pSILAC mass spectrometry reveals ZFP91 as IMiD-dependent substrate of the CRL4\(^{\text{CRBN}}\) ubiquitin ligase

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Thalidomide and its derivatives lenalidomide and pomalidomide (IMiDs) are effective treatments of haematologic malignancies. It was shown that IMiDs impart gain-of-function properties to the CUL4-RBX1-DDB1-CRBN (CRL4\(^{\text{CRBN}}\)) ubiquitin ligase that enable binding, ubiquitination and degradation of key therapeutic targets such as IKZF1, IKZF3 and CSNK1A1. While these substrates have been implicated as efficacy targets in multiple myeloma (MM) and 5q deletion associated myelodysplastic syndrome (del(5q)-MDS), other targets likely exist. Using a pulse-chase SILAC mass spectrometry-based proteomics approach, we demonstrate that lenalidomide induces the ubiquitination and degradation of ZFP91. We establish ZFP91 as a \textit{bona fide} IMiD-dependent CRL4\(^{\text{CRBN}}\) substrate and further show that ZFP91 harbours a zinc finger (ZnF) motif, related to the IKZF1/3 ZnF, critical for IMiD-dependent CRBN binding. These findings demonstrate that single time point pulse-chase SILAC mass spectrometry-based proteomics (pSILAC MS) is a sensitive approach for target identification of small molecules inducing selective protein degradation.
The ubiquitin-proteasome system (UPS) constantly remodels the proteome to regulate a plethora of cellular processes. In the UPS, ubiquitin is covalently coupled to a substrate lysine by activity of an E1 (ubiquitin activating enzyme), E2 (ubiquitin conjugating enzyme) and E3 (ubiquitin ligase) enzyme cascade. Specificity in the UPS is largely conferred by E3 ligases, of which there are more than 600 in the human genome.

Thalidomide and its derivatives lenalidomide and pomalidomide (collectively known as IMiDs for immune-modulatory drugs), while initially marketed as sedatives found to cause severe teratogenic side effects, are widely used to treat haematologic malignancies, including multiple myeloma (MM) and 5q deletion associated myelodysplastic syndrome (del(5q) MDS). Lenalidomide binds to the ubiquitin ligase CRL4<sub>CRBN</sub> and exhibits dual activity by inhibiting CRL4<sub>CRBN</sub> from ubiquitinating endogenous substrates such as MEIS2 (refs 9, 10), while simultaneously promoting the CRL4<sub>CRBN</sub>-dependent ubiquitination and degradation of IKZF1 (Ikaros), IKZF3 (Aiolos) and Casein kinase 1 alpha (CSNK1A1). Recent reports have further demonstrated that lenalidomide disrupts the ubiquitin-independent regulation of CD147 and MCT1 by CRBN and that glutamine synthetase (GS) is ubiquitinated by CRL4<sub>CRBN</sub> in an IMiD-independent fashion. However, the pleiotropic effects of IMiDs and the largely unexplained adverse effects such as the profound teratogenicity and neurotoxicity suggest that other substrates likely exist. To identify novel CRL4<sub>CRBN</sub> substrates regulated by lenalidomide, we performed proteomics studies in non-hematopoietic cell lines to avoid masking of results with the dominant presence of IKZF1/3 in hematopoietic tissues.

Mass spectrometry-based methods have become standard for the identification of E3 ubiquitin ligase substrates and were successfully applied to various classes of ligases, such as the Cullin RING ligase family SCF/CRL1 (refs 17–19). However, other ligase families have proven recalcitrant to commonly used methods such as affinity-purification mass spectrometry, di-Gly proteomics or whole proteome relative quantification. Common complications for the identification of ubiquitin ligase substrates are secondary effects, such as transcriptional response or changes to translation. For profiling of protein stability-directed drugs, such as lenalidomide, most secondary effects can be reduced through relatively short drug treatments (<24 h) compared to the typical treatment times for RNAi or genetic inactivation (typically >48 h). However, faced with the challenge of relatively small changes to total protein levels, we reasoned that an approach directed at measuring changes to protein turnover in response to drug treatment would provide increased sensitivity compared to relative quantification of total protein abundance. Such a sensitive method would also be valuable to a growing field of drug development efforts directed at protein degradation.

Pulse-chase experiments, performed with 35S isotope labelling or through blockade of translation by treatment with cycloheximide (CHX), are considered the gold standard method to investigate protein stability. Modern mass spectrometry combined with stable isotope labelling (SILAC) can principally be used to perform pulse-chase experiments on a proteome-wide scale, and this has been applied to interrogate global protein synthesis and decay rates. However, such experiments typically involve multiple time points.

Here we show that direct measurement of changes in protein stability in a single time point mass spectrometry experiment is a sensitive and robust alternative to identify targets of lenalidomide. The method is based on temporally controlled incorporation of heavy amino acids by changing the growth medium (pSILAC) simultaneously with applying the desired perturbation, such as treatment with a drug. Samples are subjected to shotgun proteomics utilizing liquid chromatography coupled tandem mass spectrometry (LC-MS/MS). The pre-existing (light amino acid containing) and newly synthesized (heavy amino acid containing) protein species can be differentially quantified based on their characteristic mass difference to derive heavy to light (H/L) protein ratios. In such an experimental design, and assuming constant incorporation rate, logarithmic H/L ratios should increase linearly over time. H/L protein ratios represent a quantitative measure of protein stability and differences between treated and control cells can be used to identify drug-induced changes to protein turnover, such as induced degradation of target proteins. We demonstrate that a single time point pSILAC experiment is sufficient to identify CSNK1A1 along with new substrate candidates. We further show that the zinc finger (ZnF) protein, ZFP91, a putative ubiquitin ligase, is a bona fide lenalidomide-dependent CRL4<sub>CRBN</sub> substrate.

Results

Identification of novel CRL4<sub>CRBN</sub> targets through pSILAC MS.

We sought to explore a single time point pSILAC approach, which would increase the depth of proteome profiling and significantly reduce the required machine time compared to multi time point experiments. To understand the scopes and limitations of this approach, we performed a direct comparison with multi time point pSILAC and quantified the changes to total protein abundance using tandem mass tags (TMT).

To generate a reference data set for direct comparison, we first designed a multi time point pSILAC experiment. We used metabolic pulse labelling with heavy amino acids to measure protein turnover in growing HEK293T cells treated with 30 μM lenalidomide or dimethyl sulfoxide (DMSO) control (Fig. 1a). The high concentration of lenalidomide was chosen to achieve maximum response in a short time window and thereby aid substrate identification. HEK293T cells were chosen to bias the experiment towards targets beyond the well-studied transcription factors IKZF1 and IKZF3 that are found explicitly in hematopoietic lineages. Cells were collected in duplicates at three time points, 6 h (T6), 10 h (T10) and 16 h (T16) post treatment with lenalidomide or DMSO and parallel metabolic labelling. To prevent arginine to proline conversion, a pervasive problem in SILAC, the cell culture medium was supplemented with excess unlabelled proline. Label swap experiments at the T6 and T10 time points were performed and analysed to exclude the possibility of SILAC label artefacts (growth medium changing from H/L isotope label on treatment, see also Supplementary Figs 1a–c and 2a–c). While we observe a small systematic offset in H/L protein ratios comparing forward to reverse experiments (most pronounced at early time points), the overall correlation between forward and reverse experiments and random spot checks suggest that label artefacts can be neglected for further analysis. Whole-cell lysates were pre-fractionated into 15 fractions by SDS–PAGE and tryptic peptides subjected to LC-MS/MS analysis on an Orbitrap VELOS platform (see Methods section). Mass spectrometry data were analysed with the MaxQuant software. We identified 100,763 peptide sequences, which were assigned to 6,328 unique protein groups (false discovery rate <1% on peptide and protein level; requiring min one unique peptide quantified for a protein to be included).

Polypeptides synthesized after drug treatment and metabolic labelling will predominantly incorporate heavy amino acids, thus the H/L protein ratios can serve as a direct readout of protein turnover (Fig. 1b–d). H/L protein ratios were quantile

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Figure 1 | Multi time point pSILAC mass spectrometry. (a) Outline of the multi time point pulse-SILAC mass spectrometry experimental design. For simplification, replicate and reverse experiments are not depicted in this figure. (b) Scatter plot depicting the change of H/L protein ratios over time. Protein ratios of the 10 and 16 h time points compared to the 6 h time point. Data in this figure are presented as means of biological replicates for T6 and T16 (n = 2) or individual data points for T10 (n = 1). (c) Frequencies of H/L protein ratios at different time points (data as in b). (d) Plots depicting the logarithmic H/L protein ratios over time are shown for the two validated targets CSNK1A1 and ZFP91 as well as for control proteins GAPDH, UBA1, DDB1 and CopS5. Differential turnover for CSNK1A1 and ZFP91 is observed, while stable conditions are found for controls. Data in this figure are presented as individual data points and separate r² values for the linear regression are provided for lenalidomide and DMSO samples. (e) Scatter plot comparing protein half-lives of lenalidomide treated and control samples. Protein half-lives were obtained by fitting the H/L protein ratios of T6 (n = 2), T10 reverse (n = 1) and T16 (n = 2) time points to a decay function. For further analysis, we retained only proteins quantified in all samples and with r² > 0.9 for both linear regression fits (DMSO and lenalidomide treated), resulting in a total of 2,759 proteins. CSNK1A1 was found to exhibit a reduced half-life in presence of lenalidomide. Blue and red dotted lines indicate ± 3 and ± 5 s.d., respectively. ZFP91 was dropped from the data analysis for missing values in two replicate measurements. The depicted delta-half-life for ZFP91 is calculated using imputed values from replicates to replace the missing values. (f) Scatter plot depicting the negative correlation between differences in protein half-life (data as in e), and differences in log₂ H/L protein ratios at the T6 time point (LENA-DMSO, mean of two biological replicates). R corresponds to the Pearson’s correlation coefficient. (g) Fold change in H/L ratios comparing lenalidomide to DMSO control treatment on the x axis (HEK293 T6 SILAC samples). Moderated t-test P values were calculated using the limma package and shown as −log₁₀ values on the y axis. The vertical dashed lines indicate ± 5 s.d. log₂ fold change in H/L ratio and the horizontal dashed line indicates P value < 0.001. Data shown represents two biological replicates.
normalized (Supplementary Fig. 2a,b) and showed high correlation coefficients across replicates and time points (around 0.9, Supplementary Fig. 2c), which reflects the expected small overall perturbation induced by the drug, and the consistent rank order of H/L protein ratios over time. Of the 6,328 unique protein groups identified, the 2,837 quantified across all time points and replicates were selected to subsequently calculate protein half-lives as previously described. Additional quality filtering was applied to reject proteins for which the linear regression resulted in $r^2 < 0.9$ for a total of 2,759 remaining protein groups (Fig. 1e). Due to the experimental design geared towards drug effects, the shortest time point was set at 6 h post treatment, compromising the ability to quantify very short-lived proteins. We hypothesized that proteins differentially degraded in a lenalidomide-dependent manner should exhibit differences in protein half-lives and would thus be candidate substrates. Comparing global protein half-lives of lenalidomide treated and control cells (Fig. 1e), we identified casein kinase 1 alpha (CSNK1A1) as the most affected protein, which we validated elsewhere in parallel to a study published by Kronke et al., who linked CSNK1A1 degradation to 5q-MDS efficacy of lenalidomide. Analysis of the full data set confirmed that the effect on CSNK1A1 is more pronounced and clearly identified at the late time point (Fig. 1f), supporting the concept of single time point measurements for target identification. To test the proposed approach of a single time point pSILAC experiment, we analysed the T16 time point (two replicates) alone by calculating moderated t-test $P$ values using the limma package. We included all protein groups that were identified with at least three unique peptides and quantified with more than two peptides ($n = 2,654$). Comparison of the T16 H/L protein ratios of lenalidomide or DMSO-treated samples identifies CSNK1A1, ZFP91 and HEXB as significantly downregulated targets along with a number of proteins whose turnover is reduced on drug treatment (Fig. 1g and Supplementary Table 2).

**Single time point pSILAC mass spectrometry.** Next, we sought to assess the robustness of our approach by conducting a single time point experiment in a different cell line (Hct116), with four replicate measurements at 16 h post treatment, of which, two were designed as label swap experiments (see Supplementary Fig. 1a–c). Hct116 cells were treated with either 30 $\mu$M lenalidomide or DMSO control concomitant with transfer into medium containing amino acids labelled with heavy isotopes (similar to the first experiment). Cells were collected 16 h post treatment and split for pSILAC MS and RNA-seq experiments (Fig. 2a). Samples for MS analysis were pre-fractionated and subsequent analyses were performed on an Orbitrap Fusion mass spectrometer. LC-MS/MS data were processed with MaxQuant, identifying 108,546 peptide sequences assigned to 7,759 unique protein groups (false discovery rate $< 1\%$ on peptide and protein level; requiring a minimum of one unique peptide per protein). H/L protein ratios (H/L protein ratios of label swap experiments were inverted) were log$_2$ transformed and quantile normalized (Supplementary Fig. 3a–c). We restricted our initial analysis to protein groups quantified with at least three unique peptides in all experiments ($n = 3,352$). Comparing averaged log$_2$ H/L protein ratios of lenalidomide treated or DMSO control samples (Pearson’s correlation coefficient of 0.99), we identified CSNK1A1 as the most strongly affected protein (Fig. 2b), which is in accordance with the pSILAC experiments in HEK293T. We next assessed the statistical significance of drug-induced changes to protein turnover, which is reflected in the difference between H/L protein ratios of DMSO and lenalidomide samples. We calculated intensity binned (10 bins of equal size) $P$ values using the significance B approach (Fig. 2c), which reflects the commonly observed dependence of H/L protein variability on signal intensity. In this approach, we identified additional proteins to exhibit significantly altered H/L protein ratios, most notably ZFP91 (Fig. 2b,c and Supplementary Table 3). To exclude the possibility of the drug treatment altering RNA levels, we performed RNA-seq on the samples used for mass spectrometry. The resulting data were analysed both at the levels of exons and introns, which allows the detection of both transcriptional and post-transcriptional changes. However, very few genes displayed a significant change in transcript levels, suggesting that transcriptional changes can still be neglected at the T16 time point (Fig. 2d). Overall, the experimental design significantly reduced the required machine time compared to multi time point experiments. Notably, CSNK1A1 and ZFP91 were both also identified by analysing forward and reverse experiments individually (two replicates each), which indicates that two replicates would be sufficient to generate a list of candidates for further validation (Supplementary Fig. 1b,c).

**Comparing changes in protein stability and abundance.** Next, we directly compared single time point pSILAC MS to measuring changes in total protein levels by TMT. We treated HEK293T cells for 16 h with 30 $\mu$M lenalidomide or a DMSO control (two biological replicates, consistent with the treatment for pSILAC, except that we did not exchange the growth medium for heavy amino acids). After lysis, reduction, alkylation and tryptic digest, peptides were subjected to TMT labelling, pre-fractionation and subsequent mass spectrometry analysis using a MS3 protocol on an Orbitrap Fusion mass spectrometer (see Methods section). Following scaling and normalization of reporter ion intensities, we performed statistical analysis by calculating moderated t-test $P$ values using the limma package. While the two validated substrates CSNK1A1 and ZFP91 show induced degradation by lenalidomide, the observation, however, is not significant and neither CSNK1A1 nor ZFP91 were reliably identified as a drug target based on the TMT data set. In contrast, within the pSILAC HEK293T T16 samples as well as in the Hct116 T16 samples (see Figs 1g and 2b,c and Supplementary Fig. 1b,c), we unambiguously identify CSNK1A1 and ZFP91.

We conclude that for the target identification of small molecules that exert their effects by altering protein stability, measuring protein turnover provides a sensitive and robust alternative compared to relative quantification of changes to protein abundance.

**Combining multiple data sets can further increase robustness.** We next explored the potential for combining data from multiple data sets to enhance the robustness of target identification. To do so, we utilized a linear model approach as implemented in the limma package and frequently employed for the analysis of gene expression data (see Methods section; and Supplementary Table 1 for design matrix). This framework allows modelling of both, the general changes to log$_2$ H/L protein ratios over time and the influence of the drug treatment on these changes. When applied to H/L protein ratios of all forward pSILAC experiments (proteins with unique peptides $< 3$ or missing values were excluded, resulting in 3,353 proteins analysed), we find CSNK1A1 and ZFP91 to be the two most significant proteins ($P < 0.0005$ and log-fold change $> 0.5$), with log-fold changes of 0.72 and 0.57, respectively (Fig. 3b, Table 1 and Supplementary Table 4). Therefore, this approach is suitable to combine data from biological replicates and together with the previous results,
prompted us to further validate ZFP91 as a novel lenalidomide-induced target of CRL4\textsubscript{CRBN}.

ZFP91 is a lenalidomide-dependent CRL4\textsubscript{CRBN} substrate. ZFP91 is a ZnF protein and putative ubiquitin ligase\textsuperscript{31,32}. To recapitulate our mass spectrometry results, we treated MM.1S cells with increasing concentrations of lenalidomide, thalidomide or with DMSO control and performed western blot analysis. We found a dose-dependent decrease of ZFP91 protein when treated with lenalidomide (Fig. 4a). We next examined the effect of lenalidomide on ZFP91 half-life in cells, by performing CHX chase experiments in HEK293T and SK-N-DZ cells (SK-N-DZ were selected for higher levels of CRBN\textsuperscript{9} and to validate the results in a cell line different from the cell lines used for mass spectrometry experiments). Following treatment with 50\(\mu\)g ml\(^{-1}\) CHX and increasing concentrations of lenalidomide or thalidomide, we found that ZFP91 protein levels were largely depleted after 6h of treatment with lenalidomide in a dose-dependent manner (Fig. 4b,c). Greater depletion was observed in SK-N-DZ cells, in accordance with the higher levels of CRL4\textsubscript{CRBN} ligase. Thalidomide in contrast had very little effect on the stability of ZFP91, similar to what has been previously observed for the ZnF proteins IKZF1 and IKZF3 (refs 13,14,39).

We next confirmed that lenalidomide-dependent degradation of ZFP91 was abrogated by treatment with the proteasome inhibitor bortezomib (Fig. 4d, lanes 4–5) and the NEDD8-activating enzyme inhibitor MLN4924 (Fig. 4d, lanes 6–7), in line with a Cullin RING ligase and proteasome-dependent mechanism. To confirm the dependence of ZFP91 degradation on CRL4\textsubscript{CRBN}, we utilized CRISPR/Cas9 genome editing to generate two independent pools of HEK293T cells after 16h (T16, Hct116), identifies three genes with altered transcription, and none with altered post-transcriptional regulation. R indicates the Pearson’s correlation coefficient. Data presented are the means of biological replicates (\(n = 4\)).

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**Figure 2 | Single time point pulse-SILAC is sufficient to identify drug targets.** (a) Outline of the single time point pulse-SILAC mass spectrometry and RNA-seq experimental design. (b) Scatter plot depicting substrate candidates identified by comparing the H/L protein ratios of lenalidomide treated to DMSO control cells. Data presented are the means of protein ratios (\(n = 4\) biological replicates and R represents the Pearson’s correlation coefficient). Only protein groups are shown that were quantified with minimum of three unique peptides in each experiment (3,352). (c) Scatter plot depicting the identification of substrate candidates. Log\(_2\) changes in LENA to DMSO H/L protein ratio are shown on the x-axis and log\(_2\) sum of MS1 intensities (combined for heavy and light peptides) on the y-axis. Significance B was calculated for 10 intensity bins and proteins with Significance B P values \(< 1 \times 10^{-7}\) are shown in red\textsuperscript{35}. Only protein groups are shown that were quantified with minimum of three unique peptides in each experiment (3,352). (d) Comparison of intronic (x axis) and exonic (y axis) expression changes in lenalidomide treated to DMSO control cells after 16h (T16, Hct116), identifies three genes with altered transcription, and none with altered post-transcriptional regulation. R indicates the Pearson’s correlation coefficient. Data presented are the means of biological replicates (\(n = 4\)).
ubiquitination of ZFP91 in a dose-dependent manner (Fig. 5a, lanes 1–7). We further found that the close derivatives thalidomide and pomalidomide also promote the ubiquitination of ZFP91 in vitro.

This prompted us to characterize the binding of ZFP91 to CRL4A\(^{\text{CRBN}}\) in vitro. We measured the affinity of biotinylated ZFP91 to fluorescently labelled CRL4A\(^{\text{CRBN}}\) using time-resolved fluorescence energy transfer (TR-FRET)\(^{11}\). In presence of saturating concentrations of lenalidomide, we confirmed a tight association of ZFP91 with CRL4A\(^{\text{CRBN}}\) with an apparent dissociation constant (\(K_\text{D}\)) of 107 ± 4 nM (Fig. 5b and Table 1). In accordance with the similarity to IKZF1/3, we found that thalidomide, lenalidomide and pomalidomide promote binding of ZFP91 to CRL4A\(^{\text{CRBN}}\) in vitro, however, the potency of the compounds relative to each other slightly differs from what was observed for IKZF1/3 (ref. 11). Following this observation, we next sought to test if pomalidomide and thalidomide would also promote ZFP91 degradation in cells. We treated MM.1S cells with increasing concentrations of thalidomide, lenalidomide and pomalidomide and found that the effect of thalidomide is minimal (Fig. 5e). This is in accordance with what has been reported for IKZF1/3 and could be a result of thalidomide being significantly less stable in cell culture medium than lenalidomide and pomalidomide\(^{9,13,14}\). We conclude that ZFP91 represents

Figure 3 | Combining multiple data sets can further increase robustness and sensitivity. (a) Scatter plot depicting the fold change in relative abundance comparing lenalidomide to DMSO control treatment (HEK293 TMT samples). Log2 fold changes are shown on the x axis and negative P values are shown on the y axis. The vertical dashed lines indicate ±0.32 log2 fold change (25% up- or downregulation) and the horizontal dashed line indicates P value <0.01 (n=2 biological replicates). (b) Linear model analysis across all forward samples (multi and single time point experiments) using limma (see Methods section) finds CSNK1A1, CSTA and ZFP91 as most significant hits (P value <0.001). The data are depicted as volcano plot showing differentially stable proteins following treatment with lenalidomide as compared with DMSO. The x axis shows the log2 fold change and the y axis refers to the associated P value (−log10 scale).

Table 1 | List of proteins with altered stability.

| Gene ID | LogFC | P value | LogFC | P value | Half-life |
|---------|-------|---------|-------|---------|----------|
| CSNK1A1 | 0.62  | 2.2 × 10\(^{-85}\) | 0.72  | 7.6 × 10\(^{-5}\) | Hit |
| CA2     | 0.29  | 4.1 × 10\(^{-15}\) | 0.08  | 0.07    | No hit |
| LASP1   | 0.13  | 5.3 × 10\(^{-10}\) | 0.05  | 0.51    | No hit |
| ZFP91   | 0.34  | 7.9 × 10\(^{-6}\)  | 0.57  | 0.0004  | No hit* |
| NAA30   | −0.40 | 3.0 × 10\(^{-22}\) | −0.18 | 0.39    | No hit |
| NCSTN   | −0.39 | 2.3 × 10\(^{-21}\) | −0.32 | 0.07    | No hit |
| POMP    | −0.32 | 3.6 × 10\(^{-14}\) | −0.12 | 0.34    | No hit |
| GNL3L   | −0.29 | 3.0 × 10\(^{-12}\) | −0.41 | 0.06    | No hit |

\(^*\) ZFP91 was dropped from the half-life analysis for a missing value in one time point and is therefore designated ‘no hit’.

\(^\text{Log}_{10}\) fold changes are shown together with their respective P values. For half-life analysis, we qualitatively designated a protein a ‘hit’ or ‘no hit’.

The vertical dashed lines indicate ±0.32 log2 fold change (25% up- or downregulation) and the horizontal dashed line indicates P value <0.01 (n=2 biological replicates).
**Figure 4 | Validation of ZFP91 as bona fide lenalidomide-induced CRL4CRBN substrate.** (a) MM.1S cells were treated with increasing concentrations of lenalidomide or with DMSO. Following 24 h of incubation, ZFP91 and GAPDH levels were detected by anti-ZFP91 and anti-GAPDH western blot (shown is one representative experiment out of three replicates). (b) HEK293T cells were treated with 50 μg ml⁻¹ cycloheximide and increasing concentrations of lenalidomide, thalidomide or with DMSO, and cells were incubated for 6 h. ZFP91 and GAPDH levels were detected using anti-ZFP91 or anti-GAPDH immunoblotting (shown is one representative experiment out of five replicates). (c) as in (b) but using SK-N-DZ cells instead (shown is one representative experiment out of three replicates). (d) as in (b) but with co-treatment of bortezomib (proteasome inhibitor, lanes 4, 5) or MLN4924 (inhibitor of the NEDD8-activating enzyme, lanes 6, 7). Shown is one representative experiment out of three replicates. (e) as in (b) using parental HEK293T (lane 1–3) or two independent pools of HEK293T cells with genetic inactivation of CRBN by CRISPR/Cas9 (shown is one representative experiment out of two replicates).

**Discussion**

The increasing number of small molecules directed at altering the stability and degradation of protein targets¹⁰–²³,⁴², and natural products with potential activity towards protein stability⁴³, require large-scale and proteome-wide methods to interrogate global on- and off-target effects. Here we introduce a novel approach to detect global changes to protein turnover in a single time point pSILAC mass spectrometry-based proteomics experiment. While it should be noted that single time point pSILAC does not report precise protein half-lives, and differences to H/L ratios can be the product of multiple effects (such as induced protein degradation/stabilization, altered translation or transcriptional effects); we demonstrate that the combination of short drug treatment and RNA-seq can largely mitigate these potential sources of false positives. The combination of pSILAC with TMT⁴⁴ could further reduce the required machine time by combining replicates and multiple conditions into a single mass spectrometry experiment. However, such a strategy would likely come at the cost of sequencing depth due to necessary fragmentation of both (heavy and light) SILAC pairs to achieve TMT quantification. An alternative approach for a higher level of multiplexing could be neutron-encoded mass signatures SILAC (NeuCode-SILAC)⁴⁵, which similarly to conventional SILAC can be applied in a pulsed manner.

By applying pSILAC to the potent anti-cancer therapeutic lenalidomide, we demonstrate that a single time point experiment could identify ZFP91 as a novel lenalidomide-dependent CRL4CRBN substrate. ZFP91 and the known target CSNK1A1 were consistently detected as neo-substrate candidates in all single time point pSILAC experiments. While calculated protein half-lives from the multi time point experiment robustly identified CSNK1A1, ZFP91 was dropped from half-life calculation due to missing values at one of the time points. Characterization of the ZFP91 interaction with CRL4CRBN shows that a common structural motif in the ZnF domains of ZFP91 and IKZF1/3 mediates a high-affinity interaction with a CRL4CRBN-IMiD complex. The plethora of ZnF proteins in the human proteome, together with the tolerance for amino-acid variation increases the probability that even more targets of IMiD-induced and CRL4CRBN-mediated degradation exist.

**Methods**

**Compounds, enzymes and antibodies.** Thalidomide (HY-14658), lenalidomide (HY-A0003), pomalidomide (HY-10984), MLN4924 (HY-70062) and bortezomib (HY-10227) were purchased from MedChem Express (MCE, USA). All compounds were dissolved in DMSO at various concentrations. Due to limited stability of thalidomide in aqueous solutions, all required dilutions and cell culture media containing compounds were prepared fresh. HEK293T, SK-N-DZ, MM.1S and Hct116 cell lines were purchased from ATCC and cultured according to ATCC instructions. Cell lines were routinely tested for mycoplasma contamination using the MycoAlert detection kit (Lonza). Arginine and lysine-free cell culture medium, dialysed foetal bovine serum and amino acids for pulse-SILAC mass spectrometry were purchased from Cambridge isotope (Cambridge isotope laboratories, USA). Proteins were tested for >99% SILAC amino-acid incorporation after six–ten passages by mass spectrometry. Sequencing grade modified trypsin (V5117) was purchased from Promega (Promega, USA) and mass spectrometry grade hydrazine endopeptidase from Wako (Wako Pure Chemicals, Japan). Primary and secondary antibodies used included anti-ZFP91 at 1:100–1:500 dilution (ab30970, abcam), anti-
ZFP91 at 1:250–1:5,000 dilution (A303-245A, Bethyl Laboratories), anti-CRBN at 1:250 dilution (NBP1-91810, Novus), anti-GAPDH at 1:10,000 dilution (G8795, Sigma), IRDye680 Donkey anti-mouse at 1:10,000 dilution (926-68072, Licor) and IRDye800 Goat anti-rabbit at 1:10,000 dilution (926-32211, Licor). Ubiquitination enzymes Uba1, UbcH5a and ubiquitin were purchased from Boston Biochem.

Cell culture. HEK293T and Hct116 cells were cultured in i-arginine and i-lysine-free DMEM supplemented with dailysed foetal bovine serum, 2 mM i-glutamine and unlabelled i-arginine and i-lysine. Cells were grown to 40–50% confluency and the medium exchanged for DMEM supplemented with heavy i-lysine (13C6, 15N2–K8) to obtain heavy SILAC medium. Simultaneously, the medium was either supplemented with 30 μM lenalidomide (100 mM solution in DMSO) or the equivalent amounts of DMSO as control. Cells were incubated at 37°C, 5% CO2 and collected at various time points by two washes with phosphate-buffered saline and direct addition of lysis buffer. If samples were split for parallel RNA-seq, cells were washed twice following the manufacturer’s instructions.

Con structs and protein purification. Human DDB1, human CRBN, human CUL4A and mouse RBX1 were subcloned into pAC-derived vectors and recombinant proteins expressed as N-terminal His6 fusion proteins in Trichoplusia ni High Five insect cells using the baculovirus expression system (Invitrogen) and Recombinant CRL4ACRBN complex was purified through sequential Ni-NTA affinity, anion exchange (poros 50HQ) and size exclusion (Superdex 200) chromatography. In vitro neddylation was carried out as described, CUL4A and CRBN was incubated with a reaction mixture containing Alexa680-NEDD8(MIC) at 3.8 μM, NAE/UBA1 (E1) at 50 mM and UBC12 (E2), 1 mM ATP, 50 mM Tris pH 7.5, 100 mM NaCl, 2.5 mM MgCl2, 0.5 mM DTT and 5% (v/v) glycerol for 2 h at room temperature. Neddylated CRL4ACRBN was purified by size exclusion chromatography. Full-length human ZFP91 was obtained as synthetic codon optimized gBlock from IDT (IDT, USA). Recombinant ZFP91 was expressed as N-terminal StrepII fusion protein in Trichoplusia ni High Five insect cells using the baculovirus expression system (Invitrogen). For purification of StrepII-ZFP91, cells were lysed by sonication in lysis buffer (50 mM Tris-HCl pH 8.0, 200 mM NaCl, 5 mM 2-mercaptoethanol (BME), 1 mM phenylmethylsulfonyl fluoride, 0.1% Triton X-100 and 1 × protease inhibitor cocktail (Roche). Following ultracentrifugation, the soluble fraction was passed over Strept-Tactin Sepharose (IBA) and eluted with 2.5 mM D-desthiobiotin in wash buffer (50 mM Tris-HCl, 200 mM NaCl, 5 mM BME). The affinity purified protein was diluted to 50 mM NaCl with dilution buffer (50 mM Tris-HCl pH 8.5, 0.5 mM tris(2-carboxyethyl)phosphine (TCEP) and further purified via anion exchange chromatography (Poros 50HQ) with a linear gradient of 0–80% buffer B (50 mM Tris-HCl pH 8.5, 1 M NaCl, 0.5 mM TCEP) in buffer A (50 mM Tris-HCl pH 8.5, 25 mM NaCl, 0.5 mM TCEP) and subjected to size exclusion chromatography in 50 mM HEPES pH 7.4, 200 mM NaCl and 0.5 mM TCEP. The protein containing fractions were pooled, concentrated using ultrafiltration (Millipore) and flash frozen in liquid N2. Proteins were stored at −80°C.

Biotinylation of ZFP91. Purified StrepII-Avi-tagged wild-type or mutant ZFP91 were biotinylated in vitro at a concentration of 5–20 μM by incubation with final

![Figure 5 | ZFP91 and IKZF1/3 share a common sequence motif.](image-url)
concentrations of 2.5 μM BirA enzyme and 0.2 mM D-biotin in 50 mM HEPES pH 7.4, 200 mM NaCl, 10 mM MgCl₂, 0.25 mM TCEP and 20 mM ATP. The reaction was incubated for 4 hours at 24 °C followed by removal of the biotinylated proteins by purification via gel filtration chromatography and stored at −80 °C.

**TR-FRET binding assay.** Increasing concentrations of Alexa488-ND68-labelled CRL4C4R88 were added to pre-mixed biotinylated ZFP91 at 100 nM, terbium-coupled streptavidin at 2 nM (Invitrogen) and IMiDs at 5 μM (final concentrations) in 384-well microplates (Greiner, 784076) in a buffer containing 50 mM Tris pH 7.5, 100 mM NaCl, 0.1% pluronic acid and 1% DMSO. After excitation of terbium (Tb) fluorescence at 337 nm, emission at 490 nm (Tb) and 520 nm (Alexa488) were recorded with a 70 μs delay to reduce background fluorescence and the reaction was followed over 1 h by recording 60 technical replicates of each data point using a PHERAstar FS microplate reader (BMG Labtech). The TR-FRET signal of each data point was extracted by calculating FObs/C0 = FFree/C0 + FBound/C0 × (k-ap + FBound/C0) × (k-ap + FBound/C0) + (k-free + FBound/C0 + Gln+)/(2 × FBound/C0) 0.5.

**F**ₐ₀, **Fₐₚ** and **Fₐₜ** are the observed fluorescence intensity, and **Fₐₚ** and **Fₐₜ** bound the fluorescence intensities of the probe in its free and its bound states, respectively.

Dose response curves for IMiDs were obtained by mixing increasing concentrations of IMiD with a reservoir containing cCRL4C4R88 in 200 nM, biotin-ZFP91 or biotin-ZFP91(H141A) at 100 nM, and Tb-Streptavidin at 2 nM. The fluorescence as 520/490 nm ratio were recorded over 1 h as 60 technical replicates of each data point using a PHERAstar FS microplate reader (BMG Labtech) and resulting data analysed with GraphPad Prism 6 assuming equimolar binding of the probe to the receptor using the following equation11,17:

\[
F_{\text{Obs}} = F_{\text{Free}} + (F_{\text{Bound}} - F_{\text{Free}}) \times \frac{k_{\text{ap}} + [\text{Gln}]}{k_{\text{ap}} + [\text{Gln}] + (k_{\text{free}} + k_{\text{Bound}} + [\text{Gln}])} = \frac{4 \times C_{\text{F0}}}{C_{\text{F0}} + 2}.
\]

For the multi time point experiment, cells were lysed in RIPA buffer (50 mM Tris pH 7.3, 5 mM CaCl₂ buffer) for 6 h, and after dilution to 10–35% B in 60 min, 35–100% B in 5 min followed by 5 min wash at 100% B (high pH buffer B: 20 mM ammonium formate at pH 10.1 in 90% acetonitrile) on a YMC-Triart C18 0.5 × 250 mm, 12 µm × 3 µm column. The 36 fractions were pooled (f01 + f13 + f25, f02 + f14 + f26 and so on) to a final of 12 samples.

Peptides were analysed by nano liquid chromatography tandem mass spectrometry with an EASY-nLC 1,000 pump using a two-column set-up (Thermo Fisher Scientific). The peptides were loaded in buffer A onto a trap column (Acclaim PepMap 100, 75 µm × 2 cm, C18, 3 µm, 100 Å). They were separated, at a flow rate of 150 nl min⁻¹ with a linear gradient of 2–6% buffer B in 1 min followed by a linear increase from 6–32% B in 8 min, 22–88% B in 16 min, 88–100% B in 4 min, and by 5 min wash at 80% buffer B (buffer A: 0.1% formic acid in water, buffer B: 0.1% formic acid in acetonitrile) on a PepMap RSLC analytical column (50 µm × 15 cm, C18, 2 µm) at 45 °C mounted on a modified DPV ion source (New Objective) connected to an Orbitrap Fusion mass spectrometer (Thermo Scientific). The data were acquired using a mass range of m/z 350–1,600, resolution 120,000, AGC target 3 × 10⁶, maximum injection time 100 ms, dynamic exclusion of 60 s for the peptide measurements in the Orbitrap and a cycle time of 3 s for HCD fragmentation of the most abundant peptides and fragment ion measurement in the LTQ using AGC target 1 × 10⁶, NCE 30%, maximum injection time 250 ms.

**Mass spectrometry data analysis.** MaxQuant13 (versions 1.5.0.30 and 1.5.3.8 for multi time point and single time point, respectively) was used for RAW file processing and controlling peptide and protein level false discovery rates, assembling proteins from peptides, and protein quantification from peptides. Fragment ion spectra were searched against a human Uniprot database (downloaded on 13 November 2014 and 29 January 2015 for multi time point and single time point, respectively) and common contaminant proteins (included in the MaxQuant software). Carbamidomethylation (Cys) was set as a fixed modification; deamidation (Gln, Asn) and oxidation (Met) were selected as variable modifications. SILAC amino acids used for quantification were Ly8 and Arg10. The MaxQuant ProteomeGroup.txt output file was imported into the R framework26 for all subsequent analyses.

For the statistical analysis the data was considered as two sets (controls and IMiD-treated). The peptides and proteins only identified by site were removed from the data set. H/L protein ratios (‘Ratio H/L’ columns per time point, not normalized) were quantile normalized using the limma package26 with normalization carried out separately for each time point. The quantile normalization as described elsewhere26 and implemented in the limma package has the advantage of being a complete data method without requiring a defined baseline data set. In short, n data sets of length p are turned into a matrix X with p × n dimensions, with each data set being a column. We sort each column of X to give X_sort and now assign the mean across each row of X_sort to each element of the row to give X_sort. Next, we rearrange each column of X_sort back to the original ordering and derive X_normalized.

**Determination of protein half-lives.** While we performed the linear regression fits to obtain protein half-lives, we noted several challenges for accurate determination of protein half-lives, including low number of time points, low number of replicates, inclusion of label swap data (which display a consistent rank order but a systematic offset compared to the forward experiments, see also Supplementary Fig. 3), and also the known limitations of obtaining half-lives by SILAC mass spectrometry, such as ‘light’ amino-acid recycling27. However, we demonstrate that despite these limitations the data are informative for the identification of ligase targets. Protein half-lives were obtained as previously described24. In essence, normalized and log-transformed H/L ratios for protein groups quantified across all time points and replicates (n = 2,837) were used for linear regression (log(H/L) ~ time) for the DMSO and lenalidomide samples of each protein. Proteins that exhibited sufficient quality of both (DMSO and lenalidomide) fits (r² > 0.9, total number of proteins 2,759) were used to obtain the protein decay rate constant (kdecay) using the following equations24:

\[
k_{\text{decay}} = \frac{1}{\text{time}} \sum \frac{1}{n} \log (x_{t} + 1) + \frac{1}{t_{\text{exp}}} \log \frac{2}{t_{\text{exp}}}
\]

where m is the number of time points (t), rᵢ is the H/L ratio of a specific protein at a given time point, and tₑᵢ is the time needed to double the total amount of protein in the experiment, which is usually set to the duration of the cell cycle. The doubling time of HEK293T cells was estimated to be ≈ 22 h by counting cells growing in 96-well dishes. However, using tₑᵢ = 22 h resulted in negative decay rates for a subset of proteins, indicating that in our experimental set-up the protein doubling time was longer than 22 h. We attributed this effect to the cells growing more slowly than expected during the experiment, for example, due to increasing cell densities or the effect of heavy amino acids. To accommodate for this effect, tₑᵢ was reduced for a subset of proteins, which resulted in positive decay rates that were different from published estimates24. While the use of an incorrect tₑᵢ value leads to incorrect absolute protein decay rates, it does not limit the use of the inferred decay rates or...
protein half-lives as a relative measure in the comparison of different treatment conditions, which is the purpose in this study. Based on the decay constant, the half-life of a protein is given by:

\[ T_{1/2} = \frac{\log_{2} 2}{k_{lp}} \]

The derived protein half-lives were used for subsequent comparison.

**Data analysis for single time point pulse SILAC.** For conventional SILAC mass spectrometry, it has been shown that the distance of a ratio measured in terms of the log2 value serves as a robust estimate of the identity of the protein/peptide. Under the assumption of a normal distribution of the data, a P value can be calculated (significance A)\(^3\). Based on the observation that the error of H/L protein ratios is correlated to the summed intensity, P values are calculated for subsets of the data binned by intensity (significance B)\(^3\). Since the log-transformed LENA/DMSO protein ratios assume in approximately normal distribution, we used this approach to assess the statistical significance of proteins.

To control for potential non-linearities in some samples, log2 H/L protein ratios were quantile normalized using the normaliseBetweenArrays function from the Bioconductor library (February 2014) with both the forward and reverse sequences. Database search criteria were as follows: trypsin with two missed cleavages, a precursor mass tolerance of 5 parts per million, and fragment ion mass tolerance of 0.8 Da, static alkylation of cysteine (57.02146 Da), static TMT labelling of lysine residues and N-termini of peptides (229.16293 Da) and variable oxidation of methionine (15.99491 Da). TMT reporter ion intensities were measured using a 0.003 Da window around the theoretical m/z of each reporter ion in the TMT scan. Peptide spectral matches with poor quality MS3 spectra were excluded from quantification (<=summed signal-to-noise across 10 channels and <0.5 precursor ion specificity).

Statistical analysis was carried out using the limma package within the R framework\(^3\). RNA-seq analysis. Samples of isolated RNA were used for library preparation using ScriptSeq v2 total RNA library preparation kit and sequenced on an Illumina HiSeq 2,500 sequencer per Illumina standards. Sequencing reads were aligned to the human genome (Bsgnome.Hsapiens.UCSC.hg19 Biocoridor package, using splicedAlignment = TRUE) and quantified at the level of genes (TxDb.Hsapiens.UCSC.hg19.knownGene Biocoridor package) using the QuasR package with default parameters\(^3\). Intronic and exonic reads were separately quantified, and differentially expressed genes were identified using the edgeR Biocoridor package\(^3\) as described elsewhere\(^3\).

**Data availability**. All RNA-seq data sets have been deposited in GEO under the accession number GSE94728. Mass spectrometry raw data files have been deposited in PRIDE Archive under the accession numbers PXD005857 and PXD005849 for the HEK293T and Hc116 experiments, respectively. The data that support the findings of this study are available from the corresponding author on request.

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51. Supplementary Information accompanies this paper at http://www.nature.com/naturecommunications