APOBEC2 is a Transcriptional Repressor required for proper Myoblast Differentiation

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3132 ABSTRACT

The activation induced cytidine deaminase/apolipoprotein B editing complex (AID/APOBEC) 34 35 family comprises several nucleic acid editors with roles ranging from antibody diversification 36 to mRNA editing. APOBEC2, an evolutionarily conserved member of this family, has neither 37 an established substrate nor a mechanism of action, however genetic evidence suggests 38 functional relevance in tissues such as muscle. Here, we demonstrate that in muscle, 39 APOBEC2 does not have any of the attributed molecular functions of the AID/APOBEC 40 family, such as RNA editing, DNA demethylation, or DNA mutation. Instead, we show that 41 APOBEC2 occupies chromatin at promoter regions of certain genes, whose expression is 42 repressed during muscle cell differentiation. We further demonstrate that APOBEC2 on one 43 hand binds promoter region DNA directly and in a sequence specific fashion, while on the 44 other it interacts with HDAC transcriptional corepressor complexes. Therefore, APOBEC2, 45 by actively repressing the expression of non-myogenesis pathway genes, plays a key role in enforcing the proper establishment of muscle cell fate. 46

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49 The AID/APOBEC proteins are zinc-dependent deaminases that catalyze the 50 removal of the amino group from a cytidine base in the context of a polynucleotide chain, 51 resulting in cytidine (C) to uridine (U) transition on DNA or RNA. Members of the 52 AID/APOBEC family are closely related to one another based on homology and conservation 53 of the cytidine deaminase domain containing a zinc-dependent deaminase sequence motif ¹. However, they differ by tissue-specific expression, substrates, and biological functions 54 55 (reviewed in ²). Physiologically these proteins alter the informational content encoded in the 56 genome through a range of processes: acting as messenger RNA (mRNA) editors, affecting translation (APOBEC1) ^{3,4}, acting as DNA mutators to create novel gene variants, restrict 57 viruses and retrotransposons (AID and APOBEC3) (reviewed in ⁵) and, changing DNA 5mC 58 59 modification levels, leading to modulation of transcript abundance (AID and APOBEC1)^{6,7}.

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61 APOBEC2 is an evolutionarily conserved member of the AID/APOBEC family ⁸. 62 Substantial evidence highlights the biological relevance of APOBEC2 in metazoans. In mice, 63 APOBEC2 is highly expressed in cardiac and skeletal muscle where it affects muscle development ⁹. Specifically, in the absence of APOBEC2, there is a shift from fast to slow 64 65 muscle fiber formation, a reduction in muscle mass, and a mild myopathy with age ¹⁰. In 66 zebrafish, APOBEC2 has been implicated in muscle fiber arrangement ¹¹ and in retina and 67 optic axon regeneration ¹². In frogs, APOBEC2 is important in left-right axis specification 68 during early embryogenesis ¹³. Mutations and gene expression changes of APOBEC2 have 69 also been linked to cancer development ^{14,15}.

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71 Even though there is evidence for a biological role of APOBEC2, there are few 72 insights to the mechanism by which APOBEC2 accomplishes these. Moreover, there has 73 been no definite demonstration of its activity as a cytidine deaminase. Based on its 74 homology with the other AID/APOBEC family members, it has been hypothesized that APOBEC2 may be involved in RNA editing ^{9,14} or DNA demethylation ^{6,7,12}. It has also been 75 76 hypothesized that it has lost its deaminase activity altogether and may act biologically by a 77 different mechanism ¹⁶. However, there is currently a lack of knowledge on the direct 78 physiological targets of APOBEC2, and its mechanism of action.

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To answer some of these questions, we performed knockdown studies of APOBEC2 during the differentiation of the C2C12 murine myoblast cell line to systematically characterize the transcriptome and DNA methylation patterns of APOBEC2 deficient C2C12 cells. Our results confirm the requirement of APOBEC2 for myoblast to myotube differentiation. Additionally, we demonstrate the requirement of its amino-terminal disordered region for nuclear localization and myotube differentiation. While our results do not support

86 APOBEC2 roles on RNA editing and on DNA methylation, we find that APOBEC2 87 downregulation leads to substantial gene expression changes affecting programs associated 88 with myogenesis. Moreover, genomic occupancy experiments demonstrate that APOBEC2 89 interacts with chromatin at promoters of genes that are repressed during myoblast 90 differentiation. Furthermore, these targets are derepressed with reduced abundance of 91 APOBEC2, which allude to APOBEC2 acting as a transcriptional repressor. Notably, these 92 target derepressed genes are not directly involved in myogenesis or muscle differentiation; 93 instead they seem enriched for genes in the innate immune / inflammatory pathway. Finally, 94 we show that APOBEC2 directly interacts with DNA as well as Histone Deacetylase 1 95 (HDAC1) repressor complexes, which supports the molecular function of APOBEC2 as a 96 transcriptional repressor. Taken together, our data suggest that APOBEC2 has a direct role in regulating active gene transcription during myoblast differentiation as a transcriptional 97 98 repressor.

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APOBEC2 is required for myoblast to myotube differentiation

102 The C2C12 myoblast cell line was derived from mouse satellite cells activated to 103 proliferate after muscle injury in adult mice ¹⁷. C2C12 myoblasts are thought to recapitulate 104 the first steps of muscle differentiation in culture and upon differentiation induce APOBEC2 105 expression ¹³ (Supplementary Fig. 1A), making them a suitable model to investigate putative 106 roles of this suspected cytidine deaminase in situ.

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To explore the role of APOBEC2 during myogenesis, we reduced APOBEC2 protein levels with short hairpin RNA (shRNA) against APOBEC2 mRNA. Protein reduction was efficient and coincided with reduced myoblast-to-myotube differentiation, evidenced by the decrease in expression of TroponinT and myosin heavy chain (MyHC), protein markers of late differentiation (Fig. 1A). At the cellular level downregulation of APOBEC2 protein levels coincided with reduced myotube formation (Fig. 1B). These observations match those previously reported using mouse embryonic stem cell-derived myogenic precursors ¹⁸.

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We then restored APOBEC2 protein in these knockdown cells through overexpression. This led to an increase in MyHC levels, which confirms the essential role of APOBEC2 in myoblast differentiation in this model (Fig.1C). Additionally, we produced truncated APOBEC2 (residues 41-224 mouse APOBEC2) based on its published structures ^{19,20}. Truncated APOBEC2 (del(2-40)A2), was unable to restore MyHC expression levels. Interestingly, the truncated form of APOBEC2 was only found in the cytoplasmic fraction of differentiated C2C12 myoblasts (Supplementary Fig. 1C). We speculated that the nuclear

localization of APOBEC2 was necessary for its role in muscle differentiation and pointed to amolecular function within the nuclear compartment.

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APOBEC2 loss leads to gene expression changes related to muscle differentiation

To study how APOBEC2 loss leads to problems in C2C12 myoblast-to-myotube 128 129 differentiation, we performed RNA sequencing (RNA-Seq) to compare the transcriptome 130 dynamics of APOBEC2 knockdown and control cells during differentiation. We observed that 131 reduced APOBEC2 levels led to substantial gene expression changes (Fig. 1D,E). Notably, 132 genes downregulated during differentiation were enriched for muscle development Gene 133 Ontology (GO) terms, whereas genes upregulated were enriched for GO terms related to 134 immune response (Fig. 1C; Supplementary File 1). The decreased expression of muscle 135 differentiation related genes reflects the observed reduced myotube formation. Though 136 undetectable on the immunoblot (Fig. 1A), perturbation of APOBEC2 levels prior to inducing 137 differentiation seems to affect the potential of C2C12 to differentiate into myotubes. Genes 138 involved in muscle development or function were also downregulated at day 0 prior to 139 inducing the cells to differentiate.

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Furthermore, genes related to cell development or differentiation GO terms, particularly immune system development, blood vessel development, and nervous system development, are among those overrepresented in the list of upregulated genes with APOBEC2 deficiency (Supplementary File 2). These spurious non-muscle transcriptional programs possibly reflect transdifferentiation events, which have been observed for C2C12 myoblasts ²¹.

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148 We next investigated possible molecular mechanisms of how APOBEC2 causes these gene expression changes. Due to the conserved cytidine deaminase domain within 149 150 the AID/APOBEC family, APOBEC2 is posited to be an RNA editor ^{9,14} and a DNA 151 demethylase ^{6,7,12}. Upon comparing the transcriptomes of the APOBEC2 knockdown and 152 control C2C12 cells for instances of C-to-U RNA editing, using a previously validated 153 pipeline ²², we could not identify C-to-U or A-to-I RNA editing events that were APOBEC2 154 dependent (Supplementary Fig. 2A). Similarly, using bisulfite sequencing, we were unable to observe significant methylation differences between the APOBEC2 knockdown and control 155 C2C12 cells that could account for the gene expression changes (Supplementary Fig. S2B-156 157 C). Altogether, these results strongly indicate that APOBEC2 is neither involved in mRNA 158 deamination nor DNA demethylation in differentiating muscle.

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161 APOBEC2 binds promoter regions

162 Cytidine deaminases of the AID/APOBEC family can bind and mutate DNA either at 163 gene bodies, e.g. exons of the immunoglobulin locus, as catalyzed by AID, reviewed in ²³ or 164 165 at promoter regions, e.g. local hypermutations as catalyzed by APOBEC3 family members -166 reviewed in ²⁴. To assess whether APOBEC2 could also bind genomic DNA, we first 167 determined the subcellular localization of APOBEC2 in muscle cells. We observe that 168 APOBEC2 is present in both the cytoplasm and nucleus of differentiated C2C12 myotubes (Fig. 2A, Supplementary Fig. 1C). A weak nuclear localization signal (NLS) can be predicted 169 170 by cNLS Mapper ²⁵ (APOBEC2 residues 26 to 57, with a score of 3.7) that could explain the 171 lack of nuclear localization of truncated APOBEC2, del(2-40)A2. However, full length 172 APOBEC2 does not show NLS activity but is homogenously distributed throughout the cell. 173 presumably through passive diffusion ²⁶. To then assess whether nuclear APOBEC2 could 174 bind chromatin, we utilized sequential salt extraction based on the principle that loosely bound proteins will be dissociated at low salt concentration, while tightly bound ones will not 175 176 ²⁷. Using this technique, we found that a fraction of APOBEC2 within differentiating C2C12 cells, remains bound to chromatin even at high salt concentrations (up to 1 M NaCl). As a 177 178 comparison histone H4 dissociates completely from DNA at 0.75 M NaCl (Fig. 2B). These 179 data suggest a strong association of nuclear APOBEC2 with chromatin in differentiating 180 myoblasts.

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182 To determine APOBEC2 binding localization within chromatin, we performed 183 chromatin immunoprecipitation-sequencing (ChIP-Seq) experiments, and calculated 184 enrichment of APOBEC2 at specific loci over input using MACS2 (see Methods). We 185 performed each experiment in triplicate, and only peaks that were called in at least 2 out of 3 186 replicates were analyzed. Importantly, we queried APOBEC2 occupancy at two different time 187 points, 14- and 34-hours post-differentiation, that precede the RNA-Seq time points, where 188 we observed changes in gene expression and represent time points of low and higher 189 APOBEC2 protein abundance. Overall, the signal around peak summits of transcription start 190 sites (TSS) is higher at 34 versus 14 hours, reflecting an increase of APOBEC2 in 191 chromatin. In contrast, input controls show lower enrichment over the peak summits (Fig. 192 2C).

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Annotating the APOBEC2 peaks by genomic feature showed that for both time points most of the peaks fall within promoters, defined as regions -2 kilobases (kb) to +2 kb around the TSS (Fig. 2D). Next, we wanted to determine whether APOBEC2 occupies specific motifs within promoter regions. The members of the AID/APOBEC family that bind to DNA

198 have some local sequence preference with regard to sites they mutate, but do not display 199 rigorous sequence specificity, e.g. akin to transcription factor (TF) binding sites ²⁸. To assess 200 the motif signatures in those regions we queried through motif enrichment the over-201 representation of TF 8-mers sequences associated to main TF modules ²⁹. Among 108 TF 202 modules tested against a controlled background of negative sequences (see Methods), we 203 observed 19 of those significantly enriched at APOBEC2 regions in at least one 204 differentiation time point. SP and KLF motifs were among the top enrichments observed (Fig. 205 2E), suggesting a co-regulatory role between these factors and APOBEC2. In general, TF 206 specificity groups increase their significance between 14 and 34 hours but with lower overall 207 effect sizes or only at the later time points, suggesting that the interplay between TFs and 208 APOBEC2 occupied regions is specific at 14 hours but broader at later time points, likely as 209 consequence of events related to its early binding. Additionally, we did not observe 210 APOBEC2 related DNA mutation at the occupied peaks, indicating that APOBEC2 is not a 211 DNA mutator like other AID/APOBEC family members (Supplementary Fig. 3).

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APOBEC2 represses expression of occupied genes

215 Focusing on the promoter bound genes we determined that there are ~1500 genes that are bound by APOBEC2 near their transcription start sites in any of the time points, 590 216 217 of which are occupied at both time points (Fig. 3A). Overall, about 79% of the genes that are 218 bound by APOBEC2 in the 14-hour time point remain bound at the 34-hour time point. 219 Further, using Gene Set Enrichment Analysis to determine the distribution of APOBEC2 220 occupied genes at both time points in a list of ranked expression changes either through 221 differentiation (day 0 to 2) or upon APOBEC2 knockdown (A2 shRNA vs GFP shRNA at day 222 2). Our results show that APOBEC2 occupied genes are significantly enriched at genes 223 downregulated through differentiation (Fig. 3B). Moreover, upon APOBEC2 knockdown, 224 APOBEC2 occupied genes are instead enriched at upregulated genes (Fig. 3C). 225 Interestingly, overall expression of APOBEC2 bound genes through C2C12 differentiation 226 from day 0 to 2 is significantly decreased during differentiation compared to genome wide 227 expression changes (-0.3 versus 0.1 mean log2 fold changes; t = -8.2; adjusted P-values < 228 0.0001; Fig. 3D). Furthermore, expression of APOBEC2 occupied genes increase upon 229 APOBEC2 knockdown (t-statistic = 4.0; adjusted P-value < 0.001) highlighting a global 230 repressive role of APOBEC2 during differentiation. Altogether, these results suggest that the 231 observed transcriptional changes linked to APOBEC2 deficiency are due to APOBEC2 232 functioning as a transcriptional repressor.

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To validate its possible role as a transcriptional repressor, we selected candidate genes repressed by APOBEC2 occupancy. We narrowed it down to a list of genes occupied

236 by APOBEC2 which are downregulated during differentiation (day 2) and upregulated with 237 APOBEC2 knockdown. In this gene list, we did not find an overrepresentation of GO terms 238 relating to muscle differentiation; but rather, we found terms related to development of other 239 lineages (Supplementary Table S1) similar to the non-muscle genes upregulated with 240 APOBEC2 deficiency. Furthermore, a list of 31 genes related to differentiation (GO: 0045595 241 regulation of differentiation) in other tissue contexts were upregulated with APOBEC2 loss 242 (Fig. 3E). We speculate that APOBEC2 is acting as a repressor to direct C2C12 243 differentiation into the muscle-lineage by repressing specious gene networks related to other 244 lineages.

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APOBEC2 binds directly to single- and double-stranded DNA

Thus far, the results suggest that APOBEC2 has a molecular role unique to the AID/APOBEC family. While it does not have the capacity to modify nucleic acids (RNA or DNA) through deamination ^{30,31}, it is capable of binding chromatin to regulate transcription. This implies either that APOBEC2 interacts directly with DNA at promoter regions, or that it interacts with transcription regulators that do so, or both.

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254 To assess whether APOBEC2 is capable of binding DNA directly, we generated 255 recombinant APOBEC2 (expressed in insect cells, Supplementary Fig. 4A) and tested it in 256 vitro for its capacity to bind DNA oligonucleotides that represent APOBEC2 bound 257 sequences in vivo. We tested an SP/KLF DNA motif as it was highly represented in the 258 ChIP-Seg data (Figure 4A). As a negative control, we tested an A-tract motif that based on 259 the ChIP-Seq was not bound by APOBEC2 but which occurred frequently at promoters of 260 differentially regulated genes. Using microscale thermophoresis (MST), purified APOBEC2 261 bound both single-stranded and double-stranded DNA containing the SP/KLF motif with 262 reasonable affinity (calculated Kd = 253 ± 86 nM and 709 ± 186 nM, respectively, Figure 4A, 263 S4B). In contrast, measurements for the A-tract motif showed large variances between trials, 264 indicating lack of specific binding (Figure S4C). This ability to selectively bind both ssDNA 265 and dsDNA in a sequence-specific manner in vitro (Fig. 4A) and in vivo (Fig. 2E) makes APOBEC2 unique among AID/APOBEC family members. These results demonstrate that 266 267 APOBEC2 can target specific genes at promoter regions, and through this binding repress 268 transcription.

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Using gel electromobility shift assays (EMSA), we checked for APOBEC2 cooperativity with the transcription factor SP1 on its cognate motif. As expected, purified SP1 protein alone shifted dsDNA with the SP/KLF motif (lane 7, Fig. 4B). Unexpectedly, we were unable to see a shift by APOBEC2 alone on either ssDNA or dsDNA unlike with MST (lanes

274 3 and 6, Fig. 4B). This could be due to the running conditions which were unable to preserve 275 the presumably weaker intermolecular affinity. However, APOBEC2 together with SP1 276 produced a shift with stronger intensity indicating cooperativity between the two DNA binding 277 proteins (lane 7 vs lane 8, Fig. 4B). Interestingly, the two do not interact directly 278 (Supplementary Fig. 4D), suggesting that the enhanced SP1 binding is mediated through 279 DNA – possibly through changes in DNA conformation given that APOBEC2 has no 280 detectable deaminase activity. This cooperativity suggests a molecular mechanism wherein 281 APOBEC2 alters transcription factor affinity for specific motifs leading to transcriptional 282 repression.

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APOBEC2 interacts with corepressor complexes in vivo

286 To ascertain the mechanism of action for the observed transcriptional repression, we 287 used proximity-dependent protein biotinylation (BioID) to confirm and identify other putative 288 APOBEC2 proximal protein complexes that may indicate direct interactions. After statistical 289 curation we identified 124 proteins that were significantly more tagged by APOBEC2-BirA 290 and/or BirA-APOBEC2 than GFP-BirA controls in C2C12 cells (Supplementary Table S2). 291 Functional annotation showed that many APOBEC2 neighboring proteins are related to cell 292 membrane and cytoskeleton organization processes, in line with its high cytoplasmic 293 abundance (Fig. 2B), but terms related to chromatin modification and histone deacetylation 294 were also enriched (Supplementary Fig. 5A and Supplementary Table S3). Of special 295 interest was the identification the histone deacetylase 1 (HDAC1) and Chromodomain-296 helicase-DNA-binding protein 4 (CHD4), both components of the nucleosome remodeling 297 and histone deacetylation (NuRD) transcriptional corepressor complex ³², (Fig. 4C and 298 Supplementary Fig. 5B). Using co-IP, we validated that APOBEC2 interacts with HDAC1 in 299 differentiated C2C12 myoblasts (Fig. 4D). Together with the observation that APOBEC2 300 interacts directly with chromatin, this suggests that APOBEC2 plays a role in gene regulation 301 through epigenetic nucleosome modification with these transcriptional corepressor 302 complexes.

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Taken together, our results suggest a direct role of APOBEC2 in repressing specific transcribed genes, likely mediated by an HDAC1 co-repressor complex during C2C12 myoblast differentiation (Fig. 4F).

308 DISCUSSION

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There have been many hypothesized roles for the cytidine deaminase APOBEC2. Here we show that the expression of APOBEC2 during myoblast differentiation has consequential effects on myotube formation owing to at least one unexpected molecular

313 function: transcriptional control. We discovered that APOBEC2 loss, leads to faulty myoblast 314 differentiation and concomitant gene expression changes. We show that these gene 315 expression changes come about through direct chromatin interaction of APOBEC2 at 316 promoters – with no observed APOBEC2 related changes in RNA editing, DNA methylation, 317 or DNA mutation. We find instead that APOBEC2 is capable of directly binding DNA and 318 interacting with corepressor complexes. Indeed, through this interaction, APOBEC2 319 specifically represses transcriptional programs unrelated to muscle differentiation thus 320 indirectly supporting proper muscle differentiation.

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322 The deaminase domain of APOBEC2 appears to have lost catalytic activity ^{30,31}. 323 However, it seems to have retained its ability to bind nucleic acids. Binding ssDNA is nothing 324 new to members of the AID/APOBEC family ³³. However, the demonstration that APOBEC2 325 directly binds to dsDNA is groundbreaking for the family. There has been evidence for 326 APOBEC3-driven dsDNA mutation but no evidence for direct dsDNA binding; instead, the 327 mutational mechanism is likely through ssDNA ^{34,35}. Prior work has linked APOBEC2 328 overexpression to RNA editing of specific transcripts as observed in the healthy livers of 329 transgenic mice which eventually develop hepatocellular carcinoma ¹⁴. Notably, RNA editing 330 was only detectable in the liver at specific transcripts for the transgenic mice. However, 331 based on our transcriptome analysis, we were unable to find evidence of such RNA editing 332 in our myoblast differentiation models. Prior work has also reported mild effects of 333 APOBEC2 on DNA methylation specifically at the MYOG promoter ¹⁸; yet from our ChIP-Seq 334 data, we do not find occupancy at the MYOG gene. Furthermore, there is conflicting data on 335 the role of the AID/APOBEC family in active DNA demethylation; and, APOBEC2 dependent demethylation has not been found in other cellular contexts ^{36,37}. A role in transcriptional 336 337 repression such as we propose, would also explain these prior data, as indirect effects of 338 improper differentiation.

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Previous studies suggested that recombinant APOBEC2 is incapable of binding DNA or deaminating it ^{16,30,31}. Our experiments using recombinant protein produced in eukaryotic cells show that APOBEC2 is capable of binding DNA with reasonable affinity, and when it does so, it alters the ability of proximal transcription factors (such as SP1) to interact with their cognate motifs. Similar findings have also been observed with the transcription factor 9006F2, suggesting but not proving a role for APOBEC2 in transcription ¹⁶.

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Furthermore, our proximity ligation experiments reveal that APOBEC2 protein directly interacts with HDAC1 and CHD4 containing repressor complexes providing a direct epigenetic mechanism for the repression of APOBEC2 occupied genes. HDAC1 and CHD4

are components of the NuRD corepressor complex, whose involvement has already been proposed in skeletal or myocardial muscle fate determination ³⁸. Additionally, the abundance of APOBEC2 in the cytoplasm and its proximity to proteins that participate in cell morphology and the cytoskeleton could either reflect a mechanism limiting APOBEC2 access to the nucleus, or perhaps an additional function that remains to be studied.

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356 Uniquely amongst AID/APOBEC family members, the amino-terminus of APOBEC2 357 contains a region that is glutamate-rich and intrinsically disordered ²⁰. Loss of this 358 unstructured domain results in inability of the protein to rescue the knockdown phenotype 359 likely through the loss of its nuclear retention. Proteins with similar disordered regions have 360 been shown to form liquid phase separated membrane-less compartments and this function 361 appears to be especially important for transcriptional regulation as transcription factors with 362 disordered regions have been shown to form such compartments ³⁹. Even though, no 363 enzymatic activity has been detected for APOBEC2, potential substrates could be present in 364 these chromatin compartments: transient RNA species, eluding detection limits by our RNA-365 Seq method, or, potentially, ssDNA structures at melted promoters. We propose that through 366 its N-terminal unstructured region APOBEC2 is retained in the nucleus where on one hand it 367 binds DNA at promoter regions in a sequence specific fashion, and on the other, it recruits 368 corepressor complexes to repress transcription.

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We hypothesize that APOBEC2 acts as a modulator of its bound promoters during the myogenic program – fine-tuning it for muscle differentiation and repressing other lineage programs. MYOD1 has been shown to bind and activate lineage programs outside the muscle lineage; however, this is mitigated by corepressors ⁴⁰. Furthermore, as APOBEC2 expression under healthy conditions is mostly confined to muscle tissue (both skeletal and heart), where it might be acting as a 'many-but-muscle' lineage repressor – similar to MYT1L in neuronal differentiation ⁴¹.

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378 The discovery that APOBEC2 has a direct role in transcriptional control impacts how 379 we interpret the phenotypes that have been attributed to it in the mouse knockout models 380 and other biological systems, well beyond muscle differentiation - for example zebrafish 381 retina and optic nerve regeneration, Xenopus embryo development, and cancer 382 development ^{12–14}. In the zebrafish models, APOBEC2 loss leads to similar defective muscle 383 phenotypes but it is deemed essential in the retinal regeneration model - where cellular reprogramming is a key step ¹². Directly or indirectly, these prior observations likely relate to 384 385 aberrant transcriptional programs, normally silenced in the context of tissue development or 386 cell differentiation due to APOBEC2 transcriptional control. Taken together, we postulate that

- 387 APOBEC2 is a transcriptional repressor that modulates transcriptional programs during cell
- 388 differentiation or reprogramming.
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390 Methods

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392 Data availability393

High throughput sequencing datasets are all found in: GSE117732 and more specifically:
 RNA-Seq (GSE117730); ChIP-Seq (GSE117729); ERRBS (GSE117731). Mass
 spectrometry data for BioID performed in Flp-In 293 T-REx cells have been deposited in
 MassIVE under ID.

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399 C2C12 cell culture

401 C2C12 cells (CRL-1772, ATCC) were maintained in DMEM (30-2002, ATCC) with 10% fetal 402 bovine serum and fed every two days. To differentiate equal number of cells (2.5 x 10⁵) 403 were seeded in 6-well plates followed by media change to DMEM with 2% horse serum after 404 12 hours. For generating single cell clones for RNA- Seq and RRBS experiments C2C12s 405 were sorted using fluorescence-activated cell sorting (FACS) and seeded into a 96 well 406 plate. Each clone was expanded and tested for successful knockdown through 407 immunoblotting.

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409 APOBEC2 knockdown and overexpression

411 C2C12 cells were infected with lentiviruses carrying shRNA, targeting either APOBEC2 or 412 GFP. All APOBEC2 shRNAs were obtained from The Broad Institute's Mission TRC-1 413 mouse library and present in pLKO.1-puro construct. Plasmids used: pLKO.1 - TRC cloning 414 vector (Addgene, # 10878)⁴²; pLKO.1 puro GFP siRNA (Addgene, # 12273)⁴³. The design 415 of shRNAs and cloning in pLKO.1-TRC, were done according to the Addgene protocol 416 (Protocol Version 1.0. December 2006). The following shRNAs sequences were used for 417 APOBEC2 knockdown: A2 shRNA: GCTACCAGTCAACTTCTTCAA; GFP shRNA: 418 GCAAGCTGACCCTGAAGTTCAT.

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420 Virions were produced by co-transfection of pLKO.1-puro shRNA containing construct, 421 packaging plasmid psPAX2 (Addgene, #12260) and envelope plasmid pMD2.g (Addgene, 422 #12259) in 293T cells (CRL-3216, ATCC). Transfections were done using Lipofectamine 423 2000 Reagent (Invitrogen) as per manufacturer instructions. Supernatants with lentiviral 424 particles were collected at 24 and 48 hours after transfection, passed through a 0.45 mm 425 filter and applied to C2C12s. For APOBEC2 constitutive knockdown, C2C12 cells were 426 infected with pLKO.1 containing lentiviruses in growth media containing 8 µg/mL polybrene 427 for 12 hours. Two days after lentiviral infection cells were cultured with 4 µg/ml puromycin 428 containing media for two more days to select cells stably expressing the shRNA.

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Rescue constructs, mouse APOBEC2 and del(1-41)APOBEC2 with silent mutations to
escape shRNA knockdown, and tagged constructs, APOBEC2 and del(1-41)APOBEC2 with
C-terminal 3xHA-tags, were cloned into pMXs-GFP/Puro retroviral vectors (Cell Biolabs,
Inc.). Virions were produced in 293T cells by co-transfection with pMXs construct and pCLEco (Novus Biologicals) using Lipofectamine 2000.

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437 *Immunoblotting and co-immunoprecipitation*

For immunoblotting experiments, C2C12 cells were first washed with cold PBS and lysed in
 100 μl RIPA lysis buffer (Santa Cruz, sc-24948) in 6-well plates. They were incubated at 4°C

441 for 15 minutes and then extracts were scrapped into a microfuge tube. Lysates were snap 442 frozen in liquid nitrogen. After thawing the lysates on ice and clearing out cell debris by 443 centrifugation, equal amounts of total protein (ranges between 10-30 µg) were boiled in 444 SDS-PAGE sample buffer and loaded onto each lane of a polyacrylamide gel (Criterion XT 445 Bis-Tris Gel 12%, Bio-Rad). Following electrophoresis, the resolved protein was transferred 446 to a nitrocellulose membrane and subjected to western blot analysis. The source and dilution 447 for each antibody used were: polyclonal rabbit-APOBEC2 (gift from Alin Vonica MD, PhD), 448 1:1000; monoclonal mouse-APOBEC2 (clone 15E11, homemade), 1:5000; TroponinT clone 449 JLT- 12 (T6277, Sigma-Aldrich), 1:500; alpha-tubulin DM1A (Abcam, ab7291), 1:5000; 450 MyHC MF-20 (DSHB), 1:20; rabbit anti-SP1 (Merck, 07-645), 1:1000; and rabbit anti-HDAC1 451 antibody (ab7028), 1:2000.

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For co-immunoprecipitation experiments, 4 x 10⁶ C2C12 cells were plated and lysed after 4 453 454 days in differentiation medium. Cells were trypsinized, washed with cold PBS, and lysed in 1 455 mL cell lysis buffer: 0.5% Tween 20, 50 mM Tris pH 7.5, 2 mM EDTA, and freshly added 1X 456 Halt protease and phosphatase inhibitor cocktail EDTA-free (Thermo, 78441). Mixture was 457 vortexed and incubated on ice for 10 min. twice. Nuclei were separated from the cytoplasmic 458 fraction by centrifugation (6000 g, 1 minute, 4°C). Nuclei were washed with 1 mL cold PBS 459 before lysing in 250 µL high salt nuclear lysis buffer: 800 mM NaCl, 1% NP40 (Igepal CA-460 640), 50 mM Tris pH 7.5, and freshly added 1X protease and phosphatase inhibitor cocktail, 461 EDTA-free. Mixture was vortexed and incubated on ice for 10 min, twice. Nuclear lysate was 462 then diluted to a final salt and detergent concentration of 400 mM NaCl and 0.5% NP40 463 using 250 µL dilution buffer: 50 mM Tris pH 7.5 and 1X protease and phosphatase inhibitor 464 cocktail, EDTA-free. Nuclear lysates were treated with Benzonase (Merck-Millipore, 70664). 465 Nuclear lysates were pre-cleared on 25 µL Dynabeads M-280 Sheep anti-mouse or anti-466 rabbit IgG (Thermo, 11201D/12203D), depending on primary IgG antibody. Pre-cleared 467 nuclear lysate was then added to 50 μ L beads conjugated with 2-4 μ g primary IgG antibody: 468 rabbit anti-APOBEC2 (Sigma, HPA017957) or rabbit lgG isotype control. 469 Immunoprecipitation was done overnight at 4C with rotation. Beads were thoroughly washed 470 before resuspending and boiling in SDS-PAGE sample buffer.

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472 Immunofluorescence staining and fusion index of C2C12 cells

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474 C2C12 cells (5x10⁴ cells) were seeded in collagen-coated coverslips (BD Biosciences, 475 356450) in 12- well plate the day before inducting of differentiation with 2% horse serum. 476 They were washed with cold PBS and fixed with paraformaldehyde (4%) in PBS for 10 477 minutes at 4 °C. This was followed by 2 washes, 5 minutes each at room temperature and 478 blocking solution (0.5% BSA, 1% gelatin, 5% normal goat serum, 0.1% Triton) in PBS for 1 479 hour at room temperature. This was followed by overnight stain with antibodies in a 480 humidified chamber at 4°C, three washes with cold PBS 5 min each at room temperature. 481 Coverslips were then incubated with secondary antibodies for 1 hour at room temperature, 482 followed by three washes with PBS 5 min at room temperature. Immunofluorescence 483 staining of C2C12 cells was carried out with primary antibodies: MyHC MF20 (DSHB) and 484 FLAG M2 (Sigma, F1804). Nuclei were counterstained and coverslips were mounted with 485 VECTASHIELD Antifade Mounting Medium with DAPI (Vector Laboratories, H-1200). 486 Images were taken using confocal Leica TCS SP5 II or widefield Zeiss Cell Observer and 487 image analysis was done with Fiji/ImageJ.

488

490

489 Chromatin salt-extraction profiling

491 C2C12 cells were seeded in equal numbers (2x10⁶ cells) and induced to differentiate after
492 12 hours. Five days after differentiation they were lysed in the plate with 100 µl sucrose lysis
493 buffer (320 mM sucrose, 0.5% NP-40, 10 mM Tris pH 8.0, 3 mM CaCl₂, 2 mM Mg acetate,
494 0.1 mM EDTA). Extracts were incubated for 5 minutes on ice and spun at 500 g for 5
495 minutes to collect the nuclear pellet and supernatant as the cytosol fraction. Nuclear pellets

496 were washed with no-salt Nuclei Buffer (50 mM Tris pH 8, 1% NP-40, 10% glycerol). 497 Following the washes the nuclear proteins were extracted at increasing concentrations of 498 NaCl from 250 mM up to 2 M in Nuclei Buffer during which they are homogenized using 499 dounce tissue grinders (Fisher, K8853000000), incubated on ice for 10 minutes and spun at 500 4°C for 10 additional minutes. Eluted material was collected, resolved on polyacrylamide gel 501 electrophoresis (Criterion XT Bis-Tris Gel 12%, Bio- Rad) and immunoblotted with specific 502 antibodies: Histone H4 (Merck, 05-858R), 1:5000; monoclonal mouse-APOBEC2 (clone 503 15E11, homemade), 1:5000; α-tubulin DM1A (Abcam, ab7291), 1:5000.

504

505 RNA expression analysis

506

507 Library preparation and sequencing were done by Rockefeller University Genomics 508 Resource Center [https://www.rockefeller.edu/genomics/] using TruSeq Stranded mRNA 509 Sample Prep kit as per manufacturer's instruction. The procedure includes purification of 510 poly-adenylated RNAs. Libraries were sequenced with 50bp paired-read sequencing on the 511 HiSeq2500 (Illumina). Paired end read alignments and gene expression analysis were 512 performed with the Bioinformatics Resource Center at Rockefeller University, Paired-end reads were aligned to mm10 genome using the subjunc function in the Bioconductor 513 514 Rsubread ⁴⁴ package and bigWig files for visualization were generated from aligned reads using the Bioconductor rtracklayer ⁴⁵ and GenomicAlignments packages ⁴⁶. For analysis of 515 differential expression, transcript quantifications were performed using Salmon ⁴⁷ in quasi-516 517 mapping mode. Gene expression values were calculated from transcript quantifications 518 using tximport ⁴⁸. Gene expression changes were identified at a cut off of 5% FDR 519 (benjamini-hockberg correction) using the Wald test implemented in DESeg2⁴⁹. Annotation 520 files used: BSgenome.Mmusculus.UCSC.mm10(v1.4.0);org.Mm.db(v3.5.0); 521 TxDb.Mmusculus.UCSC.mm10.knownGene.gtf.gz(v3.4.0)

522 523 RNA editing analysis

524 RNA editing analysis was performed as previously reported elsewhere ²². Editing detection 525 526 was performed by comparing C2C12 control samples (GFPsh) to APOBEC2 knockdown 527 samples using RNA-Seq datasets in triplicates for each sample. Minimum filters include 528 quality control thresholds (minimum of five reads covering the putative site with at least two 529 reads supporting the editing event; filtering of reads that contain indels or support an edit in 530 the first or last two base pairs of a read). Stringent filters applied to the APOBEC1 531 dependent C-to-U edited sites include all of the above and additionally the magnitude of the 532 control vector was at least 15 and the angle between the wild-type and knockout vectors was 533 at least 0.11 radians, as described in the paper referenced in this section.

- 534
- 535 *Enhanced reduced representation bisulfite sequencing (ERRBS)* 536

537 ERRBS library preparation, sequencing and read alignment was performed by the 538 Epigenomics Core Facility of Weill Cornell Medicine [epicore.med.cornell.edu/] as previously 539 described ^{50,51}. The procedure includes bisulfite conversion of the DNA. Libraries were 540 sequenced with 50bp single reads (SR) in HiSeq2500 (Illumina). Reads were aligned to a 541 bisulfite converted reference mouse genome with Bismark ⁵². The methylation context for 542 each cytosine was determined with scripts from the core facility.

543

Here coverage of specific genomic regions by ERRBS dataset, refers to the percent of features (eg percent of promoters, CpG islands) that contain at least one CpG that is well covered (> 10x). For gene specific annotations the mm10 UCSC knownGene annotations from the UCSC table browser were used and for CpG islands the mm10 cpgIslandExt track of the UCSC table browser. Genomic features were defined as: CpG islands, CpG island shores were defined as 2kb upstream and downstream of a CpG island; Gene promoters 550 (region 2kb upstream and 2kb downstream of the TSS), exons, introns and intergenic 551 regions.

552

553 *Differential methylation analysis* 554

MethylKit (v1.3.8) ⁵³ was used to identify differentially methylated cytosines (DMCs) with q-555 556 value less than 0.01 and methylation percentage difference of at least 25% after filtering 557 ERRBS dataset by coverage, normalizing by median and including CpG sites that are 558 covered >10x, in 3 out of 5 biological replicates (lo.count = 10, lo.perc = NULL, hi.count = 559 1000, hi.perc = 99.9), (destrand=TRUE,min.per.group=3L). eDMRs (v0.6.4.1) ⁵⁴ was used to 560 empirically determine differentially methylated regions, using the DMCs identified with 561 methylKit. In order for a region to be defined as a DMR, default parameters (num.DMCs=1, 562 num.CpGs=3, DMR.methdiff=20) of eDMR were used, so that each region has: (1) at least 1 563 DMC in the region, as determined using methylKit, (2) at least 3 CpGs included in the region 564 and (3) absolute mean methylation difference greater than 20%. For a region to be defined as a significant DMR, default parameters were used (DMR.qvalue = 0.001, mean.meth.diff = 565 566 20, num.CpGs = 5, num.DMCs = 3) so that each significant DMRs has (1) 5 CpGs where at least 3 of them are significant DMCs as determined by methylKit (2) have a minimum 20% 567 568 methylation change for the region.

569

570 Chromatin immunoprecipitation method

571 572 C2C12s were plated at ~70% confluence 12 hours prior to inducing differentiation (seed 573 ~2x10⁶ cells) maintained in DMEM (ATCC, 30-2002) with 10%FBS. This was followed by 574 media change to DMEM with 2% horse serum (Life Biotechnologies, 26050-088) to induce 575 differentiation. The cells (~5x10⁶ /10cm plate) were harvested at 24-hour or 34-hour after 576 inducing differentiation. They were fixed on plate with 1% PFA in PBS for 10 minutes at RT. 577 Glycine was added to a final concentration of 125mM. Cells were washed 2 times with 1x PBS with protease inhibitor cocktail (Roche, 11836170001). They were lysed on the plate 578 579 with cold Farnham lysis buffer to ~10x10^6 cells /mL (5mM PIPES pH 8.0, 0.5% NP-40, 580 85mM KCI,1mM EDTA, PIC) and incubated rotating for 15min at 4°C. Lysates were scraped 581 off the plates, pelleted and resuspended in LB2 (10 mM Tris pH 8.0, 200 mM NaCl, 1 mM 582 EDTA, 0.5 mM EGTA, PIC) and incubated rotating for 15 minutes at 4°C and then 583 centrifuged. Pellets were resuspended to 5x10⁷ cells/mL in LB3 (10 mM Tris pH 8.0, 100 584 mM NaCl, 1 mM EDTA, 0.5 mM EGTA, 0.1% sodium-deoxycholate, 0.5% sodium lauroyl 585 sarcosinate, PIC) until suspension was homogenized. Samples were then sonicated using 586 Covaris ultrasonicator model S220 for 15 minutes with the following settings: 140W peak power, 5% duty, 200 cycles per burst. TritonX-100 to a final concentration of 1% was added 587 588 to the samples. Samples were clarified by centrifugation at 20,000 g for 10 minutes at 4°C. 589 The supernatant is the soluble chromatin extract. The soluble fragmented chromatin from 590 ~2.5x10^7 was used for each IP. For each IP 100ul Dynabeads (Thermofisher anti-rabbit M280, 11203D) were mixed with 10ul polyclonal rabbit-APOBEC2 antibodies (gift from Alin 591 592 Vonica MD, PhD) incubating overnight (~16 hours). A magnetic stand was used to separate 593 beads from the lysate and beads were washed one time each with for 5min in: low salt wash 594 (0.1%SDS, 1%Triton X-100, 2 mM EDTA, 20 mM Tris pH8, 150 mM NaCl, PIC), high salt 595 wash (0.1%SDS, 1% Triton X- 100, 2mM EDTA, 20mM Tris pH8, 500mM NaCl, PIC), lithium 596 chloride wash (150mM LiCl, 1% NP-40, 1% NaDOC, 1mM EDTA, 10mM TrispH8, PIC), TE 597 wash (10mM Tris-HCl pH8, 1mM EDTA, 50mM NaCl, PIC). Beads were resuspended in 52 598 ul of elution buffer (50mM Tris-HCl pH8, 10mM EDTA, 1%SDS) and incubated at 30min at 599 65°C while shaking to prevent beads from settling. The eluate was transferred to a new tube, inputs of the same volume were incubated for 8 hours at 65°C to reverse the crosslink. The 600 samples were treated with RNAse (Roche, 11119915001) for 1 hour at 37°C, and with 601 602 Proteinase K for 2 hours at 55°C. Fragmented DNA was purified with Ampure beads 603 (Agencourt AMPure XP beads A63881) as per the manufacturer's instructions. 604

605 Chromatin immunoprecipitation sequencing and analysis

606 607 The ChIP-Seq included biological triplicates for each group. ChIP-Seq libraries were 608 prepared using NEBNext Ultra DNA Library Prep Kit as per manufacturer's instructions. Libraries were sequenced with 75 base pair single read sequencing on the NextSeg 500 609 610 (Illumina). Read alignments and initial analysis were performed with the Bioinformatics 611 Resource Center at Rockefeller University. Single-end reads were aligned to mm10 genome using the subread function in the Bioconductor Rsubread ⁴⁴ package and bigWig files for 612 613 visualization were generated from aligned reads using the Bioconductor rtracklayer ⁴⁵ and 614 GenomicAlignments packages ⁴⁶. Quality metrics for the ChIP-Seq data were assessed using ChIPQC bioconductor package ⁵⁵, according to Encyclopedia of DNA Elements 615 (ENCODE) working standards and guidelines for ChIP experiments ⁵⁶. Reads mapping to 616 617 more than one genomic location were filtered prior to peak calling using Model-based 618 Analysis of ChIP-Seq (MACS2) ^{57,58} with duplicate filtering applied and input DNA sample as 619 a control. Peaks that are reproducible (present in 2 out of 3) were filtered for known artifact 620 or blacklisted regions (The ENCODE Project Consortium, 2012). For each of the peaks a 621 weighted mean location of peak summits cross biological replicates is calculated ⁵⁹. The list 622 of binding regions 100 base pairs around the mean peak summits was used for downstream 623 analysis. Ngs.plot (v2.61) was used with specific parameters (-G mm10 -D refseq -C -L 1000 624 -FL 150 -P 4 -SC 0,1 -GO none -RB 0.05) to generate average profiles of ChIP-Seq reads 625 (Shen et al., 2014). ChIPSeeker (v1.14.2) ⁶⁰ and ChIPpeakAnno (3.12.7) ^{61,62} were used for 626 downstream analysis after peak calling for annotation of the binding regions to the nearest 627 gene. We created an APOBEC2 occupied gene set, using genes that show consistent 628 APOBEC2 occupancy at both 14-hour and 34-hour time points. GSEA (v3.0) ⁶³ was used for 629 testing the enrichment of the APOBEC2 occupied gene set in the list of genes that are 630 differentially expressed. Annotation files used: BSgenome.Mmusculus.UCSC.mm10 (v1.4.0) 631 org.Mm.db (v3.5.0) and TxDb.Mmusculus.UCSC.mm10.knownGene.gtf.gz(v3.4.0). 632

633 Gene list analysis

634 635 Gene list analyses either by statistical overrepresentation test or statistical enrichment test were done through PANTHER⁶⁴. Briefly, gene lists were filtered based on expression 636 637 (log2FoldChange, up- or downregulated at specific treatment and time point) and p-adj values (FDR< 10%) and used as input in PANTHER gene list analysis. For statistical 638 639 overrepresentation tests of upregulated genes with A2 vs GFP shRNA, genes were filtered 640 based on log2FoldChange > 0.58 and FDR < 10% at each time point and used as input list 641 with Mus musculus (all genes in database) as reference/background list. Default parameters 642 were followed for the analyses and are indicated in the corresponding output. For

- 644 Prediction of binding motifs
- 643 645

Transcription factor motifs associated to 108 TF modules ²⁹ were mapped against time-point 646 647 specific sequences harboring APOBEC2, 200 base pairs centered on peak summits. For 648 each time point, we defined a background set of negative sequences using scrambled 649 regions of the positive sequences. Using both sequence sets of positives and negatives, we 650 assessed the presence of strong 8-mers associated to each of those 108 families and their 651 ability to classify between APOBEC2 regions and negative sequences, summarizing a Receiver-Operating Characteristic Area Under the Curve (ROC-AUC) for each of those. 652 Assessment of significance in each case was done using a Wilcoxon rank sums test (one 653 654 sided). P-values were corrected through a Benjamini-Hochberg procedure.

655

656 Enrichment of ChIP-seg peaks on APOBEC2 differentially expressed target genes 657

658 Using the ChIP-Atlas as a reference, we downloaded all datasets associated to myoblast or 659 C2C12 cells (N=54). For each dataset, we intersected ChIP-seq peaks from APOBEC2 in 660 each timepoint and replicate (command *fisher* in *bedtools*), obtaining a 2 x 2 contingency 661 matrix. The number of overlaps was linked to the closest gene using 2000bp with respect to 662 TSS annotations in the mouse genome. The proportion of genes associated to a differential 663 expression comparison was done by dividing the number of APOBEC2 peaks proximal to a 664 DE-gene with a peak from a ChIP-Atlas dataset by the total number of DE-genes in that 665 comparison. This was repeated for each gene expression contrast. Mean log2 fold change 666 estimates for each ChIP-Atlas peak dataset were obtained by calculating the distribution of 667 log2 fold changes between non-target DE-genes and target DE-genes, in each time point, 668 using the three APOBEC2 ChIP-seq replicates. With those, we calculated a global mean and standard deviation across all ChIP-Atlas factors, reporting a Z-score for dataset, time point 669 670 and differentially expression comparison between APOBEC2 target and non-target genes.

671

672 BioID samples preparation 673

Constructs encoding mouse APOBEC2 fused to BirA-Flag by the N- and C-terminus were 674 675 cloned into pMX-puro. Both constructs, as well as the eGFP-BirA-Flag control, were modified 676 to encode a weak Kozak sequence (TATTGTATG) to reduce protein expression levels. Virions containing the pMX vectors were produced in the Plat-E packaging cell line 65. 677 678 C2C12 cells were spininfected with Plat-E supernatant, 8mg/ml polybrene (16,000 rpm, 90 679 min, 30C) and selected with 4 µg/ml puromycin to obtain stable cell populations. Similar bait 680 and control expression levels were confirmed by western blot, and localization of the 681 APOBEC2 constructs analyzed by immunofluorescence. C2C12 cells expressing each 682 construct were pre-cultured for 24 h with 2% low horse serum before supplementing the media to 50 µM biotin (BioBasic). Cells were harvested 24 h later, washed 3× with PBS, then 683 684 lysed in 1.5 mL of RIPA buffer and sonicated 30 s at 30% amplitude (3 × 10 s bursts with a 685 2 s break in between). Benzonase (250 U, Sigma) was added to the lysates during 686 centrifugation, 30 min at 16,000 × g, 4 °C. Forty μ L aliquots of supernatant were kept to 687 monitor expression and biotinylation and run on western blot, and the remaining lysate was 688 incubated with 70 µL of pre-washed streptavidin-sepharose beads (Sigma) for 3 h on a 689 rotator at 4 °C. Beads were then washed with 1 mL of RIPA buffer, transferred to a new tube 690 and washed again 2× with 1 mL of RIPA buffer and then 3× with 1 mL of 50 mM Ammonium 691 Bicarbonate (ABC) (Biobasic). Beads were then resuspended in 100 µL of ABC with 1 µg of 692 trypsin (Sigma) and incubated overnight at 37 °C with rotation. The following day, 1 µg of 693 trypsin was added for a further 2 h digestion. Samples were centrifuged 1 min at 2000 RPM, 694 and the supernatant was transferred to a new tube. Beads were rinsed twice with 100 µL of 695 water, and all supernatants were pooled and adjusted to 5% formic acid. Samples were then 696 centrifuged for 10 min at $16,000 \times g$ for clarification. Trypsin-digested peptides in the 697 supernatant were dried in a SpeedVac (Eppendorf) for 3 h at 30 °C. Samples were 698 resuspended in 15 µL of 5% formic acid and kept at -80 °C for mass spectrometric analysis.

699 700

701 BioID MS Data analysis 702

703 Mass spectrometry was performed at the IRCM proteomics platform. Samples were injected 704 into Q Exactive Quadrupole Orbitrap (Thermo Fisher), and raw files were analyzed with the search engines Mascot and XTandem! ⁶⁶ through the iProphet pipeline ⁶⁷ integrated in 705 706 Prohits ⁶⁸, using the mouse RefSeq database (version 73) supplemented with "common 707 contaminants" from the Max Planck Institute (http://maxguant.org/downloads.htm), the 708 Global Proteome Machine (GPM; http://www.thegpm.org/crap/index.html) and decoy 709 sequences. The search parameters were set with trypsin specificity (two missed cleavage 710 sites allowed), variable modifications involved Oxidation (M) and Deamidation (NQ). The 711 mass tolerances for precursor and fragment ions were set to 15 ppm and 0.6 Da, 712 respectively, and peptide charges of +2, +3, +4 were considered. Each search result was individually processed by PeptideProphet ⁶⁹, and peptides were assembled into proteins 713 714 using parsimony rules first described in ProteinProphet ⁷⁰ using the Trans-Proteomic

Pipeline (TPP). TPP settings were the following: -p 0.05 -x20 -PPM –d "DECOY", iprophet
 options: pPRIME and PeptideProphet: pP.

- 717
- 718 BioID interactions scoring 719

720 Six biological replicates of each bait and paired eGFP controls were done in two 721 independent experiments (3 replicates each) and combined for the analysis for maximal 722 statistical power. The estimation of interactions scorings was performed for proteins with 723 iProphet protein probability ≥ 0.9 and unique peptides ≥ 2 , by combining two algorithmic 724 approaches : SAINTexpress ⁷¹ and DESeq ⁷². The SAINTexpress (version 3.6.1) analysis 725 was performed with default settings using no compression for controls or baits. Interactions 726 displaying a BFDR \leq 0.01 were considered statistically significant. We also used DESeq2 727 (version 1.2.1335), an R package that applies negative binomial distribution to calculate 728 enrichments over controls. DESeq was run using default settings and significant preys were 729 selected by applying a ≤0.1 p-value cut-off. The combined list of significant preys obtained 730 from DEseq and SAINT was defined as potential APOBEC2 proximity interactors.

731

732 BioID annotations and network analyses

733

734 Graphical representations of protein networks were generated with Cytoscape ⁷³ (version 735 3.8.2). Prior to the importation of APOBEC2's network in Cytoscape, mouse to human 736 orthologs were extracted from the Ensembl database with the BioMart export tool (Mouse 737 genes version GRCm38.p6). Next, we extracted human prey-prey interactions from BioGrid 738 release 4.2.192 ⁷⁴and from Cytoscape's PSICQUIC built-in web service client (April 2021 release) by searching against the IntAct database ⁷⁵. Once augmented, clusters were 739 740 extracted with the Markov CLustering Algorithm (MCL) from Cytoscape's ClusterMaker2 741 application (version 1.3.1) ⁷⁶. To identify relevant complexes among clusters, the APOBEC2 742 interactome was annotated with the Gene Ontology Annotation database ⁷⁷ (GOA version 171) and CORUM (version 3.0) 78, a database of known protein complexes. 743 744 EnrichmentMaps ⁷⁹ of GO Biological Processes were generated by importing g:Profiler's ⁸⁰ 745 generic enrichment file outputs and mouse GO BP GMT file. p-values ≤ 0.05 and q-value \leq 746 0.05 were set as node cutoffs, and Edge cutoff (similarity) were set at 0.345.

747 DNA editing detection

748

749 We aligned all short reads from input and IP experiments to the mouse genome (GRCm38 750 EnsEMBL 90) using HiSAT v2.1.0 with default settings. We removed all non-unique mappers 751 and marked all read duplicates with picard.sam.markduplicates.MarkDuplicates (v 2.5.0). We 752 compared all samples to the reference genome using JACUSA v1.2.4 in call-1 mode with 753 program parameters: call-1 -s -c 5 -P UNSTRANDED -p 10 -W 1000000 -F 1024 --754 filterNM 15-T1-a D,M,Y-R. Diverging positions are reported if the LLR ratio exceeds 1.0. 755 Briefly, read count distributions at every genomic position (coverage >5) are contrasted with the expected read count based on the reference base. For the pairwise comparison of all 756 757 input samples stratified by conditions, we used JACUSA v.1.2.4 in call-2 mode with program 758 parameters: call-2 -s -c 5 -P UNSTRANDED, UNSTRANDED -p 10 -W 1000000 -F 1024 -filterNM 1 5 --filterNM 2 5 -T 1 -a D,M,Y -u DirMult:showAlpha -R. Briefly, read count 759 760 comparison from replicate input samples are contrasted with one another: A2 shRNA 761 knockdown vs GFP shRNA knockdown. Diverging positions are reported if the LLR ratio 762 exceeds 1.0.

763

764 Recombinant mouse APOBEC2 production

Recombinant His₆-tagged mouse APOBEC2 proteins were produced in Sf21 insect cells by
 the EMBL Protein Expression and Purification Core Facility. The genes encoding mouse
 APOBEC2 were cloned into the pFastBac HTa vector (Thermo) and these constructs were

used for transposition into E. coli DH10EMBacY cells (Geneva Biotech). The isolated bacmid 768 769 DNA was utilised for generation of the recombinant baculoviruses. For the mouse APOBEC2 770 protein production, 5 mL of baculovirus was used to infect 1 L of Sf21 cells at a density of 1 x 771 106 cells/ml. After 72h, the cells were harvested by centrifugation (30 min, 600 x g, 4° C) and 772 resuspended in lysis buffer (20 mM Tris pH 8.0, 800 mM NaCl, 20 mM imidazole and 5 mM 773 β-mercaptoethanol) supplemented with benzonase, 10 mM MgCl2 and cOmplete EDTA-free 774 protease inhibitors (Roche). The cells were lysed using a Dounce homogenizer and the 775 lysate was cleared by centrifugation (30 min, 20000 x g, 4°C). The cleared lysate was 776 loaded onto a 5 mL Ni-NTA column (Macherey-Nagel). After washing with a buffer consisting 777 of 20 mM Tris pH 8.0, 300 mM NaCl, 20 mM imidazole and 5 mM β-mercaptoethanol, the Ni-778 NTA column was eluted using a gradient up to 300 mM imidazole. The elution fractions 779 containing mouse APOBEC2 were pooled and dialysed overnight at 4°C to ion exchange buffer (20 mM Tris pH 8.0, 100 mM NaCl, 1 mM DTT). The dialysed sample was loaded onto 780 781 a 5 mL HiTrap Heparin HP column (Cytiva) coupled to a 5 mL HiTrap Q HP anion exchange 782 column (Cytiva). After washing, the HiTrap Heparin HP column and the HiTrap Q HP column 783 were eluted separately in a gradient ranging from 100 mM to 1M NaCl. Finally, the mouse APOBEC2 protein eluted from the HiTrap Q HP column was subjected to a size exclusion 784 785 chromatography (SEC) step using a HiLoad 16/600 Superdex 75 pg column (Cytiva) pre-786 equilibrated with SEC buffer (20 mM HEPES pH 7.5, 150 mM NaCl and 0.5 mM TCEP). 787 When removal of the His6-tag was required, His6-tagged TEV protease (produced in-house) was added to the purified mouse APOBEC2 protein. After the overnight TEV cleavage step 788 789 at 4°C, Ni-NTA beads (Qiagen) were added to the sample and incubated for 1h at 4°C. After 790 centrifugation (1 min, 100 x g, 4°C), untagged mouse APOBEC2 was recovered from the 791 flow through of the Ni-NTA beads. Recombinant mouse APOBEC2 proteins were aliquoted, 792 flash-frozen with liquid N2 and stored at -80°C. Microscale Thermophoresis (MST)

793

794 Purified APOBEC2 was labeled using Cy5 Mono NHS Ester (GEPA15101, Sigma-Aldrich) at 795 5 mg/mL protein and a 3:1 dye:protein ratio. Labeling reactions were performed in 20 mM 796 HEPES, pH 7.5, 150 mM NaCl, and 0.5 mM Tris(2-carboxyethyl)phosphine (TCEP). Reactions were incubated overnight at 4°C with constant agitation. After incubation, 797 798 reactions were deactivated using quencher buffer (ab102884, Abcam). Remaining dye was washed away by concentrating protein using Vivaspin® 500 centrifugal concentrators 799 800 (Sartorius) into MST buffer (10 mM HEPES pH 7.5, 50 mM NaCl, 5 mM MgCl₂). The degree 801 of labeling was typically at about 1.

802

803 Oligonucleotides (oligos) were ordered from Sigma-Aldrich: SP/KLF motif F: GGC GGC
804 GCG GCC CCG CCC CCT CCT CCG GC; SP/KLF motif R: GCC GGA GGA GGG GGC
805 GGG GCC GCG CCG CC; A-tract motif F: TCT CAA GAA AAA AAA AAA AAG AC; A-tract
806 motif R: GTC TTT TTT TTT TTT TCT TGA GA.

807

808 Oligonucleotides were annealed by incubating at 95°C and slow cooling to 25°C before storing at 4°C. Oligonucleotides were diluted to MST buffer + 0.05% Tween-20 for the final 809 810 reaction. MST buffer was supplemented with 0.05% Tween-20. Cy5-labeled APOBEC2 was 811 held constant at 50 nM while the oligonucleotides were titrated (1:1 dilution) between 19.53 812 to 20,000 nM. Reactions were incubated for 30 min before loading into standard glass 813 capillaries (MO-KO22, NanoTemper Technologies). MST measurements were performed 814 using a Monolith NT.115 (NanoTemper Technologies) at 85% LED power and 40% MST 815 power. Data represent 3 independent measurements. MO.Affinity Analysis v2.1.3454 (NanoTemper Technonolgies) was used for curve fitting and calculating Kd values with 816 817 Thermophoresis + T Jump settings.

818

820

819 Electromobility shift assay (EMSA)

821 Oligos used for MST were labeled with $[\gamma^{-32}P]$ ATP and annealed with complementary oligos 822 to form double-stranded (ds) oligos. Specified recombinant protein, mouse APOBEC2 and/or human SP1 (Sigma, SRP2030), and ss or ds oligos were mixed in binding reaction buffer
(10 mM HEPES pH 7.5, 50 mM NaCl, 5 mM MgCl₂, 5% glycerol and 5 mM DTT) for 30
minutes at room temperature. Reactions were resolved on 5% TBE gel (3450048, BioRad)
with 0.5X TBE buffer (1610733, Bio-Rad) for 1.5 hours at 120 volts with the tank submerged
on ice. Gels were then dried and imaged with a phosphorimager system (Azure Biosystems,
Inc.).

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829

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841

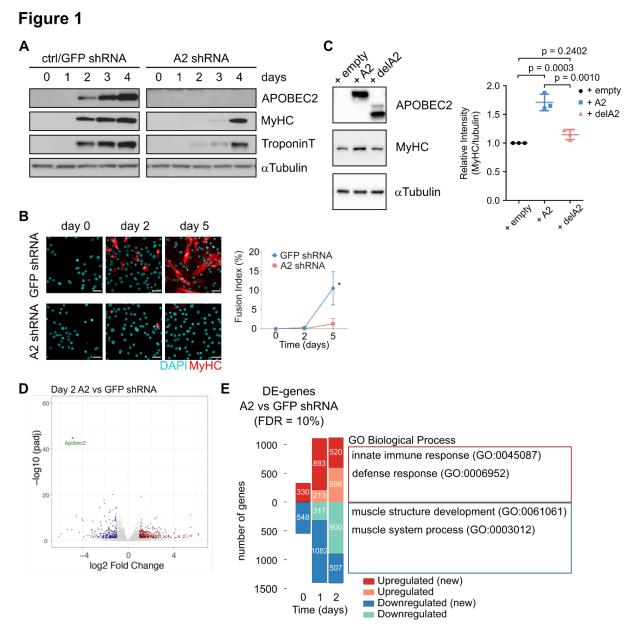
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848

849 Author Contribution

LM and JPL designed the study, performed experiments, analyzed data, and wrote the manuscript with supervision from FNP. SR for performing supplementary experiments. DH performed RNA editing analysis. ILR performed motif prediction and enrichment of ChIP-seq peaks analysis with supervision from JBZ. PGS and JR performed experimental work and JB performed analysis related to BioID data with supervision from JMDN. AV provided key reagent. CD performed DNA editing detection analysis. All authors wrote, read, and approved the final manuscript.



858

859 Figure 1. APOBEC2 expression is required for myoblast differentiation

860(A) Cell lysates from C2C12 cell lines, ctrl/GFP shRNA and A2 shRNA, at different861timepoints of myoblast differentiation (day 0 to day 4) were analyzed by Western blot.862C2C12 myoblasts were transduced either with shRNA against GFP (ctrl/GFP shRNA) or863shRNA against APOBEC2 (A2 shRNA). MyHC (myosin heavy chain) and TroponinT were864used as markers of late differentiation; αTubulin, as loading control.

866 **(B)** C2C12 cell lines were fixed and immunostained using an antibody specific to MyHC 867 (red), DAPI (cyan) was used to stain for DNA. Scale bar = 50 μ m. Line plot shows 868 quantification of fusion index. Statistics: t test. Six fields of view were measured, and data is 869 shown as means (n = 3). Error bars indicate SD; * = p < 0.05.

870
871 (C) C2C12 knockdown cell line (A2 shRNA) was transduced with retrovirus overexpressing
872 APOBEC2 or empty vector control (+ empty: transduced with empty vector, + A2: with
873 APOBEC2 vector, and + delA2 with truncated APOBEC2). Cells were collected 96 hours
874 post-transduction. Cell lysates were prepared and analyzed by Western blot. Representative
875 blot from 3 independent biological replicates. Statistics: Ratio of MyHC mean intensity to
876 aTubulin was normalized to corresponding + empty sample for each trial. Dot plot represent

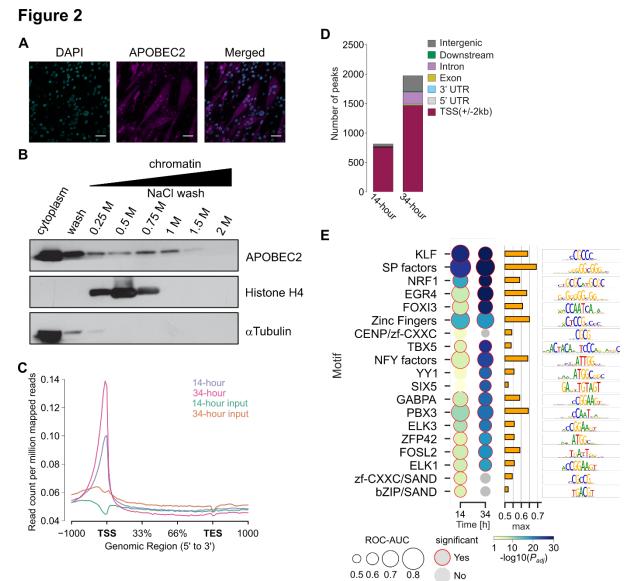
877 mean with error bars representing standard deviation (n = 3). One-way ANOVA with Tukey's 878 multiple comparisons test was performed to calculate p (adjusted P value).

879

(D) Volcano plot was generated using log2 fold changes and p-adjusted values from
 comparing gene expression differences due to APOBEC2 knockdown at day 2 following
 induction of differentiation. Significantly differentially expressed genes with p adjusted value
 <0.05 are shown in red (upregulated) or blue (downregulated) and not significantly
 differentially expressed genes in gray. APOBEC2 is shown in green.

885

(E) Number of differentially expressed genes (DE-genes) between APOBEC2 knockdown
(A2 shRNA) relative to GFP shRNA control. Colors indicate up- (red) and down-regulated
(blue) genes. A false discovery rate (FDR) cutoff of 10% was used to determine DE-genes.
Dark red and blue indicate newly differentially expressed genes at a given time point.
Common GO Biological Process terms across day 0 to 2 from statistical enrichment test
ranking genes by log2FoldChange (see Supplementary File 1 for complete output tables of
the tests).



893 894 Figure 2. APOBEC2 localization and ChIP-Seg in differentiating C2C12 myoblasts.

895 (A) Immunostaining of APOBEC2 (magenta) and DAPI-positive (blue) nuclei in differentiated 896 C2C12 myoblasts (5 days in differentiation medium). Scale bar = $50 \mu m$.

897

898 (B) The sequential salt extraction profile of endogenous APOBEC2 and histone H4 are 899 shown. alpha-tubulin is a cytoplasmic marker. The amount of indicated proteins in eluates 900 was measured by Western blotting. 901

- 902 (C) The mean normalized APOBEC2 signal (plotted as read counts per million mapped 903 reads) across all annotated genes. This plot shows the global differences in APOBEC2 904 binding between the two time points in TSS. Both time points are in biological triplicates.
- 905

906 (D) Genomic annotations of APOBEC2 consensus binding regions in each of the time points 907 (14- and 34-hour). Binding regions are annotated based on genomic feature. The priority of 908 assignment to a genomic feature when there is annotation overlap is: Promoter (2kb around 909 the TSS), 5' UTR, 3' UTR, Exon, Intron, Downstream (within 3kb downstream of the end of 910 the gene). 911

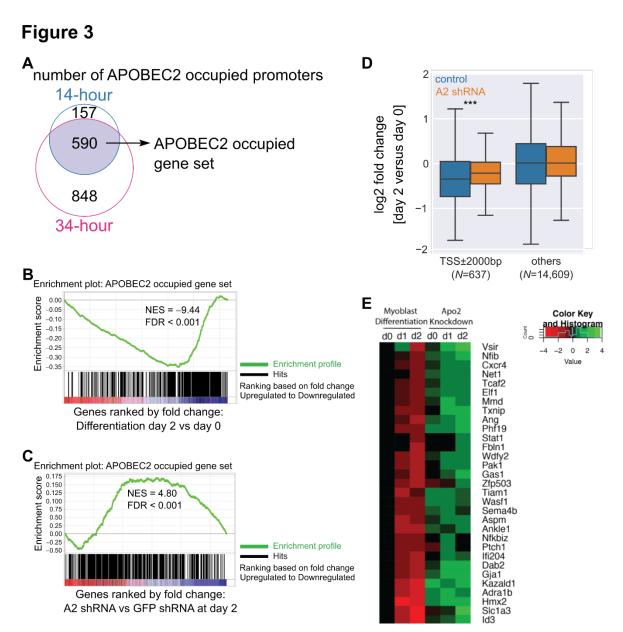
(E) Enrichment of 8-mers associated to 19 known transcription factors specificities (motifs) in 912 913 APOBEC2 ChIP-seq regions, compared against negative genomic regions (see Methods). 914

915 descendently sorted on the y-axis by minimum adjusted P value observed in both time points 916 (color). All 18 shown transcription factors are significant on at least one time point (adjusted

917 P value < 0.1, using one-sided Wilcoxon rank sum test and Benjamini-Hochberg correction.

918 Red circles). Right barplot indicates maximum ROC-AUC value, row-wise. Right sequence

919 logos depict unsupervised alignment of all 8-mers referred to that transcription factor motif ²⁹.



920 921

Figure 3. APOBEC2 binding at specific transcription factor DNA motifs.

922 923

(A) The Venn diagram represents the number of genes that have APOBEC2 occupancy on 924 925 their promoters at 14- and 34-hour time points. The genes that show consistent APOBEC2 occupancy in their promoters at both time points were used to create an APOBEC2 occupied 926 927 gene set as shown in the Venn diagram.

- 928
- (B,C) Gene set enrichment analysis (GSEA) (Subramanian et al., 2005) was used to test the 929 enrichment of the APOBEC2 occupied gene set in the list of genes that are differentially 930 931 expressed through differentiation (C) or the ones that are differentially expressed due to APOBEC2 knockdown at day 2 (D). The enrichment profile over the whole ranked gene set 932 933 is shown in green with normalized enrichment score (NES) and FDR values. Gene hits are 934 shown as black lines. A positive NES indicates gene set enrichment at the top (positive/up 935 fold change) of the ranked list; a negative NES indicates gene set enrichment at the bottom 936 (negative/down fold change) of the ranked list.
- 937

(D) Expression changes for genes in control (GFP control shRNA) and A2 knockdown (A2 938 939 shRNA) C2C12 during differentiation. Genes are grouped by the presence of an APOBEC2 940 ChIP-seq peak nearby Transcription Start Sites (TSS) at not more than 2000bp from it, or 941 genome wide background (others). Asterisks indicate adjusted P values, by two-sided t-test 942 corrected using Benjamini Hochberg procedure (***=0.001).

943

944 **(E)** Heatmap showing expression changes of genes related to differentiation (regulation of

945 differentiation, GO: 0045595) occupied by A2 during differentiation in C2C12 cells and A2

846 knockdown cells. The list of APOBEC2 repressed genes were filtered from the C2C12

947 RNASeq data of A2 occupied genes (defined as genes with A2-ChiPSeq peak in the

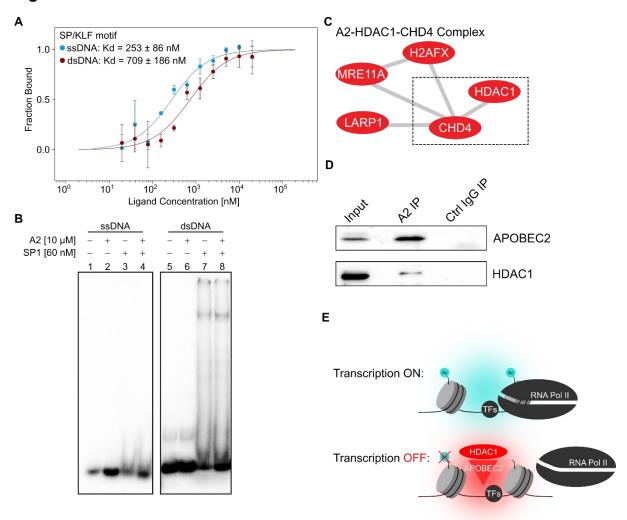
promoter region). Genes repressed during differentiation log2 fold change greater than 0.58
 (absolute fold change > 1.5) at day 1 or 2 of differentiation and upregulated in the

950 knockdown (versus control) at day 0, 1, or 2 were selected. These genes were then used as

input for statistical overrepresentation test – GO biological processes through Panther (ver

14) with the default settings using all *M. musculus* genes in database as reference list ⁸¹.

Figure 4



953 954

955 Figure 4. APOBEC2 recruits the HDAC1 co-transcriptional repressor complex.

(A) Microscale thermophoresis (MST) experiments measuring purified APOBEC2 binding to single-stranded DNA (ssDNA) or double-stranded (dsDNA) SP/KLF motif. Cy5-labeled APOBEC2 was kept constant (50 nM) while the concentration of non-labeled SP/KLF motif was titrated (1:1 dilution) between 20 nM – 20,000 nM. The calculated Kd was computed using the standard settings (thermophoresis + T jump) with the MO.Affinity Analysis v2.1 (NanoTemper Technonolgies). Values represent 3 independent measurements with error bars representing the standard deviation.

963

(B) Electrophoretic mobility shift assays (EMSA) of recombinant mouse APOBEC2 and
 human SP1 protein on either ss or dsDNA with an SP/KLF motif. Radioactively labeled ss or
 dsDNA (1 nM) was mixed with either recombinant APOBEC2 (10 uM) or SP1 (60 nM)
 protein, or both. Gel shift image is representative of at least 3 independent experiments.

968

969 (C) Selected protein complex identified by APOBEC2 proximity-dependent protein
 970 biotinylation (BioID). Each red node corresponds to a protein that was identified by BioID
 971 mass spectrometry to interact with APOBEC2. CHD4 was also identified in a BioID data
 972 comparing APOBEC2 and AID in mouse B cells ⁸². The edges denote the known interactions
 973 of these proteins with each other (see Figure S5B for other complexes).

974 **(D)** Co-immunoprecipitation (Co-IP) of APOBEC2 with HDAC1 in C2C12 myoblasts 975 differentiated to myotubes for 4 days. Nuclear protein lysates (Input) were incubated with

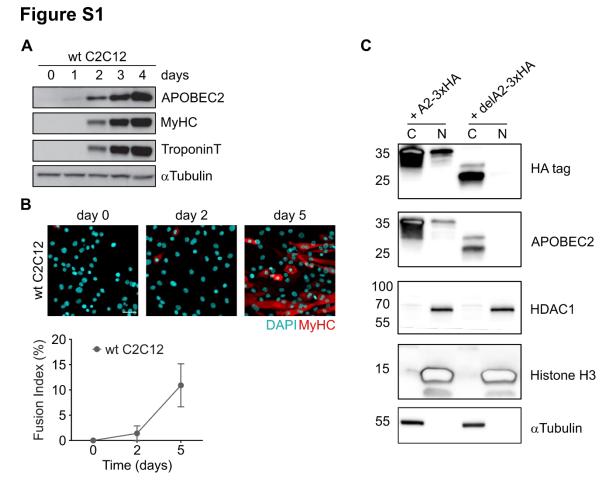
beads conjugated to either APOBEC2 antibody (A2 IP) or IgG isotype control antibody (Ctrl
IgG IP). Proteins were then eluted, ran on an SDS-PAGE gel, and blotted with APOBEC2, or
HDAC1 antibodies.

979

980 **(E)** Proposed molecular function of APOBEC2 as a co-transcriptional repressor complex that

981 acts on active/open chromatin to repress transcription through HDAC1 histone deacetylation

982 during myoblast differentiation.



983 984

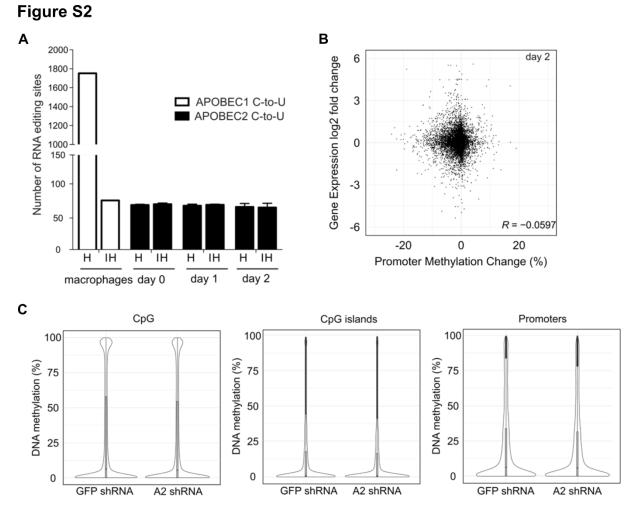
985 Supplementary Figure 1. C2C12 myoblast differentiation

986 (A) Whole cell extracts of mouse wildtype (wt) C2C12 myoblasts and myotubes were
 987 analyzed by Western blotting using anti-APOBEC2 antibodies. MyHC and TroponinT were
 988 used as markers of late differentiation, alpha-tubulin was used as loading control.

989

(B) C2C12 cells were cultured in differentiation medium for 0, 2 and 5 days, fixed, and stained with antibody to MyHC (red). Nuclei were visualized by DAPI staining (blue). Below the quantification of differentiation expressed as fusion index, which is the percentage MyHC-positive myotubes with >2 nuclei. Results are presented as means from quantification of at least 6 images/sample. Error bars indicate SD. Scale bar = 50 µm.

996 (C) Cytoplasmic and nuclear (C and N respectively) were taken from C2C12 cells
 997 transduced with vectors to overexpress 3xHA-tagged wildtype APOBEC2 (+ A2-3xHA) and
 998 truncated APOBEC2 (+ delA2-3xHA). Western blots were performed with respective lysates
 999 using antibodies against HA-tag, APOBEC2, HDAC1, Histone H3, and αTubulin.



1000 1001

1002 Supplementary Figure 2. mRNA editing and DNA demethylation in C2C12 myoblasts

1003 (A) Candidate C-to-U RNA editing sites called from APOBEC2 knockdown samples, control 1004 (GFPsh) at day 0, 1, and 2 in DM in C2C12s and wild-type and APOBEC1-/- macrophages 1005 (positive control). Hits (H) represent candidate editing sites present in control (GFPsh) but 1006 not in APOBEC2 knockdown dataset (in positive control hits are the # of sites in APOBEC1-1007 /- dataset.) Inverse hits (IH) represent putative edited sites yielded when the inverse 1008 comparison is made, thus edit sites present in the APOBEC2 knockdown dataset but not in 1009 the control (GFPsh) (for the positive control edit sites that are present in APOBEC1-/-1010 dataset but not in wild- type). Data are represented as means ± SD using outputs of 3 RNA-1011 Seq datasets.

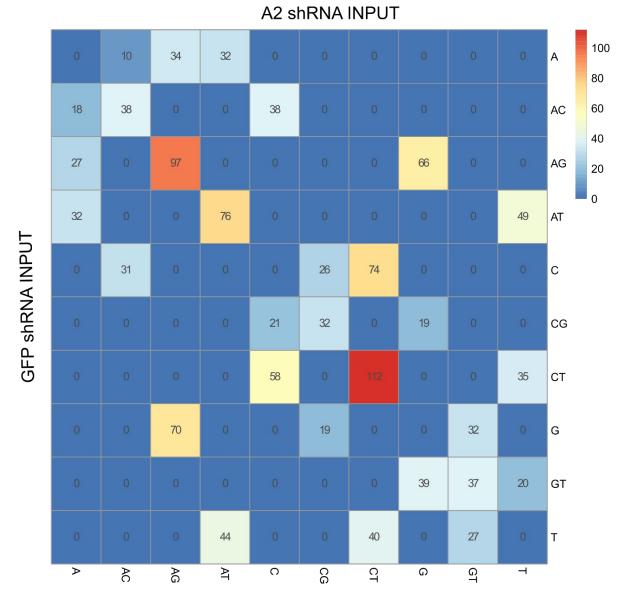
1012

(B) Methylation changes across all the represented promoters in the ERRBS dataset
 compared with the expression changes of the same genes in RNA-Seq dataset. Shown here
 are datasets from the day 2 timepoint. R = Pearson's correlation coefficient.

1016

(C) Distribution of DNA methylation frequencies in C2C12s as determined by ERRBS for
 individual CpGs, CpG islands and promoters. Promoters are defined at -/+ 2Kb around the
 TSS in Ensemble annotations. CpG islands were taken from the cpgIslandExt track of the
 UCSC table browser. Violin plots represent the distribution of DNA methylation frequencies
 for each feature. Median and first and third quartiles are shown with the box plots.

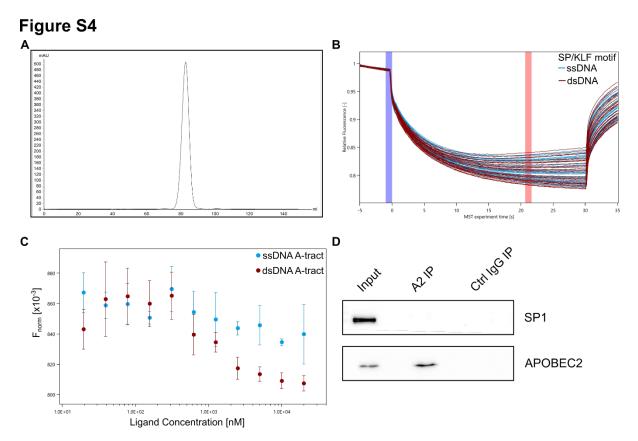
Figure S3



1022 1023

Supplementary Figure 3. DNA Editing in APOBEC2 knockdown versus control (GFP
 shRNA)
 1026

Pairwise heatmap - This is a head-to-head comparison of variant positions between the 2
input sample sets. The symmetry of the heatmap indicates that there is no preference for a
unidirectional base substitution process.



1030 1031

1035

1032 Supplementary Figure 4. Recombinant APOBEC2 electromobility shift assays

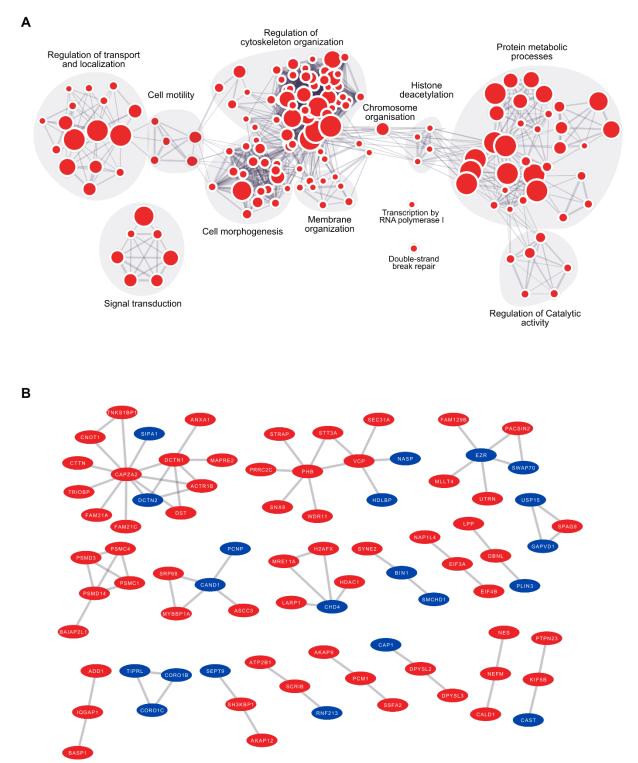
1033 **(A)** Size exclusion chromatogram (Superdex 200, GE Healthcare) of recombinant 6x-His-1034 APOBEC2 produced in High Five/Sf9 (Thermo) insect cells.

(B) MST trace for APOBEC2 interaction with ssDNA and dsDNA SP/KLF motifs. Traces
 correspond to experiment in Figure 4A. Blue and red highlighted regions represent cold and
 hot regions respectively that were used for the standard thermophoresis+ T-jump analysis.

(C) MST experiments measuring purified APOBEC2 binding to single-stranded DNA (ssDNA) or double-stranded (dsDNA) with A-tract motif. Cy5-labeled APOBEC2 was kept constant (50 nM) while the concentration of non-labeled SP/KLF motif was titrated (1:1 dilution) between 20 nM – 20,000 nM. Normalized fluorescence (Fnorm) values and graph were produced with MO.Affinity Analysis v2.1 (NanoTemper Technonolgies). Values represent 3 independent measurements with error bars representing the standard deviation.

1047 **(D)** Co-immunoprecipitation (Co-IP) of APOBEC2 with SP1 in C2C12 myoblasts 1048 differentiated to myotubes for 4 days. Nuclear protein lysates (Input) were incubated with 1049 beads conjugated to either APOBEC2 antibody (A2 IP) or IgG isotype control antibody (Ctrl 1050 IgG IP). Proteins were then eluted, ran on an SDS-PAGE gel, and blotted with APOBEC2, or 1051 SP1 antibodies.

Figure S5



1052 1053

1054 Supplementary Figure 5. BiolD GO Enrichment Map

(A) Gene ontology enrichment map based on APOBEC2 BioID hits annotated by gProfiler
 (BFDR<0.05). Gene ontology terms related to similar biological processes are clustered
 (indicated in grey).

1058

(B) Protein complexes identified by proximity-dependent protein biotinylation (BioID) in
 C2C12 expressing APOBEC2-BirA or BirA-APOBEC2. Each red node corresponds to a
 protein that was significantly enriched in either or both A2-expressing cells compared to

1062 GFP-BirA control cells. Nodes in blue indicate proteins that were preferentially labeled by 1063 APOBEC2-BirA compared to AID-BirA in mouse B lymphocytes in a previously published 1064 dataset ⁸². The edges denote the known interactions of these proteins with each as 1065 described in Methods.

1066	References		
1067 1068	1.	Conticello, S. G. The AID/APOBEC family of nucleic acid mutators. Genome Biol. 9,	
1069	0	229 (2008).	
1070 1071	2.	Salter, J. D., Bennett, R. P. & Smith, H. C. The APOBEC Protein Family: United by Structure, Divergent in Function. <i>Trends Biochem. Sci.</i> 41 , 578–594 (2016).	
1071	3.	Cole, D. C. <i>et al.</i> Loss of APOBEC1 RNA-editing function in microglia exacerbates	
1072	5.	age-related CNS pathophysiology. <i>Proc. Natl. Acad. Sci.</i> 114 , 13272–13277 (2017).	
1074	4.	Rayon-Estrada, V. <i>et al.</i> Epitranscriptomic profiling across cell types reveals	
1075	••	associations between APOBEC1-mediated RNA editing, gene expression outcomes,	
1076		and cellular function. <i>Proc. Natl. Acad. Sci.</i> 114 , 13296–13301 (2017).	
1077	5.	Harris, R. S. & Dudley, J. P. APOBECs and virus restriction. Virology vols 479–480	
1078		131–145 (2015).	
1079	6.	Rai, K. et al. DNA Demethylation in Zebrafish Involves the Coupling of a Deaminase,	
1080		a Glycosylase, and Gadd45. <i>Cell</i> 135 , 1201–1212 (2008).	
1081	7.	Guo, J. U., Su, Y., Zhong, C., Ming, G. & Song, H. Hydroxylation of 5-Methylcytosine	
1082		by TET1 Promotes Active DNA Demethylation in the Adult Brain. <i>Cell</i> 145 , 423–434	
1083	•	(2011).	
1084	8.	Krishnan, A., Iyer, L. M., Holland, S. J., Boehm, T. & Aravind, L. Diversification of	
1085		AID/APOBEC-like deaminases in metazoa: multiplicity of clades and widespread roles	
1086 1087	9.	in immunity. <i>Proc. Natl. Acad. Sci. U. S. A.</i> 115 , E3201–E3210 (2018). Liao, W. <i>et al.</i> APOBEC-2, a Cardiac- and Skeletal Muscle-Specific Member of the	
1087	9.	Cytidine Deaminase Supergene Family. <i>Biochem. Biophys. Res. Commun.</i> 260 , 398–	
1089		404 (1999).	
1090	10.	Sato, Y. <i>et al.</i> Deficiency in APOBEC2 Leads to a Shift in Muscle Fiber Type,	
1091		Diminished Body Mass, and Myopathy. J. Biol. Chem. 285, 7111–7118 (2010).	
1092	11.	Etard, C., Roostalu, U. & Strähle, U. Lack of Apobec2-related proteins causes a	
1093		dystrophic muscle phenotype in zebrafish embryos. J. Cell Biol. 189, 527–39 (2010).	
1094	12.	Powell, C., Elsaeidi, F. & Goldman, D. Injury-Dependent Muller Glia and Ganglion Cell	
1095		Reprogramming during Tissue Regeneration Requires Apobec2a and Apobec2b. J.	
1096		<i>Neurosci.</i> 32 , 1096–1109 (2012).	
1097	13.	Vonica, A., Rosa, A., Arduini, B. L. & Brivanlou, A. H. APOBEC2, a selective inhibitor	
1098		of TGF β signaling, regulates left-right axis specification during early embryogenesis.	
1099	4.4	Dev. Biol. 350 , 13–23 (2011).	
1100	14.	Okuyama, S. <i>et al.</i> Excessive activity of apolipoprotein B mRNA editing enzyme	
1101 1102		catalytic polypeptide 2 (APOBEC2) contributes to liver and lung tumorigenesis. <i>Int. J. cancer</i> 130 , 1294–301 (2012).	
1102	15.	Lohr, J. G. <i>et al.</i> Discovery and prioritization of somatic mutations in diffuse large B-	
1104	10.	cell lymphoma (DLBCL) by whole-exome sequencing. <i>Proc. Natl. Acad. Sci. U. S. A.</i>	
1105		109 , 3879–84 (2012).	
1106	16.	Powell, C., Cornblath, E. & Goldman, D. Zinc-binding Domain-dependent,	
1107		Deaminase-independent Actions of Apolipoprotein B mRNA-editing Enzyme, Catalytic	
1108		Polypeptide 2 (Apobec2), Mediate Its Effect on Zebrafish Retina Regeneration. J. Biol.	
1109		Chem. 289, 28924–28941 (2014).	
1110	17.	Yaffe, D. & Saxel, O. Serial passaging and differentiation of myogenic cells isolated	
1111		from dystrophic mouse muscle. <i>Nature</i> 270 , 725–7 (1977).	
1112	18.	Carrió, E. et al. Muscle cell identity requires Pax7-mediated lineage-specific DNA	
1113		demethylation. BMC Biol. 14, 30 (2016).	
1114	19.	Prochnow, C., Bransteitter, R., Klein, M. G., Goodman, M. F. & Chen, X. S. The	
1115		APOBEC-2 crystal structure and functional implications for the deaminase AID.	
1116 1117	20.	Nature 445 , 447–451 (2007). Krzysiak, T. C., Jung, J., Thompson, J., Baker, D. & Gronenborn, A. M. APOBEC2 Is	
1118	20.	a Monomer in Solution: Implications for APOBEC3G Models. <i>Biochemistry</i> 51 , 2008–	
1119		2017 (2012).	
1120	21.	Watanabe, Y. Conversion of myoblasts to physiologically active neuronal phenotype.	

1121		Genes Dev. 18 , 889–900 (2004).
1122	22.	Harjanto, D. <i>et al.</i> RNA editing generates cellular subsets with diverse sequence
1123	00	within populations. <i>Nat. Commun.</i> 7 , 12145 (2016).
1124	23.	Laffleur, B. <i>et al.</i> AID-induced remodeling of immunoglobulin genes and B cell fate.
1125	0.4	Oncotarget 5, 1118–1131 (2014).
1126	24.	Chan, K. & Gordenin, D. A. Clusters of Multiple Mutations: Incidence and Molecular
1127		Mechanisms. Annu. Rev. Genet. 49, 243–267 (2015).
1128	25.	Kosugi, S., Hasebe, M., Tomita, M. & Yanagawa, H. Systematic identification of cell
1129		cycle-dependent yeast nucleocytoplasmic shuttling proteins by prediction of
1130		composite motifs. <i>Proc. Natl. Acad. Sci.</i> 106 , 10171–10176 (2009).
1131	26.	Patenaude, AM. et al. Active nuclear import and cytoplasmic retention of activation-
1132		induced deaminase. <i>Nat. Struct. Mol. Biol.</i> 16 , 517–27 (2009).
1133	27.	Porter, E. G., Connelly, K. E. & Dykhuizen, E. C. Sequential Salt Extractions for the
1134		Analysis of Bulk Chromatin Binding Properties of Chromatin Modifying Complexes. J.
1135		Vis. Exp. 2017 , (2017).
1136	28.	Kohli, R. M. et al. Local Sequence Targeting in the AID/APOBEC Family Differentially
1137		Impacts Retroviral Restriction and Antibody Diversification. J. Biol. Chem. 285,
1138		40956–40964 (2010).
1139	29.	Mariani, L., Weinand, K., Vedenko, A., Barrera, L. A. & Bulyk, M. L. Identification of
1140		Human Lineage-Specific Transcriptional Coregulators Enabled by a Glossary of
1141		Binding Modules and Tunable Genomic Backgrounds. Cell Syst. 5, 187-201.e7
1142		(2017).
1143	30.	Harris, R. S., Petersen-Mahrt, S. K. & Neuberger, M. S. RNA Editing Enzyme
1144	00.	APOBEC1 and Some of Its Homologs Can Act as DNA Mutators. <i>Mol. Cell</i> 10 , 1247–
1145		1253 (2002).
1146	31.	Mikl, M. C. <i>et al.</i> Mice Deficient in APOBEC2 and APOBEC3. <i>Mol. Cell. Biol.</i> 25,
1147	01.	7270–7277 (2005).
1148	32.	Torchy, M. P., Hamiche, A. & Klaholz, B. P. Structure and function insights into the
1149	02.	NuRD chromatin remodeling complex. <i>Cell. Mol. Life Sci.</i> 72 , 2491–507 (2015).
1150	33.	Salter, J. D. & Smith, H. C. Modeling the Embrace of a Mutator: APOBEC Selection of
1151	55.	Nucleic Acid Ligands. Trends Biochem. Sci. 43, 606–622 (2018).
1152	34.	Vartanian, J., Guétard, D., Henry, M. & Wain-Hobson, S. Evidence for editing of
1152	54.	human papillomavirus DNA by APOBEC3 in benign and precancerous lesions.
1155		Science 320 , 230–3 (2008).
	35.	
1155	35.	Law, E. K. <i>et al.</i> APOBEC3A catalyzes mutation and drives carcinogenesis in vivo. <i>J.</i>
1156	26	Exp. Med. 217, (2020).
1157	36.	Nabel, C. S. <i>et al.</i> AID/APOBEC deaminases disfavor modified cytosines implicated in
1158	07	DNA demethylation. <i>Nat. Chem. Biol.</i> 8 , 751–758 (2012).
1159	37.	Powell, C., Grant, A. R., Cornblath, E. & Goldman, D. Analysis of DNA methylation
1160		reveals a partial reprogramming of the Muller glia genome during retina regeneration.
1161	00	Proc. Natl. Acad. Sci. 110, 19814–19819 (2013).
1162	38.	Gómez-Del Arco, P. et al. The Chromatin Remodeling Complex Chd4/NuRD Controls
1163		Striated Muscle Identity and Metabolic Homeostasis. <i>Cell Metab.</i> 23, 881–92 (2016).
1164	39.	Boija, A. et al. Transcription Factors Activate Genes through the Phase-Separation
1165		Capacity of Their Activation Domains. Cell 175 , 1842-1855.e16 (2018).
1166	40.	Lee, Q. Y. et al. Pro-neuronal activity of Myod1 due to promiscuous binding to
1167		neuronal genes. <i>Nat. Cell Biol.</i> 22 , 401–411 (2020).
1168	41.	Mall, M. et al. Myt1I safeguards neuronal identity by actively repressing many non-
1169		neuronal fates. <i>Nature</i> 544 , 245–249 (2017).
1170	42.	Moffat, J. et al. A Lentiviral RNAi Library for Human and Mouse Genes Applied to an
1171		Arrayed Viral High-Content Screen. Cell 124 , 1283–1298 (2006).
1172	43.	Orimo, A. et al. Stromal Fibroblasts Present in Invasive Human Breast Carcinomas
1173		Promote Tumor Growth and Angiogenesis through Elevated SDF-1/CXCL12
1174		Secretion. Cell 121, 335–348 (2005).
1175	44.	Liao, Y., Smyth, G. K. & Shi, W. The Subread aligner: fast, accurate and scalable

 with genome browsers. <i>Bioinformatics</i> 26, 1441–1842 (2009). Lawrence, M. <i>et al.</i> Software for Computing and Annotating Genomic Ranges. <i>PLoS Comput. Biol</i> 9, e1003118 (2013). Patro, R., Duggal, G., Love, M. I., Irizarry, R. A. & Kingsford, C. Salmon provides fast and bias-aware quantification of transcript expression. <i>Nat. Methods</i> 14, 417–419 (2017). Soneson, C., Love, M. I. & Robinson, M. D. Differential analyses for RNA-seq: transcript-level estimates improve gene-level inferences. <i>F1000Research</i> 4, 1521 (2015). Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. <i>Genome Biol</i>. 15, 550 (2014). Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS Genet.</i> 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a fexible aligner and methylation caller for Bisuffite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome-wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on chIP quality metrics in ChIP-seq (MACS). <i>Genome Biol.</i> 9, e201401002 (2014). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal ano chIP quality metrics and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal son. <i>BMC Bioinformatics</i> 34, 214-1-214-14 (2011). Zhang, Y. <i>et </i>	1176 1177	45.	read mapping by seed-and-vote. <i>Nucleic Acids Res.</i> 41 , e108–e108 (2013). Lawrence, M., Gentleman, R. & Carey, V. rtracklayer: an R package for interfacing
 Patrö, R., Duggal, G., Love, M. I., Irizarry, R. A. & Kingsford, C. Salmon provides fast and bias-aware quantification of transcript expression. <i>Nat. Methods</i> 14, 417–419 (2017). Soneson, C., Love, M. I. & Robinson, M. D. Differential analyses for RNA-seq: transcript-level estimates improve gene-level inferences. <i>F1000Research</i> 4, 1521 (2015). Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. <i>Genome Biol</i>. 15, 550 (2014). Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS Genet.</i> 8, e1002781 (2012). Garrett-Bakelman, F. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylkit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Kau, et al., methylkit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modEINCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinformatics</i> 314, 214.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES		46.	
 and bias-aware quantification of transcript expression. <i>Nat. Methods</i> 14, 417–419 (2017). Soneson, C., Love, M. I. & Robinson, M. D. Differential analyses for RNA-seq: transcript-level estimates improve gene-level inferences. <i>F1000Research</i> 4, 1521 (2015). Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. <i>Genome Biol</i>. 15, 550 (2014). Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS Genet.</i> 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Biolific Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome-wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-sex data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LeVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol.</i> 19, e201401002 (2014). Zhang, Y. <i>et al.</i> LeVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol.</i> 19, e201401002 (2014). Yang, Y. <i>et al.</i> LeVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct.</i>		47	
 (2017). (2017). (2017). (2015). (2016). (2017). (2018). (2014). <		47.	
 Soneson, C., Love, M. I. & Robinson, M. D. Differential analyses for RNA-seq: transcript-level estimates improve gene-level inferences. <i>F1000Research</i> 4, 1521 (2015). Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. <i>Genome Biol</i> 15, 550 (2014). Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS</i> Genet. 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma</i> 34, 2141-214.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. <i>in Methods </i>			
 transcript-level estimates improve gene-level inferences. <i>F1000Research</i> 4, 1521 (2015). Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. <i>Genome Biol.</i> 15, 550 (2014). Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS Genet.</i> 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methyliki: a comprehensive R package for the analysis of genome-wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma</i> 34, 2:14-1-2:14.14 (2011). Shang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LeVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, OY. ChIPseeker: an R/Bioconductor package for ChIP seq and <i>ChIP-Seq Dataset</i>, in <i>Methods in Molecular Biology</i> 105–124 (Human		40	
 (2015). (2015). (2016). (2016). (2017). (2017).		48.	
 Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. <i>Genome Biol.</i> 15, 550 (2014). Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS Genet.</i> 8, e1002781 (2012). Garrett-Bakelman, F. e. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylkit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2:14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq Dataset. in Methods in Molecular Biology 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- &.8. Yang, Y. <i>et al.</i> ChIP-seq MICNIS. <i>Comput. Struct. Biotechnol.</i> J. 9, e201401002 (2014). Yu, G., Wang, L6. & He, QY. ChIPseeker: an R/Bicconductor package for ChIP peak annotation, comparison and visualization. <i></i>			
 dispersion for RNA-seq data with DESeq2. <i>Genome Biol.</i> 15, 550 (2014). Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS</i> <i>Genet.</i> 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylkit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinform.</i> 34, 214.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 282–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. <i>in Methods in Molecular Biology</i> 105–124 (Humana Press, 2013).			
 Akalin, A. <i>et al.</i> Base-Pair Resolution DNA Methylation Sequencing Reveals Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS</i> <i>Genet.</i> 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 17, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Bioinformatics</i> 31, 2382–2383 (2015). Yang, Y. <i>et al.</i> ChIPpeakAnno: a Bioinformatics 11, 327 (2010). Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15		49.	
 Profoundly Divergent Epigenetic Landscapes in Acute Myeloid Leukemia. <i>PLoS</i> Genet. 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome-wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 214.1-214.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8 8. Zhu, L. J. et al. ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010).			
 Genet. 8, e1002781 (2012). Garrett-Bakelman, F. E. <i>et al.</i> Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. <i>J. Vis. Exp.</i> (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Garroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach		50.	
 Garrett-Bakelman, F. E. et al. Enhanced Reduced Representation Bisulfite Sequencing for Assessment of DNA Methylation at Base Pair Resolution. J. Vis. Exp. (2015) doi:10.3791/S2246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramaina, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the P			
 Sequencing for Assessment of DNA Methylation at Base Pair Resolution. J. Vis. Exp. (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome-wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2:14.1-2:14, 14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-seq (2005). Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Mi, H. <i>et al.</i> Protocol Update			
 (2015) doi:10.3791/52246. Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfice-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylKit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biocechnol.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8.8. Zhu, L. J. <i>et al.</i> CBioinformatics 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–7		51.	, I
 Krueger, F. & Andrews, S. R. Bismark: a flexible aligner and methylation caller for Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylkit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2:14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978.1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (· · · ·
 Bisulfite-Seq applications. <i>Bioinformatics</i> 27, 1571–1572 (2011). Akalin, A. <i>et al.</i> methylkti: a comprehensive R package for the analysis of genome-wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma</i>. 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8.8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable sy			
 Akalin, A. <i>et al.</i> methylkit: a comprehensive R package for the analysis of genome- wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2:14.1-2:14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an <i>R</i>/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8.8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable		52.	
 wide DNA methylation profiles. <i>Genome Biol.</i> 13, R87 (2012). Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Moirta, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> Prophet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1196		Bisulfite-Seq applications. <i>Bioinformatics</i> 27 , 1571–1572 (2011).
 Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2:14.1-2:14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Thu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Piat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass	1197	53.	
 methylation. <i>BMC Bioinformatics</i> 14, S10 (2013). 55. Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). 56. Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). 57. Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). 58. Zhang, Y. <i>et al.</i> Ndoel-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). 59. Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). 60. Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). 61. Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. 62. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). 63. Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). 64. Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). 75. Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). 64. Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–	1198		wide DNA methylation profiles. <i>Genome Biol.</i> 13 , R87 (2012).
 Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProtein Identification Rates and Error Estimates. <i>Mol.</i> 	1199	54.	Li, S. <i>et al.</i> An optimized algorithm for detecting and annotating regional differential
 removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. <i>Front. Genet.</i> 5, (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> Iprophet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1200		methylation. BMC Bioinformatics 14, S10 (2013).
 (2014). (2014). Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). S8. Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> IProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1201	55.	Carroll, T. S., Liang, Z., Salama, R., Stark, R. & de Santiago, I. Impact of artifact
 Landt, S. G. <i>et al.</i> ChIP-seq guidelines and practices of the ENCODE and modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> IProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1202		removal on ChIP quality metrics in ChIP-seq and ChIP-exo data. Front. Genet. 5,
 modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). St. Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> IProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1203		(2014).
 modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012). Feng, J., Liu, T. & Zhang, Y. Using MACS to Identify Peaks from ChIP-Seq Data. <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). S8. Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> IProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1204	56.	Landt, S. G. et al. ChIP-seq guidelines and practices of the ENCODE and
 <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). <i>Zhang, Y. et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProtein Identification Rates and Error Estimates. <i>Mol.</i> 	1205		modENCODE consortia. <i>Genome Res.</i> 22, 1813–1831 (2012).
 <i>Curr. Protoc. Bioinforma.</i> 34, 2.14.1-2.14.14 (2011). Zhang, Y. <i>et al.</i> Model-based Analysis of ChIP-Seq (MACS). <i>Genome Biol.</i> 9, R137 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> Irrophet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1206	57.	
 (2008). Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in</i> <i>Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1207		
 Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1208	58.	Zhang, Y. et al. Model-based Analysis of ChIP-Seq (MACS). Genome Biol. 9, R137
 Yang, Y. <i>et al.</i> LEVERAGING BIOLOGICAL REPLICATES TO IMPROVE ANALYSIS IN CHIP-SEQ EXPERIMENTS. <i>Comput. Struct. Biotechnol. J.</i> 9, e201401002 (2014). Yu, G., Wang, LG. & He, QY. ChIPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1209		
 Yu, G., Wang, LG. & He, QY. ChlPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in</i> <i>Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. et al. ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. et al. Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. et al. Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. et al. iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1210	59.	
 Yu, G., Wang, LG. & He, QY. ChlPseeker: an R/Bioconductor package for ChIP peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in</i> <i>Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. et al. ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. et al. Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. et al. Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. et al. iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1211		IN CHIP-SEQ EXPERIMENTS. Comput. Struct. Biotechnol. J. 9, e201401002 (2014).
 peak annotation, comparison and visualization. <i>Bioinformatics</i> 31, 2382–2383 (2015). Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607-8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545–15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1212	60.	
 Zhu, L. J. Integrative Analysis of ChIP-Chip and ChIP-Seq Dataset. in <i>Methods in</i> <i>Molecular Biology</i> 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 	1213		
 Molecular Biology 105–124 (Humana Press, 2013). doi:10.1007/978-1-62703-607- 8_8. 62. Zhu, L. J. et al. ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). 63. Subramanian, A. et al. Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). 64. Mi, H. et al. Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). 65. Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). 66. Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). 67. Shteynberg, D. et al. iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		61.	
 8_8. 217 62. Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). 63. Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). 64. Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). 65. Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). 66. Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). 67. Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 			
 Zhu, L. J. <i>et al.</i> ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 			
 1218 ChIP-chip data. <i>BMC Bioinformatics</i> 11, 237 (2010). 1219 63. Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 12550 (2005). 1222 64. Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). 1224 65. Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). 1226 66. Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). 67. Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		62.	
 Subramanian, A. <i>et al.</i> Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		•	
 for interpreting genome-wide expression profiles. <i>Proc. Natl. Acad. Sci.</i> 102, 15545– 15550 (2005). 64. Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). 65. Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). 66. Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). 67. Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		63	
 1221 15550 (2005). 1222 64. Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). 1224 65. Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). 1226 66. Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). 1228 67. Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 			
 Mi, H. <i>et al.</i> Protocol Update for large-scale genome and gene function analysis with the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 			
 the PANTHER classification system (v.14.0). <i>Nat. Protoc.</i> 14, 703–721 (2019). Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		64	
 Morita, S., Kojima, T. & Kitamura, T. Plat-E: an efficient and stable system for transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 			
 transient packaging of retroviruses. <i>Gene Ther.</i> 7, 1063–6 (2000). Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. <i>Bioinformatics</i> 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		65	
 1226 66. Craig, R. & Beavis, R. C. TANDEM: matching proteins with tandem mass spectra. 1227 <i>Bioinformatics</i> 20, 1466–1467 (2004). 1228 67. Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic 1229 Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		55.	
 Bioinformatics 20, 1466–1467 (2004). Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i> 		66	
122867.Shteynberg, D. <i>et al.</i> iProphet: Multi-level Integrative Analysis of Shotgun Proteomic1229Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i>		50.	U
1229 Data Improves Peptide and Protein Identification Rates and Error Estimates. <i>Mol.</i>		67	
		01.	

- 1231 68. Liu, G. *et al.* ProHits: integrated software for mass spectrometry–based interaction 1232 proteomics. *Nat. Biotechnol.* **28**, 1015–1017 (2010).
- Keller, A., Nesvizhskii, A. I., Kolker, E. & Aebersold, R. Empirical Statistical Model To
 Estimate the Accuracy of Peptide Identifications Made by {MS}/{MS} and Database
 Search. *Anal. Chem.* 74, 5383–5392 (2002).
- 1236 70. Nesvizhskii, A. I., Keller, A., Kolker, E. & Aebersold, R. A Statistical Model for
 1237 Identifying Proteins by Tandem Mass Spectrometry. *Anal. Chem.* **75**, 4646–4658
 1238 (2003).
- 1239 71. Teo, G. *et al.* SAINTexpress: Improvements and additional features in Significance 1240 Analysis of INTeractome software. *J. Proteomics* **100**, 37–43 (2014).
- 1241 72. Anders, S. & Huber, W. Differential expression analysis for sequence count data.
 1242 *Genome Biol.* 11, R106 (2010).
- 1243 73. Shannon, P. Cytoscape: A Software Environment for Integrated Models of 1244 Biomolecular Interaction Networks. *Genome Res.* **13**, 2498–2504 (2003).
- 1245 74. Chatr-Aryamontri, A. *et al.* The BioGRID interaction database: 2017 update. *Nucleic* 1246 *Acids Res.* **45**, D369–D379 (2017).
- 1247 75. Orchard, S. *et al.* The MIntAct project—IntAct as a common curation platform for 11 1248 molecular interaction databases. *Nucleic Acids Res.* **42**, D358–D363 (2014).
- 1249 76. Morris, J. H. *et al.* clusterMaker: a multi-algorithm clustering plugin for Cytoscape.
 1250 *BMC Bioinformatics* 12, 436 (2011).
- 1251 77. Huntley, R. P. *et al.* The GOA database: Gene Ontology annotation updates for 2015.
 1252 *Nucleic Acids Res.* 43, D1057–D1063 (2015).
- 125378.Giurgiu, M. *et al.* CORUM: the comprehensive resource of mammalian protein1254complexes-2019. *Nucleic Acids Res.* 47, D559–D563 (2019).
- 1255 79. Merico, D., Isserlin, R., Stueker, O., Emili, A. & Bader, G. D. Enrichment map: a
 1256 network-based method for gene-set enrichment visualization and interpretation. *PLoS*1257 *One* 5, e13984 (2010).
- 125880.Raudvere, U. *et al.* g:Profiler: a web server for functional enrichment analysis and1259conversions of gene lists (2019 update). *Nucleic Acids Res.* 47, W191–W198 (2019).
- 1260 81. Mi, H., Muruganujan, A., Ebert, D., Huang, X. & Thomas, P. D. PANTHER version 14:
 1261 more genomes, a new PANTHER GO-slim and improvements in enrichment analysis
 1262 tools. *Nucleic Acids Res.* 47, D419–D426 (2019).
- 1263 82. Methot, S. P. *et al.* A licensing step links AID to transcription elongation for 1264 mutagenesis in B cells. *Nat. Commun.* **9**, 1248 (2018).
- 1265