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# Transcriptome-wide identification of NMD-targeted human mRNAs reveals extensive redundancy between SMG6- and SMG7-mediated degradation pathways

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## ABSTRACT

**Besides degrading aberrant mRNAs that harbor a premature translation termination codon (PTC), nonsense-mediated mRNA decay (NMD) also targets many seemingly "normal" mRNAs that encode for full-length proteins. To identify a *bona fide* set of such endogenous NMD targets in human cells, we applied a meta-analysis approach in which we combined transcriptome profiling of knockdowns and rescues of the three NMD factors UPF1, SMG6 and SMG7. We provide evidence that this combinatorial approach identifies NMD-targeted transcripts more reliably than previous attempts that focused on inactivation of single NMD factors. Our data revealed that SMG6 and SMG7 act on essentially the same transcripts, indicating extensive redundancy between the endo- and exonucleolytic decay routes. Besides mRNAs, we also identified as NMD targets many long non-coding RNAs as well as miRNA and snoRNA host genes. The NMD target feature with the most predictive value is an intron in the 3' UTR, followed by the presence of upstream open reading frames (uORFs) and long 3' UTRs. Furthermore, the 3' UTRs of NMD-targeted transcripts tend to have an increased GC content and to be phylogenetically less conserved when compared to 3' UTRs of NMD insensitive transcripts.**

## INTRODUCTION

Nonsense-mediated mRNA decay (NMD) was initially described as a quality control mechanism clearing transcripts harbouring a premature termination codon (PTC) from the cell (Losson and Lacroute 1979; Maquat et al. 1981). Since a PTC can be caused by a mutation in the gene sequence or by aberrant pre-mRNA splicing, NMD was always associated with abnormal or pathological conditions. More recently however, a large number of mRNAs with no PTC were found to be downregulated, often only moderately, by the NMD pathway (Mendell et al. 2004; Rehwinkel et al. 2005; Tani et al. 2012; Wittmann et al. 2006; Yepiskoposyan et al. 2011). An emerging view of NMD

53  is therefore that of a post-transcriptional mechanism contributing to the fine-
54  tuning of gene expression.

55  The molecular mechanism of NMD is only partially understood and depends
56  on the interplay of many factors. On-going translation is a pre-requisite for
57  NMD to take place (Carter et al. 1995; Thermann et al. 1998) and there is an
58  emerging consensus that NMD results as a consequence of aberrant
59  translation termination (He and Jacobson 2015; Karousis and Mühlemann
60  2016; Lykke-Andersen and Jensen 2015), which can be detected as ribosome
61  stalling at NMD-eliciting termination codons (TCs) (Amrani et al. 2004;
62  Peixeiro et al. 2012). The ATP-dependent RNA helicase UPF1 is central for
63  NMD activation. UPF1 binds RNA rather unspecifically and independent of
64  translation (Hurt et al. 2013; Zünd et al. 2013; Hogg and Goff 2010).
65  Activation of NMD in metazoans involves phosphorylation of UPF1 by SMG1
66  (Kurosaki et al. 2014; Yamashita et al. 2001). In mammalian cells, two
67  mechanistically distinct pathways have been described to execute the
68  degradation of the target mRNAs (Mühlemann and Lykke-Andersen 2010).
69  The endonuclease SMG6 is recruited to NMD-targeted transcripts by
70  activated UPF1 (Okada-Katsuhata et al. 2012; Nicholson et al. 2014;
71  Chakrabarti et al. 2014) and cleaves them in the vicinity of the TC (Huntzinger
72  et al. 2008; Eberle et al. 2009; Lykke-Andersen et al. 2014; Schmidt et al.
73  2015; Boehm et al. 2014). For the second pathway, the SMG5/SMG7
74  heterodimer binds to phosphorylated SQ epitopes in the C-terminal part of
75  UPF1 and recruits through the C-terminus of SMG7 the deadenylase
76  CCR4/NOT (Jonas et al. 2013; Loh et al. 2013; Chakrabarti et al. 2014).
77  Whether the SMG6- and the SMG7-mediated decay pathways act
78  independently of each other and maybe even target a distinct subpopulation
79  of mRNAs has so far not been addressed on endogenous targets at a
80  genome-wide level.

81  The characteristics that render a seemingly "normal" endogenous
82  mRNA to be degraded by NMD by and large still need to be elucidated, owing
83  to our fragmented understanding of the mechanism underlying NMD-eliciting
84  aberrant translation termination. The best-characterized NMD-inducing
85  feature is the presence of an exon junction complex (EJC) farther than 50
86  nucleotides downstream of the TC, a situation that is very common also in

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87    PTC-containing transcripts. However, the presence of a downstream EJC

88    seems to enhance the efficiency of NMD (i.e. it increases the extent of mRNA

89    level reduction) rather than being an essential signal to trigger NMD (Bühler et

90    al. 2006; Singh et al. 2008). Several studies have also shown that long 3'

91    UTRs and short ORFs located upstream of the main ORF (uORFs) can

92    activate NMD (Mendell et al. 2004; Yepiskoposyan et al. 2011; Bühler et al.

93    2006; Eberle et al. 2008; Hansen et al. 2009; Singh et al. 2008). But these

94    transcript characteristics have only a limited predictive value and many

95    mRNAs with long 3' UTRs or uORFs are in fact insensitive to NMD

96    (Yepiskoposyan et al. 2011; Boehm et al. 2014; Hogg and Goff 2010;

97    Burroughs et al. 2010). It has been shown that poly(A) binding protein C1

98    (PABPC1) promotes translation termination *in vitro* (Ivanov et al. 2016) and

99    effectively antagonizes NMD when tethered close to PTCs in cells (Ivanov et

100    al. 2008; Eberle et al. 2008; Behm-Ansmant et al. 2007). It is thought that

101    UPF1 competes with PABPC1 for interacting with the eukaryotic release

102    factor 3 (eRF3) (Singh et al. 2008) and that an extended physical distance

103    between the terminating ribosome and the poly(A) tail-associated PABPC1,

104    which is a typical configuration of transcripts with uORFs, PTCs or long 3'

105    UTRs, increases the chance for UPF1 to win this competition and elicit NMD

106    (Mühlemann and Jensen 2012). On the other hand, RNA-binding proteins that

107    prevent UPF1 from accessing the mRNA just downstream of the TC can tilt

108    the balance towards proper termination and inhibit NMD, as recently shown

109    for Rous sarcoma virus (RSV) by polypyrimidine tract binding protein 1

110    (PTBP1) (Ge et al. 2016).

111         In order to better identify the NMD-eliciting features of mRNAs, a high

112    confidence set of endogenous NMD-sensitive transcripts is needed. Previous

113    genome-wide studies showed little agreement regarding the endogenous

114    NMD targets, and depletion of different NMD factors affected different sets of

115    transcripts, which prevents a reliable meta-analysis of these data. Partly,

116    these discrepancies can be explained by technical biases inherent to the

117    different methods used. To overcome these limitations, we performed RNA-

118    seq experiments following the latest best practices of the field. We carried out

119    knockdowns of three well-characterized NMD factors (UPF1, SMG6 and

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120 SMG7) and also operated the respective rescues, which allowed us to
121 increase the stringency and accuracy of the analysis.

122        Our results show that despite the existing individual differences, UPF1,
123 SMG6 and SMG7 similarly affect the abundance of a large number of
124 mRNAs. Among the NMD-targeted genes, we found a significant enrichment
125 of miRNA host genes, in addition to the already reported snoRNA host genes
126 (Lykke-Andersen et al. 2014). Many non-coding RNAs also appear to be
127 targeted by NMD, depending on the presence of open reading frames (ORFs)
128 in their sequence, consistent with recent reports showing ribosome
129 association of many supposedly non-coding transcripts (Ingolia et al. 2014).
130 Furthermore, we also obtained evidence of transcription upstream of
131 canonical start sites, which appears to be partially cleared by the NMD
132 pathway, but to a lesser extent than has been reported for yeast (Malabat et
133 al. 2015). Already known NMD-inducing features were also enriched among
134 our NMD targets, including 3' UTR introns, uORFs and long 3' UTRs.
135 Additionally, 3' UTRs of NMD targets have on average a higher GC content
136 and are phylogenetically less conserved than NMD-immune transcripts.

137 **RESULTS**

138 **Experimental setup to identify a high-confidence set of NMD-targeted**
139 **transcripts**

140 Based on the current literature it is unclear if what is commonly termed NMD
141 constitutes a single biochemical pathway or a blend of several ones. In
142 mammalian cells, evidence for UPF2- as well as for UPF3-independent NMD
143 has been reported (Gehring et al. 2005; Chan et al. 2007) and it has further
144 been suggested that SMG6 and SMG7 might represent two independent
145 branches of NMD to initiate target RNA degradation (Mühlemann and Lykke-
146 Andersen 2010). Given these uncertainties, we decided to operationally
147 define NMD as an RNA degradation pathway that depends on UPF1 and
148 SMG6 or SMG7. Accordingly, we performed shRNA-mediated knockdowns
149 (KD) in HeLa cells for these three NMD factors and also operated the
150 respective rescues by expressing an RNAi-resistant version of the respective

151  protein (Fig. 1A). To generate a reference and control dataset (Ctrl), a

152  knockdown with a scrambled shRNA sequence was performed. To address

153  the extent of redundancy between SMG6 and SMG7, we also performed

154  double knockdowns (dKD) of both factors and rescued this condition with

155  either SMG6 or SMG7. Western blotting showed efficient depletion of the

156  respective NMD factors in all knockdowns and rescue protein levels were

157  comparable or higher than the endogenous levels of the respective NMD

158  factor (Figure 1A). Checking the relative RNA levels of several previously

159  identified NMD-targeted transcripts showed increased RNA levels under our

160  knockdown conditions and a partial to complete rescue under our rescue

161  conditions (Fig. 1B). Thus, our experimental conditions resulted in the

162  attempted inhibition and at least partial rescue of NMD activity.


163  **mRNA-seq and principal component analysis**

164  Three independent biological replicates of the aforementioned 10 different

165  conditions were enriched for poly(A)+ RNA and subjected to high throughput

166  sequencing. The obtained reads were mapped to the human genome

167  (GRCh38) using TopHat (Kim et al. 2013) and gene counting was performed

168  with the program featureCounts (Liao et al. 2014). Since the library

169  preparations and sequencing was carried out in two batches (batch A: Ctrl,

170  UPF1 KD, SMG6 KD, and respective rescue samples; batch B: Ctrl, SMG7

171  KD, dKD and respective rescue samples), a total of 6 Ctrl reference samples

172  was produced. Although the principle component analysis (PCA) showed the

173  absence of a significant batch effect, indicated by the close clustering of the

174  Ctrl samples (Supplemental Fig. S1), we nevertheless opted to separately

175  compare every sample to the controls of the respective batch. The PCA also

176  revealed that the KD samples of the different factors do not cluster as close to

177  each other as one might have expected, which can have different

178  explanations. One reason is that the PCA maximizes the components of

179  variation between all the samples and are equally influenced by both

180  upregulated (i.e. NMD targeted) and downregulated genes (i.e. non-NMD-

181  related effects), of which the downregulated ones are not expected to be

182  conserved for different NMD factors. It should also be noted that the first two

183 principal components only report a fraction of the total variation present in the

184 dataset (in this case 49%). Additionally, UPF1 is known to be involved in

185 several biological processes other than NMD (Isken and Maquat 2008), which

186 may explain why UPF1 KDs cluster farther away from the other samples. The

187 fact that the rescue samples are all closer to their respective KDs than to the

188 control (Ctrl) confirms our observation from individual transcripts (Fig. 1B) that

189 we only achieved a partial rescue in our experiments, despite of an overall

190 higher-than-endogenous expression of the RNAi-resistant constructs (Fig. 1A

191 and Supplemental Table S1). This can be attributed to the fact that a

192 puromycin selection marker on the shRNA encoding plasmids enabled us to

193 achieve the knockdown in essentially every cell surviving the selection, which

194 was not the case for the cells expressing the rescue construct. Therefore a

195 fraction of the cells in the rescue conditions was depleted for the endogenous

196 NMD factor yet did not express the corresponding rescue construct, resulting

197 in the observed partial rescue.

198 **Bioinformatics approach to identify *bona fide* NMD targeted genes**

199 A first differential expression analysis was conducted at the gene level using

200 DESeq2 (Love et al. 2014) to compute log2 fold changes (log2FC) between

201 two conditions. To represent the knockdown and the rescue effect for each

202 gene as a single value, the KD/Ctrl log2FC and the inverse of the rescue/KD

203 log2FC were averaged, thus resulting in a positive value for NMD targets (see

204 methods). The respective *p*-values of these two comparisons were combined

205 using a method called "sum of *p*-value" (Edgington 1972) (Supplemental Fig.

206 S2A), which allows the detection of differentially expressed genes with

207 enhanced sensitivity and confidence. The overall good negative correlation

208 between these two log2FC values justifies this approach (Supplemental Fig.

209 S2B). From here on, when describing targets of e.g. UPF1 or the dKD

210 rescued with SMG6, we will refer exclusively to these combined log2FC (KD-

211 rescue log2FC) and *p*-values (KD-rescue *p*-values). Using this approach and

212 Fisher's method (see Supplemental Methods), we computed a global list of

213 significant differentially expressed genes (Supplemental Table S2). Since the

214 meta-analysis procedure tends to inflate the number of significant results and

215   since it was already observed that RNA-seq methods cannot control their true

216   false discovery rate (FDR) with few replicates (Soneson and Delorenzi 2013),

217   we decided to focus on the top 1000 most significant genes of our analysis,

218   defining this as our high-confidence set of NMD targets. A comparison with

219   the transcripts annotated as NMD sensitive in the Ensembl database proved

220   the benefit of this combinatorial approach over relying on data from individual

221   knockdown or rescue conditions (Supplemental Fig. S3). This comparison

222   also revealed that the rescue conditions (log2FC rescue/KD) generally

223   identified more annotated NMD transcripts than the corresponding knockdown

224   conditions (log2FC KD/Ctrl), despite being only partial (see above). Besides

225   genes that show the pattern expected for NMD targets (RNA increase upon

226   KD and decrease in rescue; see Supplemental Table S3, "positive results"),

227   comparable numbers of genes were affected in the opposite way (RNA

228   decrease upon KD and increase in rescue; see Table S3, "negative results").

229   However, many more of the former are characterized by a higher change in

230   expression (log2FC > 0.5, Table S3). Negative results are also less shared

231   among the different factors (Supplemental Fig. S4), suggesting that they are

232   more likely to arise from transcriptional noise. These observations together

233   indicate that indirect gene expression variation caused by NMD inactivation is

234   secondary to the changes in RNA levels that can be attributed to active NMD

235   decay. Finally, it's important to underline the fact that the single SMG7 KD

236   inhibited NMD to a much lesser extent than the KD of all other NMD factors,

237   an observation that we already reported in an earlier study (Yepiskoposyan et

238   al. 2011).

239   **Identified NMD targets show expected properties**

240   We expected the rescue experiments to increase the accuracy of identifying

241   NMD targeted RNAs because of the reduction of off-target hits. To test the

242   actual benefits of the rescues, we compared our data to a previous published

243   dataset that was also produced by RNA-seq from RNA of HeLa cells (Tani et

244   al. 2012). The correlation of the simple log2FC obtained in UPF1 KD is very

245   low (Supplemental Fig. S5A). This is a common situation when comparing

246   RNA-seq datasets of different publications, which is mainly caused by

247 environmental and technological variables that strongly impact the results.

248 However, the overlap between the two datasets improves considerably when

249 the UPF1 rescue is included in the analysis (Supplemental Fig. S5B). The

250 correlation score increases from 0.06 to 0.21 and the 100 top targets of our

251 meta-analysis (red dots) are also more evidently among the most strongly

252 reacting genes in this comparison.

253 To further assess the quality of our data, we complemented it with a

254 UPF1 CLIP dataset that we generated by expressing C-terminally FLAG-

255 tagged UPF1 in HeLa cells depleted for the endogenous UPF1. Consistent

256 with previously observed preferential steady-state association of UPF1 with

257 NMD targets (Johansson et al. 2007; Johns et al. 2007; Kurosaki and Maquat

258 2013; Silva et al. 2008; Hwang et al. 2010; Lee et al. 2015), the top 1000

259 significant NMD targets are overall enriched in UPF1 CLIP tags

260 (Supplemental Fig. S6), demonstrating a strong correlation between the two

261 datasets.

262 We have further compared our top NMD targets with the SMG6

263 cleavage sites determined by Schmidt and colleagues (Schmidt et al. 2015).

264 In this study, the authors employed parallel analysis of RNA ends (PARE) to

265 determine 5' termini of RNA decay intermediates produced by SMG6 and

266 dependent on UPF1. The determined SMG6 targets in this study are strongly

267 enriched in our list of significant NMD targets (Supplemental Fig. S7). The

268 overlap is nevertheless far from complete, as already observed in the original

269 publication (Schmidt et al. 2015).

270 **UPF1, SMG6 and SMG7 define a homogenous pathway**

271 To have a comprehensive view of the entire data, we performed a cluster

272 analysis without adding any *a priori* information (Fig. 2A). The most significant

273 cluster (cluster 1), of the two we determined, comprises ~ 40'000 genes and

274 shows no particular trend of differential expression, consistent with the

275 expectation that the majority of poly(A)+ RNAs are not targeted by NMD. The

276 second cluster (cluster 2) comprises ~ 4'000 genes and shows the differential

277 expression pattern expected for NMD targets, in which the log2FC is positive

278 for KD/Ctrl and negative for rescue/KD conditions. This shows that NMD is

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279    responsible for the most relevant pattern in our data, having a stronger impact
280    than any other effect or bias. It is also interesting to note that among the less
281    significant additional clusters, none shows any factor-specific trend
282    (Supplemental Fig. S8), highlighting the overall uniformity of our data and
283    indicating the absence of significant sub-pathways, i.e. branched or
284    alternative NMD routes.

285        One of the key features of our dataset is the extent of overlap between
286    the different NMD factors studied. A high overlap would first of all confirm that
287    our approach indeed yields a high confidence list of NMD targeted transcripts
288    and at the same time justifies the definition of NMD as a UPF1, SMG6 and
289    SMG7-dependent pathway. To get an overview on the extent of overlap in our
290    data, we selected for every condition the top significant 1000 targets and
291    examined what percentage of those targets was also identified in the other
292    conditions (Fig. 2B). Overall, we find an extensive overlap between most
293    conditions. For example, at least 90% of the top 1000 UPF1 targets also
294    showed a positive KD-rescue log2FC in the other conditions, albeit not
295    necessarily significant, with the only exception of SMG7. It is indeed evident
296    that the SMG7 condition correlates least with the other ones by a large
297    margin. At least in part, this is due to a smaller number of genes being
298    statistically significantly affected by SMG7, resulting in a higher proportion of
299    false positive hits among the top 1000 targets, which do not correlate with the
300    targets of the other conditions. It is interesting to observe that SMG7 KD is
301    leading to a notable upregulation of both UPF1 and SMG6 levels (Fig. 2C).
302    This autoregulatory feedback might explain why SMG7 KD only weakly
303    impaired NMD. This effect was also observed, albeit less pronounced, in
304    UPF1 KD but not in SMG6 KD. The comparison of SMG6 and SMG7 single
305    KDs is therefore biased by the autoregulation phenomenon. For this reason,
306    the double KD of these two factors is essential to disentangle their individual
307    contributions to the NMD pathway.

308    **SMG6 and SMG7 act on the same target genes**

309    With regards to the previously proposed independent pathways to degrade
310    mammalian NMD targets (Mühlemann and Lykke-Andersen 2010), the

311    extremely high overlap of the most significant targets between dKD SMG6

312    and dKD SMG7 is intriguing (Fig. 2B). Since these lists are computed taking

313    into account the same dKD and Ctrl samples, they are not independent and

314    their similarity could be overestimated. We therefore directly compared the

315    effects caused by SMG6 and SMG7 rescues in the dKD cells (Fig. 3A). As

316    can be seen from the spindle-shaped cloud of dots, most genes were affected

317    in the same way by the SMG7 rescue and the SMG6 rescue, with an overall

318    stronger effect caused by the SMG6 rescue. This correlation is highly

319    significant (Pearson's correlation coefficient 0.842) and there is no substantial

320    group of genes reacting in only one of the two conditions. To have a more

321    rigorous statistical view on the variance present in this comparison, we also

322    performed some simulations. In these simulations we generated a

323    hypothetical SMG7 rescue dataset, in which we reproduced the changes in

324    RNA levels observed in SMG6 rescue. We could thus measure how much

325    these two rescue conditions would correlate, if there were absolutely no

326    difference in the specificity of the two factors. The correlation of this new

327    variable with the SMG6 log2FC was 0.883 ± 0.001, close to the one observed

328    for the real data. This shows that most of the variation present in this

329    comparison is caused by transcriptional noise and that only a small amount of

330    it is actually caused by the different activity of SMG6 and SMG7. To estimate

331    this additional small variation, we applied some variability to the simulated

332    gene expression and compared the resulting correlation with the observed

333    one (Fig. 3B). This allowed us to estimate the average extent of gene

334    expression specificity between SMG6 and SMG7 to 8.19% ± 0.15. From

335    these analyses we can conclude that SMG7 activity has generally a weaker

336    effect than SMG6, but they act on the same genes.

337    **Long "non-coding" RNAs, small-RNA host genes and pervasive**
338    **transcripts are targeted by NMD**

339    To get a first overview on what kind of RNAs we have among our 1000 most

340    significant NMD targets, we categorized them according to their biotype (Fig.

341    4A). As expected, the majority (78%) of the genes codes for proteins.

342    However, there are also a considerable proportion of various non-coding

343    genes, with the main sub-classes being pseudogenes (9%), long intergenic

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344  non-coding RNAs (lincRNAs; 6%) and antisense transcripts (4%). Given that
345  NMD is a translation-dependent process, it might be surprising at first sight
346  that several genes annotated as "non-coding" are affected. However, many
347  pseudogenes are known to give rise to PTC-containing mRNAs (Mitrovich and
348  Anderson 2005) and recent ribosome profiling studies found many transcripts
349  categorized as lincRNAs to be associated with ribosomes (Ingolia et al. 2011;
350  Calviello et al. 2015; Carlevaro-Fita et al. 2016). In few cases, the short
351  polypeptides encoded by these lincRNAs were even detected (Ingolia et al.
352  2014) thus revealing them as a misnomer. Given their documented evidence
353  for associating with ribosomes, one would in fact predict that these mostly
354  short ORFs, similar to uORFs, would terminate translation in an mRNP
355  context that leads to NMD activation. Supporting this view, we find a strong
356  correlation between the number of predicted ORFs (minimal length of 3
357  codons) on a non-coding RNA and its likelihood to be identified as an NMD
358  target in our study (Fig. 4B).

359      An interesting group of genes that is significantly enriched in our data
360  comprises host genes for snoRNAs and miRNAs (Fig 4C). Consistent with a
361  previous study reporting an overrepresentation of snoRNA host genes among
362  NMD targets (Lykke-Andersen et al. 2014), snoRNA host genes are 3-fold
363  enriched among our top 1000 hits, when compared to genes that neither host
364  snoRNAs nor miRNAs (non hosts) ($p$-value = 6e-6, Fisher exact test).
365  Similarly, we also detected a 3-fold enrichment of miRNA host genes ($p$-value
366  = 4e-14). The snoRNAs and miRNAs are encoded in the introns of these
367  genes and processed from the excised introns of the pre-mRNAs. In many
368  cases, the spliced RNAs do not encode a functional protein and the non-
369  conserved short ORFs occurring in these spliced RNAs will presumably be
370  translated and trigger NMD because of aberrant translation termination in the
371  same way as proposed for lincRNAs and mRNAs containing uORFs. Thus,
372  while high transcription rates of the snoRNA and miRNA host genes are
373  required to produce sufficient amounts of these small RNAs, NMD ensures
374  that the spliced host RNAs, which can represent waste products for the cell,
375  are quickly degraded.

376      In *S. cerevisae*, another group of transcripts has recently been
377  revealed to be abundant among the NMD targets, namely transcripts whose

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378 transcription starts upstream of the canonical transcription start sites (TSS)
379 (Malabat et al. 2015). Owing to their additional sequence upstream of the
380 main ORF, they have an increased likelihood to contain uORFs that will
381 activate NMD. To see if such pervasive transcripts are also present among
382 the NMD targets in human cells, we searched in our data for sequence
383 coverage 200 bp upstream of annotated TSS, ignoring intervals where other
384 genetic annotations were present. The log2FCs of these "upstream TSS"
385 sequences, normalized with the expression level of the corresponding gene,
386 follow an NMD-sensitive distribution, even though the effect size is rather
387 small (Fig. 4D). We conclude that although detectable, pervasive transcripts
388 are not commonly present in our data, possibly because pervasive
389 transcription occurs at a much lower frequency in human cells than in yeast.

390 **Different 3' UTR features are associated with NMD sensitivity**

391 In the absence of a detailed understanding of the mechanism of NMD,
392 empirical identification of transcript features that can differentiate NMD targets
393 from non-targets is an important and active area of research. Even though
394 such feature searches only yield correlations without implying a causal
395 connection, they can help characterize the pathway and formulate hypotheses
396 on the molecular mechanism. To have the best-possible correlation between
397 mRNA properties and their level of degradation by NMD, we performed a
398 transcript-level analysis. In this analysis, the expression of all the different
399 splicing isoforms of each gene is estimated independently, providing
400 information on the behaviour of specific mRNA molecules. The difficulty of
401 uniquely assigning reads to single transcripts, however, determines a lower
402 accuracy, compared to a gene-level study (Soneson et al. 2015). The
403 combination of all conditions into a single measure was carried out in the
404 same way as the gene-level analysis.

405 By far, the most prominent and significant NMD feature in our analysis
406 is the presence of an intron in the 3' UTR located more than 50 nucleotides
407 downstream of the stop codon ($p$-value < 2e-16, Fisher exact test). 40% of the
408 significant targets are characterized by this property (Fig. 5A). This confirms
409 many previous studies and supports the model that an EJC downstream of a

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410  stop codon highly facilitates the decay process on the mRNA. To analyze

411  additional features, we focused only on the transcripts that do not contain a 3'

412  UTR intron. This is motivated by the fact that this property can mask the

413  presence of other features. For example, it was reported that mRNAs with an

414  intron in the 3' UTR are more strongly degraded when the 3' UTR is short

415  (Hurt et al. 2013). Focusing on this filtered set we analyzed the presence of

416  uORFs. For this analysis, uORFs were defined as ≥ 3 codon-long ORFs with

417  both an AUG and a stop codon upstream of the main ORF, being aware that

418  *in vivo* not all of these uORFs will actually be translated. Notwithstanding this

419  oversimplification, we observed a highly significant enrichment of uORFs

420  among the NMD targets (*p*-value: 2e-10). All mRNAs with a uORF were

421  similarly discarded from further analyses.

422      Besides the presence of an intron, additional characteristics of the 3'

423  UTR are also important in determining if an RNA is targeted by NMD (Fig.

424  5B). In our data, we found that NMD targets show a longer 3' UTR than a

425  control group that we defined as a set of mRNAs with similar Ctrl expression

426  levels as the NMD sensitive ones (*p*-value: 2e-5, permutation test). This

427  feature, despite extensive experimental validation, shows only a limited

428  statistical significance in our data. Nevertheless, with a median 3' UTR length

429  of 836 nucleotides, NMD targets tend to have on average ~ 50% longer 3'

430  UTRs than transcripts of the control group (median length of 561 nucleotides).

431  In addition, the GC content of the 3' UTRs of NMD targets is also significantly

432  higher than in NMD-insensitive transcripts (*p*-value: 2e-10). This finding is in

433  line with the higher UPF1 propensity to GC rich regions we observe in our

434  CLIP experiment (Supplemental Fig. S9). It was suggested that UPF1

435  ATPase and helicase activity was reduced when associated with GC motifs

436  (Bhattacharya et al. 2000), which could lead to its enrichment on GC-rich 3'

437  UTRs and thereby promote NMD. Furthermore, we also determined the

438  phylogenetic conservation of the 3' UTRs in NMD targets and the control

439  group (PhyloP score) (Pollard et al. 2010) and found that 3' UTRs of NMD

440  targets are significantly less conserved (*p*-value: 7e-13). The biological

441  meaning of this lower phylogenetic conservation of the 3' UTRs of NMD

442  targets is unknown and possible explanations remain speculative at this point

443  (see Discussion).

14

444 **DISCUSSION**

445 This study presents an attempt to determine a high-confidence set of

446 endogenous transcripts targeted by NMD. Several such attempts have been

447 previously reported, using either microarrays (Mendell et al. 2004; Wittmann

448 et al. 2006; Viegas et al. 2007; Yepiskoposyan et al. 2011) or deep

449 sequencing (Tani et al. 2012; Hurt et al. 2013; Schmidt et al. 2015). However,

450 the overlap among the hits in these different studies was minimal, questioning

451 the robustness of the results. Our approach differs from these previous

452 studies in the high sequencing quality and an experimental design that

453 combines data from 10 different experimental conditions. The rescue

454 conditions substantially improved the accuracy of the differential expression

455 detection by controlling for indirect effects and possible biases introduced by

456 the shRNA procedure (Supplemental Fig. S3). Furthermore, the meta-analysis

457 of UPF1, SMG6 and SMG7 also resulted in increased statistical power and

458 helped to filter out individual false positive hits. In particular, studies

459 investigating only the effect of UPF1 depletion on the transcriptome are prone

460 to yield many false positives because UPF1 is involved in additional pathways

461 beside NMD (Isken and Maquat 2008). Finally, the SMG6/SMG7 double KDs

462 were crucial to reveal the redundancy in the activity of SMG6 and SMG7. The

463 single SMG7 KD caused a 77% upregulation of SMG6 mRNA (Fig. 2C), which

464 prevents an unbiased evaluation of the extent of SMG7 activity in normal

465 cells, since many of its targets are likely masked by an increased SMG6

466 activity. If instead the cells were depleted of both factors, it became evident

467 that rescue plasmids of either gene had very similar effects on the

468 transcriptome. Although the magnitude of these changes was clearly higher

469 for SMG6, we determined that SMG7 acts on essentially the same targets,

470 with an estimated variability of only 8.2% (Fig. 3B). We therefore conclude

471 that SMG6- and SMG7-mediated degradation routes appear to be two highly

472 redundant branches of the mammalian NMD pathway. This is consistent with

473 previous observations made with reporter genes (Luke et al. 2007; Jonas et

474 al. 2013; Metze et al. 2013) and in line with what was observed in a study

475 focusing on SMG6 endonucleolytic activity (Schmidt et al. 2015). In this work,

476 the authors observed an accumulation of decapped transcripts upon depletion

477   of SMG6, indicating the presence of a complementary decay mechanism that
478   most likely involves SMG7 mediated recruitment of the CCR4/NOT
479   deadenylase followed by decapping of the deadenylated RNAs.

480        Our approach allowed us to confirm and expand several previous
481   observations. For example the GAS5 transcript, which is not associated to
482   any known peptide sequence, has been previously discovered to be stabilized
483   upon NMD inactivation (Weischenfeldt et al. 2008; Tani et al. 2013). In our
484   study, we found that a substantial fraction of the genes affected by NMD are
485   associated in the databases to a non-coding biotype (Fig. 4A). The
486   dependency of NMD on translation indicates that this classification as "non-
487   coding" may be inaccurate and that in reality these RNAs engage the
488   translation machinery. This hypothesis is supported by a clear correlation
489   between the number of possible ORFs on these transcripts and the likelihood
490   of being subject to NMD (Fig. 4B) and by the fact that recent ribosome
491   profiling studies and polysome analyses found many non-coding RNAs to be
492   associated with ribosomes (Ingolia et al. 2011, 2014; Carlevaro-Fita et al.
493   2016).

494        In our list of NMD targets, we found a significant enrichment of snoRNA
495   and miRNA host genes (Fig. 4C). SnoRNA host genes have already been
496   described to be frequently targeted by NMD and it was proposed that this
497   uncouples the expression of the snoRNAs and the corresponding host gene
498   (Lykke-Andersen et al. 2014). The same regulation appears to apply for
499   miRNA host genes, which we also show to frequently undergo NMD. We
500   speculate that from many miRNA host genes a cell only requires high
501   numbers of the specific mature miRNA but not of the cognate spliced
502   transcript. Splicing the pre-mRNA of a miRNA host gene to an NMD-sensitive
503   transcript ensures low levels of that transcript despite a high transcription rate
504   of the gene.

505        In yeast, it has been shown that RNA polymerase II transcription often
506   initiates upstream the usual transcription start site (a phenomenon called
507   pervasive transcription) and it was shown that NMD plays an important role
508   in clearing these spurious transcripts (Malabat et al. 2015). In our data we
509   could also observe a similar activity, even though the phenomenon appeared
510   to be much less common than in yeast (Fig. 4D). It should however be noted

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511    that Malabat and colleagues used an experimental method aimed at

512    specifically identifying transcriptional start sites (TSS sequencing), which is

513    much more sensitive in detecting even very low abundant pervasive

514    transcripts than a normal RNA-seq like ours. Nevertheless, our data indicates

515    that unlike in yeast, such pervasive transcripts only constitute a small fraction

516    of the NMD-targeted transcriptome in mammalian cells, presumably because

517    the frequency of spurious transcription initiation is much lower than in yeast.

518          An accurate list of *bona fide* NMD targets may help to uncover

519    common features among NMD-sensitive transcripts, which in turn can give

520    further insights into NMD target identification and eventually allow the

521    computational prediction of NMD targets. The most prominent feature we

522    could determine in our data is, as expected, the presence of an intron in the 3'

523    UTR farther than 50 nucleotides downstream of the TC (Fig. 5A). The NMD-

524    stimulating effect of EJCs is well known and characterized (Karousis and

525    Mühlemann 2016). However, there is a significant portion of transcripts

526    among the NMD targets that lacks a 3' UTR intron, proving the existence of

527    other NMD-triggering signals. Among the NMD targets without 3' UTR introns

528    we could observe a significant enrichment for the presence uORFs as the

529    second most relevant feature. If we then focus on targets without either of

530    these two characteristics, we observe longer 3' UTRs to be significantly

531    correlated with NMD susceptibility (Fig. 5B). Models proposing a common

532    mechanism through which these different features lead to NMD have been put

533    forward (Amrani et al. 2004; Stalder and Mühlemann 2008; Schweingruber et

534    al. 2013) and there is ample supporting evidence for them. In quantitative

535    terms however, we want to emphasize that the presence of a 3' UTR intron is

536    by far the most important criteria with the most predictive power, whereas the

537    predictive power of uORFs and long 3' UTRs is rather limited. Many

538    transcripts with long 3' UTRs or predicted uORFs are in fact known to escape

539    NMD and for some of them the NMD-protecting factors are known. For

540    example, PTBP1 binding to the RNA stability element in Rous sarcoma virus

541    protects the viral genomic RNA from NMD by preventing interaction of UPF1

542    with the 200 nucleotide region downstream of the TC (Ge et al. 2016).

543    Furthermore, a high AU content within the first 200 nucleotides downstream of

544    the TC of several mRNAs with long 3' UTRs has also been reported to confer

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545  resistance to NMD, but no trans-acting factor was identified (Toma et al.
546  2015). These findings point towards a complex interplay between NMD-
547  promoting and NMD-inhibiting determinants that to a large extent still remain
548  to be elucidated.

549       In line with AU-rich elements in the 3' UTR correlating with NMD
550  resistance, the GC content in the 3' UTRs of our 660 NMD targets lacking a
551  uORF or 3' UTR intron was significantly higher than in a control group of
552  NMD-immune transcripts. This finding is also consistent with our CLIP data,
553  showing a higher propensity of UPF1 for GC-rich regions, and with recent
554  reports of a significant enrichment for guanosine residues in UPF1 binding
555  regions (Hurt et al. 2013). G-rich and GC-rich sequences have a higher
556  propensity to form secondary structures and it has been speculated that such
557  secondary structure might slow down the helicase/translocase activity of
558  UPF1, thereby resulting in its enrichment in these regions (Hurt et al. 2013).

559       Finally, analysis of the phylogenetic conservation of the 3' UTR
560  sequences revealed a significantly lower conservation of the 3' UTRs of NMD
561  targets compared to NMD-insensitive 3' UTRs (Fig. 5B). We can currently
562  only speculate about the biological meaning of this intriguing finding. Novel
563  transcripts, arising for example from gene duplications, transposons insertions
564  or viral infections, could often be detrimental for the cell. We hypothesize that,
565  at this stage, the transcripts often would present features that render them
566  susceptible to NMD, which in fact may be beneficial for the cell. If however
567  such a transcript acquires a new function that educes a selective pressure for
568  increased gene expression, transcript variants escaping NMD will now confer
569  a selective advantage. This could occur for example by evolving binding sites
570  in the 3' UTR that would protect the transcripts from NMD (as discussed
571  above). Such NMD-avoiding motifs would then be conserved in the future,
572  since they provide an evolutionary advantage. According to this view, NMD
573  might have an important role for the evolution of genomes in that it enables
574  cells to entertain an evolutionary playground by reducing the detrimental
575  effects that could be caused by young and not yet fully functional genes. This
576  scenario could explain our observation that younger genes appear to be more
577  susceptible to NMD than evolutionary more ancient ones. Notably, the recent
578  observation that some RNA virus genomes are recognized and degraded by

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579 NMD would be consistent with this scenario (Balistreri et al. 2014; Garcia et
580 al. 2014).

581      In summary, we believe that the set of endogenous NMD-targeted
582 transcripts that we have identified herein will provide a highly valuable
583 resource and reference for the scientific community for further investigations
584 into both the biological role and the mechanism of NMD.


# MATERIALS AND METHODS

585


## Experimental methods

586

587 For knockdowns, $2x10^5$ HeLa cells were seeded into six-well plates and 24
588 hours later the cells were transiently transfected using Dogtor (OZ
589 Biosciences). For single factor knockdowns, 400 ng of pSUPERpuro plasmids
590 expressing shRNAs against UPF1, SMG6, SMG7 or control plasmids were
591 transfected. For the knockdown and rescue conditions 400 ng pcDNA3-NG-
592 UPF1-WT-Flag, pcDNA3-SMG6-FLAG or pcDNA3-SMG7-FLAG were
593 included in the transfection mixtures. For double knock-down experiments 400
594 ng of each pSUPERpuro plasmid was added and the rescue of each factor
595 was achieved by including 400 ng of pcDNA3-SMG6-FLAG or pcDNA3-
596 SMG7-FLAG accordingly. The cells were split into a T25-cm$^2$ cell culture flask
597 and selected with puromycin at a concentration of 1.5 µg/µL. 24 hours prior to
598 harvesting the cells were washed with PBS and the puromycin-containing
599 medium was exchanged with normal DMEM-FCS medium. Cells were
600 harvested 4 days after transfection.
601 The shRNA target sequence for UPF1 and SMG6 were described in
602 (Paillusson et al. 2005) and SMG7 was described in (Metze et al. 2013). Total
603 RNA was extracted using the GenElute Mammalian Total RNA Miniprep Kit
604 (Sigma-Aldrich).
605 Cell harvesting for protein samples (derived from the same sample as RNA
606 preparation) and measurement of relative mRNA levels by reverse
607 transcription quantitative polymerase chain reaction (RT-qPCR) were done as
608 described in (Nicholson et al. 2012). Briefly, $2x10^5$ cell equivalents were
609 analyzed on a 10% PAGE and detection was performed using Anti-RENT1

610 (UPF1) (Bethyl, A300-038A), anti- EST1 (SMG6) (Abcam, ab87539), Anti-

611 SMG7 (Bethyl, A302-170A) and Anti-CPSF73 (custom made) antibodies.

612 qPCR assays have been described elsewhere (Yepiskoposyan et al. 2011),

613 except for the assays to measure the following genes:

614 GAS5 (5´-GCACCTTATGGACAGTTG-3´, 5´-GGAGCAGAACCATTAAGC-3´)

615 CDKN1A (5´-GACCAGCATGACAGATTTCTAC3´, 5´-CAAACTGAGACTAAG

616 GCAGAAG), TMEM183A (5´-TGCTCCGGCCGAGTGA-3´, 5´-

617 ACCGCCGGAT

618 CCGAGTT-3´), RP9P (5´- CAAGCGCCTGGAGTCCTTAA-3´, 5´-

619 AGGAGGTTT

620 TTCATAACTCGTGATCT-3´), GADD45B (5´-TCAACATCGTGCGGGTGTCG-

621 3´, 5´-CCCGGCTTTCTTCGCAGTAG-3´), ATF4 (5´- TCAACATCGTGCGGGT

622 GTCG-3´, 5´- CCCGGCTTTCTTCGCAGTAG-3´)

623 A total of 33 samples were sequenced: control knockdowns (Ctrl) in 6

624 replicates, all other conditions in triplicates. TruSeq Stranded mRNA kit

625 (chemistry v3) was used in the preparation of the library and in the poly(A)

626 enrichment step. The first batch was sequenced on an Illumina HiSeq2500

627 and the second on an Illumina HiSeq3000 machine. Reads are single-end

628 and 100bp long. The sequencing depth of every sample is reported in

629 Supplemental Table S4.


630 **UV cross-linking and immunoprecipitation (CLIP) of UPF1-Flag**

631 Knockdown of endogenous UPF1 was induced in HeLa tTR-KRAB-shUPF1

632 cells (Metze et al. 2013) by addition of 5 μg/mL doxycycline and 8x10$^6$ cells

633 were transiently transfected with 4 μg of a pcDNA3 expression plasmid

634 encoding a C-terminally Flag-tagged, RNAi-resistant version of UPF1 using

635 30 μL of Lipofectamine 2000. 44 hours post transfection, cells were washed

636 and cross-linked in ice-cold PBS applying 150 mJ/cm$^2$ UV-C light (Bio-Link

637 BLX-E, 254 nm). After irradiation, cells were scraped of the culture dish,

638 collected by centrifugation, flash-frozen in liquid nitrogen and stored at -80 °C.

639 After cell lysis in 3 mL hypotonic lysis buffer (10 mM Tris-HCl pH 7.5, 10 mM

640 NaCl, 2 mM EDTA, 0.5% (v/v) Triton X-100, Halt Protease Inhibitor Cocktail)

641 and removal of cell debris by centrifugation, the supernatant was adjusted to

642    160 mM NaCl and incubated with 30 U RNase I (Ambion) and 15 U Turbo

643    DNase (Ambion) at 37 °C for 7.5 minutes. 160 µL Dynabeads Protein G were

644    incubated with 18 µg of mouse anti-FLAG M2 antibody (Sigma Aldrich),

645    washed and resuspended in 1 mL hypotonic lysis buffer and incubated with

646    the cell lysate at 4 °C for 1.5 hours. The beads were then washed three times

647    with IP-buffer (50 mM HEPES-NaOH pH 7.5, 300 mM KCl, 0.05% (v/v) NP-

648    40, Halt Protease Inhibitor Cocktail). To label the co-precipitated RNA

649    fragments, dephosphorylation with Antarctic phosphatase was followed by

650    incubation with 22.5 µL γ-32P-ATP (10 mCi/mL), 5 µL 100 mM ATP, and 100

651    U T4 Polynucleotide Kinase in a total volume of 150 µL at 37 °C for 45

652    minutes. The protein-RNA adducts were heat-eluted from the beads, resolved

653    at 70 °C on a 2x NuPAGE Novex buffer 4-12% Bis-Tris Midi Gel (Life

654    Technologies), transferred to nitrocellulose membrane using the iBlot system,

655    and visualized by phosphorimager scanning. The section of the membrane

656    harboring the RNA-UPF1 adducts was excised and the RNA fragments were

657    retrieved by Proteinase K digestion followed by phenol/chloroform extractions

658    and ethanol precipitation. cDNA library preparations and Illumina sequencing

659    was performed at Fasteris (Geneva, Switzerland) according to their standard

660    small RNA sequencing protocol. All bioinformatics analyses were performed

661    in the same way as with the RNA-seq data, which is described in the following

662    paragraphs.

663    **Gene counting and differential expression analysis**

664    For the gene-level analysis, sequencing reads were processed with

665    Trimmomatic (Bolger et al. 2014) to remove low quality regions, poly(A) tails

666    and adapter sequences. The reads were then mapped to the human genome

667    (GRCh38) with TopHat (Kim et al. 2013), version 2.0.13. This step was aided

668    by the use of the Ensembl gene annotation, release 81. The gene counting

669    was performed with the program featureCounts (Liao et al. 2014), version

670    1.4.6. DESeq2 (Love et al. 2014) was employed for the differential expression

671    computation, version 1.6.3. All comparisons were corrected with the sva

672    package, version 3.12.0, in order to compensate for secondary biases in the

673    data. The transcript-level analysis was based on the isoform abundance

674  estimation provided by RSEM (Li and Dewey 2011), version 1.2.19. The
675  annotation used was Ensembl, release 84.


676  **Meta-analysis of the data**

677  Our final goal is to provide a unique score for every gene, to estimate its
678  likelihood of being an NMD target. The first step to achieve such synthetic
679  result was to combine KD and rescue conditions for every factor. A joint gene-
680  specific log2FC value was generated by computing the average between the
681  KD/Ctrl log2FC and the inverse of the rescue/KD log2FC. So if a gene is
682  upregulated in UPF1 KD compared to Ctrl and it's downregulated in the
683  rescue compared to the KD will have a high positive combined log2FC. We
684  called this quantity KD-rescue log2FC. The significance of this combined
685  log2FC was computed by a technique called sum of *p*-value (Edgington
686  1972). All genes downregulated in a KD or upregulated in a rescue, were
687  assigned a p-value of 1 before applying this algorithm. We called this quantity
688  KD-rescue *p*-value. In all cases in which we will refer to the significance or the
689  log2FC of a single condition, like UPF1 or dKD_SMG6, we refer to these
690  meta-analysis computations. Next, we aimed at finding the genes that
691  complied with our definition of NMD target: gene reacting to UPF1 and at least
692  one between SMG6 and SMG7. We therefore combined all conditions in a
693  single list of significant results, using a set of *p*-value meta-analysis methods
694  (Supplemental Table S2). The results from SMG6 and dKD_SMG6 were
695  combined with Fisher's method in a single meta_SMG6 score. The same
696  comparison was done for SMG7. A meta_SMGs significance score was then
697  computed with a sum of *p*-value from meta_SMG6 and meta_SMG7. The final
698  significance parameter used to determine the list of most significant NMD
699  targets was calculated with a Fisher's method from meta_SMGs and
700  UPF1_FDR (meta_meta).


701  **Differential expression simulations**

702  To rigorously compare two perturbed transcriptional conditions we performed
703  a number of simulations. First, we recreated a theoretical dataset in which the
704  transcriptional changes were based on the differences observed between

705 SMG6 rescue and the dKD. We generated gene counts by sampling from a
706 negative binomial distribution whose parameters were estimated with
707 DESeq2. This simulation was meant to produce a hypothetical SMG7 rescue
708 condition that behaved in the same exact way as SMG6 rescue. The only
709 exception was that the intensity of the log2FC was decreased by a factor of
710 0.57, which was the difference we estimated in SMG6 and SMG7 intensity
711 from a linear model of the real data. Since this initial simulation showed a
712 correlation with SMG6 rescue log2FC slightly higher than the observed one
713 for SMG7, we executed a series of additional simulations to which we added
714 more variability. The expected log2FC were multiplied by a confounding factor
715 of different intensities. This method allowed us to estimate the percent level of
716 specificity between the transcriptional effects of SMG6 and SMG7.

## DATA AVAILABILITY

718 All sequencing data from this study are available on the Gene Expression
719 Omnibus (GEO) with the id: GSE86148. In an attempt to allow complete
720 reproducible research, all scripts used in this work are available online on
721 GitHub at the address: https://github.com/Martombo/NMDseq.

## ACKNOWLEDGMENTS

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974

975 **FIGURE LEGENDS**

976

977 **FIGURE 1.**

978 Monitoring of UPF1, SMG6 and SMG7 knockdown (KD), double knockdown

979 (dKD) and rescue experiments compared to a control knockdown (Ctrl).

980 A) Lysates corresponding to $2x10^5$ cell equivalents of HeLa cells transiently

981 transfected with the indicated knockdown and rescue constructs were

982 analyzed by western blotting. After electrophoretic separation of the proteins

983 on 10% SDS-PAGE and transfer to nitrocellulose membranes, membrane

984 sections were incubated with antibodies against UPF1, SMG6, SMG7 and

985 CPSF73, the latter serving as loading control. The anti-SMG7 antibody gives

986 a double band of which only the upper band (arrow) corresponds to SMG7. B)

987 Relative mRNA levels of known endogenous NMD-targeted mRNAs (GAS5,

988 RP9P, SMG5, ATF4), normalized to β-actin mRNA (ACTB), were determined

989 for all conditions 72 hours post transfection by RT-qPCR. Mean values and

990 standard deviations of three independent experiments are shown, with the

991 samples in the control knockdown (Ctrl) set to 1.0.

992

993 **FIGURE 2.**

994 High overlap of putative NMD targets identified in the different conditions.

995 A) Bar plot displaying the results of a k-means clustering procedure performed

996 on the log2FC measured for each gene in the indicated conditions. The

997 number of clusters was set to two. The y-axis shows the average log2FC of all

998 the genes present in cluster 1 (comprising 40'000 genes) and cluster 2 (4'000

999 genes). KD refers to the log2FC (KD/Ctrl) and rescue to the log2FC

1000 (rescue/KD), respectively. The double knockdown of SMG6 and SMG7 (dKD)

1001 was rescued either with SMG6 (dKD SMG6 rescue) or with SMG7 (dKD

1002 SMG7 rescue). B) Histogram of genes with positive KD-rescue log2FC, which

1003 is expected from NMD targets. The top 1000 targets in each of our conditions

1004 were determined. Each of these sets corresponds to a cluster of columns in

1005 the plot. The y-axis shows the fraction of these targets that have a positive

1006 log2FC in the other conditions. The log2FC used for this analysis is the

1007 average between the log2FC (KD/Ctrl) and the inverse of the log2FC

1008    (rescue/KD) (see methods). C) Bar plots showing the RNA levels of the three

1009    factors under study upon each single KD.

1010

1011    **FIGURE 3.**

1012    SMG6 and SMG7 dKD and individual rescues reveal the highly redundant

1013    activity of these two NMD factors.

1014    A) Scatter plot comparing SMG6 and SMG7 rescues from the dKD. The

1015    picture shows the log2FC of the analysis of SMG6 rescue vs dKD (x-axis) and

1016    SMG7 rescue vs dKD (y-axis). Coloured in red are genes significantly

1017    downregulated in *either* of the two conditions. The histograms on the y-axis

1018    (on the right side) and x-axis (on the top) show the SMG7 log2FC distribution

1019    of all the significant downregulated targets in SMG6 rescue and *vice versa*,

1020    respectively. B) Simulation to estimate the variation between SMG6 and

1021    SMG7 results. Based on the negative binomial parameters computed from our

1022    data, new counts datasets were simulated. Additional variation was added to

1023    provide an accurate estimate of the individual difference between dKD_SMG6

1024    and dKD_SMG7 rescues. The picture compares the correlation scores (y-

1025    axis) found in the simulations at different levels of variation (x-axis) with the

1026    observed one in our dataset (black horizontal line).

1027

1028    **FIGURE 4.**

1029    NMD targets transcripts classified as non-coding, small-RNA host RNAs and

1030    products of pervasive transcription.

1031    A) Pie chart illustrating the top 1000 NMD targets categorized according to

1032    their biotype. 78% of these NMD targets code for protein, 9% are

1033    pseudogenes, 6% lincRNAs, and 4% antisense transcripts. B) The number of

1034    possible ORFs in non-coding RNAs correlates with the likelihood of

1035    undergoing NMD. The expressed non-coding genes are partitioned in different

1036    bins depending on how many theoretical ORFs can be predicted on their

1037    sequence. The y-axis reports the percentage of NMD targets (top 1000) of all

1038    the genes in each bin (e.g. 8% of genes with 3 ORFs are NMD targets). C)

1039    Top NMD targets are enriched in snoRNA and miRNA host genes. Each bar

1040    shows the percentage of genes that are among the NMD targets (top 1000) in

1041    every class of genes. D) Transcripts initiating upstream of the canonical

1042   transcription start site (TSS) are partially cleared by NMD. The number of

1043   reads upstream of every annotated TSS has been computed in the indicated

1044   conditions. This quantity was divided by the total counts of every gene and

1045   every condition was analyzed comparing KDs to Ctrl and rescues to KDs, as

1046   in the normal analysis (see methods). A box plot showing the log2FC of these

1047   quantities is displayed.

1048

1049   **FIGURE 5.**

1050   NMD targets are enriched in known and novel characteristic features.

1051   A) Bar plot showing the enrichment of mRNAs with introns in the 3' UTR and

1052   uORFs among NMD targets. All mRNAs with introns in the 3' UTR have been

1053   removed from the analysis of uORFs. B) Box plots comparing 3' UTR features

1054   between NMD targets and a matched control group. All mRNAs with 3' UTR

1055   introns or uORFs have been removed from the analysis. This resulted in a

1056   total of 660 significant isoforms. The control set is always a set of mRNAs that

1057   have the same expression levels of the NMD mRNAs in the Ctrl condition.

1058

A)

B)
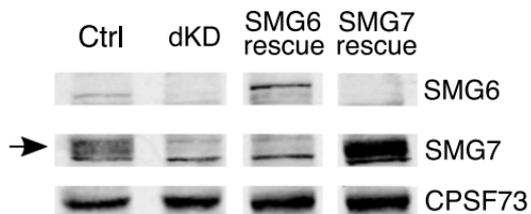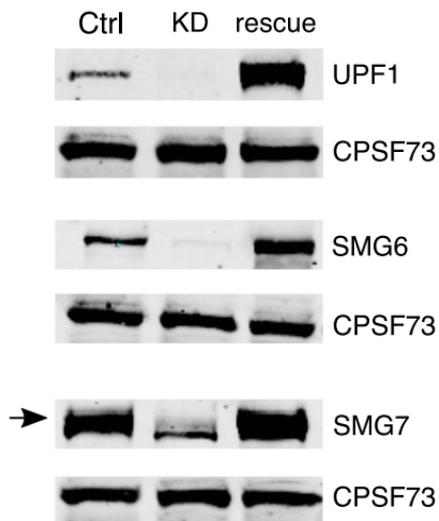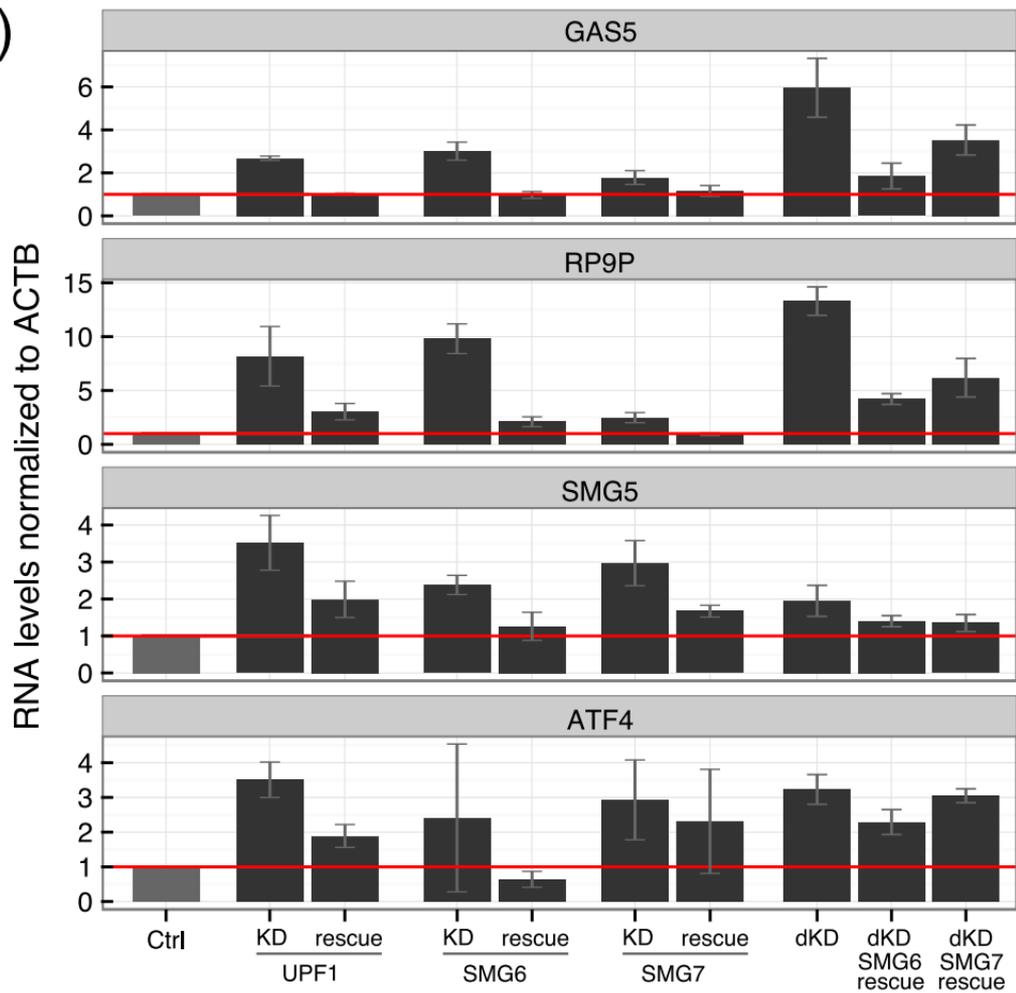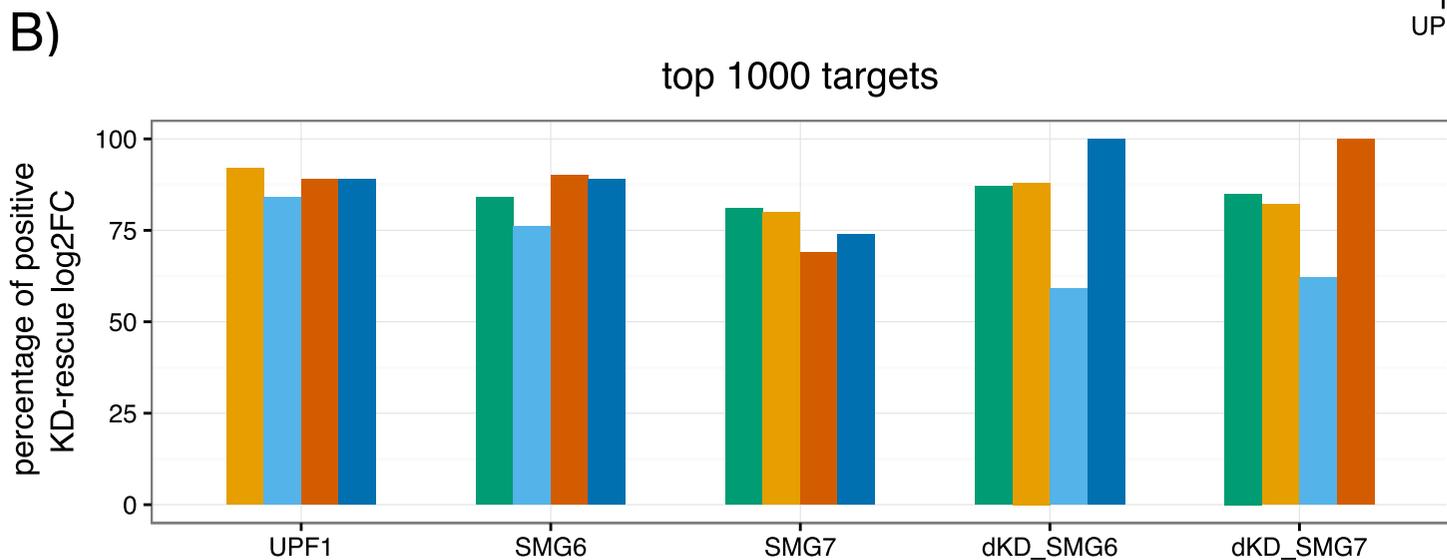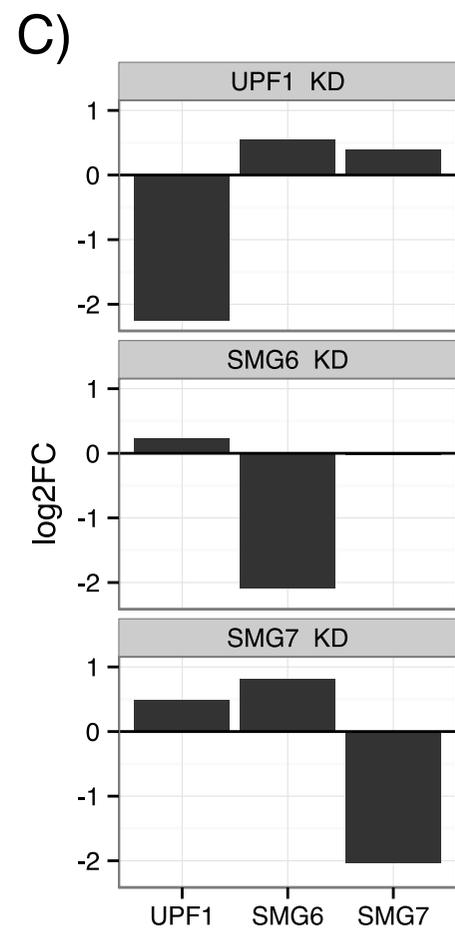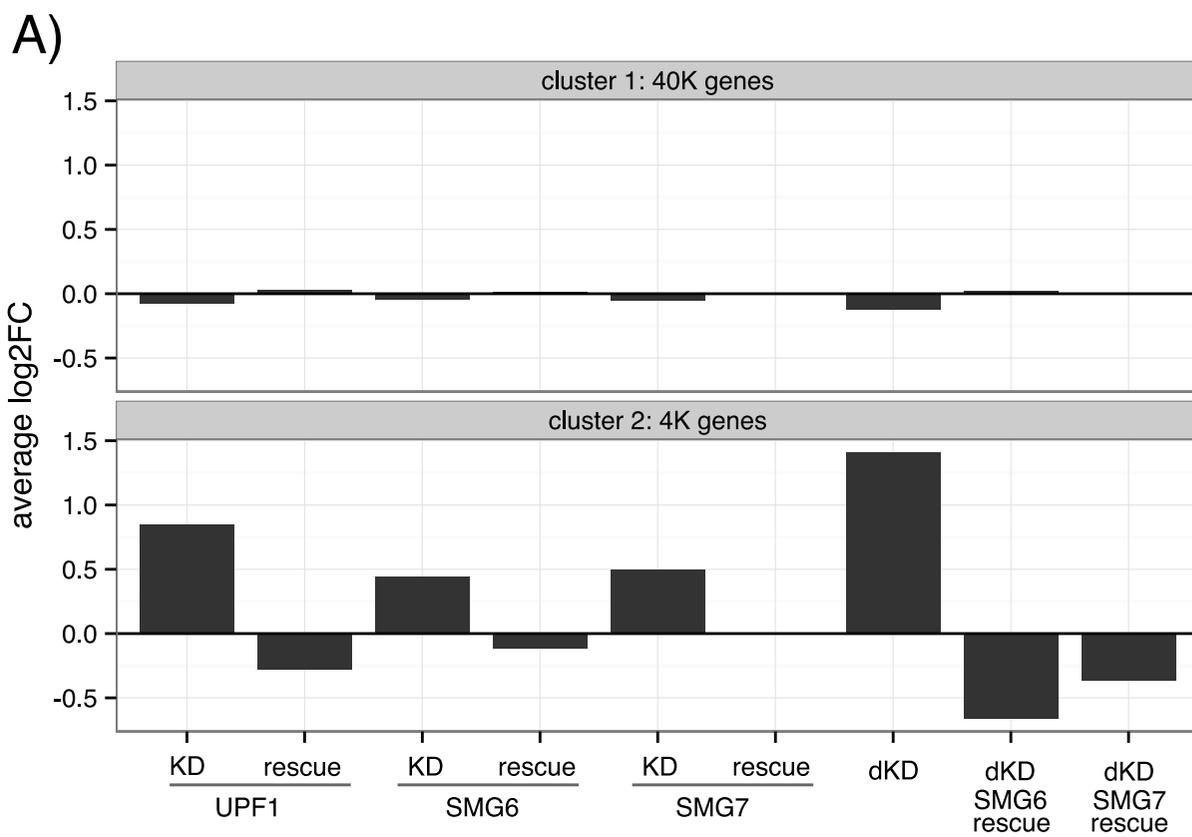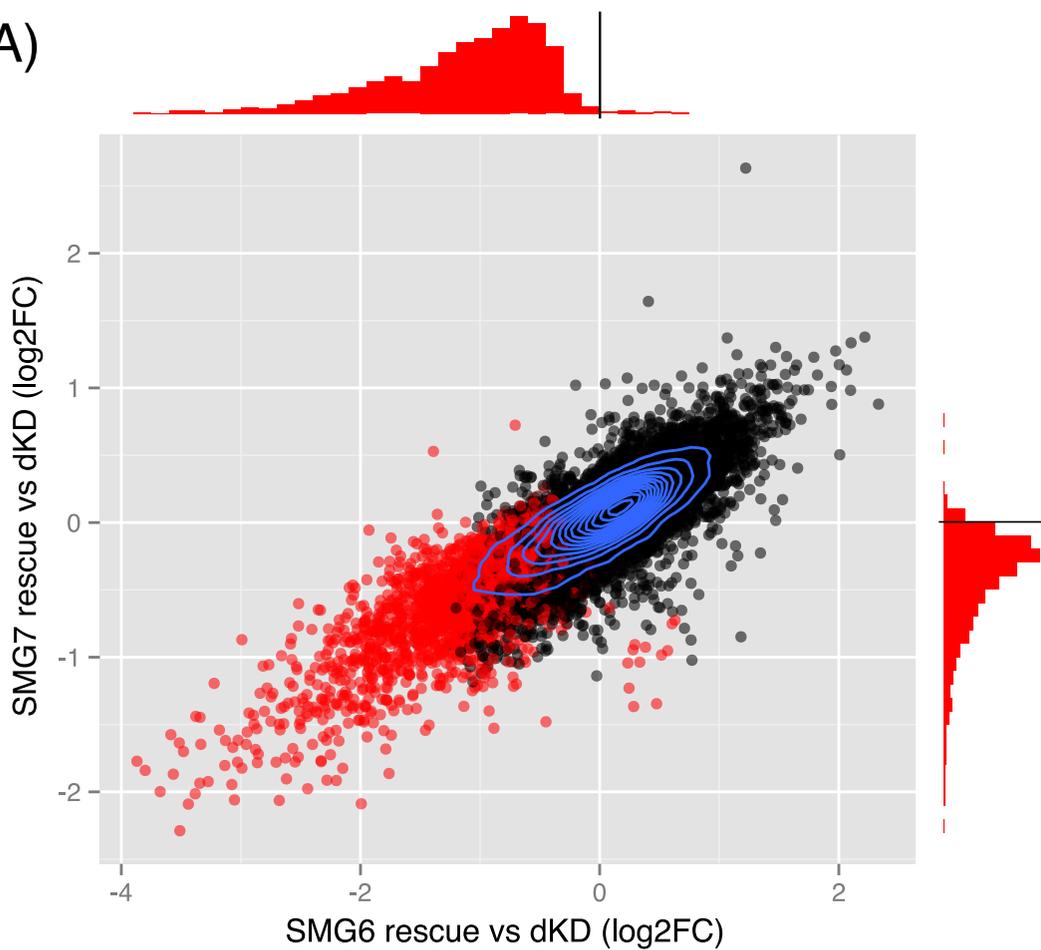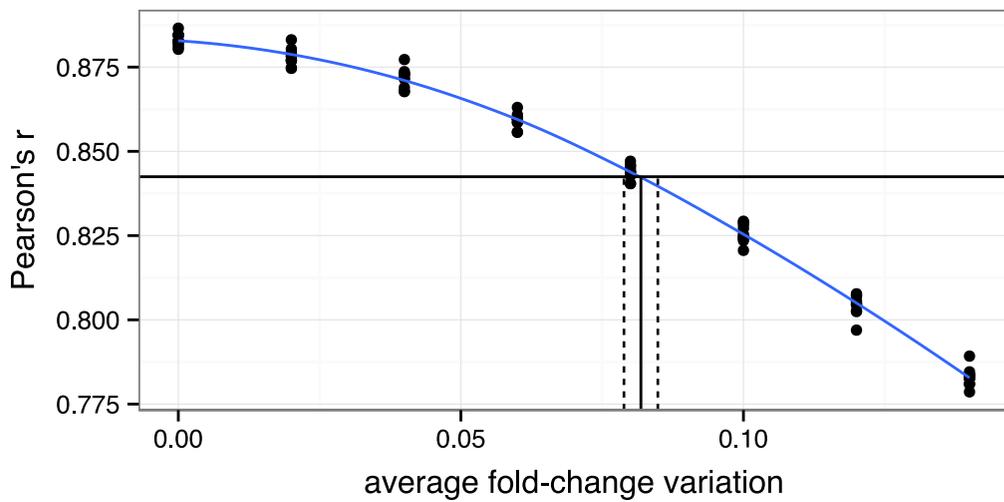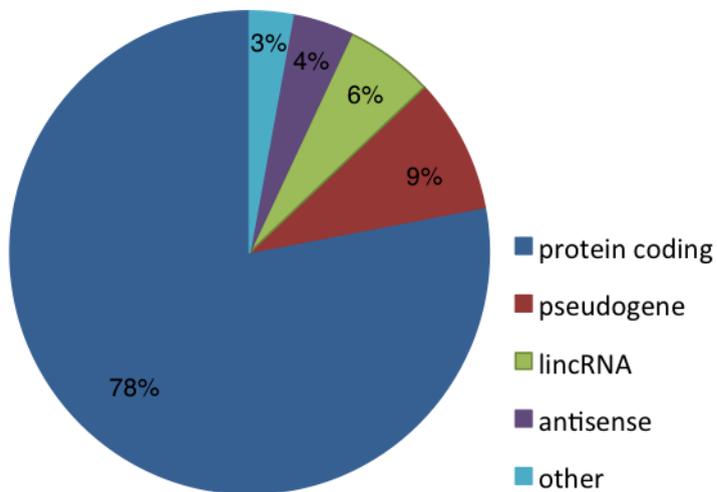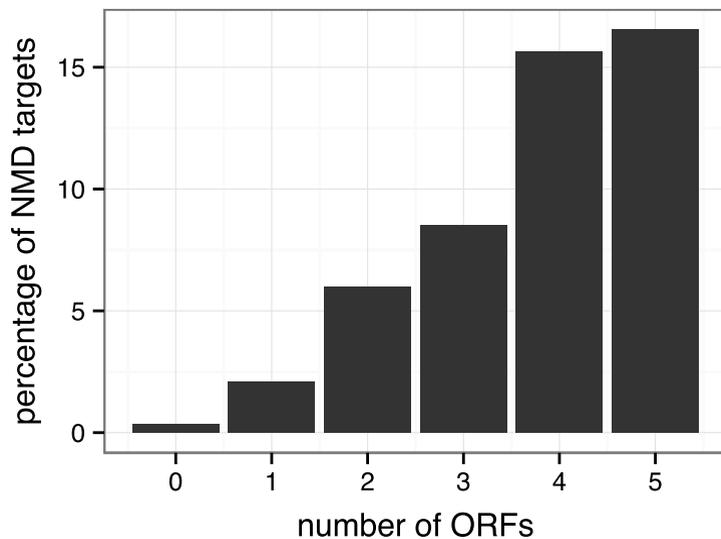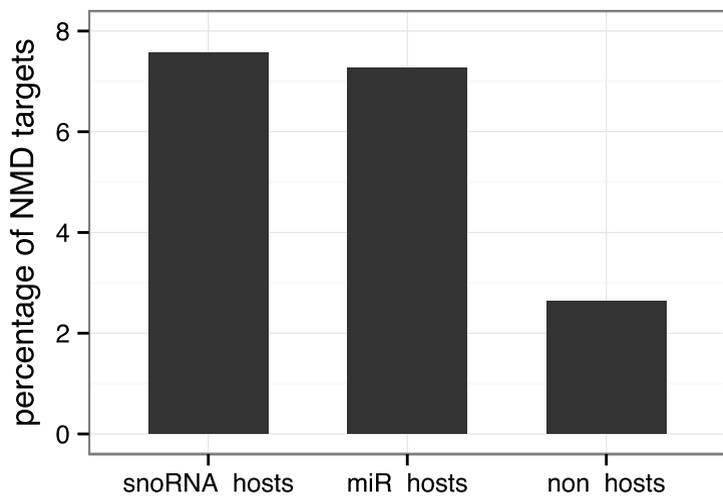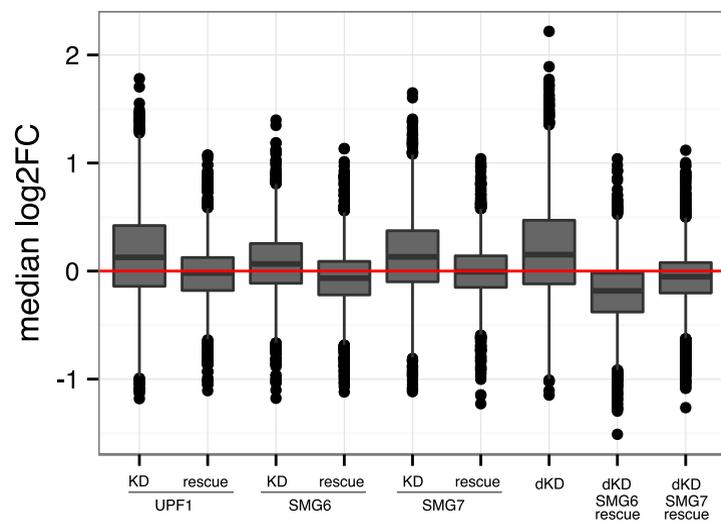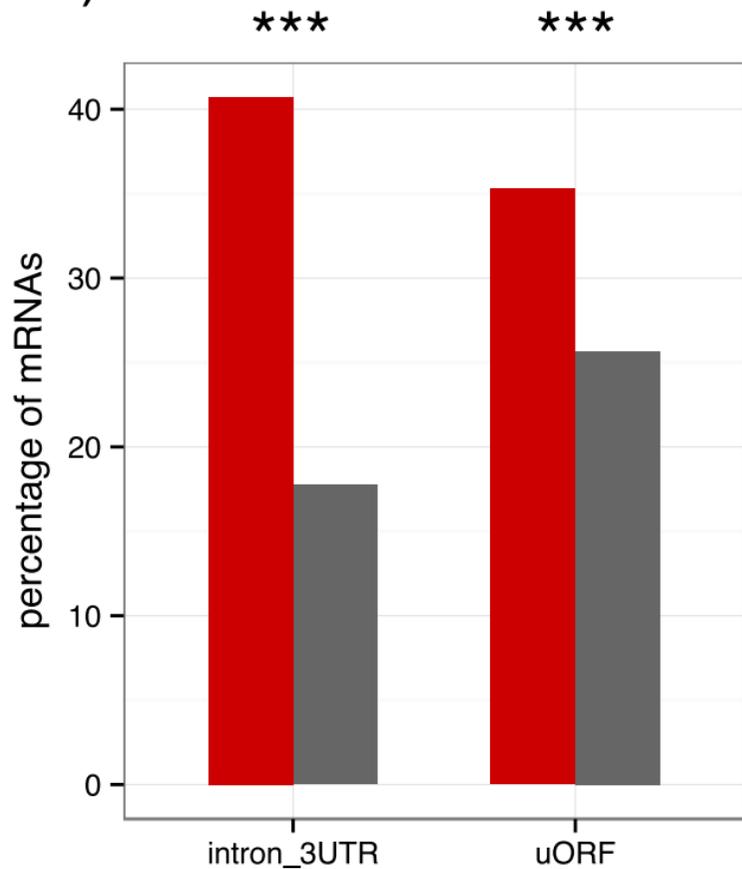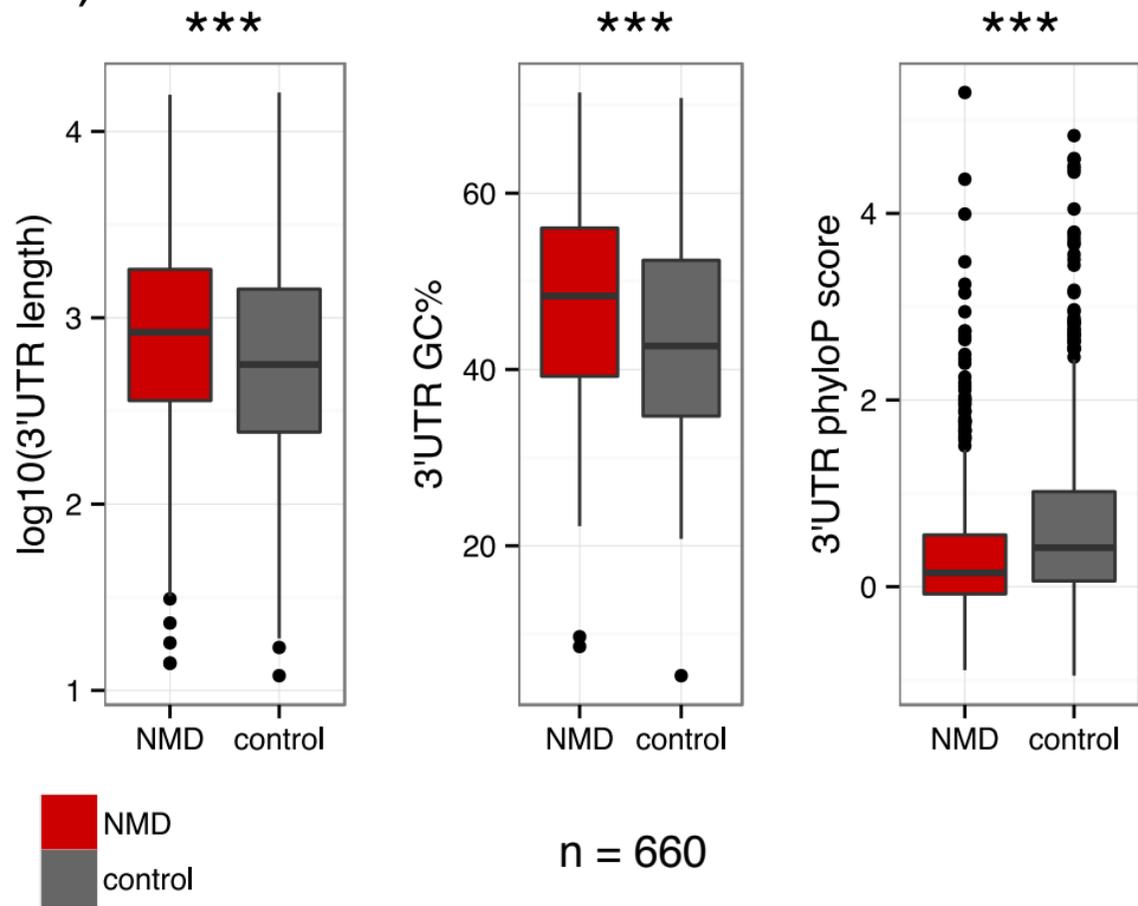
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# Transcriptome-wide identification of NMD-targeted human mRNAs reveals extensive redundancy between SMG6- and SMG7-mediated degradation pathways

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