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2	Identification of highly penetrant Rb-related synthetic lethal
3	interactions in triple negative breast cancer
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34 Abstract

35 Although defects in the *RB1* tumour suppressor are one of the more common 36 driver alterations found in triple negative breast cancer (TNBC), therapeutic approaches that exploit this have not been identified. By integrating molecular 37 38 profiling data with data from multiple genetic perturbation screens, we 39 identified candidate synthetic lethal (SL) interactions associated with RB1 40 defects in TNBC. We refined this analysis by identifying the highly penetrant effects, reasoning that these would be more robust in the face of molecular 41 42 heterogeneity and would represent more promising therapeutic targets. A 43 significant proportion of the highly penetrant RB1 SL effects involved proteins 44 closely associated with RB1 function, suggesting that this might be a defining 45 characteristic. These included nuclear pore complex components associated with the MAD2 spindle checkpoint protein, the kinase and bromodomain 46 containing transcription factor TAF1 and multiple components of the SCF^{SKP} 47 Cullin F box containing complex. Small molecule inhibition of SCF^{SKP} elicited 48 an increase in p27^{Kip} levels, providing a mechanistic rationale for RB1 SL. 49 Transcript expression of SKP2, a SCF^{SKP} component, was elevated in *RB1* 50 51 defective TNBCs, suggesting that in these tumours, SKP2 activity might buffer 52 the effects of *RB1* dysfunction.

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54 Keywords: Rb, synthetic lethality, triple negative breast cancer, SKP2, TAF1,

penetrance

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58 Introduction

59 Patients who develop triple negative breast cancer (TNBC), i.e. those breast 60 cancers that lack amplification of the ERBB2 gene as well as expression of both the estrogen and progesterone receptors, tend to have a relatively poor 61 62 prognosis and represent a significant area of unmet clinical need, where novel therapeutic approaches are acutely needed (recently reviewed in ⁵). Although 63 64 some targeted approaches have been proposed for molecularly defined subsets of TNBC patients, for most, classical chemotherapy regimens still 65 66 represent the mainstay of treatment, making the requirement to identify novel 67 targets in this disease critical. One approach to this problem has been to 68 define the molecular composition of TNBCs and then to use this information to help identify therapeutic vulnerabilities that might operate in the disease. 69 Already, the delineation of genomic, transcriptomic and proteomic profiles of 70 71 tumours has identified a series of distinct molecular subtypes of TNBC, as 72 well as identifying likely cancer driver gene mutations that operate in the disease⁵¹. 73

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75 One of the more frequent driver alterations in TNBCs involves deleterious 76 mutations (e.g. truncating mutation, gene deletions etc.) in the 77 Retinoblastoma tumour suppressor gene (RB1 aka Rb). In non-tumour cells, 78 Rb's canonical role is in cell cycle progression, a function mediated in part by the repressive effect Rb has on the E2F family of transcription factors²⁹. A 79 80 somewhat reductionist model of Rb's role in tumour suppression, suggests 81 that loss of Rb's E2F repressive function allows cells to prematurely transition 82 through the G₁ cell cycle checkpoint; it also seems likely that loss of Rb 83 function in breast cancer also influences additional processes that contribute 84 to the development of the disease including the differentiation of stem and 85 progenitor cells and the transition of cells from an epithelial to a mesenchymal phenotype²⁹. 86

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Commensurate with its key role in cell cycle control, genomic alterations in the *RB1* gene are relatively common in TNBCs^{31, 32, 49} as well as in a series of other malignancies^{17, 27, 60}. In TNBC, loss of Rb protein expression is found in > 40 % of cases^{58, 61} (and reviewed in³⁰). Although patients with Rb defective tumours (as defined by immunohistochemistry and/or gene expression) tend to have a relatively favorable response to classical chemotherapy^{20, 26, 64} many either fail to respond to therapy or later relapse with therapy resistant disease, suggesting that novel therapeutic approaches are required to target this patient subset.

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98 One approach to identifying novel therapeutic targets that could be exploited 99 in patients with specific tumour suppressor gene defects has been to identify 100 synthetic lethal interactions associated with these genes. For example, the 101 identification of synthetic lethal interactions between BRCA1 or BRCA2 102 tumour suppressor genes and inhibition of the PARP1 DNA repair protein, has driven the clinical development and approval for use of PARP inhibitors for the 103 treatment of cancer³⁹. One notable feature of the BRCA/PARP1 synthetic 104 lethal effect, that contributes to its translational value, is that it is highly 105 penetrant²: i.e. in otherwise molecularly diverse pre-clinical models, and 106 107 cancer patients, the presence of the predictive biomarker, in this case BRCA1 108 or BRCA2 mutation, more often than not predicts a profound anti-tumour cell 109 response to a PARP inhibitor. This highly penetrant nature suggests that this 110 particular synthetic lethal effect is robust in the face of the molecular 111 heterogeneity that exists between different human cancers. Here, we describe 112 efforts to identify highly penetrant synthetic lethal effects associated with deleterious Rb alterations in TNBC; we reasoned that those with greatest 113 114 penetrance will be more likely to operate in the diverse molecular contexts within the TNBC subtype and thus represent more promising therapeutic 115 targets. Although genes such as $TSC2^{54}$ and elements of the Dicer pathway⁴⁴ 116 have been shown to be synthetic lethal with Rb defects, as far as we are 117 aware, the penetrance of these effects, or whether these operate in models of 118 119 TNBC, has not as yet been assessed. The availability of several, large-scale, shRNA and siRNA screens^{8, 9, 42, 62}, conducted in multiple tumour cell lines, 120 some of which are derived from TNBCs, now make a detailed identification of 121 122 highly penetrant RB1-related synthetic lethal effects now possible. For this 123 reason, we describe here a detailed molecular analysis of Rb status in tumour 124 cell lines derived from TNBC. We then designed a straightforward data

125 analysis pipeline that allowed us to use this Rb annotation to interrogate both in-house and publically available large-scale, shRNA and siRNA screens to 126 127 identify candidate Rb-related synthetic lethal effects. Included within this data 128 analysis pipeline, we included an estimate of penetrance. In triaging the 129 candidate Rb-synthetic lethal effects that operated in TNBC tumour cells to identify those with greatest penetrance, we identified a series of 130 131 pharmacologically tractable effects, one of which, SKP2, we validated using 132 both genetic and pharmacological methods. We also noted that a significant 133 proportion of the highly penetrant Rb SL effects in TNBC involved proteins 134 closely associated with Rb function, suggesting that this might be a defining 135 characteristic.

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137 **Results**

138 Annotation of Rb status in TNBC tumour cell lines

139 To identify highly robust synthetic lethal effects associated with Rb defects in TNBC, we classified a molecularly diverse panel of TNBC tumour cell lines 140 (TCLs) according to Rb status and then used this Rb classification to 141 142 interrogate publically available genetic screen data using a data analysis 143 pipeline (described later) that identified highly penetrant synthetic lethal 144 effects. To do this, we first classified TNBC tumour cell lines using a 145 combination of genomic, transcriptomic and proteomic data to identify those 146 with Rb genetic or epigenetic defects. We used western blotting to assess Rb protein expression in 30 breast TCLs, including 16 TNBC (Figure 1A,B). In 147 148 this analysis, we found that TCLs with deleterious mutations in the RB1 gene 149 (BT549, RB1 c.265_607del343, MDAMB468, RB1 c.265_2787del2523, DU4475, RB1 c.1_2787del2787, MDAMB436, RB1 c.607_608ins107) lacked 150 151 Rb protein expression, suggesting that the mutational status of RB1 152 correlated with protein expression (Figure 1A,B). We also found that SUM149 cells exhibited low Rb protein expression, an observation we found to be 153 154 consistent with reduced RB1 mRNA levels (Supplementary Data 1). To 155 assess Rb protein status by orthogonal means, we compared our western blot data with publically available mass spectrometry data describing the 156 proteomic profiles of 16 TNBC tumour cell lines³³. Using average intensity-157

158 based absolute protein abundance (iBAQ) data for Rb from mass spectrometry profiling³³ (Supplementary Data 1), we found that TNBC TCLs 159 160 classified by western blotting as being Rb defective exhibited no Rb peptides 161 (MDAMB468, MDAMB436, HCC1937, DU4475, BT549) when compared to 162 those tumour cell lines we had classified as Rb proficient (Figure 1C, p=0.0002), giving us some confidence in our classification. Examination of 163 transcriptomic profiles of TNBC TCLs³ revealed that TCLs with reduced levels 164 165 of RB1 mRNA exhibited low Rb protein expression (Figure 1D, p=0.0075), 166 suggesting that *RB1* mRNA expression levels could be used as a reasonable 167 proxy for protein expression. Taking proteomic, genomic and transcriptomic data into consideration (Figure 1E), we then classified a total of 42 TNBC 168 169 TCLs according to Rb status, identifying 12 with one or more defects in Rb ("Rb defective" e.g. low protein expression, truncating mutation/gene deletion, 170 171 reduced mRNA levels: BT549, HCC1937, DU4475, MDAMB436, MDAMB468, CAL148, HDQP1, MB157, SUM149, HCC1187, HCC3153 and CAL851) and 172 173 30 TNBC TCLs as not exhibiting such defects ("Rb not altered" TCLs).

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175 To further assess the validity of our Rb classification, we assessed the 176 transcriptomic profiles of Rb defective TNBC TCLs to assess whether these 177 reflected loss of Rb function (Figure 1F). Using the Rb defective and Rb not altered classification described above for 42 TNBC TCLs, we identified 839 178 179 differentially expressed genes (452 with reduced expression in Rb defective TCLs, 387 with elevated expression (p<0.05, Limma analysis, Supplementary 180 181 Data 2)). As expected, we found RB1 itself to be the fifth most down-regulated gene in the Rb defective TCLs compared to Rb not altered TCLs (log fold 182 change of -2.4, p value = 1.6×10^{-6} (Limma analysis); Supplementary Data 2 183 and Figure 1F). We also found that Rb defective TCLs exhibited elevated 184 expression of cyclins associated with G₁ checkpoint control and S phase 185 progression (Cyclin E1 (p=0.03), E2 (p=0.02), CDKN2A (p16; p=0.002)), as 186 well as elevated expression of the Rb regulated E2F3 transcription factor and 187 its binding partner TFDP1 (Figure 1F (p=0.005 and p=0.0001, respectively)). 188 Using the ENRICHR annotation tool¹² to identify pathways rather than 189 individual genes that were differentially expressed in the Rb defective group, 190 191 we found that pathways related to Rb and G₁ to S cell cycle control to be

192 among the most significantly dysregulated in Rb defective TNBC TCLs, including "Retinoblastoma (RB) in Cancer WP2446" $p=1.7 \times 10^{-11}$ and "G₁ to S 193 cell cycle control WP45" p=1.4 x 10^{-9} , both of which included genes such as 194 MCM4,6 and 7, TFDP1, CCNE1, CCNE2, CHEK1, E2F3, and RBP1. Using 195 196 the same annotation tool we also searched for the key transcription factors that regulated genes that were differentially expressed in Rb defective TNBC 197 198 TCLs. We found that a significant proportion of the differentially expressed 199 genes in Rb defective TNBC TCLs were targets of Rb-related E2F-family transcription factors including E2F7, E2F4 and E2F1 (e.g. p =0.001; 0.005; 200 9.3×10^{-7} for E2F7; E2F4 and E2F1 respectively). 201

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203 To compare the observations made in TCLs with TNBC tumours, we used the 204 same approach of exploiting RB1 gene mutation/copy number status and RB1 205 mRNA expression profiles to classify 140 TCGA triple negative breast tumors¹⁰ according to Rb status; this approach identified 48 Rb defective 206 207 TNBC tumours and 92 where Rb was not altered. Assessing the transcriptomic profiles of these TNBCs we again found that genes associated 208 209 with Rb and Rb-related G₁ to S cell cycle control were frequently dysregulated in Rb defective TNBCs including CDKN2A, TFDP1, CCNE1, CCNE2, E2F3, 210 CHEK1 and DYRK1A, a recently identified RB1 synthetic lethal gene⁹ (Figure 211 1G, Supplementary Data 3), consistent with the observations made in TNBC 212 TCLs. We also applied the same approach to classify 182 Metabric TNBC 213 tumours⁴⁶ according to Rb status; this approach identified 55 Rb defective 214 215 and 132 Rb not altered TNBC samples. Assessing the transcriptomic profiles 216 of these TNBCs we again found that genes associated with Rb and Rb-217 related G₁ to S cell cycle control were frequently dysregulated in Rb defective TNBCs (Figure 1H, Supplementary Data 4). These global transcriptional 218 219 patterns in Rb defective TNBC TCLs and human tumours suggested that our 220 Rb classification approach was somewhat valid.

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A pipeline for the identification of highly-penetrant Rb synthetic lethal effects in TNBC

225 Using the Rb classification of TNBC TCLs described above, we re-analysed 226 publically available genetic screen data (e.g. genome-wide shRNA screen data^{42, 62}, gene subset shRNA screen data⁴³ or gene subset siRNA data^{8, 9}) 227 from TNBC TCLs, using a series of iterative data processing steps designed 228 229 to identify highly-penetrant synthetic lethal effects (Figure 2A). In summary, 230 these steps involved: (i) step one - identification of candidate synthetic lethal effects: using shRNA screen data⁴² for 12 Rb defective TNBC cell 231 lines and 30 Rb not altered cell lines we used the siMEM algorithm⁴² to 232 233 identify those genes whose inhibition appeared to target the Rb defective 234 models to a greater extent than Rb proficient TNBC TCLs (p<0.05). (ii) steps 235 two and three – apply Z score thresholds to identify profound cell 236 inhibitory effects: although step one allowed us to identify genes whose 237 inhibition selectively targeted Rb defective TCLs to a greater extent than Rb 238 not altered TCLs, we reasoned that the scale of cell inhibition in these two 239 TCL cohorts might also be critical in identifying the most suitable therapeutic 240 targets. For example, for the purposes of identifying novel therapeutic targets 241 we were less interested in genes whose inhibition profoundly inhibited both 242 Rb defective and Rb not altered groups, even if the Rb defective models exhibited significantly greater sensitivity; we assumed that inhibition of these 243 244 targets would likely cause significant normal cell toxicity. Similarly, we also 245 assumed that candidate synthetic lethal effects that did not elicit profound cell 246 inhibitory effects in Rb defective TCLs would be less likely to elicit profound 247 anti-cancer therapeutic effects when exploited clinically. For these reasons, 248 we triaged the list of candidate synthetic lethal effects identified in step 1 to 249 remove from further analysis: (a) those genes that appeared to be synthetic 250 lethal with Rb but whose targeting elicited profound cell inhibition in Rb not 251 altered TCLs; and (b) those genes that appeared to be synthetic lethal with 252 Rb but whose targeting did not elicit profound cell inhibitory effects in Rb 253 defective TCLs. To do this we used shRNA/siRNA Z score thresholds to 254 estimate the effect of each RNAi reagent; (iii) steps four and five - identify 255 effects with greatest penetrance: We reasoned that the most clinically 256 effective synthetic lethal targets are likely to be those that have complete or

257 highly penetrant effects, i.e. the presence of the predictive biomarker (in this case an Rb defect) is more often than not associated with profound sensitivity 258 259 to target inhibition. For this reason, we finally triaged synthetic lethal effects by 260 calculating synthetic lethal penetrance scores; in this case, synthetic lethal 261 penetrance (SLP) for each synthetic lethal effect was calculated as the 262 percentage of Rb defective TCLs that exhibited a Z score of less than -1 when 263 targeted with shRNA. We also calculated synthetic lethal coverage (SLC) scores, i.e. the percentage of TNBC TCLs that were sensitive to shRNA that 264 265 were Rb defective, to estimate the specificity of synthetic lethal effects for Rb 266 defects.

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To illustrate the concepts of penetrance and coverage as applied to functional 268 genomic screens in tumour cell lines, we examined a well-characterised and 269 270 therapeutically actionable oncogene addiction effect that operates in breast 271 cancer, namely that associated with amplification of the epidermal growth 272 factor receptor oncogene, ERBB2. Amplification and overexpression of ERBB2 is used clinically to stratify breast cancer patients for treatment with 273 ERBB2-targeting agents such as the monoclonal antibody trastuzumab⁵⁹. In 274 examining the Colt2/Marcotte et al shRNA dataset (78,432 shRNAs targeting 275 16,056 genes in 77 breast TCLs representing TNBC and non-TNBC 276 subtypes⁴², we found that, as shown before⁴², that shRNA targeting of *ERBB*2 277 selectively targeted ERBB2 amplified breast TCLs (siMEM, p<0.0001), elicited 278 minimal inhibition in ERBB2 non-amplified TCLs (median Z -1), profound 279 inhibition in ERBB2 amplified TCLs (median Z -3) and had a SLP 280 (penetrance) score of 80% and a SLC (coverage) score of 28% (Figure 2B); 281 282 the high penetrance score in this case, taken together with ERBB2 fulfilling 283 the other selection criteria, reconfirms *ERBB2* as a suitable therapeutic target in appropriately stratified breast cancer patients. 284

To identify Rb synthetic lethal effects, we first used genome-wide shRNA data 287 from 42 TNBC TCLs described in the Colt2/Marcotte et al study⁴². In this 288 dataset, the combined effects of multiple shRNAs targeting a single gene are 289 described as Z normalised Gene Activity Ranking Profile (zGARP) scores⁴¹. 290 291 We re-analysed this shRNA screen data using the siMEM mixed effect linear 292 model method (see step one, above), which identifies synthetic lethal effects 293 by estimating the depletion or "drop-out" rate of individual shRNAs in cohorts of cell lines classified according to a molecular feature⁴², in this case Rb 294 deficiency. Using the siMEM approach with our Rb classification of 42 TNBC 295 296 TCLs, we identified 1065 Rb-specific dependencies (p<0.05, siMEM, Figure 297 2C): 437 genes where shRNA preferentially inhibited Rb defective TNBC (i.e. Rb synthetic lethal effects) and 628 genes where shRNAs preferentially 298 299 targeted Rb proficient TNBC TCLs (Supplementary Data 5). Amongst these, we noted that shRNAs targeting CDK4, CDK6 or the CDK4,6 cyclin partner 300 301 gene Cyclin D1 (CCND1) preferentially inhibited Rb proficient TNBC TCLs (Figure 2C), consistent with the hypothesis that inhibition of CDK4,6 activity 302 303 restores cell cycle control in Rb proficient TNBC tumour cells and elicits cell inhibition¹. In terms of identifying Rb dependencies, we noted that the siMEM 304 305 analysis of the Colt2 dataset identified shRNA targeting the E2F family transcription factor, E2F3, as being one of the most significant Rb synthetic 306 307 lethal effects (Figure 2C). These Rb-related observations thus gave us some confidence in the approach. We also carried out similar analyses in 308 siRNA/shRNA datasets that included TNBC TCLs, from other sources: the 309 DRIVE dataset⁴³, the Achilles dataset⁶² and the ICR-Intercell dataset^{8, 9} and 310 provide the lists of Rb dependencies identified in these datasets in 311 312 Supplementary Data 6, 7, 8, respectively.

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Application of Z score thresholds identifies profound cell inhibitory effects

317 As described above, we were interested in identifying Rb synthetic lethal 318 effects that had profound effects in Rb defective TNBCs but had minimal 319 effects in cells without Rb defects. Although approaches such as siMEM are 320 useful in identifying putative vulnerabilities, they often do not, when used in 321 isolation, identify synthetic lethal effects with these favoured characteristics. 322 For example, Figure 2D and 2E illustrate a pair of p<0.05 Rb dependencies 323 identified by siMEM analysis (Step One), that had either profound cell inhibitory effects in both Rb not altered as well as Rb defective TNBC TCLs 324 e.g. *PSMD1* (Figure 2D, siMEM $p = 3 \times 10^{-5}$, median Z in Rb defective of -6, 325 median Z in Rb not altered of -4) or relative paucity of profound cell inhibition 326 effects in the Rb defective TCL cohort, e.g. KLF16 (Figure 2E, siMEM p = 327 0.0002, median Z in Rb defective of -0.8). To eliminate such effects from 328 329 further study, we applied a pragmatic approach that removed from further 330 assessment p<0.05 synthetic lethal effects where the median zGARP score in 331 the Rb defective TCLs was >-1 (i.e. effects where profound cell inhibition in 332 Rb TCLs not observed) and those effects where median zGARP score in Rb 333 proficient effects was <-2 (i.e. dependencies which still elicited profound cell 334 inhibition in Rb proficient TCLs); three examples that fulfilled these criteria, GPS1, SNRPF and SNW1, are shown in Figure 2F,G,H. This triage step 335 336 identified 122 Rb synthetic lethal effects in the Colt2 dataset that fulfilled these criteria (Supplementary Data 9). Similarly, triaged dependencies were 337 identified in the DRIVE⁴³, Achilles⁶² and ICR-Intercell datasets^{8, 9} 338 (Supplementary Data 10, 11 and 12, respectively). 339

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Highly penetrant Rb synthetic lethal effects in TNBC include TAF1, TAF1 target genes, nuclear pore complex proteins and the SCF^{SKP}-COP9 signalosome complexes

We next calculated synthetic lethal penetrance (SLP) and coverage (SLC) scores for each Rb synthetic lethal effect triaged in steps two and three. In our analysis of the *Colt2* dataset, we calculated SLP and SLC scores for 122 Rb synthetic lethal effects, identifying 54 effects that had greater than 80 % penetrance (Figure 3A (SLP>90% shown), Supplementary Data 9) and also 349 identified highly penetrant effects from our analysis of the DRIVE and Achilles and ICR-Intercell datasets (Supplementary Data 10, 11 and 12, respectively). 350 351 Amongst the highly penetrant Rb SL candidates (>80% penetrant), a 352 significant number are known to be involved in RNA splicing (Supplementary Figure 1, p=5.14x10⁻¹², GO Biological Processes 2017, "mRNA splicing, via 353 spliceosome", Enrichr¹², Supplementary Data 13), including: ISY1 (SLP = 354 355 92%); SON (SLP = 100%); CWC22 (SLP = 100%); DDX23 (SLP = 100%); POLR2E (SLP = 100%); GEMIN5 (SLP = 92%); POLR2F (SLP = 100%); 356 TFIP11 (SLP = 92%); SRRM2 (SLP = 83%); LSM2 (SLP = 83%); SNW1 (SLP 357 = 83%); SART3 (SLP = 83%); FIP1L1 (SLP = 83%) and SNRPD3 (SLP = 92). 358 Interestingly, a RNAi screen in *C.elegans* also identified a synthetic lethal 359 360 interaction between Rb loss and components of the splicing machinery¹¹. Similarly, a number of proteins involved in the regulation of translation were 361 also identified as candidate Rb SL targets ($p=1.7 \times 10^{-7}$, GO Biological 362 Processes 2017, "regulation of translation by machinery localisation", 363 Enrichr¹². Supplementary Figure 1. Supplementary Data 13). These included: 364 RPS24 (SLP = 83%); RPS27 (SLP = 100%); RPS28 (SLP = 100%); RPL18A 365 366 (SLP = 92%); RPL13A (SLP = 92%); RPL10 (SLP = 92%); TCOF1 (SLP = 92%): GEMIN5 (SLP = 92%) and RPL38 (SLP = 92%). 367

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We also identified two nuclear pore complex (NPC) components^{4, 22}, NUP88 369 (SLP = 100%) and NUP214 (SLP = 83%) as highly penetrant Rb synthetic 370 lethal partners (Supplementary Figure 1, Supplementary Data 13). NPCs are 371 372 responsible for trafficking proteins between the nucleus and the cytoplasm. In 373 particular, NPCs control spindle assembly, faithful chromosome segregation and mitotic progression by controlling the temporal and spatial localisation of 374 proteins^{6, 45, 66} including the E2F transcriptional target and spindle assembly 375 checkpoint (SAC) protein, MAD2, whose elevated expression is required for 376 chromosomal instability in Rb defective cells^{50, 52}. It seems possible that the 377 highly penetrant synthetic lethalities between Rb and NUP88 or NUP214 378 could be related to the temporal and spatial localisation of MAD2, perhaps by 379 380 causing a level of excessive genomic instability that is incompatible with cell 381 viability.

383 When taking all of the highly penetrant Rb synthetic lethal effects identified in 384 the analysis of the Colt2 dataset into account, a significant number of these 385 included genes/proteins known to be associated with Rb function including the 386 Rb interacting protein TAF1 (siMEM p=0.016, SLP 92% and SLC 38%, Figure 4A)⁵⁶. TAF1 encodes the major 250 kDa subunit of the transcription initiation 387 factor. TFIID⁵⁷ which binds core promoter regions, including promoter start 388 sites in genes implicated in cell cycle control³⁶. We found that over half (n = 389 390 33) of the highly penetrant Rb synthetic lethal genes encompassed putative TAF1 binding sites, as defined by TAF1 Chip-Seq data^{12, 16} a significant 391 enrichment over chance alone ($p = 2.72 \times 10^{-13}$, Figure 4B). One mechanistic 392 393 explanation for these observations might be that many of the highly penetrant 394 Rb synthetic lethal interactions that operate in TNBC are caused by aberrant TAF1 activity, a downstream effect of Rb dysfunction. TAF1 encompasses 395 396 two protein domains, a kinase and bromodomain, which in principle are pharmacologically tractable. When taken together with the highly penetrant 397 398 Rb synthetic lethal interaction, this might make TAF1 an attractive target for cancer drug discovery^{7, 18, 53}. 399

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401 SKP1 and SKP2 were also identified as highly penetrant Rb synthetic lethal effects (Figure 4C,D, SLP 100% and 92% for SKP1 and SKP2, respectively). 402 Both SKP1 and SKP2 are part of an E3 ubiquitin ligase complex, SCF^{SKP2} 403 (Skp, Cullin F-box containing complex), whose activity is greatest during late 404 $G_1 / early \; S \;$ phase of the cell cycle. Using ubiquitin ligation, the ${\sf SCF}^{{\sf SKP2}}$ 405 406 complex targets proteins for proteasomal degradation including the cyclin dependent kinase inhibitors, p21 and p27⁶⁷. Three other highly penetrant Rb 407 408 synthetic lethal effects identified were also associated with the SCF complex 409 (Figure 4E-G); COPS1 (aka GPS1, SLP 83%, Figure 4E), COPS3 (SLP 83%, 410 Figure 4F) and COPS4 (SLP 92%, Figure 4G) encode components of the COP9 (Constitutive photomorphogenesis 9) signalosome complex (CSN^{35, 37}). 411 412 CSN regulates the ubiquitin ligase activity of SCF complexes via the deneddylation of the ring finger subunits (e.g. Rbx1) within SCF¹⁵ (Figure 4I). 413 We also noted that CKS1B (CDC28 protein kinase regulatory subunit 1B) also 414 415 represented a penetrant Rb synthetic lethal partner (SLP 58%, Figure 4H); 416 together with its co-factor SKP2, CKS1B provides the substrate specific 417 targeting of p27 by SCF^{SKP2 47}. When we compared transcriptomic data from 418 RB1 defective TNBCs to those from TNBC with no apparent RB1 defect 419 (Supplementary Data 3 and 4), we noted that RB1 defective TNBCs 420 expressed significantly elevated levels of *SKP2* and *COPS1 (GPS1)* mRNA 421 (Figure 4J,K,L, p<0.05 for both TCGA¹⁰ and Metabric⁴⁶ data, Wilcox rank sum 422 test), suggesting that in these particular tumours, elevated SKP2 activity might 423 buffer the effects of RB1 dysfunction.

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SKP2 directly interacts with Rb²⁸ and has previously been shown to be 425 required for the hyper-proliferative phenotype of Rb-depleted human 426 427 retinoblastoma cells, via its regulatory control over p27 levels⁶⁷. In cells with normal G₁/S cell cycle control, Rb binds SKP2, impairing its activity within the 428 SCF^{SKP2} complex; loss of Rb however, results in elevated SKP2 activity, a 429 430 resultant reduction in p27 and p21 protein levels, loss of Cyclin E-CDK2 and Cyclin D-CDK4,6 inhibition and thus progression of cells through the G₁ 431 restriction point and into S phase⁴⁰. To directly test whether Rb loss in a 432 breast epithelial cell could cause synthetic lethality with SKP2 inhibition, we 433 434 silenced SKP2 using two different siRNAs in non-tumour breast epithelial 435 MCF10A cells expressing either a shRNA silencing RB1 or a non-silencing 436 (NS) control shRNA (Figure 5A). Silencing of SKP2 (Figure 5B) elicited synthetic lethality in Rb defective cells but not Rb proficient cells (Figure 5C, 437 438 Student's t test p < 0.0001, Supplementary Figure 2). We also found that the toolbox SKP2 inhibitor SKPinC1, which impairs the binding of phosphorylated 439 p27 to the substrate recognition pocket formed by SKP2 and CKS1B⁶⁵, had a 440 profound synthetic lethal effect on Rb defective MCF10A cells, but minimal 441 442 effects in Rb wild type cells (p<0.001 two-way ANOVA, Figure 5D) and 443 induced apoptosis in Rb defective cells (Supplementary Figure 3). We also 444 assessed SKPinC1 sensitivity in 13 TNBC TCLs, and found that Rb defective TCLs were more sensitive than TCLs with no apparent Rb defect (Figure 5E 445 and p < 0.0022, Student's t test). In addition, we confirmed that exposure of 446 MCF10A cells with SKPinC1 inhibitor led to a stabilization of p27 protein 447 levels (Figure 5F). Targeting of p27 by siRNA also partially reversed the 448 449 inhibitory effect of SKPinC1 (Supplementary Figure 4), suggesting that the 450 effect of SKPinC1 was p27 dependent. Taken together, these observations

451 suggested that SKP2 small molecule inhibition, could in principle, elicit the
452 highly penetrant Rb synthetic lethal effect seen in TNBC tumour cells with
453 RNA interference reagents.

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455 We also assessed whether other highly penetrant synthetic lethal effects 456 operated in the RB1 isogenic MCF10A system. Using an arrayed siRNA 457 screen we tested all 54 genes that we identified as highly penetrant (> 80% penetrance) RB synthetic lethal effects in TNBC cell lines in the Colt2 study 458 459 along with an additional four controls (e.g. E2F3) for a total of 58 genes. We 460 found that over half of the identified dependencies (55%) were observed in 461 the isogenic system (Supplementary Figure 5), including profound synthetic 462 lethal effects associated with TIMELESS, PCDH1, PITRM1, E2F3, SMN2 and TCOF1. This suggests that these effects can be specifically associated with 463 RB loss, and that they are not an artefact of either the shRNA library used in 464 465 Colt2 or the pooled screening approach. It seemed possible that differences in the proliferative rate of RB1 defective vs. wild type MCF10A cells could 466 467 account for the synthetic lethal effects observed. However, we found that the 468 proliferative rate of RB1 defective and wild type MCF10A cells was similar (Supplementary Figure 6), suggesting this might not have a significant bearing 469 470 on the synthetic lethal effects identified. Moreover, analysis of previously published doubling times for 17 TNBC cell lines identified no significant 471 472 difference between RB1 defective and RB1 proficient models (Mann-Whitney U-test p=0.4)²⁵. 473

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475 We also assessed which of the highly penetrant Rb synthetic lethal effects 476 identified in our analysis of the Colt2 data were also identified as highly penetrant effects in the TNBC TCLs in two other shRNA screens: Achilles⁶² 477 and DRIVE⁴³. Comparing the p<0.05 Rb penetrant synthetic lethal effects 478 479 between the three datasets, we noted that SKP2 was one of two synthetic 480 lethal effects identified in all three screens, the other effect being associated with SART3, the RNA splicing factor (Figure 6A-C, Supplementary Data 9, 10 481 482 and 11). In each screen, SKP2/Rb synthetic lethality was highly penetrant with 483 SLP scores of 92% (Colt2), 75% (Achilles) and 100% (DRIVE dataset, Figure 484 6D-F). Whilst the TNBC TCLs described in these three datasets are partially

overlapping, the shRNA libraries used were independently designed and
synthesised and the screens independently carried out. As such, identifying
SKP2 as a highly penetrant Rb synthetic lethal effect in all three datasets
suggested that this effect was somewhat independent of the shRNA reagents
used.

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491 Rb/SKP2 synthetic lethality operates in tumour cells from lung and other 492 cancer histologies

493 Rb defects are relatively prevalent in TNBC, but are not unique to this cancer 494 subtype and are also evident in in many tumour types including, for example, small-cell lung cancers, bladder carcinomas, osteosarcomas, glioblastomas, 495 endometrial carcinomas and retinoblastomas^{14, 17, 23, 27, 55}. To estimate 496 whether the highly penetrant Rb/SKP2 synthetic lethality in TNBC extended to 497 other cancer histologies, we re-analysed the results of two recent large-scale 498 499 shRNA screens that encompass TCLs derived from non-breast cancer 500 histologies. Project DRIVE includes 373 non-breast tumour cell lines with available Rb mutation and copy number status that were derived from 18 501 different cancer histologies⁴³. In an analysis of the DRIVE dataset that 502 excluded the breast tumour cell lines, we found a significant association 503 504 between *RB1* mutation/deletion and sensitivity to *SKP2* shRNA (Figure 7A,B; MP-test p<0.0001, SLP=75%). Of over 6,000 genes tested, only E2F3 (MP-505 506 test p<0.0001) showed a stronger association with Rb status (Figure 7A). Similarly, project Achilles includes 467 non-breast tumour cell lines with 507 508 available Rb mutation and copy number status, derived from 20 different cancer histologies⁶². In an analysis of the Achilles dataset that excluded the 509 510 breast tumour cell lines, we also found a significant association between RB1 511 mutation/deletion and sensitivity to SKP2 shRNA (Figure 7C,D; MP-test 512 p<0.0001, SLP=74%). Of over 17,000 genes, SKP2 synthetic lethality was the 513 second most significant effect associated with Rb loss, after CDK2 (MP-test p<0.0001) with E2F3 being the third most significant effect (Figure 7C). The 514 515 elevated penetrance of the Rb/SKP2 synthetic lethal effect in tumour cell 516 models other than those derived from breast cancer suggested that this 517 genetic dependency might operate in multiple histologies. When we 518 considered the specific cancer histology of tumour cell lines, rather than

519 analysing these *en mass*, we found in both the Achilles and DRIVE datasets 520 the Rb/SKP2 synthetic lethality was detectable in tumour cell lines derived 521 from lung cancers (p=0.026/SLP=71% and p=0.0008/SLP=73% for DRIVE 522 and Achilles, respectively; Figure 7E,F). However, we do note that the 523 relatively small number of Rb defective tumour cell lines from non-lung cancer 524 histologies in these datasets might impair the ability to detect the Rb/SKP2 525 synthetic lethality. For example, in both DRIVE and Achilles datsets, we noted 526 that Rb defective tumour cell lines derived from prostate cancer, 527 osteosarcoma, liver cancer and oesophageal cancer, exhibited sensitivity to 528 SKP2 shRNA (Figure 7G).

529

530 **Discussion**

531 There is now a relatively long-standing history of using genetic concepts such 532 as synthetic lethality to identify novel therapeutic targets for the treatment of cancer²⁴. In part these efforts have been successful, with synthetic lethal 533 treatments³⁹ and drugs that exploit oncogene addiction effects now being 534 approved for the treatment of the clinical disease³⁸. However, despite these 535 536 advances, one challenge to this approach has been in identifying highly 537 penetrant synthetic lethal effects that associate with the presence of a 538 molecular biomarker. Here we describe efforts to identify highly-penetrant 539 synthetic lethal effects associated with loss of the tumour suppressor Rb in 540 TNBC. Following the classification of 42 TNBC TCLs according to Rb status, 541 we interrogated genome-scale and focussed gene set shRNA screening 542 datasets, identified candidate synthetic lethal effects and then used stringent 543 filters to triage from the list of candidate synthetic lethal effects those most 544 likely to represent highly penetrant effects. This approach allowed us to 545 identify a number of highly penetrant synthetic lethal effects, many of which are associated with known functions of Rb and/or associated with Rb 546 547 interacting proteins. These included TAF1 and TAF1 target genes as well as members of the SCF^{SKP2} complex. 548

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550 We note that there are considerable caveats associated with the approach we 551 have taken to identifying highly penetrant Rb-related synthetic lethal effects in 552 TNBC and the interpretation of the data; understanding these caveats is 553 critical to the use of the information we provide in this work. Primary amongst 554 these is that we have exclusively used data from *in vitro* genetic screens to 555 identify synthetic lethal effects and assess their penetrance. It seems likely 556 that some of the effects identified in our analysis only operate in the context of 557 two-dimensional in vitro culture and are abrogated, and therefore appear less 558 penetrant, in three dimensional and/or *in vivo* settings. Such a possibility thus 559 provides the rationale for also assessing synthetic lethal effects, and assessing their penetrance, in *in vivo* models of cancer. Secondly, there is 560 561 little way of effectively estimating the true false negative rate of our approach; it is entirely possible that we have not identified real, highly penetrant, Rb 562 563 related synthetic lethal effects either because the RNA interference reagents 564 used in genetic screens ineffectively inactive the genes they are designed to target or because some other form of gene/protein inactivation (e.g. catalytic 565 inhibition of the target protein rather than gene silencing) is required to elicit a 566 567 synthetic lethal effect. Nevertheless, the identification in multiple, independently-conducted, genetic screens of the highly penetrant Rb/SKP2 568 569 synthetic lethal effect, and its recapitulation with a small molecule inhibitor, 570 suggests that this highly penetrant synthetic lethality effect is unlikely to be a 571 false positive.

572

As well as targeting Rb defective TNBC, the potential for using SKP2 573 574 inhibition in other cancer histologies associated with Rb defects is considerable. For example, the hyper-proliferative phenotype of *Rb*-depleted 575 human retinoblastoma cells and mouse melanotrophs is dependent upon the 576 SKP2^{63, 67} and the SKP2/CKS1 pocket inhibitor, SKPinC1, inhibits Rb/p53 577 defective mouse prostate tumour cell organoids⁶⁸. Finally, large-scale shRNA 578 579 screens in tumour cell lines from a variety of cancer histologies (and our 580 analysis described in Figure 6) suggest that the dependency of Rb defective 581 TCLs upon SKP2 might extend beyond models of TNBC. Whether the high penetrance of the Rb/SKP2 synthetic lethality seen in models of TNBC 582 extends to each of these histologies remains to be seen, but our initial 583 584 analysis in Figure 6 suggests that highly penetrant effects might also be 585 apparent in lung cancer.

587 It might be interesting to speculate what characteristics might differentiate a highly penetrant synthetic lethal effect that operates in cancer from less 588 589 penetrant effects; being able to understand the factors that distinguish one 590 from the other might eventually allow highly penetrant synthetic lethal effects 591 in cancer to be predicted from first principles, thus reducing the reliance upon 592 large-scale genetic screens in multiple tumour cell lines to empirically 593 establish penetrance. From our relatively unbiased analysis of genome-scale 594 shRNA screens, it is perhaps striking that many of the highly penetrant Rb 595 synthetic lethal effects we identified (e.g. SKP1, SKP2, CKS1B, COPS1, 596 COPS2, COPS3) have two characteristics: (i) they are closely, rather than 597 distally, involved in controlling an essential process in highly proliferating 598 tumour cells, namely G₁ cell cycle progression by Cyclin/CDK activity; and (ii) 599 this process is also closely, rather than distally, controlled by the synthetic lethal partner, Rb. It seems reasonable to think that synthetic lethal 600 601 interactions that control essential processes in cancer cells via small-world 602 networks (i.e. those than contain a relatively small number of nodes between 603 synthetic lethal partners and proteins involved in essential processes) might 604 be less likely to be reversed, and therefore more likely highly penetrant, than 605 synthetic lethal effects that control essential processes via distal molecular 606 mechanisms that involve larger molecular networks with multiple intervening 607 nodes. Whether this turns out to be a general principle or not remains to be 608 seen but already others have started to establish that many synthetic lethal 609 effects associated with tumour suppressor genes other than Rb can be 610 classified into a defined, and relatively small number of classes, including 611 those between paralogs and those between genes in the same molecular pathway⁶². This suggests that some of the principles that govern how 612 613 synthetic lethal effects operate in tumour cells can indeed be established.

614

615 Materials and methods

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617 Cell lines, compounds and siRNA

All cell lines were obtained from the American Type Culture Collection (ATCC)

619 and maintained according to the supplier's instructions. Cell lines were

620 routinely STR typed and mycoplasma tested. Cell lines were grown and transfected with individual and SMARTpool siGENOME siRNA (Dharmacon) 621 622 and transfected using RNAiMAX (Invitrogen) reagent as described in⁹. 623 Transfection efficiency was monitored using positive control (siPLK1) and two 624 different negative control siRNAs (siCON1 and siCON2; Dharmacon, catalogue numbers D-001210-01-20 and D-001206-14-20). SKPinC1 inhibitor 625 626 was purchased from Tocris (no. 4817) and was solubilised as a 10 mM stock 627 solution in DMSO.

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629 Western blot analysis

630 Whole cell protein lysates were extracted from cells by lysis with NP250 buffer (20mM Tris pH 7.6, 1mM EDTA, 0.5% NP40, 235mM NaCl). In gene silencing 631 experiments, cell were cultured for 48 hours after siRNA transfection, at which 632 point lysates were generated. Following a 20 minute incubation at 4^oC, 633 lysates were centrifuged and supernatents collected. Electrophoresis was 634 performed using Novex precast Bis-Tris gels (Invitrogen) and gels were 635 blotted onto nitrocellulose filters as described previously²¹. Blots were 636 637 immunoblotted in 5% (w/v) milk at $4^{\circ}C$ overnight using the following primary 638 antibodies: anti-Rb1 (1/1000 (v/v) dilution in 5 % (v/v) milk, New England Biolab, 9309), anti-p16 (1/1000, abcam), anti-SKP2 (1/1000, New England 639 Biolab, 4358), anti-p27 (1/1000, New Engand Biolab, 2552), anti-tubulin 640 (1/1000, abcam) and anti-actin (1/1000, Santa Cruz, sc-1616). After washing, 641 642 blots were incubated 1 hour at room temperature with secondary antibodies 643 (Li-COR) diluted 1/10,000 (v/v) in 5% (w/v) milk. Protein bands were visualised and guantified using the Odyssev FC imaging system and 644 ImageStudio software (Li-COR). 645

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647 **Tumour and cell line expression analysis**

Limma⁴⁸ was performed to identify differentially expressed genes RB1 defective *vs.* RB1 not altered tumour cell lines, using data from ⁴² and ³. A design matrix with the cell line RB1 classifications was created and a linear model was fitted to expression values of each gene. Next, an empirical Bayes method was used to obtain more precise gene-wise variability estimates between the two groups. Differential expression between groups was
 represented as log fold change scores with associated p-values and adjusted
 p-values. Publically available cell line mRNA expression datasets used in this
 study include Marcotte⁴² and CCLE³, as indicated.

657

For the analysis of TNBC tumours, mRNA expression, DNA copy number and 658 659 somatic mutation profiles for TCGA tumours were downloaded from GDAC (https://gdac.broadinstitute.org/), release 2016_01_28. GISTIC v2 level 4 data 660 661 were used for copy number analysis and log₂ ratios were converted to genomic gains/amplifications, neutral and loss/deletion states using threshold 662 of $\pm \log_2(3/2)$. Raw Metabric data files were downloaded from the European 663 Genome-phenome Archive (EGA; study ID EGAS0000000083). The 664 Metabric breast cancer dataset was pre-processed, summarised and quantile-665 normalised from the raw expression files generated by Illumina BeadStudio (R 666 packages: beadarray v2.4.2 and illuminaHuman v3.db 1.12.2). Probe to 667 gene-level mapping was performed by keeping the most variable (standard 668 669 deviation) probe. Metabric copy number calls were used as published in the 670 original study¹⁹.

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For TCGA breast cancer cohort, previously published³⁴ TNBC6 and TNBC4 672 calls were used resulting in 140 patients with matched mRNA, copy number 673 and mutation profiles. For the Metabric cohort, TNBC6 calls were successfully 674 derived from TNBC subtyping portal¹³ (http://cbc.mc.vanderbilt.edu/tnbc) for 675 187 patients with matched mRNA and copy number profiles. In the TCGA 676 cohort, TNBC samples with RB1 mRNA z-score < -1 were regarded as Rb-677 low, RB1 copy number \log_2 ratio < -0.585 were regarded as Rb loss and 678 samples with RB1 truncating mutations were considered as Rb inactivated. 679 This resulted in 48 Rb defective and 92 Rb not altered TNBC samples. In the 680 681 Metabric cohort, samples with RB1 mRNA z-score < -1 were regarded as Rb low, whilst samples with RB1 loss/deletions were regarded as Rb loss. This 682 683 resulted in 55 Rb defective and 132 Rb not altered TNBC samples.

685 Differential gene expression analysis on TCGA TNBC samples was
 686 performed using Limma voom⁴⁸.

687

688 Association Testing

689 The si/shRNA Mixed Effect Model (siMEM) was used to measure the essentiality of genes in the Colt2 genome-wide shRNA dropout screen⁴². 690 691 siMEM uses hierarchical linear regression and considers level of each shRNA 692 to be a regression function of its initial abundance over time, baseline trend in 693 abundance over time and difference in abundance trend between samples 694 sharing a common feature. siMEM results are represented as the difference 695 between the dropout rate of hairpins in the two cell line groups being 696 compared.

697 For analysis of the DRIVE and Achilles shRNA screen datasets, where only a single screen time-point was collected, a one-sided median permutation (MP) 698 699 test was used to identify associations between RB1 status of cell lines and 700 their sensitivities to shRNA targeting of genes. For each gene, the observed 701 difference between the median of scores of RB1 defective and RB1 not 702 altered groups of tumour cell lines was compared. A total of one million 703 random samples with the same sample size as in the two groups were 704 created. The differences in the medians of the groups were calculated for 705 each random sample and the statistical significance of the difference was 706 determined.

707 Cell viability assays

500 cells per well were seeded into 384-well plates. After 24 hours cells were 708 709 exposed to increasing concentrations of SKPinC1 inhibitor diluted in DMSO 710 using an Echo 550 liquid handler (Labcyte). Cells were incubated with the 711 inhibitor for five days after which cell viability was estimated with CellTitre-Glo 712 reagent (Promega). For gene silencing experiments cells were incubated for 713 five days following siRNA transfection prior to viability assessment. 714 Luminescence readings from drug exposed cells were normalised to those 715 from DMSO exposed cells to calculate Surviving Fractions (SF). SFs were 716 used to derive four parameter logistic regression dose/response curves using

720 Statistical analysis

Unless otherwise stated all data is represented here as mean \pm standard error using Graphpad Prism Software (La Jolla, CA). All conditions were performed in triplicate in at least triplicate experiments. Statistical significance was measured using either Student's t-test, pearson coefficient correlation, Mann-Whitney U test or two-way ANOVA. *p*<0.05 was considered significant.

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728 **Conflict of interest statement**

- There are no conflicts of interest to declare.
- 730

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747 **References**

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771

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783

786

- Asghar US, Barr AR, Cutts R, Beaney M, Babina I, Sampath D *et al.* SingleCell Dynamics Determines Response to CDK4/6 Inhibition in Triple-Negative
 Breast Cancer. Clin Cancer Res 2017; 23: 5561-5572.
- Ashworth A, Lord CJ, Reis-Filho JS. Genetic interactions in cancer
 progression and treatment. Cell 2011; 145: 30-38.
- Barretina J, Caponigro G, Stransky N, Venkatesan K, Margolin AA, Kim S *et al.* The Cancer Cell Line Encyclopedia enables predictive modelling of anticancer drug sensitivity. Nature 2012; 483: 603-607.
- Bernad R, Engelsma D, Sanderson H, Pickersgill H, Fornerod M. Nup214Nup88 nucleoporin subcomplex is required for CRM1-mediated 60 S
 preribosomal nuclear export. J Biol Chem 2006; 281: 19378-19386.
- 5 Bianchini G, Balko JM, Mayer IA, Sanders ME, Gianni L. Triple-negative
 breast cancer: challenges and opportunities of a heterogeneous disease. Nat
 Rev Clin Oncol 2016; 13: 674-690.
- 7686Blower MD, Nachury M, Heald R, Weis K. A Rae1-containing769ribonucleoprotein complex is required for mitotic spindle assembly. Cell7702005; 121: 223-234.
- 772 7 Bouche L, Christ CD, Siegel S, Fernandez-Montalvan AE, Holton SJ, Fedorov
 773 O *et al.* Benzoisoquinolinediones as Potent and Selective Inhibitors of BRPF2
 774 and TAF1/TAF1L Bromodomains. J Med Chem 2017; 60: 4002-4022.
- Brough R, Frankum JR, Sims D, Mackay A, Mendes-Pereira AM, Bajrami I *et al*. Functional viability profiles of breast cancer. Cancer Discov 2011; 1: 260-273.
- 780 9 Campbell J, Ryan CJ, Brough R, Bajrami I, Pemberton HN, Chong IY *et al.*781 Large-Scale Profiling of Kinase Dependencies in Cancer Cell Lines. Cell Rep
 782 2016; 14: 2490-2501.
- Cancer Genome Atlas N. Comprehensive molecular portraits of human breast tumours. Nature 2012; 490: 61-70.
- Ceron J, Rual JF, Chandra A, Dupuy D, Vidal M, van den Heuvel S. Largescale RNAi screens identify novel genes that interact with the C. elegans
 retinoblastoma pathway as well as splicing-related components with synMuv
 B activity. BMC Dev Biol 2007; 7: 30.
- Chen EY, Tan CM, Kou Y, Duan Q, Wang Z, Meirelles GV *et al.* Enrichr:
 interactive and collaborative HTML5 gene list enrichment analysis tool. BMC
 Bioinformatics 2013; 14: 128.

Subtyping Tool for Triple-Negative Breast Cancer. Cancer Inform 2012; 11: 147-156. Choi W, Ochoa A, McConkey DJ, Aine M, Hoglund M, Kim WY et al. Genetic Alterations in the Molecular Subtypes of Bladder Cancer: Illustration in the Cancer Genome Atlas Dataset. Eur Urol 2017; 72: 354-365. Chung D, Dellaire G. The Role of the COP9 Signalosome and Neddylation in DNA Damage Signaling and Repair. Biomolecules 2015; 5: 2388-2416. Consortium EP. The ENCODE (ENCyclopedia Of DNA Elements) Project. Science 2004; 306: 636-640. Corney DC, Flesken-Nikitin A, Choi J, Nikitin AY. Role of p53 and Rb in ovarian cancer. Adv Exp Med Biol 2008; 622: 99-117. Crawford TD, Tsui V, Flynn EM, Wang S, Taylor AM, Cote A et al. Diving into the Water: Inducible Binding Conformations for BRD4, TAF1(2), BRD9, and CECR2 Bromodomains. J Med Chem 2016; 59: 5391-5402. Curtis C, Shah SP, Chin SF, Turashvili G, Rueda OM, Dunning MJ et al. The genomic and transcriptomic architecture of 2,000 breast tumours reveals novel subgroups. Nature 2012; 486: 346-352. Ertel A, Dean JL, Rui H, Liu C, Witkiewicz AK, Knudsen KE et al. RB-pathway disruption in breast cancer: differential association with disease subtypes, disease-specific prognosis and therapeutic response. Cell Cycle 2010; 9: 4153-4163. Farmer H, McCabe N, Lord CJ, Tutt AN, Johnson DA, Richardson TB et al. Targeting the DNA repair defect in BRCA mutant cells as a therapeutic strategy. Nature 2005; 434: 917-921. Fornerod M, van Deursen J, van Baal S, Reynolds A, Davis D, Murti KG et al. The human homologue of yeast CRM1 is in a dynamic subcomplex with CAN/Nup214 and a novel nuclear pore component Nup88. EMBO J 1997; 16: 807-816. Goldhoff P, Clarke J, Smirnov I, Berger MS, Prados MD, James CD et al.

Chen X, Li J, Gray WH, Lehmann BD, Bauer JA, Shyr Y et al. TNBCtype: A

- Goldnoff P, Clarke J, Smirnov I, Berger MS, Prados MD, James CD *et al.*Clinical stratification of glioblastoma based on alterations in retinoblastoma tumor suppressor protein (RB1) and association with the proneural subtype. J
 Neuropathol Exp Neurol 2012; 71: 83-89.
- 840 24 Hartwell LH, Szankasi P, Roberts CJ, Murray AW, Friend SH. Integrating
 841 genetic approaches into the discovery of anticancer drugs. Science 1997; 278:
 842 1064-1068.
 843

851

855

859

863

867

871

875

879

884

- 844 25 Haverty PM, Lin E, Tan J, Yu Y, Lam B, Lianoglou S *et al.* Reproducible
 845 pharmacogenomic profiling of cancer cell line panels. Nature 2016; 533: 333846 337.
- 848 26 Herschkowitz JI, He X, Fan C, Perou CM. The functional loss of the
 849 retinoblastoma tumour suppressor is a common event in basal-like and luminal
 850 B breast carcinomas. Breast Cancer Res 2008; 10: R75.
- Higashiyama M, Doi O, Kodama K, Yokouchi H, Tateishi R. Retinoblastoma
 protein expression in lung cancer: an immunohistochemical analysis.
 Oncology 1994; 51: 544-551.
- Ji P, Jiang H, Rekhtman K, Bloom J, Ichetovkin M, Pagano M *et al.* An RbSkp2-p27 pathway mediates acute cell cycle inhibition by Rb and is retained
 in a partial-penetrance Rb mutant. Mol Cell 2004; 16: 47-58.
- Jiang Z, Jones R, Liu JC, Deng T, Robinson T, Chung PE *et al.* RB1 and p53
 at the crossroad of EMT and triple-negative breast cancer. Cell Cycle 2011;
 10: 1563-1570.
- 30 Johnson J, Thijssen B, McDermott U, Garnett M, Wessels LF, Bernards R.
 865 Targeting the RB-E2F pathway in breast cancer. Oncogene 2016; 35: 4829866 4835.
- 31 Jones RA, Robinson TJ, Liu JC, Shrestha M, Voisin V, Ju Y *et al.* RB1
 deficiency in triple-negative breast cancer induces mitochondrial protein
 translation. J Clin Invest 2016; 126: 3739-3757.
- Knudsen ES, McClendon AK, Franco J, Ertel A, Fortina P, Witkiewicz AK.
 RB loss contributes to aggressive tumor phenotypes in MYC-driven triple
 negative breast cancer. Cell Cycle 2015; 14: 109-122.
- 876 33 Lawrence RT, Perez EM, Hernandez D, Miller CP, Haas KM, Irie HY *et al.*877 The proteomic landscape of triple-negative breast cancer. Cell Rep 2015; 11:
 878 630-644.
- Lehmann BD, Jovanovic B, Chen X, Estrada MV, Johnson KN, Shyr Y *et al.*Refinement of Triple-Negative Breast Cancer Molecular Subtypes:
 Implications for Neoadjuvant Chemotherapy Selection. PLoS One 2016; 11:
 e0157368.
- Li P, Xie L, Gu Y, Li J, Xie J. Roles of Multifunctional COP9 Signalosome
 Complex in Cell Fate and Implications for Drug Discovery. J Cell Physiol
 2017; 232: 1246-1253.
- Lin CY, Tuan J, Scalia P, Bui T, Comai L. The cell cycle regulatory factor
 TAF1 stimulates ribosomal DNA transcription by binding to the activator
 UBF. Curr Biol 2002; 12: 2142-2146.

899

906

910

914

919

924

928

932

- Lingaraju GM, Bunker RD, Cavadini S, Hess D, Hassiepen U, Renatus M *et al.* Crystal structure of the human COP9 signalosome. Nature 2014; 512: 161-165.
- 897 38 Lord CJ, Ashworth A. Biology-driven cancer drug development: back to the
 898 future. BMC Biol 2010; 8: 38.
- 900 39 Lord CJ, Ashworth A. PARP inhibitors: Synthetic lethality in the clinic.
 901 Science 2017; 355: 1152-1158.
 902
- 90340Lu Z, Bauzon F, Fu H, Cui J, Zhao H, Nakayama K *et al.* Skp2 suppresses904apoptosis in Rb1-deficient tumours by limiting E2F1 activity. Nat Commun9052014; 5: 3463.
- 90741Marcotte R, Brown KR, Suarez F, Sayad A, Karamboulas K, Krzyzanowski908PM *et al.* Essential gene profiles in breast, pancreatic, and ovarian cancer909cells. Cancer Discov 2012; 2: 172-189.
- 911 42 Marcotte R, Sayad A, Brown KR, Sanchez-Garcia F, Reimand J, Haider M *et al.* Functional Genomic Landscape of Human Breast Cancer Drivers,
 913 Vulnerabilities, and Resistance. Cell 2016; 164: 293-309.
- McDonald ER, 3rd, de Weck A, Schlabach MR, Billy E, Mavrakis KJ,
 Hoffman GR *et al.* Project DRIVE: A Compendium of Cancer Dependencies
 and Synthetic Lethal Relationships Uncovered by Large-Scale, Deep RNAi
 Screening. Cell 2017; 170: 577-592 e510.
- 920 44 Nittner D, Lambertz I, Clermont F, Mestdagh P, Kohler C, Nielsen SJ *et al.*921 Synthetic lethality between Rb, p53 and Dicer or miR-17-92 in retinal
 922 progenitors suppresses retinoblastoma formation. Nat Cell Biol 2012; 14: 958923 965.
- 925 45 Orjalo AV, Arnaoutov A, Shen Z, Boyarchuk Y, Zeitlin SG, Fontoura B *et al.*926 The Nup107-160 nucleoporin complex is required for correct bipolar spindle
 927 assembly. Mol Biol Cell 2006; 17: 3806-3818.
- 929 46 Pereira B, Chin SF, Rueda OM, Vollan HK, Provenzano E, Bardwell HA *et al.*930 The somatic mutation profiles of 2,433 breast cancers refines their genomic 931 and transcriptomic landscapes. Nat Commun 2016; 7: 11479.
- 933 47 Reed SI. Ratchets and clocks: the cell cycle, ubiquitylation and protein
 934 turnover. Nat Rev Mol Cell Biol 2003; 4: 855-864.
 935
- Ritchie ME, Phipson B, Wu D, Hu Y, Law CW, Shi W *et al.* limma powers
 differential expression analyses for RNA-sequencing and microarray studies.
 Nucleic Acids Res 2015; 43: e47.
- 940 49 Robinson TJ, Liu JC, Vizeacoumar F, Sun T, Maclean N, Egan SE *et al.* RB1
 941 status in triple negative breast cancer cells dictates response to radiation
 942 treatment and selective therapeutic drugs. PLoS One 2013; 8: e78641.

943		
944	50	Rodriguez-Bravo V, Maciejowski J, Corona J, Buch HK, Collin P, Kanemaki
945		MT et al. Nuclear pores protect genome integrity by assembling a premitotic
946		and Mad1-dependent anaphase inhibitor. Cell 2014; 156: 1017-1031.
947		
948	51	Russnes HG, Lingiaerde OC, Borresen-Dale AL, Caldas C, Breast Cancer
949	• -	Molecular Stratification: From Intrinsic Subtypes to Integrative Clusters Am J
950		Pathol 2017: 187: 2152-2162
951		Tullof 2017, 107. 2152 2162.
952	52	Schvartzman IM Duijf PH Sotillo R Coker C Benezra R Mad2 is a critical
953	52	mediator of the chromosome instability observed upon Rh and p53 pathway
953 954		inhibition Cancer Cell 2011: 10: 701-714
055		minoriton. Cancer Cen 2011, 19: 701-714.
056	53	Sdelai S. Lardaou CH. Tallant C. Klansch F. Klaibar P. Bonnatt I. at. al.
950	55	Manning the chamical chromatin reactivation landscape identifies PDD4
957		TAE1 group talk. Not Chem Diel 2016, 12, 504 510
938		1AF1 closs-talk. Nat Chefn Bloi 2010, 12. 304-310.
939	51	Searly IC, Li D, Dy W. Torrecting Dh mytant concerns hy inactivating TSC2
900	54	Searle JS, LI B, Du W. Targeting RD mutant cancers by inactivating TSC2.
961		Uncotarget 2010; 1: 228-232.
962	~ ~	
963	22	Semczuk A, Schneider-Stock R, Miturski R, Skomra D, Tomaszewski J,
964		Roessner A <i>et al.</i> RB protein expression in human endometrial carcinomasan
965		immunohistochemical study. Pathol Res Pract 2000; 196: 41-46.
966		
967	56	Shao Z, Ruppert S, Robbins PD. The retinoblastoma-susceptibility gene
968		product binds directly to the human TATA-binding protein-associated factor
969		TAFII250. Proc Natl Acad Sci U S A 1995; 92: 3115-3119.
970		
971	57	Shao Z, Siegert JL, Ruppert S, Robbins PD. Rb interacts with
972		TAF(II)250/TFIID through multiple domains. Oncogene 1997; 15: 385-392.
973		
974	58	Stefansson OA, Jonasson JG, Olafsdottir K, Hilmarsdottir H, Olafsdottir G,
975		Esteller M et al. CpG island hypermethylation of BRCA1 and loss of pRb as
976		co-occurring events in basal/triple-negative breast cancer. Epigenetics 2011; 6:
977		638-649.
978		
979	59	Takada M, Ishiguro H, Nagai S, Ohtani S, Kawabata H, Yanagita Y et al.
980		Survival of HER2-positive primary breast cancer patients treated by
981		neoadjuvant chemotherapy plus trastuzumab: a multicenter retrospective
982		observational study (JBCRG-C03 study). Breast Cancer Res Treat 2014; 145:
983		143-153.
984		
985	60	Thomas DM, Carty SA, Piscopo DM, Lee JS, Wang WF, Forrester WC et al.
986		The retinoblastoma protein acts as a transcriptional coactivator required for
987		osteogenic differentiation. Mol Cell 2001: 8: 303-316.
988		
989	61	Trere D, Brighenti E, Donati G, Ceccarelli C, Santini D, Taffurelli M et al.
990	-	High prevalence of retinoblastoma protein loss in triple-negative breast
991		cancers and its association with a good prognosis in patients treated with
992		adjuvant chemotherapy. Ann Oncol 2009: 20: 1818-1823.

993		
994	62	Tsherniak A, Vazquez F, Montgomery PG, Weir BA, Kryukov G, Cowley GS
995		et al. Defining a Cancer Dependency Map. Cell 2017; 170: 564-576 e516.
996		
997	63	Wang H, Bauzon F, Ji P, Xu X, Sun D, Locker J et al. Skp2 is required for
998		survival of aberrantly proliferating Rb1-deficient cells and for tumorigenesis
999		in Rb1+/- mice. Nat Genet 2010; 42: 83-88.
1000		
1001	64	Witkiewicz AK, Ertel A, McFalls J, Valsecchi ME, Schwartz G, Knudsen ES.
1002		RB-pathway disruption is associated with improved response to neoadjuvant
1003		chemotherapy in breast cancer. Clin Cancer Res 2012; 18: 5110-5122.
1004		
1005	65	Wu L, Grigoryan AV, Li Y, Hao B, Pagano M, Cardozo TJ. Specific small
1006		molecule inhibitors of Skp2-mediated p27 degradation. Chem Biol 2012; 19:
1007		1515-1524.
1008		
1009	66	Wu Z, Jin Z, Zhang X, Shen N, Wang J, Zhao Y et al. Nup62, associated with
1010		spindle microtubule rather than spindle matrix, is involved in chromosome
1011		alignment and spindle assembly during mitosis. Cell Biol Int 2016; 40: 968-
1012		975.
1013		
1014	67	Xu XL, Singh HP, Wang L, Qi DL, Poulos BK, Abramson DH et al. Rb
1015		suppresses human cone-precursor-derived retinoblastoma tumours. Nature
1016		2014; 514: 385-388.
1017		
1018	68	Zhao H, Lu Z, Bauzon F, Fu H, Cui J, Locker J et al. p27T187A knockin
1019		identifies Skp2/Cks1 pocket inhibitors for advanced prostate cancer.
1020		Oncogene 2017; 36: 60-70.
1021		
1022		
1023		

Figure and Table legends

Figure 1. Rb status in TNBC tumour cell lines. A. Western blot illustrating 1025 1026 Rb and p16 protein expression in 30 breast tumour cell lines (TCLs). BT549, 1027 MDAMB436, MDAMB468 and DU4475 each possess loss of function 1028 mutations in the RB1 gene. SUM149 cells express reduced RB1 mRNA. B. 1029 Scatter plot illustrating quantification of Rb protein levels delineated from (A). 1030 Protein expression in Rb defective vs. not altered, p=0.0484, Student's t test. C. Scatter plot illustrating Rb protein expression in defective and proficient 1031 TNBC TCLs estimated by mass spectrometry data from³³. TNBC TCLs were 1032 classified into "Rb defective" and "not altered" groups according to western 1033 1034 blot data from (A); using this classification, normalised iBAQ Rb peptide scores were compared and are shown. p=0.0002, Fishers exact test. **D**. 1035 Scatter plot illustrating the correlation between Rb protein and mRNA 1036 transcript levels in TNBC TCLs. Rb protein levels from (A,B) are compared to 1037 1038 Rb mRNA transcript levels obtained from the CCLE database³. Correlation r=0.7, p=0.0075, Pearson's correlation. E. Oncoprint illustrating Rb and breast 1039 1040 cancer subtype status amongst 42 TNBC TCLs. F. Volcano plot illustrating mRNAs that are differentially expressed (limma analysis p<0.05) in Rb 1041 1042 defective (vs. Rb not altered) TNBC TCLs. Genes functionally related to Rb 1043 are highlighted, as is Rb itself. G. Volcano plot of mRNAs that are differentially expressed (limma analysis p<0.05) in 48 Rb defective (vs. Rb not altered) 1044 triple negative breast tumours from the TCGA study¹⁰. Genes functionally 1045 related to Rb are highlighted, as is Rb itself. H. Volcano plot of mRNAs that 1046 1047 are differentially expressed (limma analysis p<0.05) in 55 Rb defective (vs. Rb 1048 not altered) triple negative breast tumours from the Metabric study⁴⁶. Genes 1049 functionally related to Rb are highlighted, as is Rb itself.

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1053 Figure 2. Identifying highly penetrant Rb synthetic lethal effects that 1054 operate in TNBC. A. Schematic illustrating the data analysis workflow used. 1055 In the first instance, 16,056 gene zGARP scores from shRNA screens in 42 1056 TNBC cell lines described in the Colt2 dataset were analysed; parallel 1057 analyses were carried out using data from the DRIVE and Achilles datasets 1058 (see main text). B. Scatter plot illustrating ERBB2 Z GARP scores in 77 breast 1059 tumour cell lines partitioned according to effects in ERBB2 amplified and non-1060 amplified TCLs. ERBB2 shRNA elicits a p<0.0001 oncogene addiction effect (siMEM) with 80% penetrance in ERBB2 amplified tumour cell lines (red). 1061 Coverage is also shown. C. Scatter plot illustrating 1065 p<0.05 significant 1062 siMEM Rb synthetic lethal effects identified from the siMEM analysis of TNBC 1063 1064 TCLs in the *Colt2* study (step one in (A)). p<0.05 effects are ranked ordered by siMEM p value. E2F3 (synthetic lethal in Rb defective), CDK6, CDK4 and 1065 the CDK4,6 cyclin partner, Cyclin D1 (CCND1) (dependencies in Rb not 1066 altered) are highlighted. D,E. Scatter plots illustrating Z scores in 42 TNBC 1067 1068 TCLs for two siMEM p<0.05 candidate Rb synthetic lethal effects, PSMD1 (D) 1069 and KLF16 (E), removed from further analysis by the use of Z score filters (step two and three in (A)). zGARP scores for PSMD1 (preferentially target Rb 1070 defective, siMEM p value 3 x 10⁻⁵) indicate all but two Rb not altered tumour 1071 cell lines exhibit Z score of <-2 (median Z in not altered group <-4). zGARP 1072 scores for *KLF16* (preferentially target Rb defective, siMEM *p* value 0.0002) 1073 1074 indicate that the majority of Rb defective TCLs exhibit Z score of >-1 (median Z in defective group = -0.8), despite median Z scores being significantly 1075 1076 different in Rb not altered vs. deficient TCLs. F-H. Scatter plots illustrating Z 1077 scores in 42 TNBC TCLs for three siMEM p<0.05 candidate Rb synthetic 1078 lethal effects, GPS1, SNRPF and SNW1 where median Z in defective group 1079 <-1 and median Z in proficient group >-2 (step two and three in (A)).

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Figure 3. Highly penetrant Rb synthetic lethal effects. Scatter plots
illustrating Z scores in 42 TNBC TCLs for 30 candidate Rb synthetic lethal
effects which pass Z score threshold assessment and demonstrate a
penetrance score of greater than 90%, as summarised in steps 1-5 of Figure
2A.

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1088 Figure 4. TAF1 and SKP2 as central nodes in highly-penetrant Rb 1089 synthetic lethal networks. A. Scatter plot illustrating Z scores in 42 TNBC TCLs for TAF1 from the data analysis illustrated in Figure 2A. B. Cytoscape 1090 network plot illustrating 33 highly penetrant (>80% penetrance) Rb synthetic 1091 1092 lethal effects identified as TAF1 transcription factor target genes, as annotated by ENCODE data^{12, 16}. **C-H.** Scatter plots illustrating Z scores in 42 1093 1094 TNBC TCLs for SKP1, SKP2, COPS1,3,4 and CKS1B from the data analysis 1095 in Figure 2A. I. Pathway diagram highlighting the role of multiple highly penetrant Rb synthetic lethal effects in the control of p27 activity. J. Volcano 1096 plot illustrating mRNAs from highly penetrant Rb SL genes that are 1097 1098 differentially expressed (limma analysis p<0.05) in 48 Rb defective vs. 92 Rb not altered triple negative breast tumours from the TCGA study¹⁰. Four highly 1099 SCF^{SKP}/COP9 complex genes, highlighted in red, demonstrate significantly 1100 higher mRNA expression levels in the Rb defective cell lines. K. As per panel 1101 (J), using data from the Metabric study⁴⁶ L. Box plots illustrating elevated 1102 SKP2 or GPS1 (COPS1) mRNA expression in Rb defective TNBC from both 1103 the TCGA ¹⁰ and Metabric studies⁴⁶. *p* values shown calculated with Wilcox 1104 1105 rank sum test.

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1108 Figure 5. Small molecule inhibition of SKP2 in Rb defective breast cell 1109 lines is synthetic lethal. A. Western blot illustrating loss of Rb expression in 1110 isogenic MCF10A non-tumour breast epithelial cells with constitutive 1111 expression of either a control non-targeting shRNA (shCONTROL) or an Rb-1112 targeting shRNA (shRB1). B. Western blot illustrating loss of SKP2 protein 1113 expression in MCF10A cells 48 hours after transfection with SKP2 or control, 1114 non-targeting, siRNAs (siCON1 or siCON2). C. Bar chart illustrating cell 1115 inhibitory effects in isogenic MCF10A cells with/without stable expression of 1116 Rb shRNA transfected with SKP2 siRNA. Cells were reverse transfected with 1117 siRNAs as shown and cultured for five continuous days, at which point cell viability was assessed by use of CellTitre Glo reagent. SKP2 siRNA caused 1118 1119 significant cell inhibition (p<0.001, Student's t test) in cells with stable Rb 1120 silencing, but not in cells with wild type Rb expression. D. Dose response survival curves illustrating the cell inhibitory effects of the SKP2 small 1121 molecule inhibitor, SKPinC1, in isogenic MCF10A cells with/without stable 1122 1123 expression of Rb shRNA. Cells were exposed to SKPinC1 for five continuous 1124 days at which point cell viability was assessed by use of CellTitre Glo reagent. 1125 Rb defective cells demonstrated a profound sensitivity, compared to Rb wild 1126 type cells (p<0.0001, two-way ANOVA, SF₅₀ = 1 μ M and >10 μ M in Rb defective and wild type cells, respectively). E. Tumour cell inhibitory effect of 1127 1128 SKPinC1 in 13 TNBC TCLs classified according to Rb status. Cells were exposed to 1 μ M SKPinC1 as in (H). Surviving fractions are shown (p=0.0022, 1129 Student's t test). F. Western blot illustrating p27 protein levels in Rb defective 1130 1131 cells exposed to 0.5 μ M and 1 μ M SKPinC1 for 24 hours.

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1134 Figure 6. SKP2 identified as a highly-penetrant Rb synthetic lethal effect in multiple, independently derived, datasets. A-C. Scatter plots comparing 1135 p values (-log₁₀) for Rb synthetic lethal effects identified in Colt2⁴², Achilles⁶² 1136 and DRIVE⁴³ datasets. p<0.05 effects in any single screen are shown. p<0.05 1137 1138 effects in two screens are shown the top right hand quadrant of each plot. 1139 SKP2 and SART3, which were identified in all three screens as synthetically 1140 lethal with Rb defects, are highlighted in red. p<0.05 effects in Colt2 data 1141 were identified by siMEM, followed by Z score and penetrance filtering (see Figure 2A); p<0.05 effects in DRIVE and Achilles data were identified by 1142 1143 median permutation t test, followed by Z score and penetrance filtering (see Figure 2A) **D.** Scatter plots illustrating Z scores in 42 TNBC TCLs for SKP2 1144 1145 from the Colt2 data analysis. E. Scatter plots illustrating Z scores in 16 TNBC 1146 TCLs for SKP2 from the Achilles data analysis. F. Scatter plots illustrating Z 1147 scores in 12 TNBC TCLs for SKP2 from the Drive data analysis.

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1149 Figure 7. SKP2 identified as a highly-penetrant Rb synthetic lethal effect 1150 in other histologies in two independently derived datasets. A. Scatter plot 1151 of 775 p<0.05 significant Rb synthetic lethal effects identified from the MP-test 1152 analysis of 373 non-breast cancer TCLs in the Drive study (step one in Figure 1153 1A). All 775 p<0.05 effects are ranked ordered by MP test p value. SKP2 and 1154 E2F3 are highlighted. **B.** Scatter plot illustrating RSA scores in 373 non-breast 1155 TCLs with Rb annotation for SKP2 sensitivity from the Drive data analysis. C. Scatter plot of 1467 p<0.05 significant Rb synthetic lethal effects identified 1156 1157 from the MP-test analysis of 467 non-breast TCLs in the Achilles study (step 1158 one in Figure 1A). All 1467 p<0.05 effects are ranked ordered by MP test p 1159 value. SKP2 and E2F3 are highlighted. **D.** Scatter plots illustrating Demeter 1160 scores in 1467 non-breast TCLs with Rb annotation for SKP2 sensitivity from 1161 the Achilles data analysis. E,F. Scatter plots illustrating RSA and Demeter 1162 scores in 63 and 115 lung TCLs with Rb annotation for SKP2 sensitivity from 1163 the Drive and Achilles studies, respectively. G. Scatter plot of intersect of cell 1164 line between the two datasets showing SKP2 sensitivity in Drive RSA scores 1165 (x axis) and Achilles Demeter scores (y axis) for selected histologies with only 1166 a single Rb defective line. This graph illustrates a trend between Rb defects 1167 and sensitivity to SKP2 shRNA across seven different histotypes.