

A Kinetic Scout Approach Accelerates Targeted Protein Degradation Development

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Abstract:

Bifunctional molecules such as targeted protein degraders induce proximity to promote gain-of-function pharmacology. These powerful approaches have gained broad traction across academia and the pharmaceutical industry, leading to an intensive focus on strategies that can accelerate their identification and optimization. We and others have previously used chemical proteomics to map degradable target space, and these datasets have been used to develop and train multiparameter models to extend degradability predictions across the proteome. In this study, we now turn our attention to develop generalizable chemistry strategies to accelerate the development of new bifunctional degraders. We implement lysine-targeted reversible-covalent chemistry to rationally tune the binding kinetics at the protein-of-interest across a set of 25 targets. We define an unbiased workflow consisting of global proteomics analysis, IP/MS of ternary complexes and the E-STUB assay, to mechanistically characterize the effects of ligand residence time on targeted protein degradation and formulate hypotheses about the rate-limiting step of degradation for each target. Our key finding is that target residence time is a major determinant of degrader activity, and this can be rapidly and rationally tuned through the synthesis of a minimal number of analogues to accelerate early degrader discovery and optimization efforts.

Introduction:

Bifunctional targeted protein degraders have revolutionized drug discovery over the past decade, providing a means to target previously undruggable functions of liganded targets, overcome inhibitor resistance mechanisms and provide enhanced selectivity over inhibition.¹ However, targeted protein degraders can be challenging to develop, with much of the design process remaining empirical.² Medicinal chemistry represents a

38 resource-intensive component of targeted protein degrader development.³ Consequently, the mechanisms of
39 targeted protein degradation, and impacts of biological and chemical variables have been intensively studied
40 over the past 5 years.⁴

41 Rapid progress in understanding how biological variables contribute to a given target's degradability in
42 different cellular contexts has been made. We previously mapped the degradable kinome via large-scale
43 proteomic profiling, allowing us to assign a degradability score to each degraded kinase.⁵ These data were used
44 in combination with other proteomic, genomic and molecular modeling datasets to generate predictive machine-
45 learning based models of degradability that can be applied proteome-wide to inform target selection.⁶ Elegant
46 work from the Schulman lab has shown that both the activation (neddylation) state of the E3-ligase⁷, and the
47 expression level of the hijacked E2-enzyme⁸ drive cell line and cell state dependent degradation pharmacology,
48 informing model selection.

49 Chemical variables, defined as those we can change via small molecule design, have also been
50 rigorously examined via focused experimental and theoretical studies with the aim of accelerating the
51 optimization process and providing rational design strategies for degraders (Fig 1A).⁹⁻¹⁵ These models differ in
52 their underlying structure, but agree in their predictions that the stability of the functional ternary complex has a
53 profound impact on the expected degradation outcome. These models are used to predict how features such as
54 the target and E3-ligase ligand binding kinetics (affinity, residence time), as well as the co-operativity (α) of the
55 ternary complex, impact degradation activity under different biological conditions. The optimal values for each
56 system depend on biological features such as ubiquitination and deubiquitination rate, target and E3-ligase
57 abundance, target sequence length and proteasomal degradation rate.⁹⁻¹⁵

58 Experimentally, the targeted protein degradation field has invested substantial efforts in understanding
59 how the co-operativity of ternary complexes can be influenced by changes to the degrader linker length,
60 composition and rigidity.^{4,9} In a typical targeted protein degrader medicinal chemistry campaign, a number of
61 linker variants are synthesized and evaluated with the goal of influencing ternary complex formation and stability.
62 Rational design strategies informed by structural biology of degrader-bound complexes have been successfully
63 implemented to increase co-operativity and improve the degradation of BRD4.⁹ Recently, Ichikawa and
64 colleagues have shown that ternary complex co-operativity of CRBN-recruiting BRD4 degraders can also be
65 modulated by varying the component of CRBN binders oriented towards the protein-protein interface.¹⁶ However,

66 the extent to which the co-operativity of a complex can be influenced by linker optimization is also determined
67 by the features of the protein-protein interface formed in the productive ternary complex. In many cases,
68 methodical exploration across the full range of ternary complex co-operativity values is not easily achievable for
69 a given degrader series. Instead, structure-cooperativity relationships are empirically determined through
70 iterative synthesis and testing.³

71 The impact of systematically fine-tuning the target residence time has also been examined, primarily in
72 the context of the kinase target BTK, due to its relevance in cancer and the availability of matched reversible,
73 reversible-covalent and irreversible covalent ligands targeting cysteine 481.¹⁷⁻²² These studies demonstrated that
74 target residence time profoundly impacts degradation, and that covalent bond formation can be compatible with
75 targeted protein degradation in some contexts. Whilst these reports support the hypothesis that fine-tuning target
76 residence time is a viable strategy for degrader optimization, they focus on a narrow target scope, as potent
77 cysteine-directed binders are not available for many targets-of-interest. How large of an effect fine-tuning target
78 residence time is anticipated to have on targeted protein degraders across a wide target space remains
79 underexplored. A rigorous understanding of how kinetic binding variables influence TPD could aid in designing
80 more optimal initial test libraries when developing degraders for new targets, in the same way that 'linkerology'
81 is now systematically explored. These data are also crucial for validation and training the next iteration of
82 mathematical models of targeted protein degradation.⁹⁻¹⁵ In this study, we sought to map the impact of target
83 residence time on ternary complex formation, ubiquitination and target protein degradation across numerous
84 degradable targets (Figure 1A).

85

86 Results:

87 To systematically investigate the impact of target residence time on TPD, we required a method to vary
88 ligand target residence time across numerous targets. Targeted protein degraders have high molecular weight,
89 and compounds generated early in the discovery pipeline are likely to have physicochemical properties that
90 negate efficient cellular washout, and impede accurate measurement of live-cell off-rates. To circumvent the
91 need for large analogue libraries and live cell off-rate measurements, we leveraged a series of pan-kinase binder
92 analogues which differ only in the functional group that interacts with the catalytic lysine (Figure 1B). Compound
93 3, the parent molecule, is a pan-kinase binder that interacts with the catalytic lysine via reversible hydrogen bond

94 formation with the nitrile group.²³ In Compound 1, a benzaldehyde participates in reversible-covalent imine
95 formation, which is readily hydrolyzed resulting in a residence time below 10 minutes for the majority of engaged
96 kinases, as measured by washout chemical proteomics.²⁴ Stabilization of this imine by incorporation of a
97 salicaldehyde moiety (YTP-2137, PDB: 7FIC), extends cellular residence time kinome-wide, ranging from ~10
98 minutes to over 6 hours depending on the individual kinase.²⁴ Finally, incorporation of a phenyl sulfonylfluoride
99 results in irreversible covalent bond formation at the catalytic lysine kinome-wide (XO44).²⁵ Importantly, each of
100 these analogues retains a similar pan-kinome binding profile in live cells.²³⁻²⁵ We incorporated these well-
101 characterized pan-kinome binders into VHL-recruiting targeted protein degraders, varying the regiochemistry of
102 the lysine-targeting moiety and the linker, to form a library of compounds we termed kinetic scout degraders
103 (Figure 1B, Figure S1). This approach allows predictable variance of the target residence time across the kinome,
104 using a minimal set of three covalent analogues and a reversible control.

105 To characterize the relative cell permeability of the kinetic scout degraders we evaluated cellular VHL
106 target engagement using a previously described dual luciferase assay where the test compound competes with
107 dTAG^V-1 for occupancy of VHL, resulting in a rescue of NLuc-FKBP12^{F36V} degradation (Figure S2). Promising
108 series were re-tested in a more accurate VHL NanoBRET assay (Figure 1C).²⁶ In parallel, we tested our library
109 in cell viability assays, using isogenic parental and VHL knockout MOLT4 cells to identify compounds with
110 degradation-dependent viability effects, observed as a reduction in cell toxicity in the VHL knockout lines (Figure
111 1D, Figure S3). Based on these data, we selected a set of 4 degraders that had matched regiochemistry and
112 linker length, varying only in their lysine-targeting functional group (ERS-01-006, JWJ-01-290, LS-1-037 and
113 JWJ-01-293 respectively). These degraders show comparable cell permeability in NanoBRET VHL engagement
114 assays (Figure 1C) and VHL-dependent effects in viability assays (Figure 1D).

115 To test the impact of residence time on degradation, we performed global proteomics analysis of MOLT4
116 cells following treatment with ERS-01-006, JWJ-01-290, LS-1-037 or JWJ-01-293 for 5 hrs at 1 μ M and 10 μ M
117 concentrations (Figure 2A). A short time point was selected to minimize the likelihood of indirect target
118 downregulation, particularly important when profiling pan-kinase binders that profoundly change cell signaling
119 and proteome composition at longer time points.⁵ MOLT4 cells were selected as they have high levels of active
120 VHL, allowing for fast targeted protein degradation to occur.⁷ We observed profound differences in the
121 degradation profiles of the kinetic scout degraders with different lysine-targeting functional groups. A total of 25

122 kinases were degraded by at least one compound in the series (Figure 2A). Plotting the targets that met statistical
123 cutoffs for each molecule at one or both evaluated concentrations, we observe overlapping targets predominantly
124 occur between compounds with closely matched reactivity at lysine, with the exception of PRKAA1 (Figure 2B).
125 We performed orthogonal validation of the proteomics data by immunoblotting, confirming that only LS-1-037
126 and JWJ-01-293 degrade CDK6, at concentrations matching the proteomic profiling data (Figure 2D). As a
127 second example, we used immunoblotting to validate that only JWJ-01-293 degraded NEK9 (Figure 2D). To
128 ensure these effects were due to binding kinetics and not influenced by the fluoro-sulfone group in JWJ-01-293,
129 we synthesized methyl sulfone JWJ-01-291 as a more closely matched negative control to JWJ-01-293 and
130 confirmed no degradation. To enable cell-based studies of binary degrader-target binding, we synthesized
131 negative control degraders where the kinase binding warhead and linker remained intact, but a diastereomer of
132 the VHL-binding ligand was used to prevent VHL recruitment and degradation from occurring on the timescale
133 of the assay (Figure S4A). We next used a NanoBRET cellular target engagement assay to validate that all four
134 negative control compounds bind CDK6 in live cells with ATF-01-129, ATF-01-074, ATF-01-076 demonstrating
135 comparable cell permeability and IC₅₀ values at equilibria, and ATF-01-075 demonstrating greater potency, as
136 expected for an irreversible compound (Figure 2E). We next performed the CDK6 NanoBRET assay to measure
137 cellular washout kinetics, for validating that our degraders engaged with CDK6 in a comparable fashion to the
138 parent compounds. Here, cells were treated with 10 μM degrader, washed 4 x with media, and then treated with
139 K-10 tracer. Signal recovery was then measured over 2 hrs. The non-covalent ATF-01-129 and ATF-01-162
140 analogues were rapidly washed out in the CDK6 NanoBRET washout assay, indicated by a rapid return of BRET
141 signal to baseline (DMSO). The reversible covalent (ATF-01-074, ATF-01-076) analogues demonstrated partial
142 displacement on the timescale of the assay, consistent with a longer residence time / slower K_{off}, and the
143 irreversible covalent analogue (ATF-01-075) was not displaced consistent with covalent binding (Figure 2F,
144 Figure S4B).

145 To examine how occupancy correlates with degradation, we next evaluated cellular target occupancy of the VHL-
146 negative diastereomers of ERS-01-006, JWJ-01-290, LS-1-037 or JWJ-01-293 (ATF-01-129, ATF-01-074, ATF-
147 01-076, or ATF-01-075) at equilibria across a panel of 192 kinases using NanoBRET assays in live cells at 1 μM
148 (Figure S5). Overall, low target engagement of kinases at 1 μM degrader concentrations was observed across
149 the series relative to control molecule XO44, consistent with reduced cell permeability of the bifunctional

150 molecules compared to parent inhibitor controls. Expected enhancement of target engagement by irreversible
151 molecule ATF-01-075 was observed. The CDK subfamily showed highest target occupancy by ATF-01-075,
152 however this trend is not reflected in the degradation data. This discrepancy agrees with findings from our labs
153 and others demonstrating that binary target occupancy does not correlate with degradation outcomes, due to
154 the additional requirement of forming a productive ternary complex.⁵

155 To gain a mechanistic understanding of the drivers of differential degradation across the series, we used
156 a sensitive VHL affinity-purification coupled to mass spectrometry method where cell lysates are supplemented
157 with exogenous VHL-EloB-EloC to evaluate the ternary-complexes formed proteome-wide with ERS-01-006,
158 JWJ-01-290, LS-1-037 and JWJ-01-293 (1 μ M) in complex with VHL-EloB-EloC (Figure 3A). We observed
159 enrichment of a large proportion of the kinome, reflecting the multitargeted nature of the kinase binding
160 warheads, and indicating that the four compounds can each promote formation of ternary complexes between
161 VHL and the majority of the kinome when VHL is not limiting (Figure 3B-C). Examining trends across the series,
162 we observed an increase in the number and fold-change of kinase complexes enriched as the residence time of
163 the kinase-targeting ligand increased, consistent with the predicted increase in ternary complex stability (Figure
164 3C).

165 To evaluate which of these complexes result in ubiquitination, we performed the E-STUB assay coupled
166 to mass spectrometry to identify proteins ubiquitinated by VHL in the presence of 1 μ M ERS-01-006, JWJ-01-
167 290, LS-1-037 and JWJ-01-293 (Figure 4, Figure S6). To preserve the integrity of the proteome, the E-STUB
168 assay was performed under proteasome inhibition and cells were treated with the degraders for 45 minutes
169 followed by a 15 min biotin labeling. Consequently, not all degraded targets were enriched in the E-STUB assay.
170 Additionally, E-STUB is validated only in easily transfectable HEK293T cells, and thus protein abundance
171 differences between HEK293T and MOLT4 may also account for reduced target detection. Despite these
172 caveats, we considered the E-STUB data a useful snapshot of the acute effects of the four kinetic scout
173 degraders on targeted ubiquitination. In addition to ubiquitination of degraded targets, a number of ubiquitinated
174 kinases were identified whose abundance did not change in our global proteomics analysis, such as CDK13.
175 These may represent kinases that are insufficiently ubiquitinated to promote degradation, or kinases that are
176 degraded slowly, and therefore not captured at the 5 hrs time point.

177 The majority of kinases complexed by VHL in the IP/MS assay were neither ubiquitinated nor degraded.
178 We have previously shown that pan-kinase degraders can form non-productive ternary complexes that do not
179 support degradation, which is in part supported by our ubiquitination data (Figure 4). Furthermore, unlike in our
180 experiment, where exogenously expressed VHL was used to enhance detection of ternary complexes, in the
181 cell, the potential kinase targets compete for a limited pool of VHL and therefore the most efficient E3-
182 ligase:degrader:target interactions dominate the degradation pharmacology at short time points. Finally, these
183 complexes may lead to slow ubiquitination and degradation, that occurs on a time scale beyond what we can
184 measure with acute proteomic approaches.

185 To better understand how targeted protein degradation of individual kinases is impacted by differences
186 in the kinase-binder residence time, we examined the relationship between target degradation, target occupancy,
187 ternary complex formation, and productive ubiquitination for targets detected in multiple assays (Figure 5, Figure
188 S7). Grouping the targets by residence-time preference allowed us to identify trends and formulate data-driven
189 mechanistic hypotheses. Kinases such as AAK1 and STK17B were only productively ubiquitinated and degraded
190 by fully reversible molecule ERS-01-006, but formed ternary complexes with all kinetic scout degraders,
191 indicating for these kinases, ternary complex dissociation may be the rate limiting step for degradation.
192 Conversely, kinases that were preferentially degraded by irreversible compound JWJ-01-293 are exemplified by
193 AURKA, ITK and NEK9. These kinases show a greater fold-change in the enrichment of both ternary complexes
194 and ubiquitination with longer kinase binder residence time, indicating that a greater ternary complex stability is
195 required to achieve productive ubiquitination. CDK1 formed ternary complexes that resulted in ubiquitination with
196 all compounds, but was preferentially degraded by JWJ-01-293, indicating that polyubiquitination may be the
197 rate-limiting step for CDK1 degradation in this compound series. Many kinases showed profiles suggestive of
198 both effects at work, for example CDK6 where increased enrichment of the ternary complex occurs across ERS-
199 01-006, JWJ-01-290 and LS-1-37 resulting in degradation by LS-1-37 at 10 μ M, but degradation activity is
200 reduced in the covalent analogue JWJ-01-293 at 10 μ M. These results are in broad agreement with those
201 predicted by mathematical modeling approaches, where a balance of ternary complex co-operativity and binary
202 target residence time must be achieved for optimal degradation, and these effects vary with protein ubiquitination
203 rate, protein abundance, and protein length. However, the residence time requirements of a given kinase for
204 optimal degradation do not correlate with experimental or predicted scores of degradability, indicating that

205 currently the optimal ternary complex stability for a given productive E3-ligase:degrader:target complex must be
206 determined empirically.

207

208 Discussion:

209 Systematic tuning of the target residence time is an under explored strategy in early targeted protein
210 degrader development projects reported in the academic literature. Our key finding is that alterations to the off-
211 rate of the target-binding ligand have an outsized effect on degrader efficacy and selectivity across a wide target
212 space, comparable to varying the linker length.⁵ These findings indicate prospective degrader discovery efforts
213 can be accelerated by incorporation of systematic ligand off-rate variance in initial degrader designs. Kinetic
214 scout degrader libraries combined with the unbiased proteomic workflow we outline here, can identify the optimal
215 ligand binding properties for successful degradation of a target of interest. In our study, both ternary complex
216 formation, and the extent of ubiquitination within these complexes, were impacted by changes in ligand off-rate.
217 Together, these data can be used to define a rational degrader optimization strategy.

218 We propose that the implementation of tunable reversible – covalent chemistry at lysine is an efficient
219 method for rapid modulation of ternary complex dynamics, that can complement linkerology, when embarking
220 on new degrader-discovery efforts. Unlike cysteine, lysine is commonly found in or adjacent to drugged enzyme
221 active sites, and ligand binding pockets.²⁷ Recent work has characterized a substantial library of lysine-targeting
222 pharmacophores, demonstrating broad ligandability of lysines proteome-wide, indicating our approach has the
223 potential to be applied broadly.²⁷ Furthermore, although here we explore modulating ternary complex stability via
224 systematic tuning of the target ligand residence time, a similar conceptual approach could be applied to the E3-
225 ligase ligand.

226 Our previous work has shown that the creation and open sharing of unbiased chemical proteomic
227 degradation datasets can accelerate fundamental research into degrader mechanism-of-action, by providing
228 uniform training and test sets for machine-learning, model building and evaluation.¹² We anticipate the datasets
229 in this manuscript will serve as useful in training and evaluation for the mathematical modeling of targeted protein
230 degradation and induced proximity drugs. When combined with existing models of biological degradability and
231 *in silico* structural biology pipelines, we believe this dataset and approach will ultimately aid development of
232 better predictive modeling with which to guide the development of bifunctional induced proximity molecules.

233

234

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245

246 Author contributions:

247 A T F, J J, L H S, E R S, G J P, K D-D, performed small molecule synthesis. G E G, P G, A J T, performed
248 cellular assay development and evaluation of molecules. GEG, ATF, AJT performed data analysis. ATF, GEG
249 performed figure preparation. H-T H, performed the E-STUB assay. K A D performed proteomics experiments
250 and data analysis. W S, E S F, F M F supervised the study, and performed funding acquisition. KAD and FMF
251 conceived the study. FMF wrote the manuscript with input and edits from all authors.

252

253 Conflict of Interest Statement:

254 K.A.D receives or has received consulting fees from Kronos Bio and Neomorph Inc. H.-T.H has no conflict of
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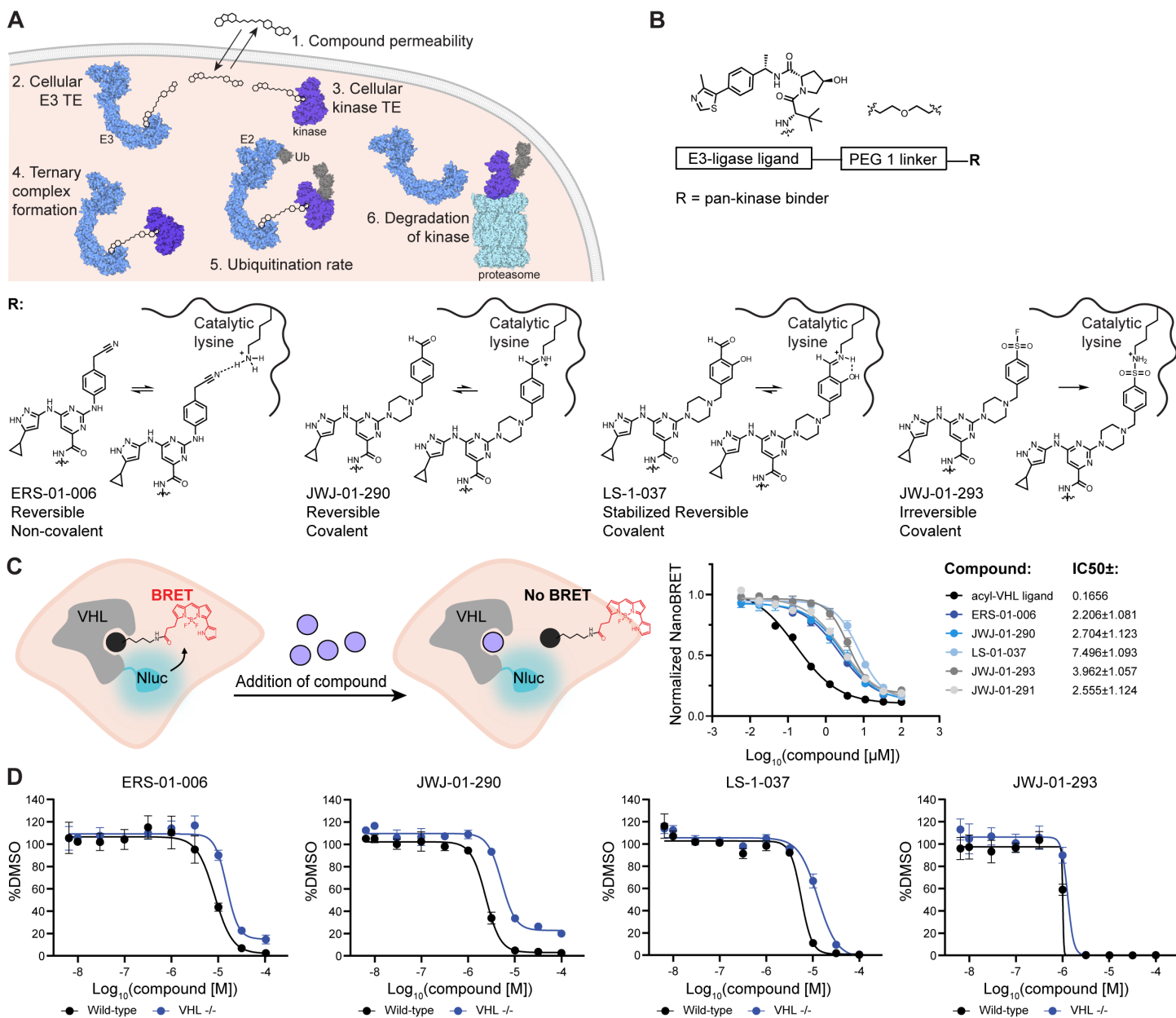
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Figure 1 | Overview of the Kinetic Scout Degradation Approach. A. Schematic showing variables that impact targeted protein degradation measured in this study. B. Design of kinetic scout degrader library. Pan-kinase binders with comparable kinome-wide selectivity profiles but distinct binding off-rates, mediated by reversible, reversible covalent and covalent interactions at the conserved lysine were employed. C. VHL NanoBRET target engagement assay to measure permeability of kinetic scout degrader library. HEK293 cells expression NanoLuc-VHL were treated with 1 μ M tracer and indicated concentration of kinetic scout degrader for 2 hrs. BRET signal was normalized to DMSO BRET signal. D. Viability assay in MOLT4 and MOLT4 VHL^{-/-} cells. Cells were treated with DMSO or indicated concentration of compound for 72 hr and luminescence was measured after addition of CellTiter-Glo reagents. C.-D. Data shown as the average of $n = 3$ replicates \pm standard deviation.

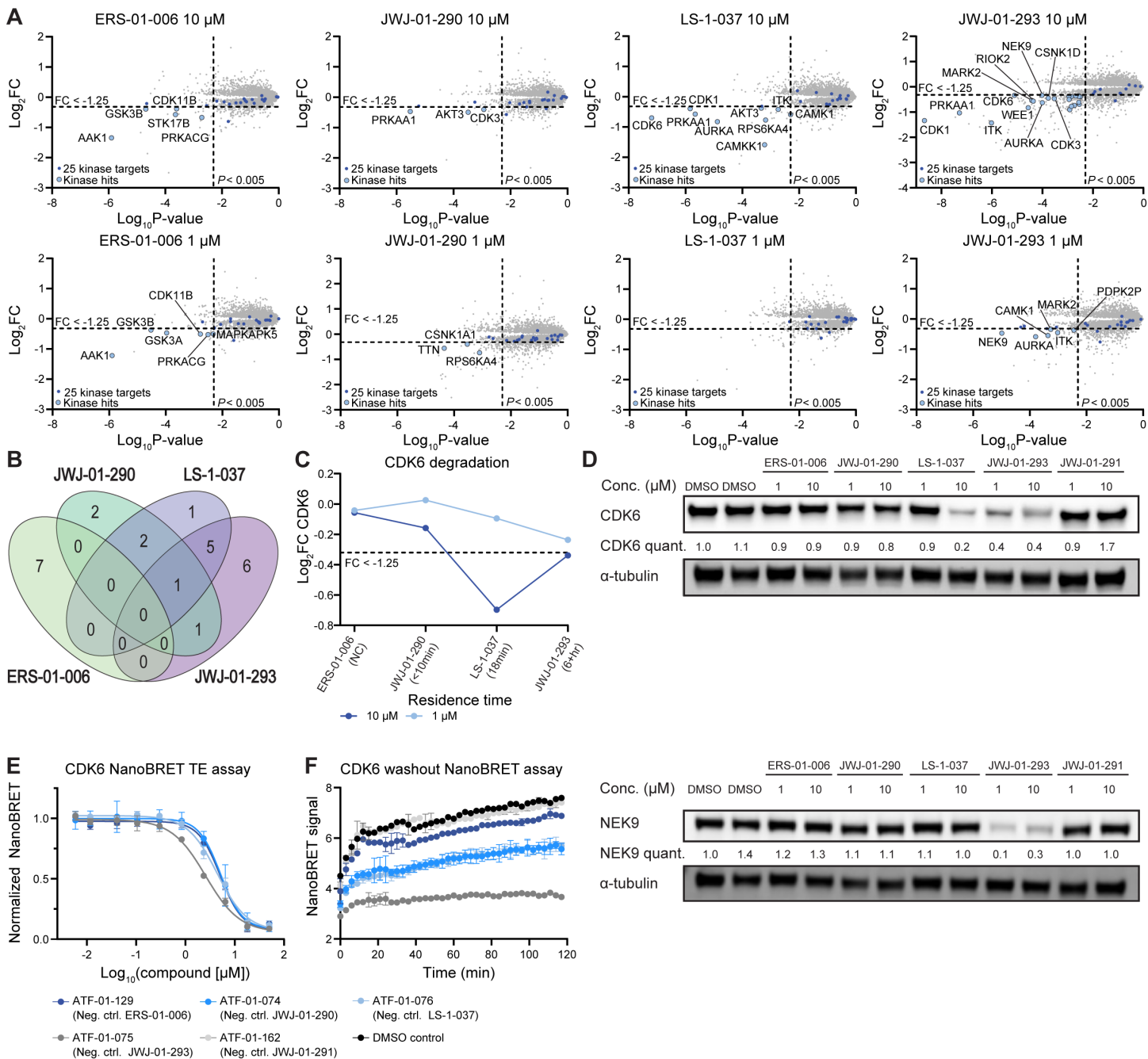
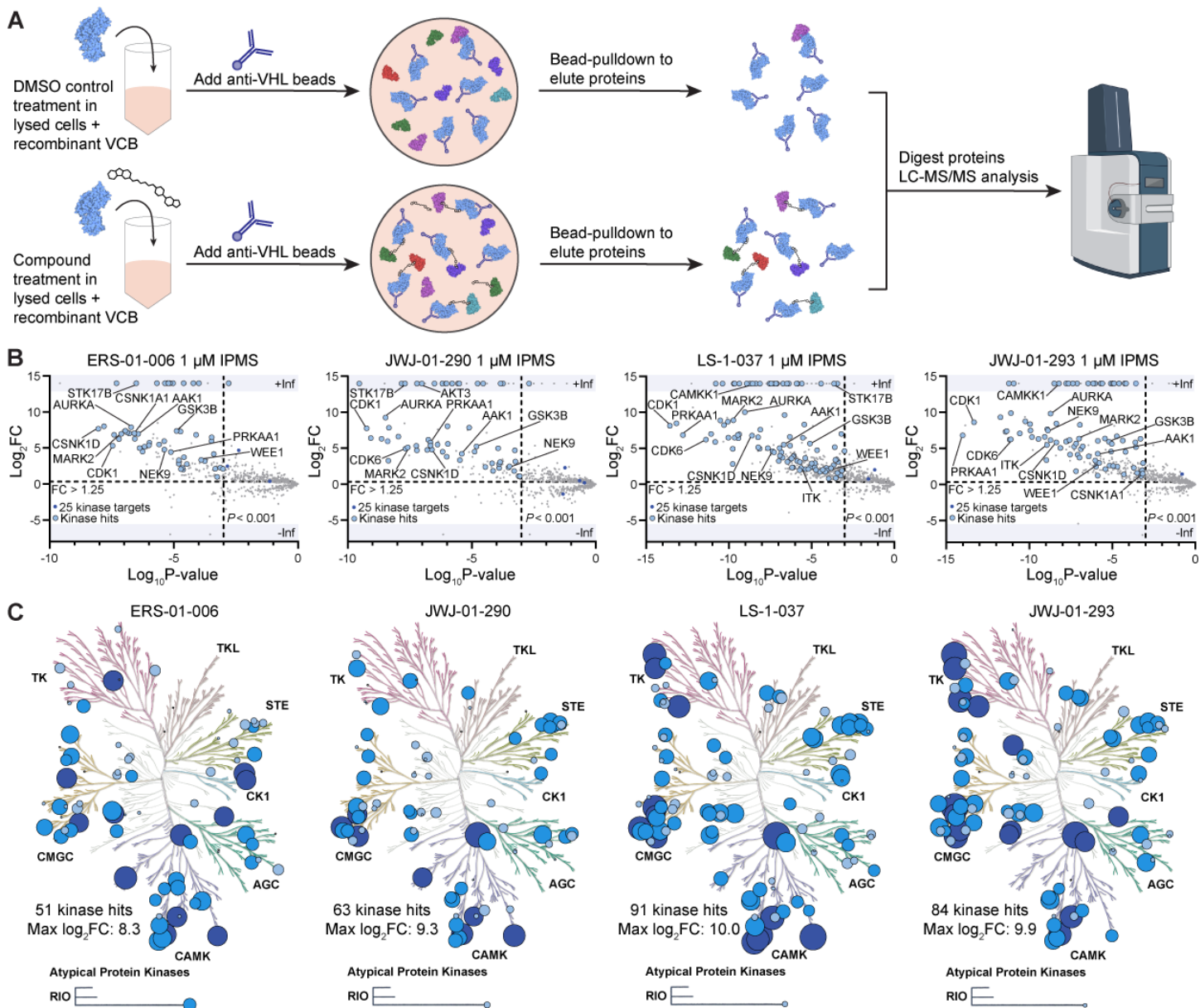
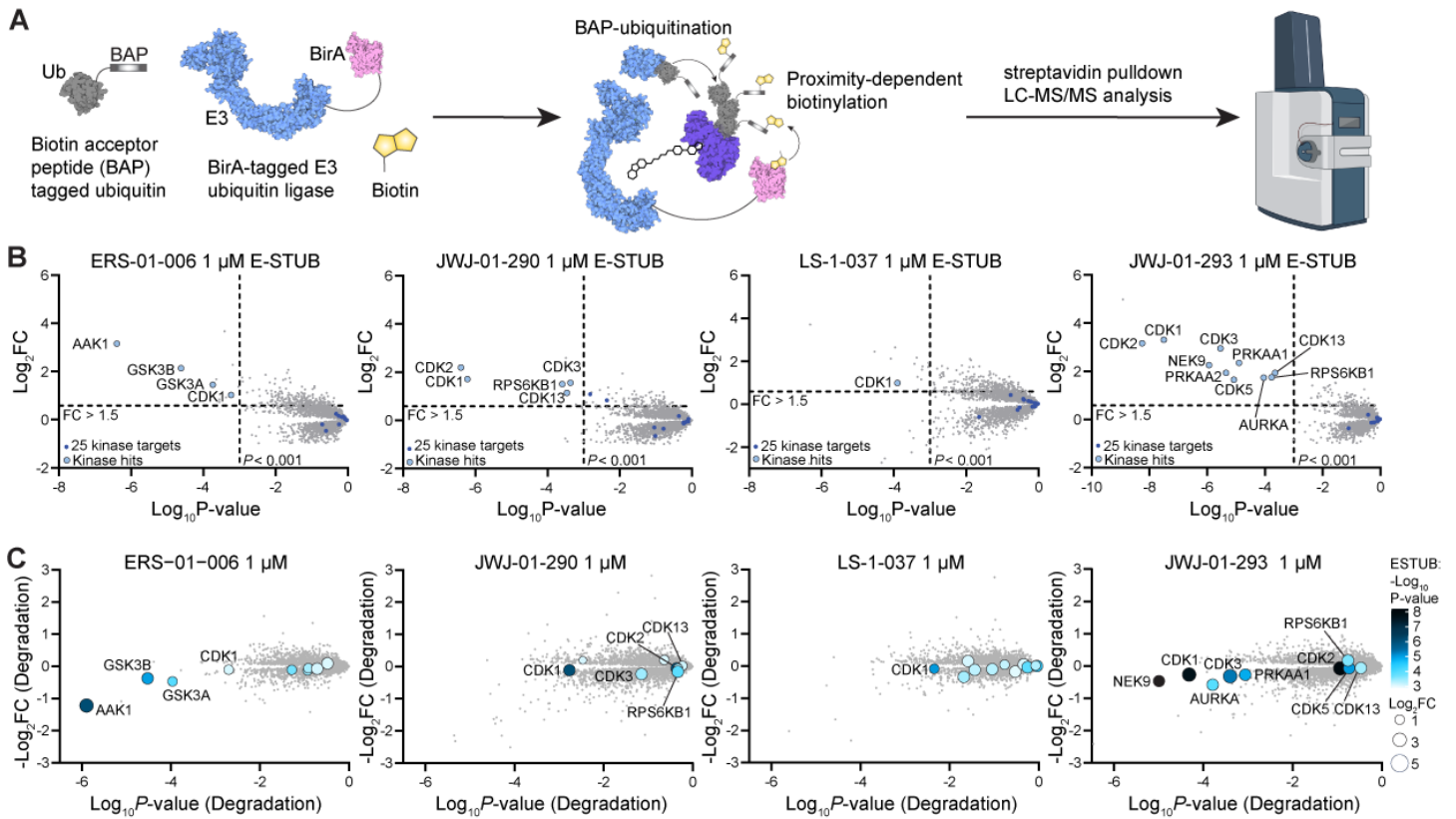


Figure 2 | Degraded target space strongly influenced by residence time of pan-kinase recruiter. A. Global proteomics analysis of MOLT4 cells treated with the indicated compound for 5 hrs. B. Venn diagram illustrating the overlap of degraded kinases by each compound. C. Influence of kinase binder residence time on CDK6 degradation. D. Immunoblot validating compound-specific CDK6 and NEK9 degradation. MOLT4 cells were treated for 6 hrs with the indicated compounds at the indicated concentrations. Data representative of $n = 3$ biological replicates. E. CDK6 NanoBRET assay showing minimal equilibrium IC_{50} differences between reversible kinetic scout degraders. HEK293 cells expressing CDK6-NanoLuc were treated with 0.5 μ M K-10 tracer and indicated compound at indicated concentrations for 2 hrs. The data was background corrected by subtracting BRET signal in the absence of tracer and then normalized to DMSO BRET signal. Data shown as the average of $n = 3$ replicates \pm standard deviation. F. CDK6 NanoBRET washout assay. Cells were treated with 10 μ M of the indicated compound for 2 hrs, then washed with 2 x opti-MEM+10% FBS, followed by 2 x opti-MEM. 0.5 μ M K-10 tracer was added and BRET signal read over a 2 hr time course.

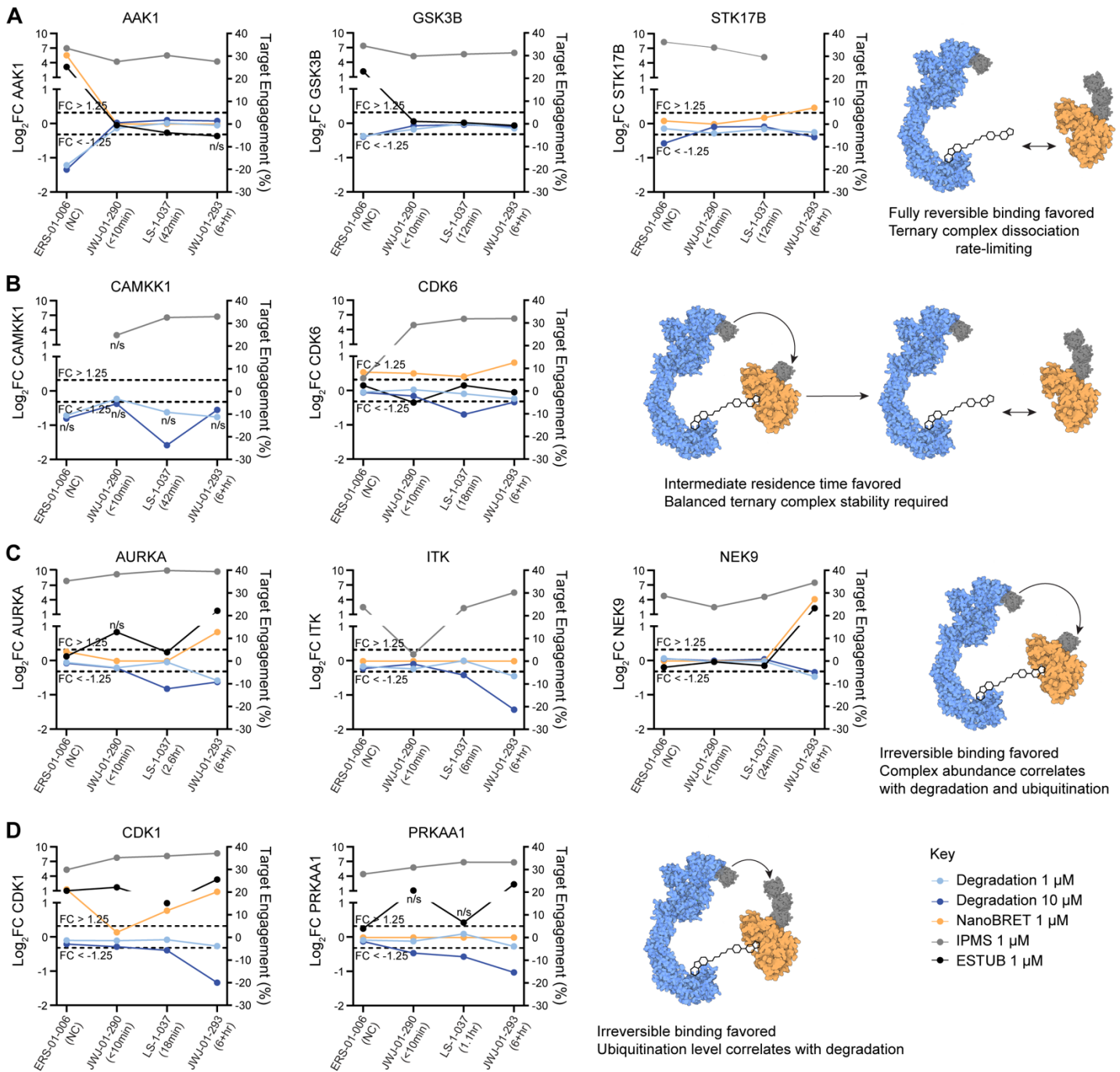
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360 Key: • log₂FC < 1.25 (does not meet FC cut-off) • log₂FC = 1.25-3.99 • log₂FC = 4.00-6.99 • log₂FC = 7.00-10.00
 361 **Figure 3 | Kinetic scout degraders form ternary complexes with VHL kinome-wide.** A. Schematic depicting
 362 the VHL IP/MS assay. B. Scatterplot demonstrating relative protein abundance following Flag-VHL enrichment
 363 from in-lysate treatment with 1 μ M of each molecule. Scatterplots display fold change in abundance to DMSO.
 364 Significant changes were assessed by moderated t-test as implemented in the limma package⁷⁸ with log₂FC
 365 shown on the y-axis and negative log₁₀P-value on the x-axis. C. Compound-dependent VHL IP/MS enrichment
 366 of kinases plotted on a kinome tree. Illustration reproduced courtesy of Cell Signaling Technology, Inc.
 367



368 **Figure 4 | E-STUB reveals rapidly ubiquitinated kinases across the proteome.** A. Schematic depicting the
 369 E-STUB assay. B. E-STUB data showing fold change in abundance of streptavidin-enriched proteins following
 370 1 hr compound treatment in 293T VHL^{-/-} cells expressing VHL-BirA and A3-ubiquitin. C. Comparison of target-
 371 dependent degradation and ubiquitination. E-STUB data showing fold change in abundance (represented by
 372 size of dots) and statistical significance (represented by saturation of color) of B is overlaid onto a volcano plot
 373 (from Fig. 2A) of the global proteomics analysis of MOLT4 cells treated with the indicated compound for 5 hrs.
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Figure 5 | Multiparameter analysis reveals drivers of residence-time based degradation outcomes.

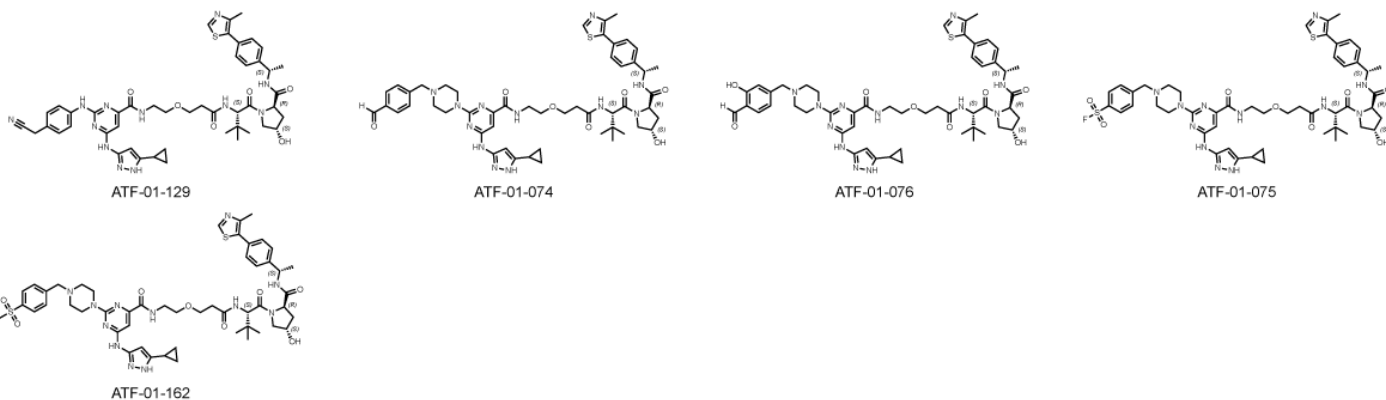
Multiparameter profiles for representative kinases that are preferentially degraded by degraders which

incorporate A. fully reversible binders. B. reversible covalent binders C.-D. irreversible covalent binders. K192

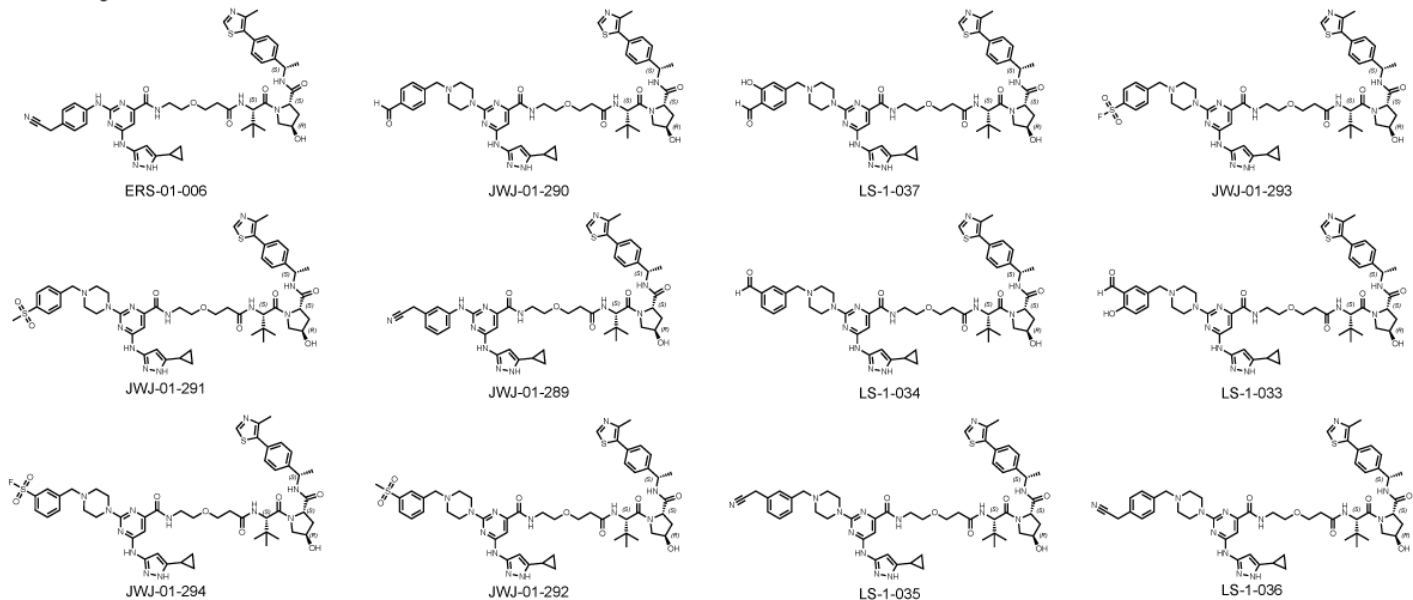
NanoBRET measurements used negative control compounds.

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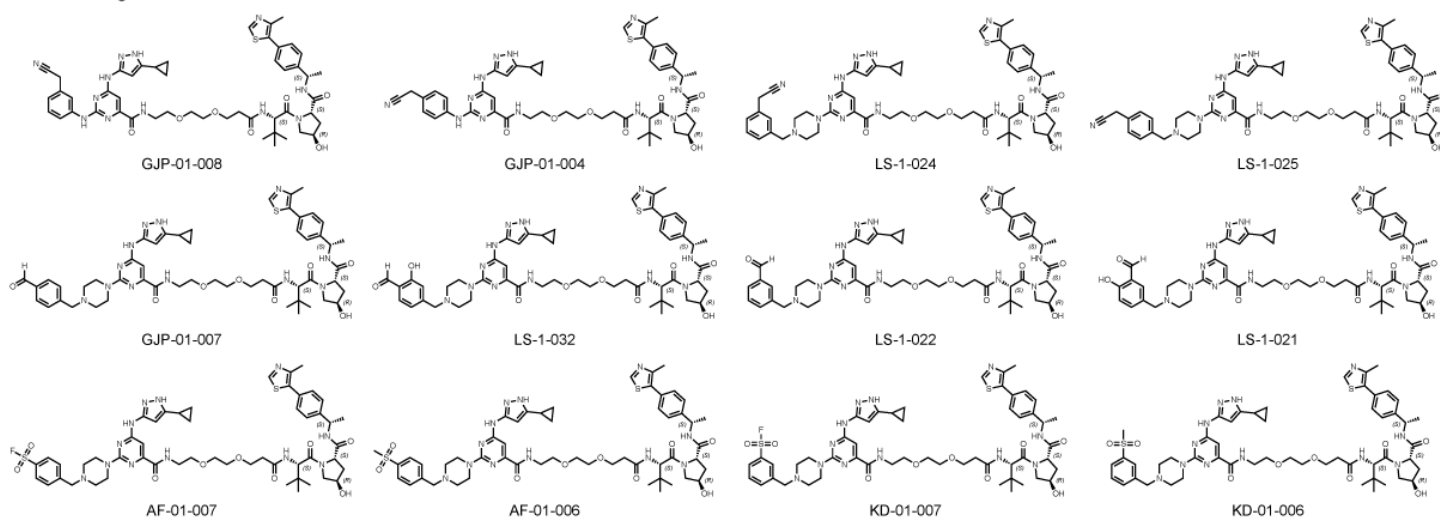
Negative controls - Inactive VHL ligand



PEG1 analogs

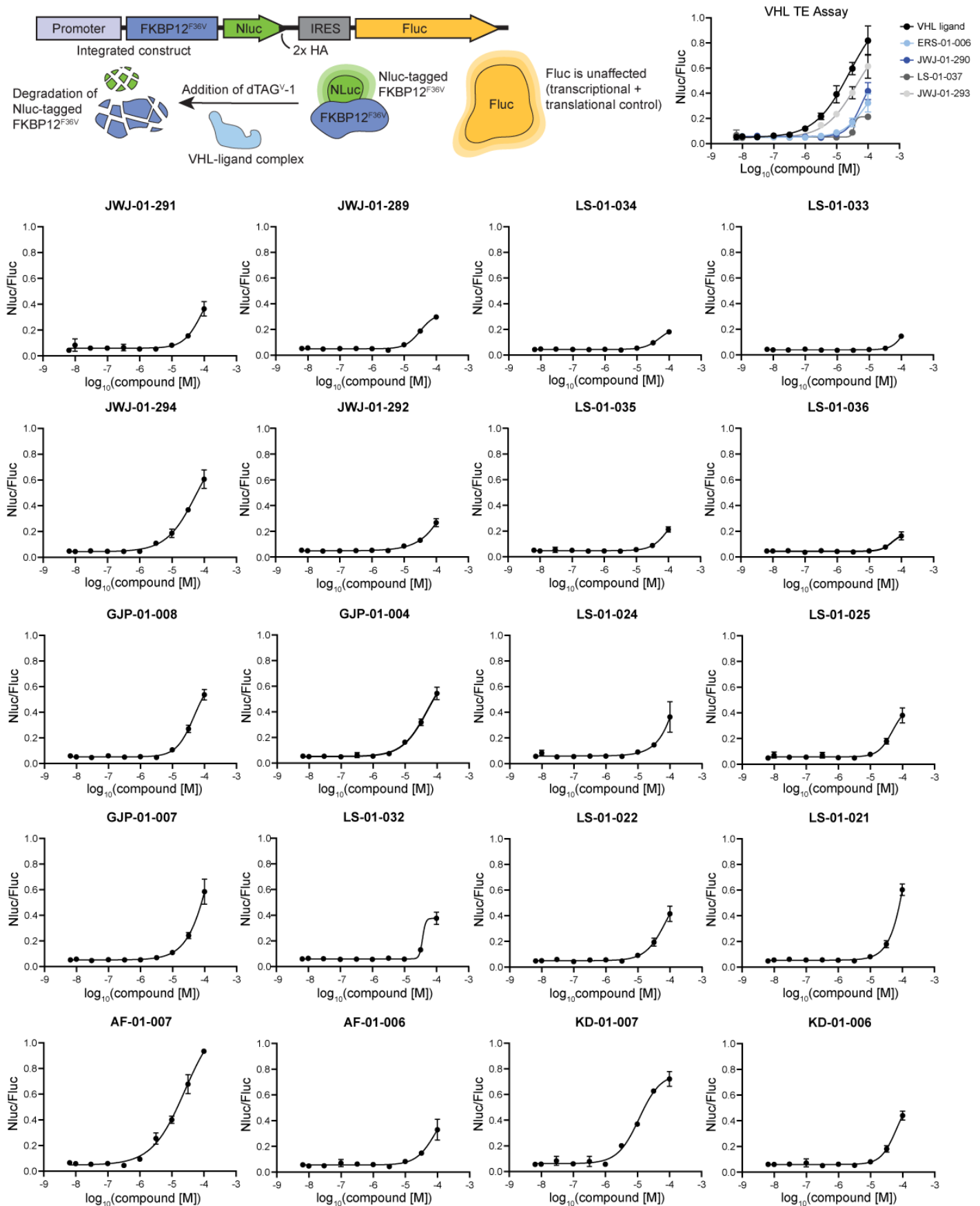


PEG2 analogs



Supporting Figure 1 | Chemical Structures of Kinetic Scout Degradер Library Compounds.

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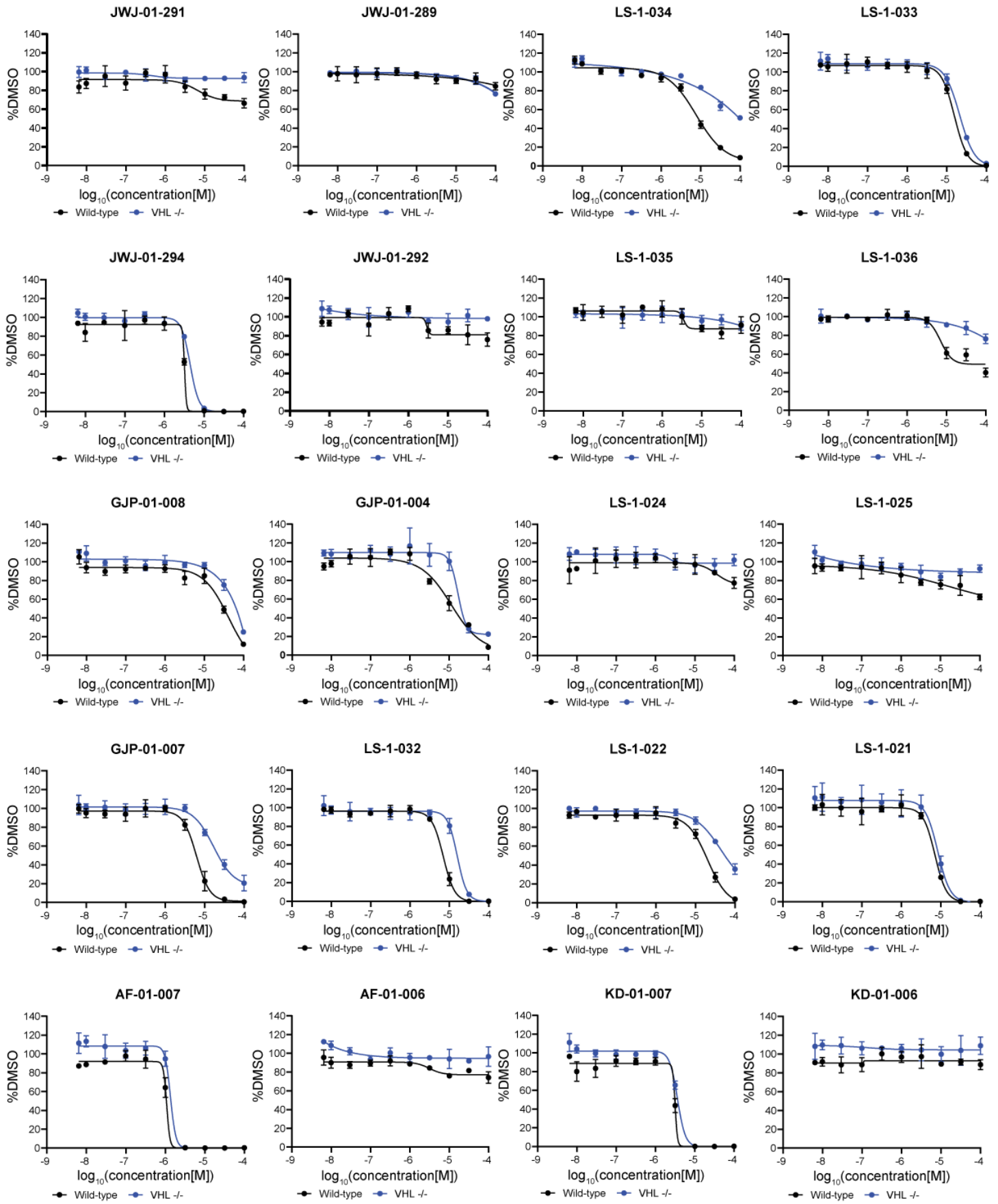


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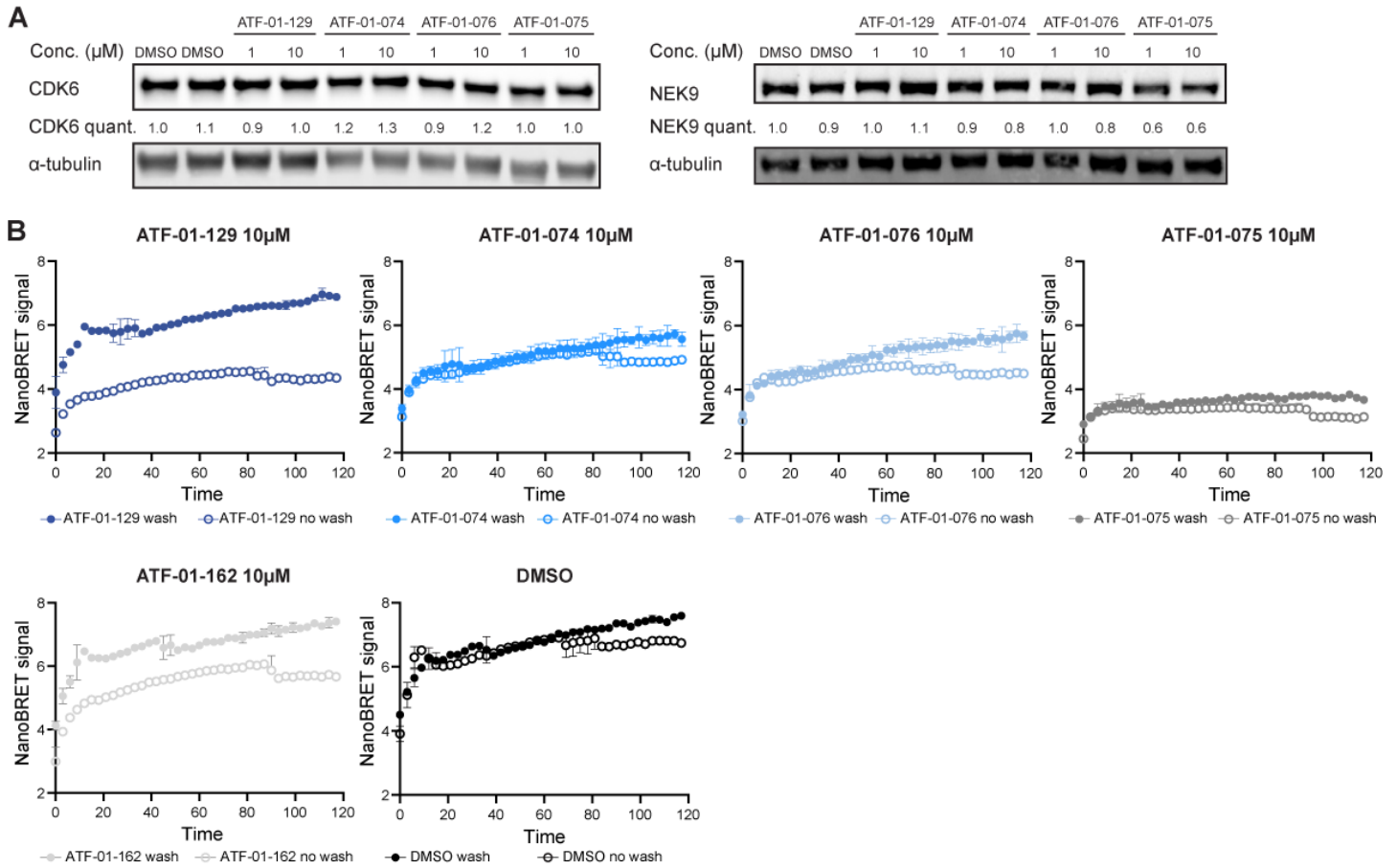
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Supporting Figure 2 | Cellular VHL Engagement of Kinetic Scout Degradable Library. VHL cellular target engagement assay measures competitive displacement of dTAG^{V-1} in cells. HEK293FT-FKBP12^{F36V}-NanoLuc cells were treated with 100 nM dTAG^{V-1} and indicated concentration of compound for 6 hr and luminescence was measured using a ClarioSTAR Plus microplate reader. Data shown as the average +/- S.D. of $n = 3$ replicates.

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391 **Supporting Figure 3 | VHL-dependent Cell Viability Effects of Kinetic Scout Degraders.** Viability assay in
392 MOLT4 and MOLT4 VHL^{-/-} cells. Cells were treated with DMSO or indicated concentration of compound for 72
393 hr and luminescence was measured after addition of CellTiter-Glo reagents. Data shown as the average +/- S.D.
394 of *n* = 3 replicates.
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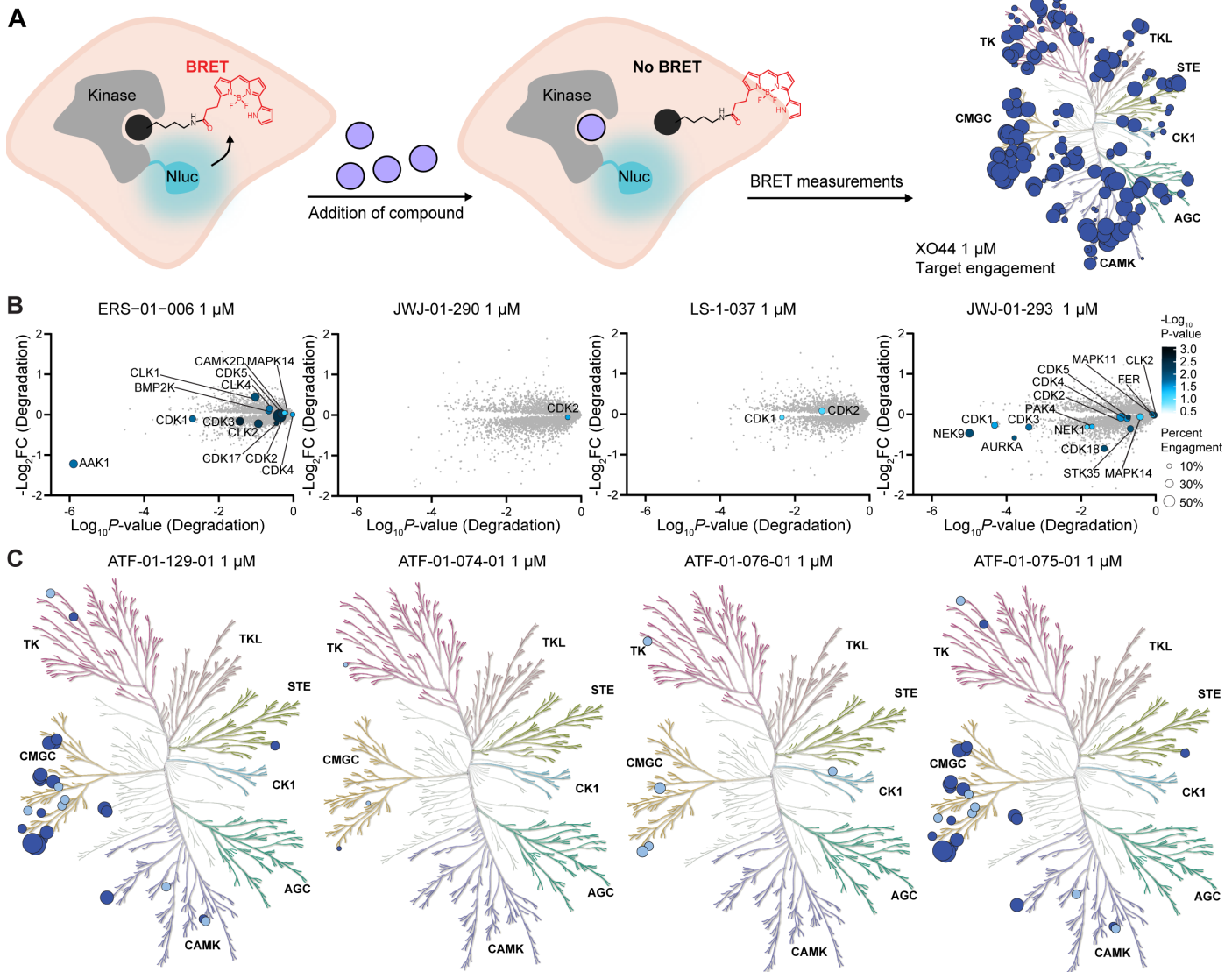
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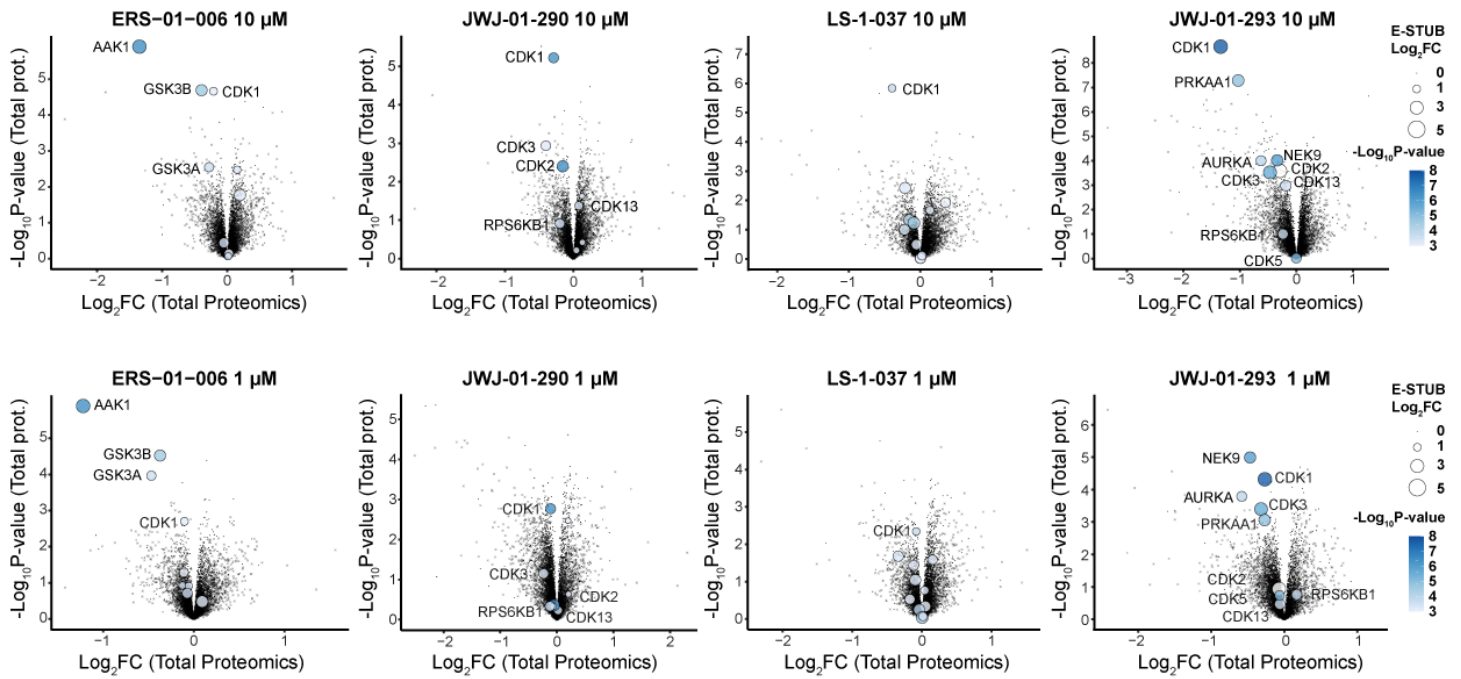
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Supporting Figure 4 | CDK6 NanoBRET washout experiment. A. Immunoblot validating no degradation of CDK6 and NEK9 with VHL inactive compounds. MOLT4 cells were treated for 6 hrs with the indicated compounds at the indicated concentrations. B. Cells expressing CDK6-NanoLuc construct were treated with DMSO or indicated compound for 2 hr and then media was replaced twice with Opti-MEM + 10% FBS and then twice with Opti-MEM for wells intended to be washed. 0.5 μM K-10 tracer and Complete Substrate Plus Inhibitor Solution was added then NanoBRET signal was measured. Data shown as the average \pm S.D. of n = 2 replicates.

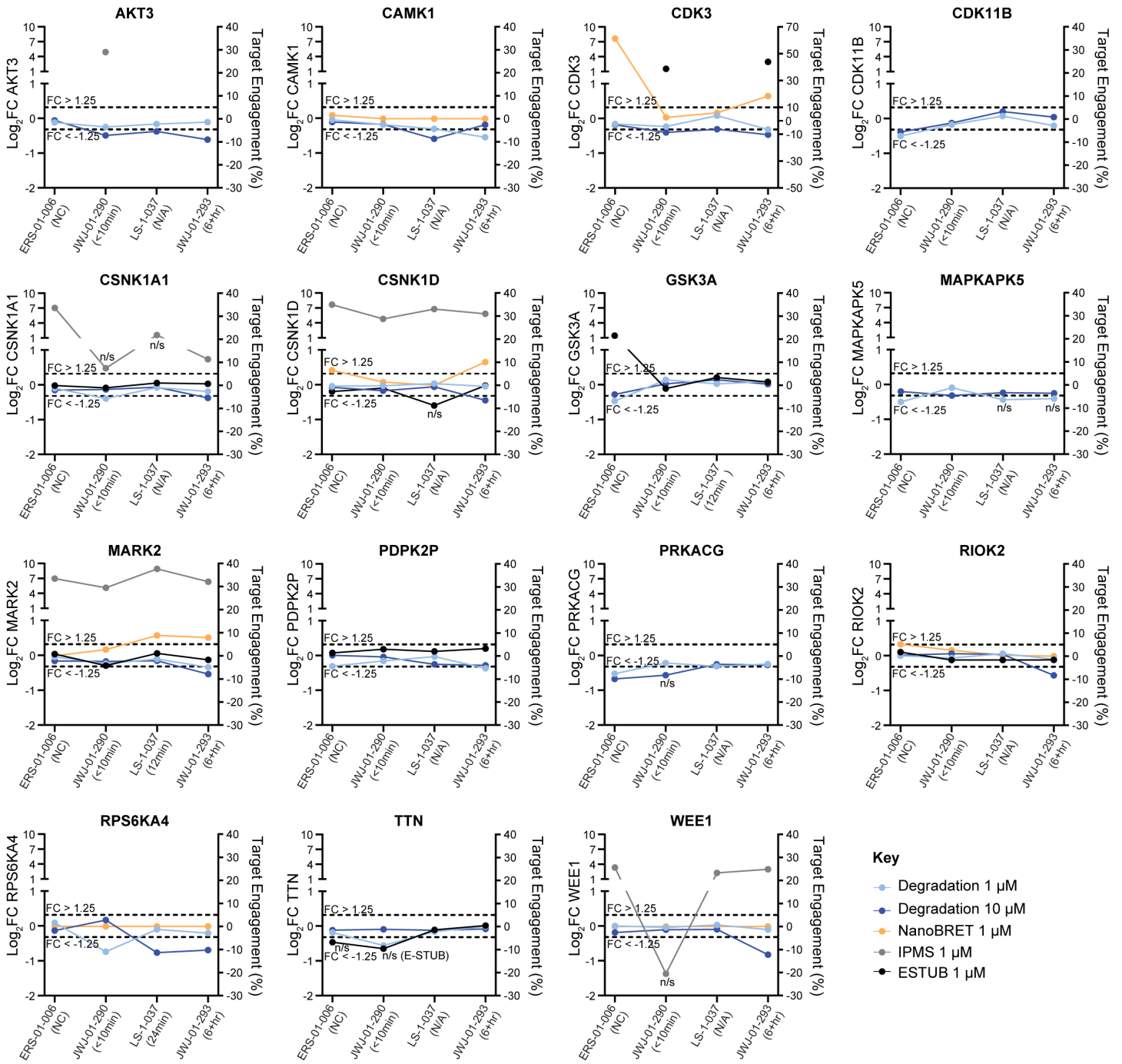


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 405 **Supporting Figure 5 | Equilibrium kinase target engagement is not a driver of degradation.** A. Schematic
 406 depicting the NanoBRET K192 profiling assay. B.-C. Live cell K192 NanoBRET assay used to determine percent
 407 engagement of kinases. HEK293 cells expressing kinase-NanoLuc were treated with K-10 tracer and 1 μ M
 408 compound. The percent engagement is the percent reduction of BRET signal after compound treatment
 409 compared to DMSO. Data shown as the average of $n = 3$ replicates. P -value was calculated using Student's t -
 410 test. B. Comparison of cellular target engagement and degradation. Proteomic data as in Fig. 2. Percent
 411 engagement data showing compound binding (represented by size of dots) and statistical significance
 412 (represented by saturation of color) of K192 NanoBRET assay is overlaid onto a volcano plot (from Fig. 2A) of
 413 the global proteomics analysis of MOLT4 cells treated with the indicated compound for 5 hrs. Percent
 414 engagement data was measured using negative control compounds. C. Percent target engagement plotted on
 415 KinMap. Illustration reproduced courtesy of Cell Signaling Technology, Inc.



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Supporting Figure 6 | Comparison of degraded proteins and their enrichment in the E-STUB assay. E-STUB data showing fold change in abundance (represented by size of dots) and statistical significance (represented by saturation of color) of streptavidin-enriched proteins following 1 hr compound treatment in 293T $\text{VHL}^{-/-}$ cells expressing VHL-BirA and A3-ubiquitin . The data is overlaid onto a volcano plot (from Fig. 2A) of the global proteomics analysis of MOLT4 cells treated with the indicated compound for 5 hrs. Proteins identified in global proteomics that are not detected by E-STUB are shown as a cross.



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 425 **Supporting Figure 7 | Multiparameter analysis reveals drivers of residence-time based degradation**
 426 **outcomes for additional degraded kinases.** Multiparameter profiles for kinases not shown in Figure 5. K192
 427 NanoBRET measurements used negative control compounds.
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