expressing and U2OS-TP53 knockout cells were generated previously (*34*) .

U2OS-PTBP1 and U2OS-BCL2L12 clonal knockout cells were generated using CRISPR. For each gene, the highest-scoring sgRNA was selected from the TKOv3 library (PTBP1: 5' – CGAGGTGTAGTAGTTCACCA – 3'; BCL2L12: 5' – AGAAGGAAGCCATACTGCGG – 3') and cloned into the pX330-puro plasmid using the single-step digestion-ligation protocol from the Zhang laboratory (available on the Zhang laboratory Addgene page). U2OS cells were then transiently transfected with the respective pX330-sgRNA-puro constructs using the FuGENE HD Transfection Reagent (Promega, E2311). Transfected cells were then selected with 1 µg/mL puromycin for 3 days, followed by replating and recovery for 2 days. Single cell clones were generated using limiting dilution. Clones were validated by sequencing, phenotyping, and for PTBP1 knockouts, immunoblotting.

U2OS-MCL1, U2OS-NUP93 and U2OS-EXOSC5 knockout cells were generated using a transient transfection strategy. Briefly, sgRNA (MCL1: 5' – CATGTAGAGGACCTAGAAGG – 3'; NUP93: 5' – GAACTTACAGGAGATCCAGC – 3'; EXOSC5: 5' – GAAGTTCTTACCTTGCAGGA – 3') were cloned into pX330-puro as described above. U2OS cells were plated in six-well plates (300,000 cells per well) and left to adhere overnight. The following day, cells were transiently transfected with 2 µg of respective pX330-sgRNA-puro construct using FuGENE HD Transfection Reagent. Cells were then selected using 1 µg/mL puromycin for either 2 days (for NUP93 and EXOSC5) or 3 days (for MCL1), followed by replating and expansion for either 1 day (for NUP93 and EXOSC5) or 2 days (for MCL1) prior to seeding for subsequent assays. Control cells were similarly transfected with non-targeting sgRNA (NONT: 5' – CGGCGGTCACGAACTCCAGC – 3').

For the RPB1 switchover system, a FLAG-RPB1-N792D-ΔCTD fragment was PCR amplified (Forward primer: 5' – CAATTCCACAACACTTTTGTCTTATACTTGGATCCATGGACTACAAGGACGACGATGACA - 3'; Reverse primer: 5' – TAGGGGGGGGGGAGGGAGAGGGGCCGGCCGGGGCTCAGCTGGGAGACATGGCACCAC

– 3') from FLAG-Pol2-WT (Addgene, 35175) (*55*) . The PCR fragment resulted in a RPB1

mutant that lacked the entire C-terminal domain (repeats 1-52) and included an N-terminal

FLAG tag, an α-amanitin resistance mutation, and homology regions to facilitate insertion into

pLVX-Tre3G-IRES (Clontech, 631362)—a lentiviral vector for doxycycline-inducible gene

5   expression. pLVX-Tre3G-IRES was digested with BamHI and NotI, gel extracted and purified.

The PCR product was then inserted into the MCS1 of pLVX-Tre3G-IRES using Gibson

Assembly (New England Biolabs, E2611). The resulting plasmid, pLVX-Tre3G-FLAG-RPB1-

N792D-ΔCTD, was packaged into virus using Lenti-X 293T cells (Takara, 632180), along with a

vector expressing the Tet-On 3G transactivator protein, pLVX-Tet3G (Clontech, 631358). U2OS

10   cells were infected with pLVX-Tet3G lentivirus, followed by selection with 600 µg/mL G418. Tet-

3G expressing cells were then infected with pLVX-Tre3G-FLAG-RPB1-N792D-ΔCTD virus, and

selected with 1 µg/mL puromycin. Single cell clones were generated using limiting dilution.

Clones were selected for transgene expression levels approximating endogenous Rpb1 levels,

determined using immunoblotting.

15   Compounds used in this study were as follows. SYTOX Green (Thermo Fisher Scientific,

S7020), Triptolide (Sigma-Aldrich, T3652), α-amanitin (Sigma-Aldrich, A2263), ABT-737

(ApexBio, A8193), RSL3 (ApexBio, B6095), Elesclomol (Sigma-Aldrich, SML2651), MNNG

(Biosynth Carbosynth, FM11256), TNF-α (ApexBio, P1001), SM-164 (ApexBio, A8815), IAA

(Sigma-Aldrich, I5148), Z-VAD-FMK (ApexBio, A1902), VX765 (ApexBio, A8238), TTM (Sigma-

20   Aldrich, 323446), Rucaparib (ApexBio, A4156), Necrostatin-1 (SelleckChemicals, S8037),

Ferrostatin-1 (SelleckChemicals, S7243), Doxycycline Hyclate (Sigma-Aldrich, D5207),

Actinomycin-D (SelleckChemicals, S8964, Ethynylcytidine (MedChemExpress, HY-16200),

Cordycepin (Sigma-Aldrich, C3394), Flavopiridol (SelleckChemicals, S1230), DRB (Sigma-

Aldrich, D1916), THZ1 (SelleckChemicals, S7549), THZ531 (MedChemExpress, HY-103618),

25   YKL-5-124 (MedChemExpress, HY-101257B), Abemaciclib (MedChemExpress, HY-16297A),

Palbociclib (ApexBio, A8316), Ribociclib (MedChemExpress, HY-15777), PF3600

(SelleckChemicals, S8816), PF4091 (ChemieTek, CT-PF0710), 4-NQO (SelleckChemicals, E0155), Cisplatin (SelleckChemicals, S1166), Carboplatin (Sigma-Aldrich, C2538), Camptothecin (SelleckChemicals, S1288), Topotecan HCl (SelleckChemicals, B2296), Idarubicin HCl (SelleckChemicals, S1228), Etoposide (SelleckChemicals, S1225), Teniposide (S1787 SelleckChemicals), Niraparib (SelleckChemicals, S2741), Rucaparib phosphate (SelleckChemicals, S1098), Cycloheximide (Sigma-Aldrich, C1988), Homoharringtonine (Sigma-Aldrich, SML1091), Anisomycin (Sigma-Aldrich, A9789), Bortezomib (ApexBio, A2614), MG-132 (SelleckChemicals, S2619), Brefeldin A (Sigma-Aldrich, B7651), Thapsigargin (Sigma-Aldrich, T9033), Tunicamycin (Sigma-Aldrich, T7765), ABT-199 (ApexBio, A8194), ABT-263 (ApexBio, A3007), Panobinostat (ApexBio, A8178), Vorinostat (ApexBio, A4084), Torin 2 (ApexBio, B1640), Rapamycin (ApexBio, A8167), Mirdametinib (MedChemExpress, HY-10254), Buparlisib (MedChemExpress, HY-70063), Capivasertib (MedChemExpress, HY-15431), SGI-1027 (ApexBio, B1622), JQ1 (SelleckChemicals, S7110), Staurosporine (SelleckChemicals, S1421), H_2O_2 (Fisher Scientific, BP2633500), LPS (Sigma-Aldrich, L4391), Nigericin (ApexBio, B7644), copper(II) chloride (ThermoFisher Scientific, 405845000), S63845 (SelleckChemicals, S8383).

Primary antibodies used in this study were as follows. Rpb1-NTD rabbit mAb (Cell Signaling Technology, 14958, 1:1,000 dilution), β-actin mouse mAb (Sigma-Aldrich, A2228, 1:15,000 dilution), DYKDDDDK Tag (FLAG) rabbit mAb (Cell Signaling Technology, 14793S, 1:1,000 dilution), PTBP1 rabbit mAb (Cell Signaling Technology, 57246S, 1:1,000 dilution), Histone-H3 rabbit mAb (Cell Signaling Technology, 4499S, 1:2,000 dilution), α-tubulin mouse mAb (Sigma-Aldrich, T6199, 1:10,000 dilution), pH2A.X (Ser139) rabbit mAb (Cell Signaling Technology, 9718, 1:1,000 dilution), Mcl-1 rabbit mAb (Cell Signaling Technology, 5453, 1:1,000 dilution), Apaf-1 rabbit mAb (Cell Signaling Technology, 8723S, 1:1,000 dilution).

Immunoblotting

Cells were seeded in 6-well plates (300,000 cells per well) or 10-cm dishes (1.5 x 10^6

cells per dish) and left to adhere overnight, unless otherwise stated. Cells were treated the next

morning. At indicated timepoints, media were removed and collected. Samples were washed

5   once with PBS, and the wash was added to the collected media. SDS-lysis buffer (50 mM Tris-

HCl, 2% SDS, 5% glycerol, 5 mM EDTA, 1 mM NaF, 10 mM β-glycerophosphate, 1 mM PMSF,

1 mM Na$_3$VO$_4$, protease inhibitor tablet and phosphatase inhibitor tablet) was used to lyse the

cells, along with any dead cells that were pelleted from the media and wash. Lysates were

centrifuged through an AcroPrep 96 well 3.0 µm glass fiber/0.2 µm Bio-Inert membrane filter

10   plate (Pall Laboratory, 5053). Protein content was quantified with a BCA assay (Thermo Fisher

Scientific, 23225) and normalized for equal loading. Samples were boiled for 5 minutes at 95C

in 6x Laemmli buffer. Denatured samples were run on hand-poured SDS-PAGE gels. Gels were

subsequently wet transferred onto nitrocellulose membranes and blocked in 1:1 PBS: Intercept

Blocking Buffer (LI-COR, 927-70003) for 1 hour at room temperature. Membranes were then

15   incubated overnight on a rocking shaker at 4C in primary antibody diluted in 1:1 PBS-0.1%

Tween: Intercept Blocking Buffer. The following day, membranes were incubated in primary

anti–β-actin antibody (diluted 1:15,000) for 1 hour at room temperature, followed by two 5-

minute washes with PBS-0.1% Tween. Membranes were then incubated with secondary

antibodies (diluted 1:15,000 in 1:1 PBS-0.1% Tween: Intercept Blocking Buffer) conjugated to

20   infrared dyes (LI-COR, IRDye 680RD, goat anti-mouse immunoglobulin G (IgG) secondary,

catalog no. 926-68070; IRDye 800CW, goat anti-rabbit IgG secondary, catalog no. 926-32211)

for 1 hour at room temperature. Following four 5-minute washes with PBS-0.1% Tween and one

5-minute wash with PBS, blots were visualized using a LI-COR Odyssey CLx scanner. Raw

protein expression levels were quantified in LI-COR's ImageStudio software, background

25   subtracted, divided by β-actin signals to normalize for loading differences, then normalized per

gel to a reference sample (generally, an untreated sample).

Labelling and measurement of nascent transcription

Methods for labelling and quantification of nascent transcription were adapted from methods described previously (*56*). Cells treated with transcriptional inhibitors were labeled with 1 mM 5-ethynyl uridine (Sigma-Aldrich, 909475) for 1 hour in existing media. Cells were then washed with PBS, trypsinized, pelleted, and washed with cold PBS. Cells were then fixed in 4% fresh formaldehyde in PBS at room temperature for 15 minutes. Fixed cells were washed with cold PBS, pelleted, resuspended in ice-cold 100% methanol, and stored at -20C overnight. The following day, the methanol was removed, and the cells were washed twice with PBS-0.1% Tween. Cells were then incubated for 30 minutes at room temperature with 86.5 µL Click-iT reaction buffer, 4 µL $CuSO_4$ buffer, 0.125 µL Alexa Fluor-azide-488 and 10.3 µL Click-iT reaction buffer additive (all from Thermo Fisher Scientific, C10329). Cells were washed once with 100 µL Click-iT reaction rinse buffer (from Thermo Fisher Scientific, C10329), then washed two more times with PBS-0.1% Tween. Samples were resuspended in PBS-0.1% Tween, filtered, and run on a Miltenyi MACSQuant VYB cytometer with laser and filter settings appropriate for reading Alexa-488. Data were analyzed using FlowJo software.

RNA sequencing with ERCC spike-ins for absolute RNA quantification

For evaluation of mRNA abundance following long-term triptolide exposure in DKO cells, DKO and WT cells were seeded onto six-well plates (300,000 cells per well) and left to adhere overnight. Cells were then drugged the next day (D0). To evaluate the transcriptome of WT cells after 1 day of triptolide, cells were cotreated with 50 µM Z-VAD to inhibit loss of material from cell death. Due to the long assay length, plating and drugging was staggered for each condition, such that the assay endpoint for each sample occurred simultaneously. For analyses that

defined the 50 shortest or 50 longest half-life RNAs, these lists of RNAs were based on prior studies (*57*) .

For evaluation of mRNA abundance in the context of NUP93 or EXOSC5 knockout, knockout and nontargeting cells were seeded onto six-well plates (300,000 cells per well) and left to adhere overnight. Cells were then drugged the next day.

For evaluation of mRNA abundance in Pol II switchover cells, Rpb1-N792D-ΔCTD expressing cells were seeded onto six-well plates (100,000 cells per well) and left to adhere overnight. The following day, the media was removed and replaced with fresh media with or without 2 µg/mL doxycycline. 48 hours later, drug or vehicle was gently added to the existing media.

For the extraction of mRNA for the experiments above, at assay endpoint cells were trypsinized, counted using a hemacytometer, pelleted, washed with PBS, and moved immediately to RNA extraction. Total R   NA was extracted using the RNeasy Plus Mini Kit (Qiagen, 74134) according to manufacturer protocol. To facilitate absolute RNA quantification, polyadenylated ERCC standards (Thermo Fisher Scientific, 4456740) were diluted 1:100 in nuclease-free water, and added to each sample at 1 µL/100,000 cells. ERCC spike-ins were added immediately following lysis of cells in RLT-plus buffer, and prior to all subsequent total RNA extraction steps using the RNeasy Plus kit. All RNA integrity numbers (RIN) were greater than 9, as measured on a 4150 Tapestation System (Agilent Technologies). mRNA was isolated using the NEBNext Poly(A) mRNA Magnetic Isolation Module (New England Biolabs, E7490). RNAseq libraries were prepared with the NEBNext Ultra II Directional RNA Library Prep Kit using the NEBNext Multiplex Oligos for Illumina (New England Biolabs, E7765 and E7600). Paired-end 2 x 50nt sequencing of libraries was performed in-house on the Illumina NextSeq2000 using the NextSeq 1000/2000 P2 XLEAP-SBS Reagent Kit (Illumina, 20100987).

Quality control for the RNAseq datasets were performed using FastQC. Transcript

abundance was estimated using Kallisto (0.46.1) with parameters --bootstrap-samples 30 --

single-overhang --rf-stranded (*58*). The Ensembl GRCh38 cDNA transcriptome build, appended

with ERCC control transcript sequences, was used to build a Kallisto index. Gene-level counts

were generated in R using the tximport package (*59*). Counts were normalized to ERCC spike-

ins using DESeq2, implemented with the function "estimateSizeFactors" with the option

"controlGenes" set to the identities of the 92 ERCC transcripts (*60*) . Counts were filtered for

protein coding genes in R using AnnotationHub. Differential gene expression was performed

with DESeq2, and fold change shrinkage was performed using the function lfcShrink with the

adaptive shrinkage estimator "ashr" (*61*) .

Extraction of RNA and protein from nuclear and cytoplasmic fractions

Nuclear-cytoplasmic fractionation was performed in such a way to facilitate extraction of

RNA and protein from the same cellular lysate. A similar protocol, excluding the RNA extraction

steps, was used in fractionation experiments that only examined protein expression. Cells were

plated at 6 million cells per condition (3 million cells per 15 cm plate) and left to adhere

overnight. Cells were then drugged the next day. At assay endpoint cells were trypsinized,

counted using a hemacytometer, pelleted, washed with PBS, and moved immediately to cellular

fractionation. Cells were first resuspended in 900 µL hypotonic lysis buffer (50 mM HEPES pH

7.2, 150mM NaCl, 0.5 mM EDTA, 0.15% Triton X-100, 1 mM PMSF, protease inhibitor tablet,

phosphatase inhibitor tablet, 20 U/mL SUPERase-In RNase inhibitor (Thermo Fisher Scientific,

AM2694)). Cells were pipet up and down 10 times, 1/3 of the lysate was taken as a whole-cell

lysate (WCL) sample, and the cells were left to swell on ice for 10 minutes. Samples were

centrifuged at 12,000xg for 30 seconds at 4C. The supernatant was collected as the

cytoplasmic fraction (CYT). The nuclear pellet was washed once with 500 µL wash buffer (PBS

containing 0.1% Triton X-100, 1mM EDTA and 20 U/mL SUPERase-In), centrifuged at 12,000xg

for 30 seconds at 4C, and the wash buffer was completely removed. The nuclear pellet was

then lysed in 400μL 1x nuclear lysis buffer (50 mM HEPES pH 7.2, 150mM NaCl, 0.5 mM

EDTA, 1% Triton X-100, 0.5% Sodium Deoxycholate, 0.2% SDS, 1 mM PMSF, protease

inhibitor tablet, phosphatase inhibitor tablet, 20 U/mL SUPERase-In). 10x nuclear lysis buffer

was added to WCL and CYT fractions to equalize surfactant amounts to that of the nuclear

fraction. Each resulting lysate was divided in half to facilitate either downstream protein analysis

or RNA extraction.

For protein fractions, 100x Benzonase buffer (50 mM Tris-HCl, 20 mM NaCl, 2 mM

MgCl$_2$, 2000 U/mL Benzonase Nuclease (Sigma Aldrich, E1014)) was added to lysates. Lysates

were kept on ice for 30 minutes with periodic vortexing. The BCA assay was performed on all

cellular fractions, and 10 μg of each lysate was subjected to immunoblotting as described

above. Total protein was detected using the Revert 700 Total Protein Stain Kit (LI-COR, 926-

11010) prior to incubation with primary antibody.

For extraction of total RNA from RNA fractions, samples were subjected to the RNeasy

Plus manufacturer protocol with the following modifications. 3.5x sample volume of RLT-plus

buffer was added to cellular fractions. Prior to removal of genomic DNA using the gDNA

Eliminator spin column, ERCC standards (diluted 1:10) were added to each lysate at a

concentration of 1 μL/ 1 million cells (or similarly, 1 million nuclei or 1 million cell cytoplasm's).

Following genomic DNA removal, 2.5x original lysate volume of 100% ethanol was added to

each flowthrough, and subsequent steps of the RNeasy Plus protocol were followed as

standard. mRNA was isolated as above. Sequencing libraries were prepared with the NEBNext

Ultra II Directional RNA Library Prep Kit, using manufacturer recommended modifications to

generate large (~450nt) inserts. Paired-end 2 x 150nt sequencing of libraries was performed in-

house on the Illumina NextSeq2000 using the NextSeq 2000 P3 Reagent Kit (Illumina,

20040561).

Alternative splicing analysis from RNA sequencing data

To facilitate detection of splice-level events, cytoplasmic and nuclear RNAseq FASTQ files corresponding to the same biological samples (above) were pooled. Regulated splicing events were detected and analyzed with VAST-TOOLS (*62*, *63*) . Briefly, reads were aligned to the human genome (hg38) using the align function. Differential splicing events were identified using the diff function (with parameters -S 2). Events were filtered for adequate coverage using the tidy function (with parameters -min_N 4, --noVLOW, --min_SD 5, and --p_IR). Splicing events were considered significantly different if the VAST-TOOLS reported expected absolute ΔPSI value was greater than 0.1 and if the minimum value (95% confidence) of the absolute ΔPSI value was greater than 0.

STACK assay for kinetic evaluation of drug responses

The STACK assay was used to evaluate cell proliferation and cell death continuously over time using microscopy, as previously described (*13*) . Briefly, mKate2+ cells were seeded into 96-well black-sided plates, and left to adhere overnight. The following day, cells were given drugged media containing 50 nM SYTOX Green, and imaged over time time-lapsed images were collected using an IncuCyte S3 (Essen Biosciences). Images were collected with a ×10 objective with acquisition in the green channel ex: 460 ± 20, em: 524 ± 20, acquisition time: 300ms; and red channel ex:585 ± 20, em: 635 ± 70, acquisition time: 400ms. Counts of dead (green) and live (red) cells were determined using the built-in IncuCyte software (Essen Biosciences) and exported for analysis using a custom MATLAB script.

For the long-term evaluation of BAX/BAK DKO cells following triptolide, accurate dead cell measurements using STACK as described above are insufficient, as identification of a dead cell from an image is no longer possible after the dead cell fully degrades away. To overcome

this technical confound, a modified STACK assay was used, taking advantage of the FLICK

assay (described in detail below), which does not rely on image segmentation for cell

quantification. Cells were seeded and drugged as above, at a high density (15,000 cells per

well). Separate plates were prepared for each timepoint, whereby cells were first imaged using

5   an EVOS FL Auto 2 automated microscope (ThermoFisher Scientific). Images were acquired

using a 10x objective (EVOS 10x objective, Cat #: AMEP4681). Sytox images were acquired

using a GFP filter cube (EVOS LED Cube, GFP, Cat #: AMEP4651, ex: 470/22, em: 525/50,

acquisition time: 13.5ms) Mkate2+ images were acquired using a TexasRed filter cube (EVOS

LED Cube TxRed, Cat #: AMEP4655, ex: 585/29, em: 628/32, acquisition time: 642.0ms).

10   Following imaging, FLICK was performed as an endpoint measurement to obtain lethal fraction

values, as described below. Images were linearly adjusted using a custom MATLAB script.

Quantification of drug-induced cell death using the FLICK assay

The FLICK assay was performed as described (*54*, *64*) . Briefly, 90 µL of cells were

15   seeded at a density of 2,000-5,000 cells per well in black-sided optical-bottom 96-well plates

(Greiner Bio-One, 655090) and left to adhere overnight. The following day, cells were treated

with 10 µL of media containing the indicated compound/s along with SYTOX Green, resulting in

a final SYTOX concentration of 2 µM. Fluorescence was monitored at various times following

drugging with a Tecan Spark (running SparkControl software version 2.2) microplate reader (ex:

20   503, em: 524). Gain was set to achieve a linear relationship between SYTOX signal and dead

cell number. To obtain the total cell number at the time of drugging, a duplicate "T0" plate was

lysed in 0.15% Triton X-100 and 2 µM SYTOX for 2-4 hours at 37C prior to taking a

fluorescence reading. Total cell number for each condition at the end of the assay was similarly

determined by lysing cells in 0.15% Triton X-100. From these measurements (dead cells at

25   each timepoint, along with total cell numbers at the beginning and end of the assay), live cell

numbers over time can be inferred and all calculations to determine LF, FV, RV and GR can be made (*50*):

$$LF = \frac{Dead}{Live+Dead}, \; FV = \frac{Live}{Live+Dead}, \; RV = \frac{Live_{Trt}}{Live_{Unt}}, \; GR = 2^{\left(\frac{log_2\left(\frac{Live_{Trt}}{Live_{T0}}\right)}{log_2\left(\frac{Live_{Unt}}{Live_{T0}}\right)}\right)} - 1$$

All analyses were performed using custom MATLAB scripts. Dose-response curves were fit using a 4-parameter sigmoidal equation. Lethal fraction kinetics were fit using the lag-exponential death (LED) equation (*13*):

$$LF(t) = LF_0 + (LF_P - LF_0)(1 - e^{-D_R(t-D_O)})$$

All analyses were performed using custom MATLAB scripts. All data were fit with the "fit" function in MATLAB using nonlinear least-squares.

Drug GRADE-based analysis of drug responses

True proliferation rates and death rates were calculated using the GRADE method, as previously described (*51*) . Conceptually, the GRADE method infers the true, underlying cell proliferation rates and death rates of cell population by cross-referencing experimentally measured death rates (FV or LF) and net population growth rates (GR) with computational simulations. To begin, we simulated all pairwise combinations of 500 proliferation rates and 500 death rates using the following equations where C_0 = initial cell number, t = assay length, τ = proliferation rate, and D_R = death rate.

$$Live \; Cell \; Number = \left(C_0 * 2^{\left(\frac{t}{\tau}\right)}\right) - (C_0 * 2^{\left(\frac{t}{\tau}\right)} * D_R)$$

$$Dead \; Cell \; Number = \; C_0 * 2^{\left(\frac{t}{\tau}\right)} * D_R$$

FV and GR values (defined above) were then calculated for each simulated proliferation rate and death rate pair. Next, for a given condition, experimentally determined FV and GR values

were calculated (as above). The simulated proliferation rate and death rate that yield the empirically determined FV and GR values was then found.

Quantification of Pol II half-life

Half-lives for Pol II-a and Pol II-o were determined using temporally resolved quantitating immunoblotting using an antibody recognizing the N terminal domain of Rpb1 (Cell Signaling, 14958S), as described above. Immunoblots were quantified in ImageStudio 3.1.4, normalized to β-actin, then normalized to the untreated sample (time 0) on each gel respectively, generating an abundance value relative to steady state. Data were fit to a two-term exponential decay function in MATLAB, relative to the highest Rpb1 signal observed over the time course. Half-lives determined from the fits, as the time at which half-maximal signal was reached. To correlate Pol II half-lives with cell death timing, the LF50 time was similarly found from lag-exponential death (LED) fits to STACK data. LF50 was defined as the time at which half of the maximum observed LF (measured or inferred after 72 hours) was reached.

Pol II Switch over system

For experiments involving the RPB1 switchover system, cells were seeded as described for the respective experiment. The following day, cells we treated with or without 2 µg/mL Doxycycline (Sigma, D5207) for 48 hours to induce the expression of FLAG-RPB1-N792D-ΔCTD. 10 µM α-amanitin was added to the cells without removing the conditioned media to degrade the endogenous Pol II.

Chemo-genetic profiling of triptolide-induced cell death

Conventional approaches used for these "chemo-genetic profiling" studies are unable to accurately identify death-regulatory genes due to the confounding effects of varied proliferation rates between cells harboring different gene knockouts (*34*). In general, specialized analytical methods are required to identify death-regulatory genes when drug-induced lethality occurs in a proliferative population, at low rates relative to proliferation, or via non-apoptotic mechanisms. However, in this case, we recognized that triptolide-induced death is: 1) exclusively apoptotic, 2) occurring rapidly (before the effects of varied proliferation can be observed in the population), and 3) activated at high rates (such that large numbers of dead cells accumulate before apoptotic corpses have decayed). Thus, we developed a simple experimental approach to mechanically separate live cells from apoptotic corpses, such that both populations could be sequenced independently.

The genome-wide CRISPR screen was performed using the Toronto KnockOut Library v3 (TKOv3), which contains 4 sgRNA per gene, along with 142 non-targeting sgRNA (*65*) . U2OS-Cas9 expressing cells were selected with 5 µg/mL blasticidin for 5 days. Following selection, cells were divided into two replicates, which would be handled separately for the remainder of the assay. Cells were then transduced with the TKOv3 viral library using a 'spinfection' method. To maintain sufficient library coverage, more than 200 million cells were transduced per replicate, at an MOI of 0.3. To do this, cells were divided into 12-well plates at a density of 2 million cells per along with 75 µL virus and 0.8 µg/mL polybrene (Millipore, TR1003G). Plates were centrifuged at 37 °C for 2 h at 830*g*. Following spinfection, virus-containing media was gently removed and replaced with fresh media, and cells were returned to the incubator overnight. The following morning, cells were moved out of the 12-well plates and onto 15 cm dishes. The following day, cells were treated with 1 µg/mL puromycin for 3 days, followed by a two-day expansion without selection. Cells were then seeded onto 15cm plates and let to adhere overnight prior to drugging. At the time of drugging, a T0 sample was taken

from the cell population and frozen, at a population size 650x that of the library size. The rest of the cells were treated with either 1 μM triptolide or DMSO. Cells were incubated for 32 hours in drug, resulting in approximately 50% cell death. Apoptotic corpses, which become non-adherent, were removed and saved. Plates were washed with PBS to dislodge any remaining dead cells, and PBS wash was added with the separated dead cell fraction. The adherent cells remaining on the plate were the live cell population. The live cells were trypsinized and collected. Each condition was collected such that a minimum of 650x library coverage was achieved per replicate.

Genomic DNA was extracted using the Wizard Genomic DNA Purification Kit (Fisher Scientific, PR-A1125). sgRNA sequences were PCR amplified out of the genomic DNA, and gel-extracted and purified. A second round of PCR added sequencing adaptors and multiplexing barcodes, and libraries were pooled and sequenced on a HiSeq 4000 at 500× coverage.

The FASTX-Toolkit (0.0.14) was used to trim reads to get just CRISPR guide sequences (parameters: -f 23 -l 42). Trimmed reads were mapped to the TKOv3 library with Bowtie (1.3.0) allowing for 2 mismatches (parameters: -v 2 -m 1). Samples were normalized for sequencing depth using median of ratios method implemented in DESeq2. Guides with low counts (approximately the bottom 2% of guides) were removed from the library. sgRNA-level $\log_2$ fold-changes were determined using DESeq2, and were then z-scored to the distribution of non-targeting sgRNA. The median sgRNA-level z-score was determined for each gene to collapse scores to the gene level. An empiric P value was determined for each gene by bootstrapping from the sgRNA-level fold change scores of non-targeting guides, and was FDR corrected using the Benjamini–Hochberg procedure. An FDR cutoff of 0.1 was used to determine genes that significantly alter the cell death rate following triptolide. Gene set enrichment using an FDR-corrected one-tailed Fishers Exact test was performed to test if any apoptotic-related pathways from the Molecular Signatures Database (MSigDB) were overrepresented among genes whose knockout significantly reduced triptolide-induced death.

Evaluation of drug TIS score

FLICK was used to determine lethal fraction kinetics in U2OS, PTBP1-KO, BCL2L12-KO, and *BAX/BAK* DKO cells across a panel of 46 drugs, across 7 half-log dilutions, and spanning the following 16 classes. Pol II degraders (Triptolide, 1 – 0.001 μM; α-amanitin, 10 – 0.01 μM). RNA synthesis inhibitors (Actinomycin-D, 1 – 0.001 μM; Ethynylcytidine, 10 – 0.01 μM; Cordycepin, 100 – 0.1 μM). Transcriptional CDK inhibitors (Flavopiridol, 10 – 0.01 μM; DRB, 100 – 0.1 μM; THZ1, 10 – 0.01 μM; THZ531, 10 – 0.01 μM; YKL-5-124, 31.6 – 0.0316 μM). Cell cycle CDK inhibitors (Abemaciclib, 10 – 0.01 μM; Palbociclib, 10 – 0.01 μM; Ribociclib, 10 – 0.01 μM; PF3600, 31.6 – 0.0316 μM; PF4091, 31.6 – 0.0316 μM). DNA crosslinking agents (4-NQO, 31.6 – 0.0316 μM; Cisplatin, 100 – 0.1 μM; Carboplatin, 316 – 0.316 μM). Topoisomerase I inhibitors (Camptothecin, 10 – 0.01 μM; Topotecan, 31.6 – 0.0316 μM). Topoisomerase II inhibitors (Idarubicin, 1 – 0.001 μM; Etoposide, 31.6 – 0.0316 μM; Teniposide, 31.6 – 0.0316 μM). PARP inhibitors (Niraparib, 100 – 0.1 μM; Rucaparib, 100 – 0.1 μM). Translational inhibitors (Cycloheximide, 100 – 0.1 μM; Homoharringtonine, 10 – 0.01 μM; Anisomycin 316 – 0.316 μM). Proteasome inhibitors (Bortezomib, 1 – 0.001 μM; MG132, 10 – 0.01 μM). ER- and proteotoxic-stress inducers (Brefeldin A, 31.6 – 0.0316 μM; Thapsigargin, 31.6 – 0.0316 μM; Tunicamycin, 10 – 0.01 μM). BH3 mimetics (ABT-199, 100 – 0.1 μM; ABT-263, 100 – 0.1 μM; ABT-737, 31.6 – 0.0316 μM). HDAC inhibitors (Panobinostat, 1 – 0.001 μM; Vorinostat, 100 – 0.1 μM). mTOR inhibitors (Torin2, 31.6 – 0.0316 μM; Rapamycin, 10 – 0.01 μM). MEK/PI3K/AKT inhibitors (Mirdametinib, 31.6 – 0.0316 μM; Buparlisib, 31 – 0.0316 μM; Capivasertib, 31.6 – 0.0316 μM). Others (SGI-1027, 31.6 – 0.0316 μM; JQ1, 100 – 0.1 μM; Staurosporine, 10 – 0.01 μM).

Lethal fraction kinetics were fit to a lag exponential death (LED) equation. Non-lethal conditions were defined as those having a maximum observed lethal fraction less than 0.16,

approximately double the death observed in untreated cells. Cells were further classified as

apoptotic or non-apoptotic by thresholding a 50% reduction in maximum lethal fraction in the

BAX/BAK DKO background compared to U2OS cells. To determine the "Transcription Inhibition

Similarity" (TIS) score for a particular drug-dose pair, the area under the LED fit (AUC) was

determined, and the baseline cell death was removed by subtracting out the AUC of untreated

cells. The difference in drug response between KO and U2OS cells was defined as 1 minus the

AUC ratio of KO over U2OS. This value, determined for PTBP1-KO and BCL2L12-KO

separately, was summed. Finally, scores were divided by that of 1 µM triptolide, such that a

score of one denotes a functional genetic signature equal to that of triptolide.

Probabilistic drug classifier

Drugs were classified as "triptolide-like" based on their dependency on PTBP1,

BCL2L12, BAX, and BAK1 for causing cell death. To perform this functional genetic

classification, we adapted an established approach used previously (*47*, *48*) . Briefly, for each

drug-dose pair, AUC ratios between each KO and U2OS were determined. This places each

drug-dose pair in a 3-dimensional functional genetic space. The "neighborhood" of

transcriptional inhibitors was calculated as the mean of pairwise Euclidean distances between

validated Pol II degraders (triptolide and α-amanitin) in this space. The linkage ratio describes

the similarity of a query compound to triptolide and α-amanitin and defines how much the

neighborhood expands or shrinks after the query compound is included in it. A set of negative

controls, defined as a set of drugs with known mechanism of action distinct from transcriptional

inhibition, is then forced to be included in the transcriptional inhibitor neighborhood, iteratively,

causing the neighborhood to expand or shrink. BH3 mimetics (ABT-199, ABT-263, ABT-737),

Topoisomerase inhibitors (Etoposide, Camptothecin, Idarubicin, Topotecan, and Teniposide),

and ER-stress inducers (Thapsigargin, Tunicamycin, and Brefeldin A) were used to generate

the false expansion. This process of "false expansion" generates an empirically defined null

distribution of linkage ratios. A linkage ratio for each drug-dose pair is then determined and

compared to the null distribution to derive a *P* value. *P* values are then FDR corrected using the

Benjamini–Hochberg procedure.

Data analysis and statistics

Unless otherwise noted, data analysis and statistics was performed in MATLAB (version

R2024b) using built-in functions. Bar plots were generated in GraphPad Prism 10. ImageStudio

3.1.4 was used to analyze western blots. Flow cytometry analysis was performed using FlowJo

10   version 10.8.1 software.

REFERENCES

1. A. Tsherniak, F. Vazquez, P. G. Montgomery, B. A. Weir, G. Kryukov, G. S. Cowley, S. Gill, W. F. Harrington, S. Pantel, J. M. Krill-Burger, R. M. Meyers, L. Ali, A. Goodale, Y. Lee, G. Jiang, J. Hsiao, W. F. J. Gerath, S. Howell, E. Merkel, M. Ghandi, L. A. Garraway, D. E. Root, T. R. Golub, J. S. Boehm, W. C. Hahn, Defining a Cancer Dependency Map. *Cell* **170**, 564-570.e16 (2017).

2. S. J. Vervoort, J. R. Devlin, N. Kwiatkowski, M. Teng, N. S. Gray, R. W. Johnstone, Targeting transcription cycles in cancer. *Nat. Rev. Cancer* **22**, 5–24 (2022).

3. D. R. Green, The Coming Decade of Cell Death Research: Five Riddles. *Cell* **177**, 1094–1107 (2019).

4. D. Tang, R. Kang, T. V. Berghe, P. Vandenabeele, G. Kroemer, The molecular machinery of regulated cell death. *Cell Res* **29**, 347–364 (2019).

5. G. Haimovich, D. A. Medina, S. Z. Causse, M. Garber, G. Millán-Zambrano, O. Barkai, S. Chávez, J. E. Pérez-Ortín, X. Darzacq, M. Choder, Gene Expression Is Circular: Factors for mRNA Degradation Also Foster mRNA Synthesis. *Cell* **153**, 1000–1011 (2013).

6. M. Sun, B. Schwalb, N. Pirkl, K. C. Maier, A. Schenk, H. Failmezger, A. Tresch, P. Cramer, Global Analysis of Eukaryotic mRNA Degradation Reveals Xrn1-Dependent Buffering of Transcript Levels. *Mol. Cell* **52**, 52–62 (2013).

7. S. Berry, M. Müller, A. Rai, L. Pelkmans, Feedback from nuclear RNA on transcription promotes robust RNA concentration homeostasis in human cells. *Cell Syst*, doi: 10.1016/j.cels.2022.04.005 (2022).

8. B. Slobodin, A. Bahat, U. Sehrawat, S. Becker-Herman, B. Zuckerman, A. N. Weiss, R. Han, R. Elkon, R. Agami, I. Ulitsky, I. Shachar, R. Dikstein, Transcription Dynamics Regulate Poly(A) Tails and Expression of the RNA Degradation Machinery to Balance mRNA Levels. *Mol. Cell* **78**, 434-444.e5 (2020).

9. E. Hartenian, B. A. Glaunsinger, Feedback to the central dogma: cytoplasmic mRNA decay and transcription are interdependent processes. *Crit. Rev. Biochem. Mol. Biol.* **54**, 385–398 (2019).

10. M. P. Swaffer, G. K. Marinov, H. Zheng, L. F. Valenzuela, C. Y. Tsui, A. W. Jones, J. Greenwood, A. Kundaje, W. J. Greenleaf, R. Reyes-Lamothe, J. M. Skotheim, RNA polymerase II dynamics and mRNA stability feedback scale mRNA amounts with cell size. *Cell* **186**, 5254-5268.e26 (2023).

11. D. V. Titov, B. Gilman, Q.-L. He, S. Bhat, W.-K. Low, Y. Dang, M. Smeaton, A. L. Demain, P. S. Miller, J. F. Kugel, J. A. Goodrich, J. O. Liu, XPB, a subunit of TFIIH, is a target of the natural product triptolide. *Nat Chem Biol* **7**, 182–188 (2011).

12. V. T. Nguyen, F. Giannoni, M.-F. Dubois, S.-J. Seo, M. Vigneron, C. Kédinger, O. Bensaude, In Vivo Degradation of RNA Polymerase II Largest Subunit Triggered by α-Amanitin. *Nucleic Acids Res.* **24**, 2924–2929 (1996).

13. G. C. Forcina, M. Conlon, A. Wells, J. Y. Cao, S. J. Dixon, Systematic Quantification of Population Cell Death Kinetics in Mammalian Cells. *Cell Syst* **4**, 1–18 (2017).

14. L. Galluzzi, I. Vitale, S. A. Aaronson, J. M. Abrams, D. Adam, P. Agostinis, E. S. Alnemri, L. Altucci, I. Amelio, D. W. Andrews, M. Annicchiarico-Petruzzelli, A. V. Antonov, E. Arama, E. H. Baehrecke, N. A. Barlev, N. G. Bazan, F. Bernassola, M. J. M. Bertrand, K. Bianchi, M. V. Blagosklonny, K. Blomgren, C. Borner, P. Boya, C. Brenner, M. Campanella, E. Candi, D.

Carmona-Gutierrez, F. Cecconi, F. K. M. Chan, N. S. Chandel, E. H. Cheng, J. E. Chipuk, J. A. Cidlowski, A. Ciechanover, G. M. Cohen, M. Conrad, J. R. Cubillos-Ruiz, P. E. Czabotar, V. D'Angiolella, T. M. Dawson, V. L. Dawson, V. D. Laurenzi, R. D. Maria, K.-M. Debatin, R. J. Deberardinis, M. Deshmukh, N. D. Daniele, F. D. Virgilio, V. M. Dixit, S. J. Dixon, C. S. Duckett, B. D. Dynlacht, W. S. El-Deiry, J. W. Elrod, G. M. Fimia, S. Fulda, A. J. García-Sáez, A. D. Garg, C. Garrido, E. Gavathiotis, P. Golstein, E. Gottlieb, D. R. Green, L. A. Greene, H. Gronemeyer, A. Gross, G. Hajnoczky, J. M. Hardwick, I. S. Harris, M. O. Hengartner, C. Hetz, H. Ichijo, M. Jäättelä, B. Joseph, P. J. Jost, P. P. Juin, W. J. Kaiser, M. Karin, T. Kaufmann, O. Kepp, A. Kimchi, R. N. Kitsis, D. J. Klionsky, R. A. Knight, S. Kumar, S. W. Lee, J. J. Lemasters, B. Levine, A. Linkermann, S. A. Lipton, R. A. Lockshin, C. López-Otín, S. W. Lowe, T. Luedde, E. Lugli, M. Macfarlane, F. Madeo, M. Malewicz, W. Malorni, G. Manic, J.-C. Marine, S. J. Martin, J.-C. Martinou, J. P. Medema, P. Mehlen, P. Meier, S. Melino, E. A. Miao, J. D. Molkentin, U. M. Moll, C. Muñoz-Pinedo, S. Nagata, G. Nuñez, A. Oberst, M. Oren, M. Overholtzer, M. Pagano, T. Panaretakis, M. Pasparakis, J. M. Penninger, D. M. Pereira, S. Pervaiz, M. E. Peter, M. Piacentini, P. Pinton, J. H. M. Prehn, H. Puthalakath, G. A. Rabinovich, M. Rehm, R. Rizzuto, C. M. P. Rodrigues, D. C. Rubinsztein, T. Rudel, K. M. Ryan, E. Sayan, L. Scorrano, F. Shao, Y. Shi, J. Silke, H.-U. Simon, A. Sistigu, B. R. Stockwell, A. Strasser, G. Szabadkai, S. W. G. Tait, D. Tang, N. Tavernarakis, A. Thorburn, Y. Tsujimoto, B. Turk, T. V. Berghe, P. Vandenabeele, M. G. V. Heiden, A. Villunger, H. W. Virgin, K. H. Vousden, D. Vucic, E. F. Wagner, H. Walczak, D. Wallach, Y. Wang, J. A. Wells, W. Wood, J. Yuan, Z. Zakeri, B. Zhivotovsky, L. Zitvogel, G. Melino, G. Kroemer, Molecular mechanisms of cell death: recommendations of the Nomenclature Committee on Cell Death 2018. *Cell Death Differ* **25**, 1–56 (2018).

15. C. Holze, C. Michaudel, C. Mackowiak, D. A. Haas, C. Benda, P. Hubel, F. L. Pennemann, D. Schnepf, J. Wettmarshausen, M. Braun, D. W. Leung, G. K. Amarasinghe, F. Perocchi, P.

Staeheli, B. Ryffel, A. Pichlmair, Oxeiptosis, a ROS-induced caspase-independent apoptosis-like cell-death pathway. *Nat Immunol* **19**, 130–140 (2018).

16. P. Tsvetkov, S. Coy, B. Petrova, M. Dreishpoon, A. Verma, M. Abdusamad, J. Rossen, L. Joesch-Cohen, R. Humeidi, R. D. Spangler, J. K. Eaton, E. Frenkel, M. Kocak, S. M. Corsello, S. Lutsenko, N. Kanarek, S. Santagata, T. R. Golub, Copper induces cell death by targeting lipoylated TCA cycle proteins. *Science* **375**, 1254–1261 (2022).

17. M. C. Wei, W. X. Zong, E. H. Cheng, T. Lindsten, V. Panoutsakopoulou, A. J. Ross, K. A. Roth, G. R. MacGregor, C. B. Thompson, S. J. Korsmeyer, Proapoptotic BAX and BAK: a requisite gateway to mitochondrial dysfunction and death. *Science* **292**, 727–730 (2001).

18. T. Lindsten, A. J. Ross, A. King, W.-X. Zong, J. C. Rathmell, H. A. Shiels, E. Ulrich, K. G. Waymire, P. Mahar, K. Frauwirth, Y. Chen, M. Wei, V. M. Eng, D. M. Adelman, M. C. Simon, A. Ma, J. A. Golden, G. Evan, S. J. Korsmeyer, G. R. MacGregor, C. B. Thompson, The Combined Functions of Proapoptotic Bcl-2 Family Members Bak and Bax Are Essential for Normal Development of Multiple Tissues. *Mol. Cell* **6**, 1389–1399 (2000).

19. A.-L. Valton, S. V. Venev, B. Mair, E. S. Khokhar, A. H. Y. Tong, M. Usaj, K. Chan, A. A. Pai, J. Moffat, J. Dekker, A cohesin traffic pattern genetically linked to gene regulation. *Nat. Struct. Mol. Biol.* **29**, 1239–1251 (2022).

20. J. Cidado, S. Boiko, T. Proia, D. Ferguson, S. W. Criscione, M. S. Martin, P. Pop-Damkov, N. Su, V. N. R. Franklin, C. S. R. Chilamakuri, C. S. D'Santos, W. Shao, J. C. Saeh, R. Koch, D. M. Weinstock, M. Zinda, S. E. Fawell, L. Drew, AZD4573 Is a Highly Selective CDK9 Inhibitor That Suppresses MCL-1 and Induces Apoptosis in Hematologic Cancer Cells. *Clin. Cancer Res.* **26**, 922–934 (2020).

21. M. Sancho, D. Leiva, E. Lucendo, M. Orzáez, Understanding MCL1: from cellular function and regulation to pharmacological inhibition. *FEBS J.* **289**, 6209–6234 (2022).

22. H. Widden, W. J. Placzek, The multiple mechanisms of MCL1 in the regulation of cell fate. *Commun. Biol.* **4**, 1029 (2021).

23. A. Kotschy, Z. Szlavik, J. Murray, J. Davidson, A. L. Maragno, G. L. Toumelin-Braizat, M. Chanrion, G. L. Kelly, J.-N. Gong, D. M. Moujalled, A. Bruno, M. Csekei, A. Paczal, Z. B. Szabo, S. Sipos, G. Radics, A. Proszenyak, B. Balint, L. Ondi, G. Blasko, A. Robertson, A. Surgenor, P. Dokurno, I. Chen, N. Matassova, J. Smith, C. Pedder, C. Graham, A. Studeny, G. Lysiak-Auvity, A.-M. Girard, F. Gravé, D. Segal, C. D. Riffkin, G. Pomilio, L. C. A. Galbraith, B. J. Aubrey, M. S. Brennan, M. J. Herold, C. Chang, G. Guasconi, N. Cauquil, F. Melchiore, N. Guigal-Stephan, B. Lockhart, F. Colland, J. A. Hickman, A. W. Roberts, D. C. S. Huang, A. H. Wei, A. Strasser, G. Lessene, O. Geneste, The MCL1 inhibitor S63845 is tolerable and effective in diverse cancer models. *Nature* **538**, 477–482 (2016).

24. O. Bensaude, Inhibiting eukaryotic transcription. Which compound to choose? How to evaluate its activity? *Biochem Soc Symp* **2**, 103–108 (2014).

25. J. Zhang, J. L. Corden, Phosphorylation causes a conformational change in the carboxyl-terminal domain of the mouse RNA polymerase II largest subunit. *J. Biol. Chem.* **266**, 2297–2302 (1991).

26. D. L. Cadena, M. E. Dahmus, Messenger RNA synthesis in mammalian cells is catalyzed by the phosphorylated form of RNA polymerase II. *J. Biol. Chem.* **262**, 12468–12474 (1987).

27. B. Steurer, R. C. Janssens, B. Geverts, M. E. Geijer, F. Wienholz, A. F. Theil, J. Chang, S. Dealy, J. Pothof, W. A. van Cappellen, A. B. Houtsmuller, J. A. Marteijn, Live-cell analysis of

endogenous GFP-RPB1 uncovers rapid turnover of initiating and promoter-paused RNA Polymerase II. *Proc National Acad Sci* **115**, E4368–E4376 (2018).

28. E. Y. Jacobs, I. Ogiwara, A. M. Weiner, Role of the C-Terminal Domain of RNA Polymerase II in U2 snRNA Transcription and 3′ Processing. *Mol. Cell. Biol.* **24**, 846–855 (2004).

29. M. E. Dahmus, Reversible Phosphorylation of the C-terminal Domain of RNA Polymerase II*. *J. Biol. Chem.* **271**, 19009–19012 (1996).

30. S. McCracken, N. Fong, K. Yankulov, S. Ballantyne, G. Pan, J. Greenblatt, S. D. Patterson, M. Wickens, D. L. Bentley, The C-terminal domain of RNA polymerase II couples mRNA processing to transcription. *Nature* **385**, 357–361 (1997).

31. S. Egloff, D. O'Reilly, R. D. Chapman, A. Taylor, K. Tanzhaus, L. Pitts, D. Eick, S. Murphy, Serine-7 of the RNA Polymerase II CTD Is Specifically Required for snRNA Gene Expression. *Science* **318**, 1777–1779 (2007).

32. B. Mair, J. Tomic, S. N. Masud, P. Tonge, A. Weiss, M. Usaj, A. H. Y. Tong, J. J. Kwan, K. R. Brown, E. Titus, M. Atkins, K. S. K. Chan, L. Munsie, A. Habsid, H. Han, M. Kennedy, B. Cohen, G. Keller, J. Moffat, Essential Gene Profiles for Human Pluripotent Stem Cells Identify Uncharacterized Genes and Substrate Dependencies. *Cell Rep.* **27**, 599-615.e12 (2019).

33. T. Hart, A. H. Y. Tong, K. Chan, J. V. Leeuwen, A. Seetharaman, M. Aregger, M. Chandrashekhar, N. Hustedt, S. Seth, A. Noonan, A. Habsid, O. Sizova, L. Nedyalkova, R. Climie, L. Tworzyanski, K. Lawson, M. A. Sartori, S. Alibeh, D. Tieu, S. Masud, P. Mero, A. Weiss, K. R. Brown, M. Usaj, M. Billmann, M. Rahman, M. Costanzo, C. L. Myers, B. J. Andrews, C. Boone, D. Durocher, J. Moffat, Evaluation and Design of Genome-Wide CRISPR/SpCas9 Knockout Screens. *G3 Genes Genomes Genetics* **7**, 2719–2727 (2017).

34. M. E. Honeywell, M. S. Isidor, N. W. Harper, R. E. Fontana, G. A. Birdsall, P. Cruz-Gordillo, S. A. Porto, M. Jerome, C. S. Fraser, K. A. Sarosiek, D. A. Guertin, J. B. Spinelli, M. J. Lee, Functional genomic screens with death rate analyses reveal mechanisms of drug action. *Nat. Chem. Biol.*, 1–10 (2024).

35. J. Kale, E. J. Osterlund, D. W. Andrews, BCL-2 family proteins: changing partners in the dance towards death. *Cell Death Differ.* **25**, 65–80 (2018).

36. C. Iannone, Y. Kainov, A. Zhuravskaya, F. Hamid, T. Nojima, E. V. Makeyev, PTBP1-activated co-transcriptional splicing controls epigenetic status of pluripotent stem cells. *Mol. Cell* **83**, 203-218.e9 (2023).

37. W. Zhu, B. Zhou, L. Rong, L. Ye, H. Xu, Y. Zhou, X. Yan, W. Liu, B. Zhu, L. Wang, X. Jiang, C. Ren, Roles of PTBP1 in alternative splicing, glycolysis, and oncogensis. *J Zhejiang Univ-sc B* **21**, 122–136 (2020).

38. K. Sawicka, M. Bushell, K. A. Spriggs, A. E. Willis, Polypyrimidine-tract-binding protein: a multifunctional RNA-binding protein. *Biochem Soc T* **36**, 641–647 (2008).

39. M. Bushell, M. Stoneley, Y. W. Kong, T. L. Hamilton, K. A. Spriggs, H. C. Dobbyn, X. Qin, P. Sarnow, A. E. Willis, Polypyrimidine Tract Binding Protein Regulates IRES-Mediated Gene Expression during Apoptosis. *Mol. Cell* **23**, 401–412 (2006).

40. H. A. King, L. C. Cobbold, X. Pichon, T. Pöyry, L. A. Wilson, H. Booden, R. Jukes-Jones, K. Cain, K. S. Lilley, M. Bushell, A. E. Willis, Remodelling of a polypyrimidine tract-binding protein complex during apoptosis activates cellular IRESs. *Cell Death Differ* **21**, 161–171 (2014).

41. P. Kafasla, I. Mickleburgh, M. Llorian, M. Coelho, C. Gooding, D. Cherny, A. Joshi, O. Kotik-Kogan, S. Curry, I. C. Eperon, R. J. Jackson, C. W. J. Smith, Defining the roles and interactions of PTB. *Biochem. Soc. Trans.* **40**, 815–820 (2012).

42. Y. Xue, Y. Zhou, T. Wu, T. Zhu, X. Ji, Y.-S. Kwon, C. Zhang, G. Yeo, D. L. Black, H. Sun, X.-D. Fu, Y. Zhang, Genome-wide Analysis of PTB-RNA Interactions Reveals a Strategy Used by the General Splicing Repressor to Modulate Exon Inclusion or Skipping. *Mol. Cell* **36**, 996–1006 (2009).

43. N. Keppetipola, S. Sharma, Q. Li, D. L. Black, Neuronal regulation of pre-mRNA splicing by polypyrimidine tract binding proteins, PTBP1 and PTBP2. *Crit. Rev. Biochem. Mol. Biol.* **47**, 360–378 (2012).

44. M. Llorian, S. Schwartz, T. A. Clark, D. Hollander, L.-Y. Tan, R. Spellman, A. Gordon, A. C. Schweitzer, P. de la Grange, G. Ast, C. W. J. Smith, Position-dependent alternative splicing activity revealed by global profiling of alternative splicing events regulated by PTB. *Nat. Struct. Mol. Biol.* **17**, 1114–1123 (2010).

45. S. Piñol-Roma, G. Dreyfuss, Shuttling of pre-mRNA binding proteins between nucleus and cytoplasm. *Nature* **355**, 730–732 (1992).

46. S. Piñol-Roma, G. Dreyfuss, Transcription-Dependent and Transcription-Independent Nuclear Transport of hnRNP Proteins. *Science* **253**, 312–314 (1991).

47. J. R. Pritchard, P. M. Bruno, M. T. Hemann, D. A. Lauffenburger, Predicting cancer drug mechanisms of action using molecular network signatures. *Mol. Biosyst.* **9**, 1604–1619 (2012).

48. H. Jiang, J. R. Pritchard, R. T. Williams, D. A. Lauffenburger, M. T. Hemann, A mammalian functional-genetic approach to characterizing cancer therapeutics. *Nat Chem Biol* **7**, 92–100 (2011).

49. S. Dasari, P. B. Tchounwou, Cisplatin in cancer therapy: Molecular mechanisms of action. *Eur. J. Pharmacol.* **740**, 364–378 (2014).

50. M. Hafner, M. Niepel, M. Chung, P. K. Sorger, Growth rate inhibition metrics correct for confounders in measuring sensitivity to cancer drugs. *Nat Methods* **13**, 1–11 (2016).

51. H. R. Schwartz, R. Richards, R. E. Fontana, A. J. Joyce, M. E. Honeywell, M. J. Lee, Drug GRADE: An Integrated Analysis of Population Growth and Cell Death Reveals Drug-Specific and Cancer Subtype-Specific Response Profiles. *Cell Reports* **31**, 107800 (2020).

52. A. H. Stegh, H. Kim, R. M. Bachoo, K. L. Forloney, J. Zhang, H. Schulze, K. Park, G. J. Hannon, J. Yuan, D. N. Louis, R. A. DePinho, L. Chin, Bcl2L12 inhibits post-mitochondrial apoptosis signaling in glioblastoma. *Genes Dev.* **21**, 98–111 (2007).

53. J. E. Bradner, D. Hnisz, R. A. Young, Transcriptional Addiction in Cancer. *Cell* **168**, 629–643 (2017).

54. R. Richards, H. R. Schwartz, M. E. Honeywell, M. S. Stewart, P. Cruz-Gordillo, A. J. Joyce, B. D. Landry, M. J. Lee, Drug antagonism and single-agent dominance result from differences in death kinetics. *Nat Chem Biol*, 1–10 (2020).

55. E. ROSONINA, B. J. BLENCOWE, Analysis of the requirement for RNA polymerase II CTD heptapeptide repeats in pre-mRNA splicing and 3′-end cleavage. *RNA* **10**, 581–589 (2004).

56. S. J. Vervoort, S. A. Welsh, J. R. Devlin, E. Barbieri, D. A. Knight, S. Offley, S. Bjelosevic, M. Costacurta, I. Todorovski, C. J. Kearney, J. J. Sandow, Z. Fan, B. Blyth, V. McLeod, J. H. A.

Vissers, K. Pavic, B. P. Martin, G. Gregory, E. Demosthenous, M. Zethoven, I. Y. Kong, E. D. Hawkins, S. J. Hogg, M. J. Kelly, A. Newbold, K. J. Simpson, O. Kauko, K. F. Harvey, M. Ohlmeyer, J. Westermarck, N. Gray, A. Gardini, R. W. Johnstone, The PP2A-Integrator-CDK9 axis fine-tunes transcription and can be targeted therapeutically in cancer. *Cell* **184**, 3143-3162.e32 (2021).

57. V. Agarwal, D. R. Kelley, The genetic and biochemical determinants of mRNA degradation rates in mammals. *Genome Biol.* **23**, 245 (2022).

58. N. L. Bray, H. Pimentel, P. Melsted, L. Pachter, Near-optimal probabilistic RNA-seq quantification. *Nat. Biotechnol.* **34**, 525–527 (2016).

59. C. Soneson, M. I. Love, M. D. Robinson, Differential analyses for RNA-seq: transcript-level estimates improve gene-level inferences. *F1000Research* **4**, 1521 (2016).

60. M. I. Love, W. Huber, S. Anders, Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. *Genome Biol.* **15**, 550 (2014).

61. M. Stephens, False discovery rates: a new deal. *Biostatistics* **18**, 275–294 (2017).

62. M. Irimia, R. J. Weatheritt, J. D. Ellis, N. N. Parikshak, T. Gonatopoulos-Pournatzis, M. Babor, M. Quesnel-Vallières, J. Tapial, B. Raj, D. O'Hanlon, M. Barrios-Rodiles, M. J. E. Sternberg, S. P. Cordes, F. P. Roth, J. L. Wrana, D. H. Geschwind, B. J. Blencowe, A Highly Conserved Program of Neuronal Microexons Is Misregulated in Autistic Brains. *Cell* **159**, 1511–1523 (2014).

63. J. Tapial, K. C. H. Ha, T. Sterne-Weiler, A. Gohr, U. Braunschweig, A. Hermoso-Pulido, M. Quesnel-Vallières, J. Permanyer, R. Sodaei, Y. Marquez, L. Cozzuto, X. Wang, M. Gómez-Velázquez, T. Rayon, M. Manzanares, J. Ponomarenko, B. J. Blencowe, M. Irimia, An atlas of

alternative splicing profiles and functional associations reveals new regulatory programs and genes that simultaneously express multiple major isoforms. *Genome Res.* **27**, 1759–1768 (2017).

64. R. Richards, M. E. Honeywell, M. J. Lee, FLICK: an optimized plate reader-based assay to infer cell death kinetics. *Star Protoc* **2**, 100327 (2021).

65. T. Hart, M. Chandrashekhar, M. Aregger, Z. Steinhart, K. R. Brown, G. MacLeod, M. Mis, M. Zimmermann, A. Fradet-Turcotte, S. Sun, P. Mero, P. Dirks, S. Sidhu, F. P. Roth, O. S. Rissland, D. Durocher, S. Angers, J. Moffat, High-Resolution CRISPR Screens Reveal Fitness Genes and Genotype-Specific Cancer Liabilities. *Cell* **163**, 1515–1526 (2015).

Fig. 1: Transcriptional inhibition initiates apoptosis using an active signaling process that is unrelated to the lethality caused by passive loss of mRNA. (**A**) Immunoblot of total Rpb1 protein in U2OS cells following exposure to 1 µM or 0.1 µM Triptolide. Blots are representative of three independent biological replicates. (**B**) Quantification of Rpb1 levels shown in panel (A). (**C-D**) Live and dead cell kinetics following exposure to 1 µM triptolide for Nuc::mKate2-expressing U2OS cells. Data collected using the STACK assay. mKate+ cells are live, and Sytox+ cells are dead. (C) U2OS. (D) U2OS$^{BAX-/-/BAK1-/-}$ cells (BAX/BAK DKO). Representative images for each genotype are shown above the quantification at the indicated timepoints. Scale bar, 100 µm. (**E**) Heatmap of fractional viability (FV) for cells treated with the indicated drugs in the presence/absence of death pathway specific inhibitors. z-VAD-FMK (z-VAD) targets apoptotic caspases; VX765 inhibits the pyroptotic initiator, caspase-1; TTM inhibits cuproptosis (cupro); Rucaparib (Rucap.) inhibits parthanatos (parth); Nec-1 inhibits necroptosis (necro); Fer-1 inhibits ferroptosis (ferro). RSL3, Elesclomol (Eles.), MNNG, and TSZ are canonical activators of the listed death pathways. M-157 = MDA-MB-157. (**F-G**) STACK-based analysis of live and dead cells over extended times following exposure to 1 µM triptolide. (**F**) Representative images from three biological replicates. Scale bar, 275 µm. (**G**) Lethal fraction measured using FLICK following imaging. Data are mean ± SD of 3 biological replicates. (**H**) mRNA levels following exposure to 1 µM triptolide, relative to untreated cells. Data were normalized to absolute mRNA abundance using polyadenylated ERCC spike-ins. mRNA levels were quantified only for times in which substantial lethality had not yet occurred, which was 8-hours for WT cells, and the first 10 days for DKO cells. Average mRNA abundance shown for all mRNAs. Decay kinetics for DKO were fit to a two-term exponential decay function. Range for DKO represents the 50 longest and 50 shortest half-life mRNAs. Dashed lines denote extrapolated expectation from fits. (**I**) (top) Immunoblot of total Rpb1 protein levels in HAP1 cells, or HAP1-RPB1-AID cells following exposure to 500 µg/mL auxin (IAA, 3-indoleacetic acid). (bottom) Lethal Fraction measurements for HAP1-RPB1-AID cells following 500 µg/mL IAA, in the presence or absence of the 50 µM z-VAD. Data collected using the FLICK assay. For all panels with error bars, data are mean ± SD, n ≥ 3 independent biological replicates.

Fig. 2: Loss of Pol II protein – not loss of mRNA – activates apoptosis. (**A**) Schematic describing three approaches (i – iii) for uncoupling inhibition of Pol II enzymatic activity from degradation of Pol II protein. (**B**) Histogram of spike-in normalized absolute mRNA fold changes following 8-hr exposure to 1 µM triptolide (TRP) in U2OS cells expressing sgRNA targeting NUP93 or nontargeting sgRNA. Two-sided KS test p value is shown. (**C**) Comparison of lethal fraction (LF) levels between cell types in (B) at various time points following exposure to 1 µM TRP, measured using the FLICK assay. Data normalized to max LF in cells expressing nontargeting sgRNA. Data are mean ± SD for n = 3 independent biological replicates. Wilcoxon rank sum p value shown (n.s. > 0.05). (**D**) Immunoblot of total Rpb1 levels in U2OS cells following exposure to 1 µM TRP (top), 0.316 µM THZ1 (middle), or the combination of the two (bottom). Blots shown are representative of three independent biological replicates. (**E-F**) Quantification of Rpb1 levels after 4-hr drug exposure (associated with panel D). (**E**) inactive (II-a). (**F**) active (II-o). (**G**) LF at 24-hrs, measured using the FLICK assay in identical conditions. Data are mean ± SD for n = 6 independent biological replicates. (**H-L**) Pol II switchover system used to evaluate function of inactive Pol II. (**H**) Diagram of a Pol II construct featuring a doxycycline-inducible promoter, FLAG tag, α-amanitin resistance mutation (N792D), and deletion of amino acids 1593-1970 (ΔCTD). (**I**) Experimental logic behind the Pol II switchover system. Tet-ON U2OS cells expressing the N792D/ΔCTD construct are exposed to 10 µM α-amanitin in the presence or absence of 2 µg/mL doxycycline (Dox). Dox added 48-hrs prior to α-amanitin to allow for expression of N792D/ΔCTD prior to degradation of the endogenous Pol II-o/a. (**J**) Immunoblots of Rpb1 following exposure to 10 µM α-amanitin. (**K**) Histogram of spike-in normalized absolute mRNA fold changes following 12-hr exposure to 10 µM α-amanitin. (**L**) LF kinetics following following 10 µM α-amanitin. LF measured using FLICK. Mean ± SD shown, n = 5 independent biological replicates.

Fig. 3: Apoptosis following Pol II degradation exhibits unique genetic dependencies. (**A**) Schematic of chemo-genetic profiling to identify cell death regulatory genes. (**B**) Gene-level chemo-genetic profiling data for U2OS cells treated with 1 μM triptolide. Key death regulatory genes and non-targeting controls are highlighted. (**C**) Simplified schematic of the cell intrinsic apoptotic pathway. Bcl-2* denotes the general anti-apoptotic Bcl-2 family members, whereas BH3* denotes the pro-apoptotic BH3-only family members. Regulators in blue/yellow were identified in the chemo-genetic profiling data. (**D**) Enrichment of various stress response pathways in genes whose knockout significantly suppresses triptolide-induced cell death, as identified in the chemo-genetic profiling. Significance was determined using a one-tailed Fisher's exact test corrected for multiple comparisons using the Benjamini-Hochberg (BH) FDR. (**E**) Cell death kinetics measured using FLICK following 1μM triptolide in U2OS wild-type cells, PTBP1-KO cells (red, square), BCL2L12-KO cells (red, diamond). Data are mean ± SD for n = 6 independent biological replicates. (**F**) PTBP1 and BCL2L12 dependence for drugs that degrade Pol II, or canonical apoptotic agents: triptolide (TRP): 1 μM; α-amanitin (α-ama.): 10 μM; ABT-199 (ABT): 100 μM; Staurosporine (STS): 0.5 μM. Drugs tested in U2OS-PTBP1-KO, U2OS-BCL2L12-KO, or U2OS BAX/BAK double knockout (DKO). Lethal fraction kinetics shown with area under the kinetic curve colored: grey = wild-type cells, red = DKO, PTBP1-KO, or BCL2L12-KO cells, as indicated. Areas are based on the mean of 6 independent biological replicates. (**G**) Drug-induced death rates from chemo-genetic profiling of triptolide, with genes highlighted that regulate PTBP1 translocation: nuclear import factors in yellow; nuclear export factors in blue. Non-targeting genes in black. (**H-I**) Nuclear/cytoplasmic fractionation to assess PTBP1 localization following exposure to α-ama. W = whole cell extract; C = cytoplasmic extract; N = nuclear extract. Immunoblots for Rbp1 shown to validate α-ama. function; α-Tubulin and Histone-H3 (H-H3) shown as markers of cytoplasmic and nuclear proteins, respectively. (**H**) Data are from U2OS-RBP1-N792D/ΔCTD cells exposed to 10 μM α-amanitin for 12 hours. (**I**) As in panel H, except samples collected from U2OS-RBP1-N792D/ΔCTD cells exposed to doxycycline (Dox) for 48 hours prior to exposure to α-amanitin.

Fig. 4: Commonly used anti-cancer drugs owe their lethality to Pol II degradation. (**A**) Schematic illustrating the strategy to quantify the degree to which the lethality of a given drug is dependent on the genetic dependencies observed for Pol II degraders (*i.e.*, PTBP1 and BCL2L12). (**B**) Cell death kinetics in U2OS cells for a high scoring drug (10 µM α-amanitin, TIS = 1.0) and a low scoring drug (3.16 µM Staurosporine, TIS = -0.49). Data collected using FLICK. Data are mean ± SD for n = 3 independent biological replicates. (**C**) Heatmap depicting the Transcription Inhibition Similarity (TIS) score across a 7-point half-log dose range for 46 compounds encompassing diverse established mechanisms of action and clinical utility. Data collected in U2OS cells. Grey boxes are for non-lethal doses of each drug. Compounds are ordered and colored by drug class. See methods for exact doses tested and drug class definitions. (**D**) Binary classifier, defining each drug-dose condition as "Transcriptional Inhibition-like" (TI-like) or not. Classifier results are ordered as in (**C**). Red circles denote TI-like mechanism of lethality. (**E**) Immunoblot of total Rpb1 levels in U2OS cells following exposure to high dose Abemaciclib (classified as low TIS), or a ten-fold lower dose (classified as high TIS and TI-like mechanism of lethality). (**F**) Immunoblot of total Rpb1 levels in U2OS cells following exposure to high dose Cisplatin. (**G**) Relationship between timing of cell death and timing of Pol II-a loss. Data in gray denote established transcriptional inhibitors, as shown in S10F. Red and blue data points represent tested TI-like compounds and not TI-like compounds, respectively. Black dashed line denotes a direct x = y relationship, shifted by a constant to account for a lag time between drug activity and cell death.